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Chapter 1

Introduction

GLPK (GNU Linear Programming Kit) is a set of routines written in the ANSI C programming language and organized in the form of a callable library. It is intended for solving linear programming (LP), mixed integer programming (MIP), and other related problems.

1.1 LP problem

GLPK assumes the following formulation of linear programming (LP) problem:

minimize (or maximize)

\[ z = c_1 x_{m+1} + c_2 x_{m+2} + \ldots + c_n x_{m+n} + c_0 \]  

(1.1)

subject to linear constraints

\[ x_1 = a_{11} x_{m+1} + a_{12} x_{m+2} + \ldots + a_{1n} x_{m+n} \]
\[ x_2 = a_{21} x_{m+1} + a_{22} x_{m+2} + \ldots + a_{2n} x_{m+n} \]
\[ \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \]
\[ x_m = a_{m1} x_{m+1} + a_{m2} x_{m+2} + \ldots + a_{mn} x_{m+n} \]  

(1.2)

and bounds of variables

\[ l_1 \leq x_1 \leq u_1 \]
\[ l_2 \leq x_2 \leq u_2 \]
\[ \ldots \ldots \]
\[ l_{m+n} \leq x_{m+n} \leq u_{m+n} \]  

(1.3)
where: $x_1, x_2, \ldots, x_m$ are auxiliary variables; $x_{m+1}, x_{m+2}, \ldots, x_{m+n}$ are structural variables; $z$ is the objective function; $c_1, c_2, \ldots, c_n$ are objective coefficients; $c_0$ is the constant term ("shift") of the objective function; $a_{11}, a_{12}, \ldots, a_{mn}$ are constraint coefficients; $l_1, l_2, \ldots, l_{m+n}$ are lower bounds of variables; $u_1, u_2, \ldots, u_{m+n}$ are upper bounds of variables.

Auxiliary variables are also called rows, because they correspond to rows of the constraint matrix (i.e. a matrix built of the constraint coefficients). Similarly, structural variables are also called columns, because they correspond to columns of the constraint matrix.

Bounds of variables can be finite as well as infinite. Besides, lower and upper bounds can be equal to each other. Thus, the following types of variables are possible:

<table>
<thead>
<tr>
<th>Bounds of variable</th>
<th>Type of variable</th>
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<tr>
<td>$-\infty &lt; x_k &lt; +\infty$</td>
<td>Free (unbounded) variable</td>
</tr>
<tr>
<td>$l_k \leq x_k &lt; +\infty$</td>
<td>Variable with lower bound</td>
</tr>
<tr>
<td>$-\infty &lt; x_k \leq u_k$</td>
<td>Variable with upper bound</td>
</tr>
<tr>
<td>$l_k \leq x_k \leq u_k$</td>
<td>Double-bounded variable</td>
</tr>
<tr>
<td>$l_k = x_k = u_k$</td>
<td>Fixed variable</td>
</tr>
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Note that the types of variables shown above are applicable to structural as well as to auxiliary variables.

To solve the LP problem (1.1)—(1.3) is to find such values of all structural and auxiliary variables, which:

- satisfy to all the linear constraints (1.2), and
- are within their bounds (1.3), and
- provide the smallest (in case of minimization) or the largest (in case of maximization) value of the objective function (1.1).

### 1.2 MIP problem

*Mixed integer linear programming (MIP)* problem is LP problem in which some variables are additionally required to be integer.

GLPK assumes that MIP problem has the same formulation as ordinary (pure) LP problem (1.1)—(1.3), i.e. includes auxiliary and structural variables, which may have lower and/or upper bounds. However, in case of MIP problem some variables may be required to be integer. This additional constraint means that a value of each integer variable must be only integer number. (Should note that GLPK allows only structural variables to be of integer kind.)
1.3 Using the package

1.3.1 Brief example

In order to understand what GLPK is from the user’s standpoint, consider the following simple LP problem:

\[
\begin{align*}
\text{maximize} & \quad z = 10x_1 + 6x_2 + 4x_3 \\
\text{subject to} & \quad x_1 + x_2 + x_3 \leq 100 \\
& \quad 10x_1 + 4x_2 + 5x_3 \leq 600 \\
& \quad 2x_1 + 2x_2 + 6x_3 \leq 300 \\
\end{align*}
\]

where all variables are non-negative

\[
x_1 \geq 0, \ x_2 \geq 0, \ x_3 \geq 0
\]

At first this LP problem should be transformed to the standard form (1.1)—(1.3). This can be easily done by introducing auxiliary variables, by one for each original inequality constraint. Thus, the problem can be reformulated as follows:

\[
\begin{align*}
\text{maximize} & \quad z = 10x_1 + 6x_2 + 4x_3 \\
\text{subject to} & \quad p = x_1 + x_2 + x_3 \\
& \quad q = 10x_1 + 4x_2 + 5x_3 \\
& \quad r = 2x_1 + 2x_2 + 6x_3 \\
\end{align*}
\]

and bounds of variables

\[
\begin{align*}
\infty < p \leq 100 & \quad 0 \leq x_1 < +\infty \\
\infty < q \leq 600 & \quad 0 \leq x_2 < +\infty \\
\infty < r \leq 300 & \quad 0 \leq x_3 < +\infty \\
\end{align*}
\]

where \( p, q, r \) are auxiliary variables (rows), and \( x_1, x_2, x_3 \) are structural variables (columns).

The example C program shown below uses GLPK API routines in order to solve this LP problem.\(^1\)

---

\(^1\) If you just need to solve LP or MIP instance, you may write it in MPS or CPLEX LP format and then use the GLPK stand-alone solver to obtain a solution. This is much less time-consuming than programming in C with GLPK API routines.
/* sample.c */

#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
    glp_prob *lp;
    int ia[1+1000], ja[1+1000];
    double ar[1+1000], z, x1, x2, x3;
    s1: lp = glp_create_prob();
    s2: glp_set_prob_name(lp, "sample");
    s3: glp_set_obj_dir(lp, GLP_MAX);
    s4: glp_add_rows(lp, 3);
    s5: glp_set_row_name(lp, 1, "p");
    s6: glp_set_row_bnds(lp, 1, GLP_UP, 0.0, 100.0);
    s7: glp_set_row_name(lp, 2, "q");
    s8: glp_set_row_bnds(lp, 2, GLP_UP, 0.0, 600.0);
    s9: glp_set_row_name(lp, 3, "r");
    s10: glp_set_row_bnds(lp, 3, GLP_UP, 0.0, 300.0);
    s11: glp_add_cols(lp, 3);
    s12: glp_set_col_name(lp, 1, "x1");
    s13: glp_set_col_bnds(lp, 1, GLP_LO, 0.0, 0.0);
    s14: glp_set_obj_coef(lp, 1, 10.0);
    s15: glp_set_col_name(lp, 2, "x2");
    s16: glp_set_col_bnds(lp, 2, GLP_LO, 0.0, 0.0);
    s17: glp_set_obj_coef(lp, 2, 6.0);
    s18: glp_set_col_name(lp, 3, "x3");
    s19: glp_set_col_bnds(lp, 3, GLP_LO, 0.0, 0.0);
    s20: glp_set_obj_coef(lp, 3, 4.0);
    s30: glp_load_matrix(lp, 9, ia, ja, ar);
The statement \texttt{s1} creates a problem object. Being created the object is initially empty. The statement \texttt{s2} assigns a symbolic name to the problem object.

The statement \texttt{s3} calls the routine \texttt{glp_set_obj_dir} in order to set the optimization direction flag, where \texttt{GLP_MAX} means maximization.

The statement \texttt{s4} adds three rows to the problem object.

The statement \texttt{s5} assigns the symbolic name '\texttt{p}' to the first row, and the statement \texttt{s6} sets the type and bounds of the first row, where \texttt{GLP_UP} means that the row has an upper bound. The statements \texttt{s7}, \texttt{s8}, \texttt{s9}, \texttt{s10} are used in the same way in order to assign the symbolic names '\texttt{q}' and '\texttt{r}' to the second and third rows and set their types and bounds.

The statement \texttt{s11} adds three columns to the problem object.

The statement \texttt{s12} assigns the symbolic name '\texttt{x1}' to the first column, the statement \texttt{s13} sets the type and bounds of the first column, where \texttt{GLP_LO} means that the column has an lower bound, and the statement \texttt{s14} sets the objective coefficient for the first column. The statements \texttt{s15}—\texttt{s20} are used in the same way in order to assign the symbolic names '\texttt{x2}' and '\texttt{x3}' to the second and third columns and set their types, bounds, and objective coefficients.

The statements \texttt{s21}—\texttt{s29} prepare non-zero elements of the constraint matrix (i.e. constraint coefficients). Row indices of each element are stored in the array \texttt{ia}, column indices are stored in the array \texttt{ja}, and numerical values of corresponding elements are stored in the array \texttt{ar}. Then the statement \texttt{s30} calls the routine \texttt{glp_load_matrix}, which loads information from these three arrays into the problem object.

Now all data have been entered into the problem object, and therefore the statement \texttt{s31} calls the routine \texttt{glp_simplex}, which is a driver to the

```c
s31:  glp_simplex(lp, NULL);
s32:  z = glp_get_obj_val(lp);
s33:  x1 = glp_get_col_prim(lp, 1);
s34:  x2 = glp_get_col_prim(lp, 2);
s35:  x3 = glp_get_col_prim(lp, 3);
s36:  printf("\nz = %g; x1 = %g; x2 = %g; x3 = %g\n",
          z, x1, x2, x3);
s37:  glp_delete_prob(lp);
    return 0;
}
/* eof */
```
simplex method, in order to solve the LP problem. This routine finds an optimal solution and stores all relevant information back into the problem object.

The statement s32 obtains a computed value of the objective function, and the statements s33—s35 obtain computed values of structural variables (columns), which correspond to the optimal basic solution found by the solver.

The statement s36 writes the optimal solution to the standard output. The printout may look like follows:

```
* 0: objval = 0.000000000e+00 infeas = 0.000000000e+00 (0)
* 2: objval = 7.333333333e+02 infeas = 0.000000000e+00 (0)
OPTIMAL SOLUTION FOUND
```

z = 733.333; x1 = 33.3333; x2 = 66.6667; x3 = 0

Finally, the statement s37 calls the routine glp_delete_prob, which frees all the memory allocated to the problem object.

### 1.3.2 Compiling

The GLPK package has the only header file glpk.h, which should be available on compiling a C (or C++) program using GLPK API routines.

If the header file is installed in the default location /usr/local/include, the following typical command may be used to compile, say, the example C program described above with the GNU C compiler:

```
$ gcc -c sample.c
```

If glpk.h is not in the default location, the corresponding directory containing it should be made known to the C compiler through -I option, for example:

```
$ gcc -I/foo/bar/glpk-4.15/include -c sample.c
```

In any case the compilation results in an object file sample.o.

### 1.3.3 Linking

The GLPK library is a single file libglpk.a. (On systems which support shared libraries there may be also a shared version of the library libglpk.so.)
If the library is installed in the default location `/usr/local/lib`, the following typical command may be used to link, say, the example C program described above against with the library:

```bash
$ gcc sample.o -lglpk -lm
```

If the GLPK library is not in the default location, the corresponding directory containing it should be made known to the linker through `-L` option, for example:

```bash
$ gcc -L/foo/bar/glpk-4.15 sample.o -lglpk -lm
```

Depending on configuration of the package linking against with the GLPK library may require the following optional libraries:

- `lgmp` the GNU MP bignum library;
- `lz` the zlib data compression library;
- `lltdl` the GNU ltdl shared support library.

in which case corresponding libraries should be also made known to the linker, for example:

```bash
$ gcc sample.o -lglpk -lz -lltdl -lm
```

For more details about configuration options of the GLPK package see Appendix A, page 185.
Chapter 2

Basic API Routines

This chapter describes GLPK API routines intended for using in application programs.

Library header

All GLPK API data types and routines are defined in the header file `glpk.h`. It should be included in all source files which use GLPK API, either directly or indirectly through some other header file as follows:

```c
#include <glpk.h>
```

Error handling

If some GLPK API routine detects erroneous or incorrect data passed by the application program, it writes appropriate diagnostic messages to the terminal and then abnormally terminates the application program. In most practical cases this allows to simplify programming by avoiding numerous checks of return codes. Thus, in order to prevent crashing the application program should check all data, which are suspected to be incorrect, before calling GLPK API routines.

Should note that this kind of error handling is used only in cases of incorrect data passed by the application program. If, for example, the application program calls some GLPK API routine to read data from an input file and these data are incorrect, the GLPK API routine reports about error in the usual way by means of the return code.
Thread safety
Currently GLPK API routines are non-reentrant and therefore cannot be used in multi-threaded programs.

Array indexing
Normally all GLPK API routines start array indexing from 1, not from 0 (except the specially stipulated cases). This means, for example, that if some vector $x$ of the length $n$ is passed as an array to some GLPK API routine, the latter expects vector components to be placed in locations $x[1]$, $x[2]$, $\ldots$, $x[n]$, and the location $x[0]$ normally is not used.

In order to avoid indexing errors it is most convenient and most reliable to declare the array $x$ as follows:

\[
\text{double } x[1+n];
\]

or to allocate it as follows:

\[
\begin{align*}
\text{double } &*x; \\
\ldots \\
&x = \text{calloc}(1+n, \text{sizeof(double))};
\end{align*}
\]

In both cases one extra location $x[0]$ is reserved that allows passing the array to GLPK routines in a usual way.

2.1 Problem object
All GLPK API routines deal with so called problem object, which is a program object of type glp_prob and intended to represent a particular LP or MIP instance.

The type glp_prob is a data structure declared in the header file glpk.h as follows:

\[
\text{typedef struct } \{ \ldots \} \text{ glp_prob;}
\]

Problem objects (i.e. program objects of the glp_prob type) are allocated and managed internally by the GLPK API routines. The application program should never use any members of the glp_prob structure directly and should deal only with pointers to these objects (that is, glp_prob * values).
The problem object consists of five segments, which are:

- problem segment,
- basis segment,
- interior point segment,
- MIP segment, and
- control parameters and statistics segment.

**Problem segment**

The *problem segment* contains original LP/MIP data, which corresponds to the problem formulation (1.1)—(1.3) (see Section 1.1, page 12). It includes the following components:

- rows (auxiliary variables),
- columns (structural variables),
- objective function, and
- constraint matrix.

Rows and columns have the same set of the following attributes:

- ordinal number,
- symbolic name (1 up to 255 arbitrary graphic characters),
- type (free, lower bound, upper bound, double bound, fixed),
- numerical values of lower and upper bounds,
- scale factor.

*Ordinal numbers* are intended for referencing rows and columns. Row ordinal numbers are integers 1, 2, ..., $m$, and column ordinal numbers are integers 1, 2, ..., $n$, where $m$ and $n$ are, respectively, the current number of rows and columns in the problem object.

*Symbolic names* are intended for informational purposes. They also can be used for referencing rows and columns.

*Types and bounds* of rows (auxiliary variables) and columns (structural variables) are explained above (see Section 1.1, page 12).

*Scale factors* are used internally for scaling rows and columns of the constraint matrix.

Information about the *objective function* includes numerical values of objective coefficients and a flag, which defines the optimization direction (i.e. minimization or maximization).

The *constraint matrix* is a $m \times n$ rectangular matrix built of constraint coefficients $a_{ij}$, which defines the system of linear constraints (1.2) (see Section 1.1, page 12). This matrix is stored in the problem object in both row-wise and column-wise sparse formats.
Once the problem object has been created, the application program can access and modify any components of the problem segment in arbitrary order.

**Basis segment**

The basis segment of the problem object keeps information related to the current basic solution. It includes:

- row and column statuses,
- basic solution statuses,
- factorization of the current basis matrix, and
- basic solution components.

The row and column statuses define which rows and columns are basic and which are non-basic. These statuses may be assigned either by the application program or by some API routines. Note that these statuses are always defined independently on whether the corresponding basis is valid or not.

The basic solution statuses include the primal status and the dual status, which are set by the simplex-based solver once the problem has been solved. The primal status shows whether a primal basic solution is feasible, infeasible, or undefined. The dual status shows the same for a dual basic solution.

The factorization of the basis matrix is some factorized form (like LU-factorization) of the current basis matrix (defined by the current row and column statuses). The factorization is used by the simplex-based solver and kept when the solver terminates the search. This feature allows efficiently reoptimizing the problem after some modifications (for example, after changing some bounds or objective coefficients). It also allows performing the post-optimal analysis (for example, computing components of the simplex table, etc.).

The basic solution components include primal and dual values of all auxiliary and structural variables for the most recently obtained basic solution.

**Interior point segment**

The interior point segment is automatically allocated after the problem has been solved using the interior point solver. It contains interior point solution components, which include the solution status, and primal and dual values of all auxiliary and structural variables.
MIP segment

The *MIP segment* is used only for MIP problems. This segment includes:

- column kinds,
- MIP solution status, and
- MIP solution components.

The *column kinds* define which columns (i.e. structural variables) are integer and which are continuous.

The *MIP solution status* is set by the MIP solver and shows whether a MIP solution is integer optimal, integer feasible (non-optimal), or undefined.

The *MIP solution components* are computed by the MIP solver and include primal values of all auxiliary and structural variables for the most recently obtained MIP solution.

Note that in case of MIP problem the basis segment corresponds to the optimal solution of LP relaxation, which is also available to the application program.

Currently the search tree is not kept in the MIP segment. Therefore if the search has been terminated, it cannot be continued.
2.2 Problem creating and modifying routines

2.2.1 glp_create_prob—create problem object

Synopsis

```c
glp_prob *glp_create_prob(void);
```

Description

The routine `glp_create_prob` creates a new problem object, which initially is “empty”, i.e. has no rows and columns.

Returns

The routine returns a pointer to the created object, which should be used in any subsequent operations on this object.

2.2.2 glp_set_prob_name—assign (change) problem name

Synopsis

```c
void glp_set_prob_name(glp_prob *lp, const char *name);
```

Description

The routine `glp_set_prob_name` assigns a given symbolic name (1 up to 255 characters) to the specified problem object.

If the parameter `name` is NULL or empty string, the routine erases an existing symbolic name of the problem object.

2.2.3 glp_set_obj_name—assign (change) objective function name

Synopsis

```c
void glp_set_obj_name(glp_prob *lp, const char *name);
```

Description

The routine `glp_set_obj_name` assigns a given symbolic name (1 up to 255 characters) to the objective function of the specified problem object.

If the parameter `name` is NULL or empty string, the routine erases an existing symbolic name of the objective function.
2.2.4 glp_set_obj_dir—set (change) optimization direction flag

Synopsis

void glp_set_obj_dir(glp_prob *lp, int dir);

Description

The routine glp_set_obj_dir sets (changes) the optimization direction flag (i.e. “sense” of the objective function) as specified by the parameter dir:

GLP_MIN minimization;
GLP_MAX maximization.

(Note that by default the problem is minimization.)

2.2.5 glp_add_rows—add new rows to problem object

Synopsis

int glp_add_rows(glp_prob *lp, int nrs);

Description

The routine glp_add_rows adds nrs rows (constraints) to the specified problem object. New rows are always added to the end of the row list, so the ordinal numbers of existing rows are not changed.

Being added each new row is initially free (unbounded) and has empty list of the constraint coefficients.

Returns

The routine glp_add_rows returns the ordinal number of the first new row added to the problem object.
2.2.6 glp_add_cols—add new columns to problem object

Synopsis

int glp_add_cols(glp_prob *lp, int ncs);

Description

The routine glp_add_cols adds ncs columns (structural variables) to the specified problem object. New columns are always added to the end of the column list, so the ordinal numbers of existing columns are not changed.

Being added each new column is initially fixed at zero and has empty list of the constraint coefficients.

Returns

The routine glp_add_cols returns the ordinal number of the first new column added to the problem object.

2.2.7 glp_set_row_name—assign (change) row name

Synopsis

void glp_set_row_name(glp_prob *lp, int i, const char *name);

Description

The routine glp_set_row_name assigns a given symbolic name (1 up to 255 characters) to i-th row (auxiliary variable) of the specified problem object.

If the parameter name is NULL or empty string, the routine erases an existing name of i-th row.

2.2.8 glp_set_col_name—assign (change) column name

Synopsis

void glp_set_col_name(glp_prob *lp, int j, const char *name);

Description

The routine glp_set_col_name assigns a given symbolic name (1 up to 255 characters) to j-th column (structural variable) of the specified problem object.
If the parameter name is NULL or empty string, the routine erases an existing name of j-th column.

2.2.9 glp_set_row_bnds—set (change) row bounds

Synopsis

void glp_set_row_bnds(glp_prob *lp, int i, int type, double lb, double ub);

Description

The routine glp_set_row_bnds sets (changes) the type and bounds of i-th row (auxiliary variable) of the specified problem object.

The parameters type, lb, and ub specify the type, lower bound, and upper bound, respectively, as follows:

<table>
<thead>
<tr>
<th>Type</th>
<th>Bounds</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>GLP_FR</td>
<td>$-\infty &lt; x &lt; +\infty$</td>
<td>Free (unbounded) variable</td>
</tr>
<tr>
<td>GLP_LO</td>
<td>$lb \leq x &lt; +\infty$</td>
<td>Variable with lower bound</td>
</tr>
<tr>
<td>GLP_UP</td>
<td>$-\infty &lt; x \leq ub$</td>
<td>Variable with upper bound</td>
</tr>
<tr>
<td>GLP_DB</td>
<td>$lb \leq x \leq ub$</td>
<td>Double-bounded variable</td>
</tr>
<tr>
<td>GLP_FX</td>
<td>$lb = x = ub$</td>
<td>Fixed variable</td>
</tr>
</tbody>
</table>

where $x$ is the auxiliary variable associated with i-th row.

If the row has no lower bound, the parameter lb is ignored. If the row has no upper bound, the parameter ub is ignored. If the row is an equality constraint (i.e. the corresponding auxiliary variable is of fixed type), only the parameter lb is used while the parameter ub is ignored.

Being added to the problem object each row is initially free, i.e. its type is GLP_FR.
2.2.10  glp_set_col_bnds—set (change) column bounds

Synopsis

void glp_set_col_bnds(glp_prob *lp, int j, int type,
                      double lb, double ub);

Description

The routine glp_set_col_bnds sets (changes) the type and bounds of j-th column (structural variable) of the specified problem object.

The parameters type, lb, and ub specify the type, lower bound, and upper bound, respectively, as follows:

<table>
<thead>
<tr>
<th>Type</th>
<th>Bounds</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>GLP_FR</td>
<td>$-\infty &lt; x &lt; +\infty$</td>
<td>Free (unbounded) variable</td>
</tr>
<tr>
<td>GLP_LO</td>
<td>$lb \leq x &lt; +\infty$</td>
<td>Variable with lower bound</td>
</tr>
<tr>
<td>GLP_UP</td>
<td>$-\infty &lt; x \leq ub$</td>
<td>Variable with upper bound</td>
</tr>
<tr>
<td>GLP_DB</td>
<td>$lb \leq x \leq ub$</td>
<td>Double-bounded variable</td>
</tr>
<tr>
<td>GLP_FX</td>
<td>$lb = x = ub$</td>
<td>Fixed variable</td>
</tr>
</tbody>
</table>

where $x$ is the structural variable associated with $j$-th column.

If the column has no lower bound, the parameter lb is ignored. If the column has no upper bound, the parameter ub is ignored. If the column is of fixed type, only the parameter lb is used while the parameter ub is ignored.

Being added to the problem object each column is initially fixed at zero, i.e. its type is GLP_FX and both bounds are 0.

2.2.11  glp_set_obj_coef—set (change) objective coefficient or constant term

Synopsis

void glp_set_obj_coef(glp_prob *lp, int j, double coef);

Description

The routine glp_set_obj_coef sets (changes) the objective coefficient at j-th column (structural variable). A new value of the objective coefficient is specified by the parameter coef.

If the parameter j is 0, the routine sets (changes) the constant term (“shift”) of the objective function.
2.2.12 glp_set_mat_row—set (replace) row of the constraint matrix

Synopsis

void glp_set_mat_row(glp_prob *lp, int i, int len,
                     const int ind[], const double val[]);

Description

The routine glp_set_mat_row stores (replaces) the contents of i-th row of
the constraint matrix of the specified problem object.

Column indices and numerical values of new row elements must be placed
in locations ind[1],..., ind[len] and val[1],..., val[len], respectively,
where 0 \leq len \leq n is the new length of i-th row, n is the current number
of columns in the problem object. Elements with identical column indices
are not allowed. Zero elements are allowed, but they are not stored in the
constraint matrix.

If the parameter len is 0, the parameters ind and/or val can be specified
as NULL.

2.2.13 glp_set_mat_col—set (replace) column of the constraint matrix

Synopsis

void glp_set_mat_col(glp_prob *lp, int j, int len,
                     const int ind[], const double val[]);

Description

The routine glp_set_mat_col stores (replaces) the contents of j-th column
of the constraint matrix of the specified problem object.

Row indices and numerical values of new column elements must be placed
in locations ind[1],..., ind[len] and val[1],..., val[len], respectively,
where 0 \leq len \leq m is the new length of j-th column, m is the current
number of rows in the problem object. Elements with identical row indices
are not allowed. Zero elements are allowed, but they are not stored in the
constraint matrix.

If the parameter len is 0, the parameters ind and/or val can be specified
as NULL.
2.2.14 glp_load_matrix—load (replace) the whole constraint matrix

Synopsis

void glp_load_matrix(glp_prob *lp, int ne, const int ia[],
const int ja[], const double ar[]);

Description

The routine glp_load_matrix loads the constraint matrix passed in the arrays ia, ja, and ar into the specified problem object. Before loading the current contents of the constraint matrix is destroyed.

Constraint coefficients (elements of the constraint matrix) must be specified as triplets (ia[k], ja[k], ar[k]) for $k = 1, \ldots, ne$, where ia[k] is the row index, ja[k] is the column index, and ar[k] is a numeric value of corresponding constraint coefficient. The parameter ne specifies the total number of (non-zero) elements in the matrix to be loaded. Coefficients with identical indices are not allowed. Zero coefficients are allowed, however, they are not stored in the constraint matrix.

If the parameter ne is 0, the parameters ia, ja, and/or ar can be specified as NULL.

2.2.15 glp_check_dup—check for duplicate elements in sparse matrix

Synopsis

int glp_check_dup(int m, int n, int ne, const int ia[],
const int ja[]);

Description

The routine glp_check_dup checks for duplicate elements (that is, elements with identical indices) in a sparse matrix specified in the coordinate format.

The parameters m and n specifies, respectively, the number of rows and columns in the matrix, $m \geq 0$, $n \geq 0$.

The parameter ne specifies the number of (structurally) non-zero elements in the matrix, $ne \geq 0$.

Elements of the matrix are specified as doublets $(ia[k], ja[k])$ for $k = 1, \ldots, ne$, where $ia[k]$ is a row index, $ja[k]$ is a column index.
The routine glp_check_dup can be used prior to a call to the routine glp_load_matrix to check that the constraint matrix to be loaded has no duplicate elements.

Returns
The routine glp_check_dup returns one of the following values:
- 0 — the matrix has no duplicate elements;
- \(-k\) — indices \(ia[k]\) or/and \(ja[k]\) are out of range;
- \(+k\) — element \((ia[k], ja[k])\) is duplicate.

2.2.16 glp_sort_matrix—sort elements of the constraint matrix

Synopsis

void glp_sort_matrix(glp_prob *P);

Description
The routine glp_sort_matrix sorts elements of the constraint matrix rebuilding its row and column linked lists. On exit from the routine the constraint matrix is not changed, however, elements in the row linked lists become ordered by ascending column indices, and the elements in the column linked lists become ordered by ascending row indices.

2.2.17 glp_del_rows—delete rows from problem object

Synopsis

void glp_del_rows(glp_prob *lp, int nrs, const int num[]);

Description
The routine glp_del_rows deletes rows from the specified problem object. Ordinal numbers of rows to be deleted should be placed in locations \(num[1], \ldots, num[nrs]\), where \(nrs > 0\).

Note that deleting rows involves changing ordinal numbers of other rows remaining in the problem object. New ordinal numbers of the remaining rows are assigned under the assumption that the original order of rows is not changed. Let, for example, before deletion there be five rows \(a, b, c, d,\) \(e\) with ordinal numbers \(1, 2, 3, 4, 5\), and let rows \(b\) and \(d\) have been deleted.
Then after deletion the remaining rows \( a, c, e \) are assigned new ordinal numbers 1, 2, 3.

### 2.2.18  glp_del_cols—delete columns from problem object

**Synopsis**

```c
void glp_del_cols(glp_prob *lp, int ncs, const int num[]);
```

**Description**

The routine `glp_del_cols` deletes columns from the specified problem object. Ordinal numbers of columns to be deleted should be placed in locations `num[1]`, ..., `num[ncs]`, where `ncs > 0`.

Note that deleting columns involves changing ordinal numbers of other columns remaining in the problem object. New ordinal numbers of the remaining columns are assigned under the assumption that the original order of columns is not changed. Let, for example, before deletion there be six columns \( p, q, r, s, t, u \) with ordinal numbers 1, 2, 3, 4, 5, 6, and let columns \( p, q, s \) have been deleted. Then after deletion the remaining columns \( r, t, u \) are assigned new ordinal numbers 1, 2, 3.

### 2.2.19  glp_copy_prob—copy problem object content

**Synopsis**

```c
void glp_copy_prob(glp_prob *dest, glp_prob *prob, int names);
```

**Description**

The routine `glp_copy_prob` copies the content of the problem object `prob` to the problem object `dest`.

The parameter `names` is a flag. If it is GLP_ON, the routine also copies all symbolic names; otherwise, if it is GLP_OFF, no symbolic names are copied.
2.2.20  glp_erase_prob—erase problem object content

Synopsis

void glp_erase_prob(glp_prob *lp);

Description

The routine glp_erase_prob erases the content of the specified problem object. The effect of this operation is the same as if the problem object would be deleted with the routine glp_delete_prob and then created anew with the routine glp_create_prob, with the only exception that the handle (pointer) to the problem object remains valid.

2.2.21  glp_delete_prob—delete problem object

Synopsis

void glp_delete_prob(glp_prob *lp);

Description

The routine glp_delete_prob deletes a problem object, which the parameter lp points to, freeing all the memory allocated to this object.
2.3 Problem retrieving routines

2.3.1 glp_get_prob_name—retrieve problem name

Synopsis

const char *glp_get_prob_name(glp_prob *lp);

Returns

The routine glp_get_prob_name returns a pointer to an internal buffer, which contains symbolic name of the problem. However, if the problem has no assigned name, the routine returns NULL.

2.3.2 glp_get_obj_name—retrieve objective function name

Synopsis

const char *glp_get_obj_name(glp_prob *lp);

Returns

The routine glp_get_obj_name returns a pointer to an internal buffer, which contains symbolic name assigned to the objective function. However, if the objective function has no assigned name, the routine returns NULL.

2.3.3 glp_get_obj_dir—retrieve optimization direction flag

Synopsis

int glp_get_obj_dir(glp_prob *lp);

Returns

The routine glp_get_obj_dir returns the optimization direction flag (i.e. “sense” of the objective function):

- GLP_MIN minimization;
- GLP_MAX maximization.
2.3.4  glp_get_num_rows—retrieve number of rows

Synopsis

int glp_get_num_rows(glp_prob *lp);

Returns

The routine glp_get_num_rows returns the current number of rows in the specified problem object.

2.3.5  glp_get_num_cols—retrieve number of columns

Synopsis

int glp_get_num_cols(glp_prob *lp);

Returns

The routine glp_get_num_cols returns the current number of columns the specified problem object.

2.3.6  glp_get_row_name—retrieve row name

Synopsis

const char *glp_get_row_name(glp_prob *lp, int i);

Returns

The routine glp_get_row_name returns a pointer to an internal buffer, which contains a symbolic name assigned to i-th row. However, if the row has no assigned name, the routine returns NULL.

2.3.7  glp_get_col_name—retrieve column name

Synopsis

const char *glp_get_col_name(glp_prob *lp, int j);

Returns

The routine glp_get_col_name returns a pointer to an internal buffer, which contains a symbolic name assigned to j-th column. However, if the column has no assigned name, the routine returns NULL.
2.3.8  glp_get_row_type—retrieve row type

Synopsis

int glp_get_row_type(glp_prob *lp, int i);

Returns

The routine glp_get_row_type returns the type of i-th row, i.e. the type of corresponding auxiliary variable, as follows:

- GLP_FR  free (unbounded) variable;
- GLP_LO  variable with lower bound;
- GLP_UP  variable with upper bound;
- GLP_DB  double-bounded variable;
- GLP_FX  fixed variable.

2.3.9  glp_get_row_lb—retrieve row lower bound

Synopsis

double glp_get_row_lb(glp_prob *lp, int i);

Returns

The routine glp_get_row_lb returns the lower bound of i-th row, i.e. the lower bound of corresponding auxiliary variable. However, if the row has no lower bound, the routine returns -DBL_MAX.

2.3.10 glp_get_row_ub—retrieve row upper bound

Synopsis

double glp_get_row_ub(glp_prob *lp, int i);

Returns

The routine glp_get_row_ub returns the upper bound of i-th row, i.e. the upper bound of corresponding auxiliary variable. However, if the row has no upper bound, the routine returns +DBL_MAX.
2.3.11  glp_get_col_type—retrieve column type

Synopsis

int glp_get_col_type(glp_prob *lp, int j);

Returns

The routine glp_get_col_type returns the type of j-th column, i.e. the type of corresponding structural variable, as follows:

- GLP_FR free (unbounded) variable;
- GLP_LO variable with lower bound;
- GLP_UP variable with upper bound;
- GLP_DB double-bounded variable;
- GLP_FX fixed variable.

2.3.12  glp_get_col_lb—retrieve column lower bound

Synopsis

double glp_get_col_lb(glp_prob *lp, int j);

Returns

The routine glp_get_col_lb returns the lower bound of j-th column, i.e. the lower bound of corresponding structural variable. However, if the column has no lower bound, the routine returns -DBL_MAX.

2.3.13  glp_get_col_ub—retrieve column upper bound

Synopsis

double glp_get_col_ub(glp_prob *lp, int j);

Returns

The routine glp_get_col_ub returns the upper bound of j-th column, i.e. the upper bound of corresponding structural variable. However, if the column has no upper bound, the routine returns +DBL_MAX.
2.3.14  glp_get_obj_coef—retrieve objective coefficient or constant term

Synopsis

double glp_get_obj_coef(glp_prob *lp, int j);

Returns

The routine glp_get_obj_coef returns the objective coefficient at j-th structural variable (column).

If the parameter j is 0, the routine returns the constant term (“shift”) of the objective function.

2.3.15  glp_get_num_nz—retrieve number of constraint coefficients

Synopsis

int glp_get_num_nz(glp_prob *lp);

Returns

The routine glp_get_num_nz returns the number of non-zero elements in the constraint matrix of the specified problem object.

2.3.16  glp_get_mat_row—retrieve row of the constraint matrix

Synopsis

int glp_get_mat_row(glp_prob *lp, int i, int ind[],
 double val[]);

Description

The routine glp_get_mat_row scans (non-zero) elements of i-th row of the constraint matrix of the specified problem object and stores their column indices and numeric values to locations ind[1],..., ind[len] and val[1],..., val[len], respectively, where 0 ≤ len ≤ n is the number of elements in i-th row, n is the number of columns.

The parameter ind and/or val can be specified as NULL, in which case corresponding information is not stored.
Returns

The routine \texttt{glp\_get\_mat\_row} returns the length \texttt{len}, i.e. the number of (non-zero) elements in \texttt{i}-th row.

2.3.17 \texttt{glp\_get\_mat\_col}—retrieve column of the constraint matrix

Synopsis

\begin{verbatim}
int glp_get_mat_col(glp_prob *lp, int j, int ind[],
                  double val[]);
\end{verbatim}

Description

The routine \texttt{glp\_get\_mat\_col} scans (non-zero) elements of \texttt{j}-th column of the constraint matrix of the specified problem object and stores their row indices and numeric values to locations \texttt{ind[1]}, \ldots, \texttt{ind[len]} and \texttt{val[1]}, \ldots, \texttt{val[len]}, respectively, where \(0 \leq \texttt{len} \leq \texttt{m}\) is the number of elements in \texttt{j}-th column, \texttt{m} is the number of rows.

The parameter \texttt{ind} and/or \texttt{val} can be specified as \texttt{NULL}, in which case corresponding information is not stored.

Returns

The routine \texttt{glp\_get\_mat\_col} returns the length \texttt{len}, i.e. the number of (non-zero) elements in \texttt{j}-th column.
2.4 Row and column searching routines

2.4.1 glp_create_index—create the name index

Synopsis

void glp_create_index(glp_prob *lp);

Description

The routine glp_create_index creates the name index for the specified problem object. The name index is an auxiliary data structure, which is intended to quickly (i.e. for logarithmic time) find rows and columns by their names.

This routine can be called at any time. If the name index already exists, the routine does nothing.

2.4.2 glp_find_row—find row by its name

Synopsis

int glp_find_row(glp_prob *lp, const char *name);

Returns

The routine glp_find_row returns the ordinal number of a row, which is assigned (by the routine glp_set_row_name) the specified symbolic name. If no such row exists, the routine returns 0.

2.4.3 glp_find_col—find column by its name

Synopsis

int glp_find_col(glp_prob *lp, const char *name);

Returns

The routine glp_find_col returns the ordinal number of a column, which is assigned (by the routine glp_set_col_name) the specified symbolic name. If no such column exists, the routine returns 0.
2.4.4 glp_delete_index—delete the name index

Synopsis

void glp_delete_index(glp_prob *lp);

Description

The routine glp_delete_index deletes the name index previously created by the routine glp_create_index and frees the memory allocated to this auxiliary data structure.

This routine can be called at any time. If the name index does not exist, the routine does nothing.
2.5 Problem scaling routines

2.5.1 Background

In GLPK the scaling means a linear transformation applied to the constraint matrix to improve its numerical properties.\(^1\)

The main equality is the following:

\[
\tilde{A} = RAS,
\]

(2.1)

where \(A = (a_{ij})\) is the original constraint matrix, \(R = (r_{ii}) > 0\) is a diagonal matrix used to scale rows (constraints), \(S = (s_{jj}) > 0\) is a diagonal matrix used to scale columns (variables), \(\tilde{A}\) is the scaled constraint matrix.

From (2.1) it follows that in the scaled problem instance each original constraint coefficient \(a_{ij}\) is replaced by corresponding scaled constraint coefficient:

\[
\tilde{a}_{ij} = r_{ii}a_{ij}s_{jj}.
\]

(2.2)

Note that the scaling is performed internally and therefore transparently to the user. This means that on API level the user always deal with unscaled data.

Scale factors \(r_{ii}\) and \(s_{jj}\) can be set or changed at any time either directly by the application program in a problem specific way (with the routines \texttt{glp_set_rii} and \texttt{glp_set_sjj}), or by some API routines intended for automatic scaling.

2.5.2 \texttt{glp_set_rii}—set (change) row scale factor

Synopsis

\[
\text{void glp_set_rii(glp_prob *lp, int i, double rii);}\]

Description

The routine \texttt{glp_set_rii} sets (changes) the scale factor \(r_{ii}\) for \(i\)-th row of the specified problem object.

\(^1\)In many cases a proper scaling allows making the constraint matrix to be better conditioned, i.e. decreasing its condition number, that makes computations numerically more stable.
2.5.3  

**glp_set_sjj**—set (change) column scale factor

**Synopsis**

```c
void glp_set_sjj(glp_prob *lp, int j, double sjj);
```

**Description**

The routine `glp_set_sjj` sets (changes) the scale factor $s_{jj}$ for $j$-th column of the specified problem object.

2.5.4  

**glp_get_rii**—retrieve row scale factor

**Synopsis**

```c
double glp_get_rii(glp_prob *lp, int i);
```

**Returns**

The routine `glp_get_rii` returns current scale factor $r_{ii}$ for $i$-th row of the specified problem object.

2.5.5  

**glp_get_sjj**—retrieve column scale factor

**Synopsis**

```c
double glp_get_sjj(glp_prob *lp, int j);
```

**Returns**

The routine `glp_get_sjj` returns current scale factor $s_{jj}$ for $j$-th column of the specified problem object.

2.5.6  

**glp_scale_prob**—scale problem data

**Synopsis**

```c
void glp_scale_prob(glp_prob *lp, int flags);
```

**Description**

The routine `glp_scale_prob` performs automatic scaling of problem data for the specified problem object.
The parameter flags specifies scaling options used by the routine. The options can be combined with the bitwise OR operator and may be the following:

- GLP_SF_GM: perform geometric mean scaling;
- GLP_SF_EQ: perform equilibration scaling;
- GLP_SF_2N: round scale factors to nearest power of two;
- GLP_SF_SKIP: skip scaling, if the problem is well scaled.

The parameter flags may be specified as GLP_SF_AUTO, in which case the routine chooses the scaling options automatically.

### 2.5.7 glp_unscale_prob—unscale problem data

#### Synopsis

```c
void glp_unscale_prob(glp_prob *lp);
```

The routine glp_unscale_prob performs unscaling of problem data for the specified problem object.

“Unscaling” means replacing the current scaling matrices $R$ and $S$ by unity matrices that cancels the scaling effect.
2.6 LP basis constructing routines

2.6.1 Background

To start the search the simplex method needs a valid initial basis. In GLPK the basis is completely defined by a set of statuses assigned to all (auxiliary and structural) variables, where the status may be one of the following:

- **GLP_BS**: basic variable;
- **GLP_NL**: non-basic variable having active lower bound;
- **GLP_NU**: non-basic variable having active upper bound;
- **GLP_NF**: non-basic free variable;
- **GLP_NS**: non-basic fixed variable.

The basis is valid, if the basis matrix, which is a matrix built of columns of the augmented constraint matrix (I − A) corresponding to basic variables, is non-singular. This, in particular, means that the number of basic variables must be the same as the number of rows in the problem object. (For more details see Section 4.2, page 122.)

Any initial basis may be constructed (or restored) with the API routines `glp_set_row_stat` and `glp_set_col_stat` by assigning appropriate statuses to auxiliary and structural variables. Another way to construct an initial basis is to use API routines like `glp_adv_basis`, which implement so called crashing.\(^2\) Note that on normal exit the simplex solver remains the basis valid, so in case of reoptimization there is no need to construct an initial basis from scratch.

### 2.6.2 glp_set_row_stat—set (change) row status

**Synopsis**

```c
void glp_set_row_stat(glp_prob *lp, int i, int stat);
```

**Description**

The routine `glp_set_row_stat` sets (changes) the current status of i-th row (auxiliary variable) as specified by the parameter `stat`:

- **GLP_BS** make the row basic (make the constraint inactive);
- **GLP_NL** make the row non-basic (make the constraint active);

---

\(^2\) This term is from early linear programming systems and means a heuristic to construct a valid initial basis.
GLP_NU  make the row non-basic and set it to the upper bound; if the row is not double-bounded, this status is equivalent to GLP_NL (only in the case of this routine);
GLP_NF  the same as GLP_NL (only in the case of this routine);
GLP_NS  the same as GLP_NL (only in the case of this routine).

2.6.3  glp_set_col_stat—set (change) column status

Synopsis

void glp_set_col_stat(glp_prob *lp, int j, int stat);

Description

The routine glp_set_col_stat sets (changes) the current status of j-th column (structural variable) as specified by the parameter stat:
- GLP_BS  make the column basic;
- GLP_NL  make the column non-basic;
- GLP_NU  make the column non-basic and set it to the upper bound; if the column is not double-bounded, this status is equivalent to GLP_NL (only in the case of this routine);
- GLP_NF  the same as GLP_NL (only in the case of this routine);
- GLP_NS  the same as GLP_NL (only in the case of this routine).

2.6.4  glp_std_basis—construct standard initial LP basis

Synopsis

void glp_std_basis(glp_prob *lp);

Description

The routine glp_std_basis constructs the “standard” (trivial) initial LP basis for the specified problem object.

In the “standard” LP basis all auxiliary variables (rows) are basic, and all structural variables (columns) are non-basic (so the corresponding basis matrix is unity).
2.6.5 \texttt{glp\_adv\_basis}—construct advanced initial LP basis

Synopsis

\begin{verbatim}
void glp_adv_basis(glp_prob *lp, int flags);
\end{verbatim}

Description

The routine \texttt{glp\_adv\_basis} constructs an advanced initial LP basis for the specified problem object.

The parameter \texttt{flags} is reserved for use in the future and must be specified as zero.

In order to construct the advanced initial LP basis the routine does the following:

1) includes in the basis all non-fixed auxiliary variables;
2) includes in the basis as many non-fixed structural variables as possible keeping the triangular form of the basis matrix;
3) includes in the basis appropriate (fixed) auxiliary variables to complete the basis.

As a result the initial LP basis has as few fixed variables as possible and the corresponding basis matrix is triangular.

2.6.6 \texttt{glp\_cpx\_basis}—construct Bixby’s initial LP basis

Synopsis

\begin{verbatim}
void glp_cpx_basis(glp_prob *lp);
\end{verbatim}

Description

2.7 Simplex method routines

The simplex method is a well known efficient numerical procedure to solve LP problems.

On each iteration the simplex method transforms the original system of equality constraints (1.2) resolving them through different sets of variables to an equivalent system called the simplex table (or sometimes the simplex tableau), which has the following form:

\[
\begin{align*}
    z &= d_1(x_N)_1 + d_2(x_N)_2 + \ldots + d_n(x_N)_n \\
    (x_B)_1 &= \xi_{11}(x_N)_1 + \xi_{12}(x_N)_2 + \ldots + \xi_{1n}(x_N)_n \\
    (x_B)_2 &= \xi_{21}(x_N)_1 + \xi_{22}(x_N)_2 + \ldots + \xi_{2n}(x_N)_n \\
    \quad \ldots \ldots \ldots \ldots \\
    (x_B)_m &= \xi_{m1}(x_N)_1 + \xi_{m2}(x_N)_2 + \ldots + \xi_{mn}(x_N)_n
\end{align*}
\]

(2.3)

where: \((x_B)_1, (x_B)_2, \ldots, (x_B)_m\) are basic variables; \((x_N)_1, (x_N)_2, \ldots, (x_N)_n\) are non-basic variables; \(d_1, d_2, \ldots, d_n\) are reduced costs; \(\xi_{11}, \xi_{12}, \ldots, \xi_{mn}\) are coefficients of the simplex table. (May note that the original LP problem (1.1)—(1.3) also has the form of a simplex table, where all equalities are resolved through auxiliary variables.)

From the linear programming theory it is known that if an optimal solution of the LP problem (1.1)—(1.3) exists, it can always be written in the form (2.3), where non-basic variables are set on their bounds while values of the objective function and basic variables are determined by the corresponding equalities of the simplex table.

A set of values of all basic and non-basic variables determined by the simplex table is called basic solution. If all basic variables are within their bounds, the basic solution is called (primal) feasible, otherwise it is called (primal) infeasible. A feasible basic solution, which provides a smallest (in case of minimization) or a largest (in case of maximization) value of the objective function is called optimal. Therefore, for solving LP problem the simplex method tries to find its optimal basic solution.

Primal feasibility of some basic solution may be stated by simple checking if all basic variables are within their bounds. Basic solution is optimal if additionally the following optimality conditions are satisfied for all non-basic variables:

<table>
<thead>
<tr>
<th>Status of ((x_N)_j)</th>
<th>Minimization</th>
<th>Maximization</th>
</tr>
</thead>
<tbody>
<tr>
<td>((x_N)_j) is free</td>
<td>(d_j = 0)</td>
<td>(d_j = 0)</td>
</tr>
<tr>
<td>((x_N)_j) is on its lower bound</td>
<td>(d_j \geq 0)</td>
<td>(d_j \leq 0)</td>
</tr>
<tr>
<td>((x_N)_j) is on its upper bound</td>
<td>(d_j \leq 0)</td>
<td>(d_j \geq 0)</td>
</tr>
</tbody>
</table>

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In other words, basic solution is optimal if there is no non-basic variable, which changing in the feasible direction (i.e. increasing if it is free or on its lower bound, or decreasing if it is free or on its upper bound) can improve (i.e. decrease in case of minimization or increase in case of maximization) the objective function.

If all non-basic variables satisfy to the optimality conditions shown above (independently on whether basic variables are within their bounds or not), the basic solution is called dual feasible, otherwise it is called dual infeasible.

It may happen that some LP problem has no primal feasible solution due to incorrect formulation—this means that its constraints conflict with each other. It also may happen that some LP problem has unbounded solution again due to incorrect formulation—this means that some non-basic variable can improve the objective function, i.e. the optimality conditions are violated, and at the same time this variable can infinitely change in the feasible direction meeting no resistance from basic variables. (May note that in the latter case the LP problem has no dual feasible solution.)

2.7.1 glp_simplex—solve LP problem with the primal or dual simplex method

Synopsis

int glp_simplex(glp_prob *lp, const glp_smcp *parm);

Description

The routine glp_simplex is a driver to the LP solver based on the simplex method. This routine retrieves problem data from the specified problem object, calls the solver to solve the problem instance, and stores results of computations back into the problem object.

The simplex solver has a set of control parameters. Values of the control parameters can be passed in the structure glp_smcp, which the parameter parm points to. For detailed description of this structure see paragraph “Control parameters” below. Before specifying some control parameters the application program should initialize the structure glp_smcp by default values of all control parameters using the routine glp_init_smcp (see the next subsection). This is needed for backward compatibility, because in the future there may appear new members in the structure glp_smcp.

The parameter parm can be specified as NULL, in which case the solver uses default settings.
Returns

0  The LP problem instance has been successfully solved.
   (This code does not necessarily mean that the solver has
   found optimal solution. It only means that the solution
   process was successful.)

GLP_EBADB  Unable to start the search, because the initial basis speci-
            fied in the problem object is invalid—the number of basic
            (auxiliary and structural) variables is not the same as the
            number of rows in the problem object.

GLP_ESING  Unable to start the search, because the basis matrix corre-
            sponding to the initial basis is singular within the working
            precision.

GLP_ECOND  Unable to start the search, because the basis matrix corre-
            sponding to the initial basis is ill-conditioned, i.e. its
            condition number is too large.

GLP_EBOUND  Unable to start the search, because some double-bounded
            (auxiliary or structural) variables have incorrect bounds.

GLP_EFAIL  The search was prematurely terminated due to the solver
            failure.

GLP_EOBJLL  The search was prematurely terminated, because the obj-
            ective function being maximized has reached its lower
            limit and continues decreasing (the dual simplex only).

GLP_EOBJUL  The search was prematurely terminated, because the obj-
            ective function being minimized has reached its upper
            limit and continues increasing (the dual simplex only).

GLP_EITLIM  The search was prematurely terminated, because the sim-
           plex iteration limit has been exceeded.

GLP_ETMLIM  The search was prematurely terminated, because the time
            limit has been exceeded.

GLP_ENOPFS  The LP problem instance has no primal feasible solution
            (only if the LP presolver is used).

GLP_ENODFS  The LP problem instance has no dual feasible solution
            (only if the LP presolver is used).

Built-in LP presolver

The simplex solver has built-in LP presolver. It is a subprogram that trans-
forms the original LP problem specified in the problem object to an equiva-
 lent LP problem, which may be easier for solving with the simplex method
than the original one. This is attained mainly due to reducing the prob-
lem size and improving its numeric properties (for example, by removing some inactive constraints or by fixing some non-basic variables). Once the transformed LP problem has been solved, the presolver transforms its basic solution back to the corresponding basic solution of the original problem.

Presolving is an optional feature of the routine \texttt{glp_simplex}, and by default it is disabled. In order to enable the LP presolver the control parameter \texttt{presolve} should be set to \texttt{GLP_ON} (see paragraph “Control parameters” below). Presolving may be used when the problem instance is solved for the first time. However, on performing re-optimization the presolver should be disabled.

The presolving procedure is transparent to the API user in the sense that all necessary processing is performed internally, and a basic solution of the original problem recovered by the presolver is the same as if it were computed directly, i.e. without presolving.

Note that the presolver is able to recover only optimal solutions. If a computed solution is infeasible or non-optimal, the corresponding solution of the original problem cannot be recovered and therefore remains undefined. If you need to know a basic solution even if it is infeasible or non-optimal, the presolver should be disabled.

**Terminal output**

Solving large problem instances may take a long time, so the solver reports some information about the current basic solution, which is sent to the terminal. This information has the following format:

```
nnn: obj = xxx  infeas = yyy (ddd)
```

where: ‘\texttt{nnn}’ is the iteration number, ‘\texttt{xxx}’ is the current value of the objective function (it is is unscaled and has correct sign); ‘\texttt{yyy}’ is the current sum of primal or dual infeasibilities (it is scaled and therefore may be used only for visual estimating), ‘\texttt{ddd}’ is the current number of fixed basic variables.

The symbol preceding the iteration number indicates which phase of the simplex method is in effect:

- \texttt{Blank} means that the solver is searching for primal feasible solution using the primal simplex or for dual feasible solution using the dual simplex;
- \texttt{Asterisk} (\texttt{*}) means that the solver is searching for optimal solution using the primal simplex;
- \texttt{Vertical dash} (\texttt{1}) means that the solver is searching for optimal solution using the dual simplex.
**Control parameters**

This paragraph describes all control parameters currently used in the simplex solver. Symbolic names of control parameters are names of corresponding members in the structure `glp_smcp`.

**int msg_lev** (default: GLP_MSG_ALL)
- Message level for terminal output:
  - GLP_MSG_OFF—no output;
  - GLP_MSG_ERR—error and warning messages only;
  - GLP_MSG_ON—normal output;
  - GLP_MSG_ALL—full output (including informational messages).

**int meth** (default: GLP_PRIMAL)
- Simplex method option:
  - GLP_PRIMAL—use two-phase primal simplex;
  - GLP_DUAL—use two-phase dual simplex;
  - GLP_DUALP—use two-phase dual simplex, and if it fails, switch to the primal simplex.

**int pricing** (default: GLP_PT_PSE)
- Pricing technique:
  - GLP_PT_STD—standard (textbook);
  - GLP_PT_PSE—projected steepest edge.

**int r_test** (default: GLP_RT_HAR)
- Ratio test technique:
  - GLP_RT_STD—standard (textbook);
  - GLP_RT_HAR—Harris’ two-pass ratio test.

**double tol_bnd** (default: 1e-7)
- Tolerance used to check if the basic solution is primal feasible. (Do not change this parameter without detailed understanding its purpose.)

**double tol_dj** (default: 1e-7)
- Tolerance used to check if the basic solution is dual feasible. (Do not change this parameter without detailed understanding its purpose.)

**double tol_piv** (default: 1e-10)
- Tolerance used to choose eligible pivotal elements of the simplex table. (Do not change this parameter without detailed understanding its purpose.)
double obj_ll (default: -DBL_MAX)
Lower limit of the objective function. If the objective function reaches
this limit and continues decreasing, the solver terminates the search.
(Used in the dual simplex only.)

double obj_ul (default: +DBL_MAX)
Upper limit of the objective function. If the objective function reaches
this limit and continues increasing, the solver terminates the search.
(Used in the dual simplex only.)

int it_lim (default: INT_MAX)
Simplex iteration limit.

int tm_lim (default: INT_MAX)
Searching time limit, in milliseconds.

int out_fraq (default: 500)
Output frequency, in iterations. This parameter specifies how frequently
the solver sends information about the solution process to the terminal.

int out_dly (default: 0)
Output delay, in milliseconds. This parameter specifies how long the
solver should delay sending information about the solution process to
the terminal.

int presolve (default: GLP_OFF)
LP presolver option:
GLP_ON — enable using the LP presolver;
GLP_OFF — disable using the LP presolver.

Example 1
The following main program reads LP problem instance in fixed MPS format
from file 25fv47.mps, constructs an advanced initial basis, solves the in-
stance with the primal simplex method (by default), and writes the solution
to file 25fv47.txt.

---

4 This instance in fixed MPS format can be found in the Netlib LP collection; see
/* spxsamp1.c */

#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
    glp_prob *P;
    P = glp_create_prob();
    glp_read_mps(P, GLP_MPS_deck, NULL, "25fv47.mps");
    glp_adv_basis(P, 0);
    glp_simplex(P, NULL);
    glp_print_sol(P, "25fv47.txt");
    glp_delete_prob(P);
    return 0;
}

/* eof */

Below here is shown the terminal output from this example program.

Reading problem data from '25fv47.mps'...
Problem: 25FV47
Objective: R0000
822 rows, 1571 columns, 11127 non-zeros
6919 records were read
Crashing...
Size of triangular part = 799
  0: obj =  1.627307307e+04  infeas =  5.194e+04 (23)
  200: obj =  1.474901610e+04  infeas =  1.233e+04 (19)
  400: obj =  1.343909995e+04  infeas =  3.648e+03 (13)
  600: obj =  1.756052217e+04  infeas =  4.179e+02 (7)
    775: obj =  1.789251591e+04  infeas =  4.982e-14 (1)
    800: obj =  1.663354510e+04  infeas =  2.857e-14 (1)
    1000: obj =  1.024935068e+04  infeas =  1.958e-12 (1)
    1200: obj =  7.860174791e+03  infeas =  2.810e-29 (1)
    1400: obj =  6.642378184e+03  infeas =  2.036e-16 (1)
    1600: obj =  6.037014686e+03  infeas =  0.000e+00 (1)
    1800: obj =  5.662171307e+03  infeas =  6.447e-15 (1)
    2000: obj =  5.528146165e+03  infeas =  9.764e-13 (1)
    2125: obj =  5.501845888e+03  infeas =  0.000e+00 (1)
OPTIMAL SOLUTION FOUND
Writing basic solution to '25fv47.txt'...
Example 2

The following main program solves the same LP problem instance as in Example 1 above, however, it uses the dual simplex method, which starts from the standard initial basis.

```c
/* spxsamp2.c */
#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
    glp_prob *P;
    glp_smcp parm;
    P = glp_create_prob();
    glp_read_mps(P, GLP_MPS_DECK, NULL, "25fv47.mps");
    glp_init_smcp(&parm);
    parm.meth = GLP_DUAL;
    glp_simplex(P, &parm);
    glp_print_sol(P, "25fv47.txt");
    glp_delete_prob(P);
    return 0;
}

/* eof */
```

Below here is shown the terminal output from this example program.

```
Reading problem data from '25fv47.mps'...
Problem: 25FV47
Objective: R0000
822 rows, 1571 columns, 11127 non-zeros
6919 records were read

| 0: obj = -5.394267152e+03 infeas = 1.223e+03 (516)
| 200: obj = -4.586395752e+03 infeas = 7.000e+00 (471)
| 240: obj = -4.158268146e+03 infeas = 1.106e-14 (461)
| 400: obj = -3.725320045e+03 infeas = 5.571e-16 (391)
| 600: obj = -3.104802163e+03 infeas = 1.389e-15 (340)
| 800: obj = -2.584190499e+03 infeas = 1.640e-15 (264)
| 1000: obj = -2.073852927e+03 infeas = 5.181e-15 (245)
| 1200: obj = -1.164037407e+03 infeas = 1.019e-14 (210)
| 1400: obj = -1.04826146e+03 infeas = 8.865e-15 (178)
| 1600: obj = -2.073852927e+03 infeas = 7.867e-15 (142)
| 1800: obj = -2.073852927e+03 infeas = 8.792e-15 (109)
| 2000: obj = -2.073852927e+03 infeas = 2.591e-14 (85)
| 2200: obj = -2.073852927e+03 infeas = 1.025e-13 (70)
| 2400: obj = -2.073852927e+03 infeas = 3.272e-14 (67)
```
2.7.2 glp_exact—solve LP problem in exact arithmetic

Synopsis

int glp_exact(glp_prob *lp, const glp_smcp *parm);

Description

The routine glp_exact is a tentative implementation of the primal two-phase simplex method based on exact (rational) arithmetic. It is similar to the routine glp_simplex, however, for all internal computations it uses arithmetic of rational numbers, which is exact in mathematical sense, i.e. free of round-off errors unlike floating-point arithmetic.

Note that the routine glp_exact uses only two control parameters passed in the structure glp_smcp, namely, it_lim and tm_lim.

Returns

0 The LP problem instance has been successfully solved. (This code does not necessarily mean that the solver has found optimal solution. It only means that the solution process was successful.)

GLP_EBADB Unable to start the search, because the initial basis specified in the problem object is invalid—the number of basic (auxiliary and structural) variables is not the same as the number of rows in the problem object.

GLP_ESING Unable to start the search, because the basis matrix corresponding to the initial basis is exactly singular.

GLP_EBOUND Unable to start the search, because some double-bounded (auxiliary or structural) variables have incorrect bounds.

GLP_EFAIL The problem instance has no rows/columns.

GLP_EITLIM The search was prematurely terminated, because the simplex iteration limit has been exceeded.

GLP_ETMLIM The search was prematurely terminated, because the time limit has been exceeded.
Comments

Computations in exact arithmetic are very time consuming, so solving LP problem with the routine `glp_exact` from the very beginning is not a good idea. It is much better at first to find an optimal basis with the routine `glp_simplex` and only then to call `glp_exact`, in which case only a few simplex iterations need to be performed in exact arithmetic.

2.7.3 `glp_init_smcp`—initialize simplex solver control parameters

Synopsis

```c
int glp_init_smcp(glp_smcp *parm);
```

Description

The routine `glp_init_smcp` initializes control parameters, which are used by the simplex solver, with default values.

Default values of the control parameters are stored in a `glp_smcp` structure, which the parameter `parm` points to.

2.7.4 `glp_get_status`—determine generic status of basic solution

Synopsis

```c
int glp_get_status(glp_prob *lp);
```

Returns

The routine `glp_get_status` reports the generic status of the current basic solution for the specified problem object as follows:

- `GLP_OPT` solution is optimal;
- `GLP_FEAS` solution is feasible;
- `GLP_INFEAS` solution is infeasible;
- `GLP_NOFEAS` problem has no feasible solution;
- `GLP_UNBND` problem has unbounded solution;
- `GLP_UNDEF` solution is undefined.

More detailed information about the status of basic solution can be retrieved with the routines `glp_get_prim_stat` and `glp_get_dual_stat`. 
2.7.5  glp_get_prim_stat—retrieve status of primal basic solution

Synopsis

int glp_get_prim_stat(glp_prob *lp);

Returns

The routine glp_get_prim_stat reports the status of the primal basic solution for the specified problem object as follows:

- GLP_UNDEF  primal solution is undefined;
- GLP_FEAS  primal solution is feasible;
- GLP_INFEAS primal solution is infeasible;
- GLP_NOFEAS no primal feasible solution exists.

2.7.6  glp_get_dual_stat—retrieve status of dual basic solution

Synopsis

int glp_get_dual_stat(glp_prob *lp);

Returns

The routine glp_get_dual_stat reports the status of the dual basic solution for the specified problem object as follows:

- GLP_UNDEF  dual solution is undefined;
- GLP_FEAS  dual solution is feasible;
- GLP_INFEAS dual solution is infeasible;
- GLP_NOFEAS no dual feasible solution exists.

2.7.7  glp_get_obj_val—retrieve objective value

Synopsis

double glp_get_obj_val(glp_prob *lp);

Returns

The routine glp_get_obj_val returns current value of the objective function.
2.7.8 glp_get_row_stat—retrieve row status

Synopsis

```c
int glp_get_row_stat(glp_prob *lp, int i);
```

Returns

The routine `glp_get_row_stat` returns current status assigned to the auxiliary variable associated with i-th row as follows:

- `GLP_BS` basic variable;
- `GLP_NL` non-basic variable on its lower bound;
- `GLP_NU` non-basic variable on its upper bound;
- `GLP_NF` non-basic free (unbounded) variable;
- `GLP_NS` non-basic fixed variable.

2.7.9 glp_get_row_prim—retrieve row primal value

Synopsis

```c
double glp_get_row_prim(glp_prob *lp, int i);
```

Returns

The routine `glp_get_row_prim` returns primal value of the auxiliary variable associated with i-th row.

2.7.10 glp_get_row_dual—retrieve row dual value

Synopsis

```c
double glp_get_row_dual(glp_prob *lp, int i);
```

Returns

The routine `glp_get_row_dual` returns dual value (i.e. reduced cost) of the auxiliary variable associated with i-th row.
2.7.11  glp_get_col_stat—retrieve column status

Synopsis

```c
int glp_get_col_stat(glp_prob *lp, int j);
```

Returns

The routine `glp_get_col_stat` returns current status assigned to the structural variable associated with `j`-th column as follows:

- GLP_BS basic variable;
- GLP_NL non-basic variable on its lower bound;
- GLP_NU non-basic variable on its upper bound;
- GLP_NF non-basic free (unbounded) variable;
- GLP_NS non-basic fixed variable.

2.7.12  glp_get_col_prim—retrieve column primal value

Synopsis

```c
double glp_get_col_prim(glp_prob *lp, int j);
```

Returns

The routine `glp_get_col_prim` returns primal value of the structural variable associated with `j`-th column.

2.7.13  glp_get_col_dual—retrieve column dual value

Synopsis

```c
double glp_get_col_dual(glp_prob *lp, int j);
```

Returns

The routine `glp_get_col_dual` returns dual value (i.e. reduced cost) of the structural variable associated with `j`-th column.
2.7.14  glp_get_unbnd_ray—determine variable causing unboundedness

Synopsis

int glp_get_unbnd_ray(glp_prob *lp);

Returns

The routine glp_get_unbnd_ray returns the number \( k \) of a variable, which causes primal or dual unboundedness. If \( 1 \leq k \leq m \), it is \( k \)-th auxiliary variable, and if \( m + 1 \leq k \leq m + n \), it is \( (k - m) \)-th structural variable, where \( m \) is the number of rows, \( n \) is the number of columns in the problem object. If such variable is not defined, the routine returns 0.

Comments

If it is not exactly known which version of the simplex solver detected unboundedness, i.e. whether the unboundedness is primal or dual, it is sufficient to check the status of the variable with the routine glp_get_row_stat or glp_get_col_stat. If the variable is non-basic, the unboundedness is primal, otherwise, if the variable is basic, the unboundedness is dual (the latter case means that the problem has no primal feasible solution).
2.8 Interior-point method routines

Interior-point methods (also known as barrier methods) are more modern and powerful numerical methods for large-scale linear programming. Such methods are especially efficient for very sparse LP problems and allow solving such problems much faster than the simplex method.

In brief, the GLPK interior-point solver works as follows.

At first, the solver transforms the original LP to a working LP in the standard format:

minimize
\[ z = c_1x_{m+1} + c_2x_{m+2} + \ldots + c_nx_{m+n} + c_0 \]  
subject to linear constraints
\[ a_{11}x_{m+1} + a_{12}x_{m+2} + \ldots + a_{1n}x_{m+n} = b_1 \]
\[ a_{21}x_{m+1} + a_{22}x_{m+2} + \ldots + a_{2n}x_{m+n} = b_2 \]
\[ \ldots \ldots \ldots \ldots \ldots \ldots \]
\[ a_{m1}x_{m+1} + a_{m2}x_{m+2} + \ldots + a_{mn}x_{m+n} = b_m \]

and non-negative variables
\[ x_1 \geq 0, \ x_2 \geq 0, \ldots, \ x_n \geq 0 \]

where: \( z \) is the objective function; \( x_1, \ldots, x_n \) are variables; \( c_1, \ldots, c_n \) are objective coefficients; \( c_0 \) is a constant term of the objective function; \( a_{ij} \) are constraint coefficients; \( b_1, \ldots, b_m \) are right-hand sides.

Using vector and matrix notations the working LP (2.4)—(2.6) can be written as follows:
\[ z = c^T x + c_0 \longrightarrow \min, \]
\[ Ax = b, \]
\[ x \geq 0, \]

where: \( x = (x_j) \) is \( n \)-vector of variables, \( c = (c_j) \) is \( n \)-vector of objective coefficients, \( A = (a_{ij}) \) is \( m \times n \)-matrix of constraint coefficients, and \( b = (b_j) \) is \( m \)-vector of right-hand sides.

Karush–Kuhn–Tucker optimality conditions for LP (2.7)—(2.9) are the following:
\[ Ax = b, \quad (2.10) \]
\[ A^T \pi + \lambda = c, \quad (2.11) \]
\[ \lambda^T x = 0, \quad (2.12) \]
\[ x \geq 0, \quad \lambda \geq 0, \quad (2.13) \]

where: \( \pi \) is \( m \)-vector of Lagrange multipliers (dual variables) for equality constraints (2.8), \( \lambda \) is \( n \)-vector of Lagrange multipliers (dual variables) for non-negativity constraints (2.9), (2.10) is the primal feasibility condition, (2.11) is the dual feasibility condition, (2.12) is the primal-dual complementarity condition, and (2.13) is the non-negativity conditions.

The main idea of the primal-dual interior-point method is based on finding a point in the primal-dual space (i.e. in the space of all primal and dual variables \( x, \pi, \) and \( \lambda \)), which satisfies to all optimality conditions (2.10)—(2.13). Obviously, \( x \)-component of such point then provides an optimal solution to the working LP (2.7)—(2.9).

To find the optimal point \((x^*, \pi^*, \lambda^*)\) the interior-point method attempts to solve the system of equations (2.10)—(2.12), which is closed in the sense that the number of variables \( x_j, \pi_i, \) and \( \lambda_j \) and the number equations are the same and equal to \( m + 2n \). Due to condition (2.12) this system of equations is non-linear, so it can be solved with a version of Newton’s method provided with additional rules to keep the current point within the positive orthant as required by the non-negativity conditions (2.13).

Finally, once the optimal point \((x^*, \pi^*, \lambda^*)\) has been found, the solver performs inverse transformations to recover corresponding solution to the original LP passed to the solver from the application program.

### 2.8.1 glp_interior—solve LP problem with the interior-point method

**Synopsis**

```c
int glp_interior(glp_prob *P, const glp_iptcp *parm);
```

**Description**

The routine `glp_interior` is a driver to the LP solver based on the primal-dual interior-point method. This routine retrieves problem data from the specified problem object, calls the solver to solve the problem instance, and stores results of computations back into the problem object.
The interior-point solver has a set of control parameters. Values of the control parameters can be passed in the structure `glp_iptcp`, which the parameter `parm` points to. For detailed description of this structure see paragraph “Control parameters” below. Before specifying some control parameters the application program should initialize the structure `glp_iptcp` by default values of all control parameters using the routine `glp_init_iptcp` (see the next subsection). This is needed for backward compatibility, because in the future there may appear new members in the structure `glp_iptcp`.

The parameter `parm` can be specified as `NULL`, in which case the solver uses default settings.

**Returns**

- 0 The LP problem instance has been successfully solved. (This code does *not* necessarily mean that the solver has found optimal solution. It only means that the solution process was successful.)
- `GLP_EFAIL` The problem has no rows/columns.
- `GLP_ENOCVG` Very slow convergence or divergence.
- `GLP_EITLIM` Iteration limit exceeded.
- `GLP_EINSTAB` Numerical instability on solving Newtonian system.

**Comments**

The routine `glp_interior` implements an easy version of the primal-dual interior-point method based on Mehrotra’s technique.\(^5\)

Note that currently the GLPK interior-point solver does not include many important features, in particular:

- it is not able to process dense columns. Thus, if the constraint matrix of the LP problem has dense columns, the solving process may be inefficient;
- it has no features against numerical instability. For some LP problems premature termination may happen if the matrix $ADA^T$ becomes singular or ill-conditioned;
- it is not able to identify the optimal basis, which corresponds to the interior-point solution found.

---

Terminal output

Solving large LP problems may take a long time, so the solver reports some information about every interior-point iteration, which is sent to the terminal. This information has the following format:

\[ \text{nnn: obj = fff; rpi = ppp; rdi = ddd; gap = ggg} \]

where: nnn is iteration number, fff is the current value of the objective function (in the case of maximization it has wrong sign), ppp is the current relative primal infeasibility (cf. (2.10)):

\[ \frac{\|Ax^{(k)} - b\|}{1 + \|b\|}, \]

(2.14)

ddd is the current relative dual infeasibility (cf. (2.11)):

\[ \frac{\|A^T\pi^{(k)} + \lambda^{(k)} - c\|}{1 + \|c\|}, \]

(2.15)

ggg is the current primal-dual gap (cf. (2.12)):

\[ \frac{|c^Tx^{(k)} - b^T\pi^{(k)}|}{1 + |c^Tx^{(k)}|}, \]

(2.16)

and \([x^{(k)}, \pi^{(k)}, \lambda^{(k)}]\) is the current point on k-th iteration, \(k = 0, 1, 2, \ldots\).

Note that all solution components are internally scaled, so information sent to the terminal is suitable only for visual inspection.

Control parameters

This paragraph describes all control parameters currently used in the interior-point solver. Symbolic names of control parameters are names of corresponding members in the structure glp_iptcp.

int msg_lev (default: GLP_MSG_ALL)

Message level for terminal output:

GLP_MSG_OFF—no output;
GLP_MSG_ERR—error and warning messages only;
GLP_MSG_ON —normal output;
GLP_MSG_ALL—full output (including informational messages).

6Unlike the simplex method the interior point method usually needs 30—50 iterations (independently on the problem size) in order to find an optimal solution.
Example

The following main program reads LP problem instance in fixed MPS format from file 25fv47.mps,\(^\text{7}\) solves it with the interior-point solver, and writes the solution to file 25fv47.txt.

```c
/* iptsamp.c */
#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
   glp_prob *P;
   P = glp_create_prob();
   glp_read_mps(P, GLP_MPS_DECK, NULL, "25fv47.mps");
   glp_interior(P, NULL);
   glp_print_ipt(P, "25fv47.txt");
   glp_delete_prob(P);
   return 0;
}

/* eof */
```

Below here is shown the terminal output from this example program.

```
Reading problem data from ‘25fv47.mps’...
Problem: 25FV47
Objective: R0000
822 rows, 1571 columns, 11127 non-zeros
6919 records were read
Original LP has 822 row(s), 1571 column(s), and 11127 non-zero(s)
Working LP has 821 row(s), 1876 column(s), and 10705 non-zero(s)
Matrix A has 10705 non-zeros
Matrix S = A*A' has 11895 non-zeros (upper triangle)
Minimal degree ordering...
Computing Cholesky factorization S = L'*L...
```

\(^7\)This instance in fixed MPS format can be found in the Netlib LP collection; see ftp://ftp.netlib.org/lp/data/.
Matrix L has 35411 non-zeros

Guessing initial point...

Optimization begins...

0: obj = 1.82377629e+05; rpi = 1.3e+01; rdi = 1.4e+01; gap = 9.3e-01
1: obj = 9.26045192e+04; rpi = 5.3e+00; rdi = 5.6e+00; gap = 6.8e+00
2: obj = 3.59699742e+04; rpi = 1.5e+00; rdi = 1.2e+00; gap = 1.8e+01
3: obj = 1.98827566e+04; rpi = 4.7e-01; rdi = 3.0e-01; gap = 1.9e+01
4: obj = 1.43021557e+04; rpi = 1.1e-01; rdi = 8.6e-02; gap = 1.4e+01
5: obj = 1.15716505e+04; rpi = 2.3e-02; rdi = 2.4e-02; gap = 6.8e+00
6: obj = 9.66273208e+03; rpi = 6.7e-03; rdi = 4.6e-03; gap = 3.9e+00
7: obj = 8.69438283e+03; rpi = 3.7e-03; rdi = 1.7e-03; gap = 2.0e+00
8: obj = 8.01954339e+03; rpi = 2.4e-03; rdi = 3.9e-04; gap = 1.0e+00
9: obj = 7.12267629e+03; rpi = 1.2e-03; rdi = 1.5e-04; gap = 6.6e+00
10: obj = 6.514534518e+03; rpi = 6.1e-04; rdi = 4.3e-05; gap = 4.1e-01
11: obj = 6.361572203e+03; rpi = 4.8e-04; rdi = 2.2e-05; gap = 3.0e-01
12: obj = 6.203355508e+03; rpi = 3.2e-04; rdi = 1.7e-05; gap = 2.6e-01
13: obj = 6.03294311e+03; rpi = 2.0e-04; rdi = 9.3e-06; gap = 2.1e-01
14: obj = 5.796553021e+03; rpi = 9.8e-05; rdi = 3.2e-06; gap = 1.0e-01
15: obj = 5.667032431e+03; rpi = 4.4e-05; rdi = 1.1e-06; gap = 5.6e-02
16: obj = 5.613911867e+03; rpi = 2.5e-05; rdi = 4.1e-07; gap = 3.5e-02
17: obj = 5.560572626e+03; rpi = 9.9e-06; rdi = 2.3e-07; gap = 2.1e-02
18: obj = 5.53277601e+03; rpi = 5.5e-06; rdi = 8.4e-08; gap = 1.1e-02
19: obj = 5.522746942e+03; rpi = 2.2e-06; rdi = 4.0e-08; gap = 6.7e-03
20: obj = 5.509956679e+03; rpi = 7.5e-07; rdi = 1.8e-08; gap = 2.9e-03
21: obj = 5.504571733e+03; rpi = 1.6e-07; rdi = 5.8e-09; gap = 1.1e-03
22: obj = 5.502576367e+03; rpi = 3.4e-08; rdi = 1.0e-09; gap = 2.5e-04
23: obj = 5.502057119e+03; rpi = 8.1e-09; rdi = 3.0e-10; gap = 7.7e-05
24: obj = 5.501858599e+03; rpi = 9.4e-10; rdi = 1.2e-10; gap = 2.4e-05
25: obj = 5.501852464e+03; rpi = 1.4e-10; rdi = 1.2e-11; gap = 3.0e-06
26: obj = 5.501846549e+03; rpi = 1.4e-11; rdi = 1.2e-12; gap = 3.0e-07
27: obj = 5.501845954e+03; rpi = 1.4e-12; rdi = 1.2e-13; gap = 3.0e-08
28: obj = 5.501845895e+03; rpi = 1.5e-13; rdi = 1.2e-14; gap = 3.0e-09

OPTIMAL SOLUTION FOUND

Writing interior-point solution to ‘25fv47.txt’...

2.8.2 glp_init_iptcp—initialize interior-point solver control parameters

Synopsis

int glp_init_iptcp(glp_iptcp *parm);

Description

The routine glp_init_iptcp initializes control parameters, which are used by the interior-point solver, with default values.
Default values of the control parameters are stored in the structure `glp_iptcp`, which the parameter `parm` points to.

2.8.3 `glp_ipt_status`—determine solution status

Synopsis

```c
int glp_ipt_status(glp_prob *lp);
```

Returns

The routine `glp_ipt_status` reports the status of a solution found by the interior-point solver as follows:

- `GLP_UNDEF`: interior-point solution is undefined.
- `GLP_OPT`: interior-point solution is optimal.
- `GLP_INFEAS`: interior-point solution is infeasible.
- `GLP_NOFEAS`: no feasible primal-dual solution exists.

2.8.4 `glp_ipt_obj_val`—retrieve objective value

Synopsis

```c
double glp_ipt_obj_val(glp_prob *lp);
```

Returns

The routine `glp_ipt_obj_val` returns value of the objective function for interior-point solution.

2.8.5 `glp_ipt_row_prim`—retrieve row primal value

Synopsis

```c
double glp_ipt_row_prim(glp_prob *lp, int i);
```

Returns

The routine `glp_ipt_row_prim` returns primal value of the auxiliary variable associated with i-th row.
2.8.6  glp_ipt_row_dual—retrieve row dual value

Synopsis

double glp_ipt_row_dual(glp_prob *lp, int i);

Returns

The routine glp_ipt_row_dual returns dual value (i.e. reduced cost) of the auxiliary variable associated with i-th row.

2.8.7  glp_ipt_col_prim—retrieve column primal value

Synopsis

double glp_ipt_col_prim(glp_prob *lp, int j);

Returns

The routine glp_ipt_col_prim returns primal value of the structural variable associated with j-th column.

2.8.8  glp_ipt_col_dual—retrieve column dual value

Synopsis

double glp_ipt_col_dual(glp_prob *lp, int j);

Returns

The routine glp_ipt_col_dual returns dual value (i.e. reduced cost) of the structural variable associated with j-th column.
2.9 Mixed integer programming routines

2.9.1 glp_set_col_kind—set (change) column kind

Synopsis

void glp_set_col_kind(glp_prob *mip, int j, int kind);

Description

The routine glp_set_col_kind sets (changes) the kind of j-th column (structural variable) as specified by the parameter kind:

- GLP_CV continuous variable;
- GLP_IV integer variable;
- GLP_BV binary variable.

Setting a column to GLP_BV has the same effect as if it were set to GLP_IV, its lower bound were set 0, and its upper bound were set to 1.

2.9.2 glp_get_col_kind—retrieve column kind

Synopsis

int glp_get_col_kind(glp_prob *mip, int j);

Returns

The routine glp_get_col_kind returns the kind of j-th column (structural variable) as follows:

- GLP_CV continuous variable;
- GLP_IV integer variable;
- GLP_BV binary variable.

2.9.3 glp_get_num_int—retrieve number of integer columns

Synopsis

int glp_get_num_int(glp_prob *mip);

Returns

The routine glp_get_num_int returns the number of columns (structural variables), which are marked as integer. Note that this number does include binary columns.
2.9.4 \texttt{glp\_get\_num\_bin}—retrieve number of binary columns

Synopsis

\begin{verbatim}
int glp_get_num_bin(glp_prob *mip);
\end{verbatim}

Returns

The routine \texttt{glp\_get\_num\_bin} returns the number of columns (structural variables), which are marked as integer and whose lower bound is zero and upper bound is one.

2.9.5 \texttt{glp\_intopt}—solve MIP problem with the branch-and-cut method

Synopsis

\begin{verbatim}
int glp_intopt(glp_prob *mip, const glp_iocp *parm);
\end{verbatim}

Description

The routine \texttt{glp\_intopt} is a driver to the MIP solver based on the branch-and-cut method, which is a hybrid of branch-and-bound and cutting plane methods.

If the presolver is disabled (see paragraph “Control parameters” below), on entry to the routine \texttt{glp\_intopt} the problem object, which the parameter \texttt{mip} points to, should contain optimal solution to LP relaxation (it can be obtained, for example, with the routine \texttt{glp\_simplex}). Otherwise, if the presolver is enabled, it is not necessary.

The MIP solver has a set of control parameters. Values of the control parameters can be passed in the structure \texttt{glp\_iocp}, which the parameter \texttt{parm} points to. For detailed description of this structure see paragraph “Control parameters” below. Before specifying some control parameters the application program should initialize the structure \texttt{glp\_iocp} by default values of all control parameters using the routine \texttt{glp\_init\_iocp} (see the next subsection). This is needed for backward compatibility, because in the future there may appear new members in the structure \texttt{glp\_iocp}.

The parameter \texttt{parm} can be specified as \texttt{NULL}, in which case the solver uses default settings.

Note that the GLPK branch-and-cut solver is not perfect, so it is unable to solve hard or very large scale MIP instances for a reasonable time.
Returns

0      The MIP problem instance has been successfully solved. (This code does not necessarily mean that the solver has found optimal solution. It only means that the solution process was successful.)

GLP_EBOUND  Unable to start the search, because some double-bounded variables have incorrect bounds or some integer variables have non-integer (fractional) bounds.

GLP_EROOT  Unable to start the search, because optimal basis for initial LP relaxation is not provided. (This code may appear only if the presolver is disabled.)

GLP_ENOPFS  Unable to start the search, because LP relaxation of the MIP problem instance has no primal feasible solution. (This code may appear only if the presolver is enabled.)

GLP_ENODFS  Unable to start the search, because LP relaxation of the MIP problem instance has no dual feasible solution. In other word, this code means that if the LP relaxation has at least one primal feasible solution, its optimal solution is unbounded, so if the MIP problem has at least one integer feasible solution, its (integer) optimal solution is also unbounded. (This code may appear only if the presolver is enabled.)

GLP_EFAIL  The search was prematurely terminated due to the solver failure.

GLP_EMIPGAP  The search was prematurely terminated, because the relative mip gap tolerance has been reached.

GLP_ETMLIM  The search was prematurely terminated, because the time limit has been exceeded.

GLP_ESTOP  The search was prematurely terminated by application. (This code may appear only if the advanced solver interface is used.)

Built-in MIP presolver

The branch-and-cut solver has built-in MIP presolver. It is a subprogram that transforms the original MIP problem specified in the problem object to an equivalent MIP problem, which may be easier for solving with the branch-and-cut method than the original one. For example, the presolver can remove redundant constraints and variables, whose optimal values are known, perform bound and coefficient reduction, etc. Once the transformed
MIP problem has been solved, the presolver transforms its solution back to corresponding solution of the original problem.

Presolving is an optional feature of the routine glp_intopt, and by default it is disabled. In order to enable the MIP presolver, the control parameter presolve should be set to GLP_ON (see paragraph “Control parameters” below).

Advanced solver interface

The routine glp_intopt allows the user to control the branch-and-cut search by passing to the solver a user-defined callback routine. For more details see Chapter “Branch-and-Cut API Routines”.

Terminal output

Solving a MIP problem may take a long time, so the solver reports some information about best known solutions, which is sent to the terminal. This information has the following format:

\[ +\text{nnn}: \text{mip} = \text{xxx} \ \text{<rho>} \ \text{yyy} \ \text{gap} (\text{ppp}; \text{qqq}) \]

where: ‘\text{nnn}’ is the simplex iteration number; ‘\text{xxx}’ is a value of the objective function for the best known integer feasible solution (if no integer feasible solution has been found yet, ‘\text{xxx}’ is the text ‘\text{not found yet}’); ‘\text{rho}’ is the string ‘\text{>=}’ (in case of minimization) or ‘\text{<=}’ (in case of maximization); ‘\text{yyy}’ is a global bound for exact integer optimum (i.e. the exact integer optimum is always in the range from ‘\text{xxx}’ to ‘\text{yyy}’); ‘\text{gap}’ is the relative mip gap, in percents, computed as \[ \text{gap} = \frac{|\text{xxx} - \text{yyy}|}{|\text{xxx}| + \text{DBL_EPSILON}} \cdot 100\% \] (if \text{gap} is greater than 999.9%, it is not printed); ‘\text{ppp}’ is the number of subproblems in the active list, ‘\text{qqq}’ is the number of subproblems which have been already fathomed and therefore removed from the branch-and-bound search tree.

Control parameters

This paragraph describes all control parameters currently used in the MIP solver. Symbolic names of control parameters are names of corresponding members in the structure glp_iocp.
int msg_lev (default: GLP_MSG_ALL)
   Message level for terminal output:
   GLP_MSG_OFF — no output;
   GLP_MSG_ERR — error and warning messages only;
   GLP_MSG_ON — normal output;
   GLP_MSG_ALL — full output (including informational messages).

int br_tech (default: GLP_BR_DTH)
   Branching technique option:
   GLP_BR_FFV — first fractional variable;
   GLP_BR_LFV — last fractional variable;
   GLP_BR_MFV — most fractional variable;
   GLP_BR_DTH — heuristic by Driebeck and Tomlin;
   GLP_BR_PCH — hybrid pseudocost heuristic.

int bt_tech (default: GLP_BT_BLB)
   Backtracking technique option:
   GLP_BT_DFS — depth first search;
   GLP_BT_BFS — breadth first search;
   GLP_BT_BLB — best local bound;
   GLP_BT_BPH — best projection heuristic.

int pp_tech (default: GLP_PP_ALL)
   Preprocessing technique option:
   GLP_PP_NONE — disable preprocessing;
   GLP_PP_ROOT — perform preprocessing only on the root level;
   GLP_PP_ALL — perform preprocessing on all levels.

int fp_heur (default: GLP_OFF)
   Feasibility pump heuristic option:
   GLP_ON — enable applying the feasibility pump heuristic;
   GLP_OFF — disable applying the feasibility pump heuristic.

int gmi_cuts (default: GLP_OFF)
   Gomory's mixed integer cut option:
   GLP_ON — enable generating Gomory’s cuts;
   GLP_OFF — disable generating Gomory’s cuts.

int mir_cuts (default: GLP_OFF)
   Mixed integer rounding (MIR) cut option:
   GLP_ON — enable generating MIR cuts;
   GLP_OFF — disable generating MIR cuts.
int cov_cuts (default: GLP_OFF)
Mixed cover cut option:
GLP_ON — enable generating mixed cover cuts;
GLP_OFF — disable generating mixed cover cuts.

int clq_cuts (default: GLP_OFF)
Clique cut option:
GLP_ON — enable generating clique cuts;
GLP_OFF — disable generating clique cuts.

double tol_int (default: 1e-5)
Absolute tolerance used to check if optimal solution to the current LP relaxation is integer feasible. (Do not change this parameter without detailed understanding its purpose.)

double tol_obj (default: 1e-7)
Relative tolerance used to check if the objective value in optimal solution to the current LP relaxation is not better than in the best known integer feasible solution. (Do not change this parameter without detailed understanding its purpose.)

double mip_gap (default: 0.0)
The relative mip gap tolerance. If the relative mip gap for currently known best integer feasible solution falls below this tolerance, the solver terminates the search. This allows obtaining suboptimal integer feasible solutions if solving the problem to optimality takes too long time.

int tm_lim (default: INT_MAX)
Searching time limit, in milliseconds.

int out_frq (default: 5000)
Output frequency, in milliseconds. This parameter specifies how frequently the solver sends information about the solution process to the terminal.

int out_dly (default: 10000)
Output delay, in milliseconds. This parameter specifies how long the solver should delay sending information about solution of the current LP relaxation with the simplex method to the terminal.

void (*cb_func)(glp_tree *tree, void *info) (default: NULL)
Entry point to the user-defined callback routine. NULL means the advanced solver interface is not used. For more details see Chapter “Branch-and-Cut API Routines”.

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void *cb_info (default: NULL)
Transit pointer passed to the routine cb_func (see above).

int cb_size (default: 0)
The number of extra (up to 256) bytes allocated for each node of the
branch-and-bound tree to store application-specific data. On creating a
node these bytes are initialized by binary zeros.

int presolve (default: GLP_OFF)
MIP presolver option:
GLP_ON — enable using the MIP presolver;
GLP_OFF — disable using the MIP presolver.

int binarize (default: GLP_OFF)
Binarization option (used only if the presolver is enabled):
GLP_ON — replace general integer variables by binary ones;
GLP_OFF — do not use binarization.

2.9.6 glp_init_iocp—initialize integer optimizer control pa-
rameters

Synopsis

void glp_init_iocp(glp_iocp *parm);

Description

The routine glp_init_iocp initializes control parameters, which are used
by the branch-and-cut solver, with default values.

Default values of the control parameters are stored in a glp_iocp struc-
ture, which the parameter parm points to.

2.9.7 glp_mip_status—determine status of MIP solution

Synopsis

int glp_mip_status(glp_prob *mip);

Returns

The routine glp_mip_status reports the status of a MIP solution found by
the MIP solver as follows:
GLP_UNDEF  MIP solution is undefined.
GLP_OPT     MIP solution is integer optimal.
GLP_FEAS    MIP solution is integer feasible, however, its optimality (or non-optimality) has not been proven, perhaps due to premature termination of the search.
GLP_NOFEAS  problem has no integer feasible solution (proven by the solver).

2.9.8  glp_mip_obj_val—retrieve objective value

Synopsis

double glp_mip_obj_val(glp_prob *mip);

Returns

The routine glp_mip_obj_val returns value of the objective function for MIP solution.

2.9.9  glp_mip_row_val—retrieve row value

Synopsis

double glp_mip_row_val(glp_prob *mip, int i);

Returns

The routine glp_mip_row_val returns value of the auxiliary variable associated with i-th row for MIP solution.

2.9.10 glp_mip_col_val—retrieve column value

Synopsis

double glp_mip_col_val(glp_prob *mip, int j);

Returns

The routine glp_mip_col_val returns value of the structural variable associated with j-th column for MIP solution.
2.10 Additional routines

2.10.1 lpx_check_kkt—check Karush-Kuhn-Tucker optimality conditions

Synopsis

void lpx_check_kkt(glp_prob *lp, int scaled, LPXKKT *kkt);

Description

The routine lpx_check_kkt checks Karush-Kuhn-Tucker optimality conditions for basic solution. It is assumed that both primal and dual components of basic solution are valid.

If the parameter scaled is zero, the optimality conditions are checked for the original, unscaled LP problem. Otherwise, if the parameter scaled is non-zero, the routine checks the conditions for an internally scaled LP problem.

The parameter kkt is a pointer to the structure LPXKKT, to which the routine stores results of the check. Members of this structure are shown in the table below.

The routine performs all computations using only components of the given LP problem and the current basic solution.

Background

The first condition checked by the routine is:

\[ x_R - Ax_S = 0, \quad (KKT.PE) \]

where \( x_R \) is the subvector of auxiliary variables (rows), \( x_S \) is the subvector of structural variables (columns), \( A \) is the constraint matrix. This condition expresses the requirement that all primal variables must satisfy to the system of equality constraints of the original LP problem. In case of exact arithmetic this condition would be satisfied for any basic solution; however, in case of inexact (floating-point) arithmetic, this condition shows how accurate the primal basic solution is, that depends on accuracy of a representation of the basis matrix used by the simplex method routines.

The second condition checked by the routine is:

\[ l_k \leq x_k \leq u_k \quad \text{for all} \quad k = 1, \ldots, m + n, \quad (KKT.PB) \]
where $x_k$ is auxiliary ($1 \leq k \leq m$) or structural ($m+1 \leq k \leq m+n$) variable, $l_k$ and $u_k$ are, respectively, lower and upper bounds of the variable $x_k$ (including cases of infinite bounds). This condition expresses the requirement that all primal variables must satisfy to bound constraints of the original LP problem. Since in case of basic solution all non-basic variables are placed on their bounds, actually the condition (KKT.PB) needs to be checked for basic variables only. If the primal basic solution has sufficient accuracy, this condition shows primal feasibility of the solution.

The third condition checked by the routine is:

$$\nabla Z = c = (\tilde{A})^T \pi + d,$$

where $Z$ is the objective function, $c$ is the vector of objective coefficients, $(\tilde{A})^T$ is a matrix transposed to the expanded constraint matrix $\tilde{A} = (I - A)$, $\pi$ is a vector of Lagrange multipliers that correspond to equality constraints of the original LP problem, $d$ is a vector of Lagrange multipliers that correspond to bound constraints for all (auxiliary and structural) variables of
the original LP problem. Geometrically the third condition expresses the requirement that the gradient of the objective function must belong to the orthogonal complement of a linear subspace defined by the equality and active bound constraints, i.e. that the gradient must be a linear combination of normals to the constraint planes, where Lagrange multipliers $\pi$ and $d$ are coefficients of that linear combination.

To eliminate the vector $\pi$ the third condition can be rewritten as:

$$
\begin{pmatrix}
I \\
-A^T
\end{pmatrix} \pi = \begin{pmatrix} d_R \\ d_S \end{pmatrix} + \begin{pmatrix} c_R \\ c_S \end{pmatrix},
$$

or, equivalently:

$$
\pi + d_R = c_R, \\
-A^T \pi + d_S = c_S.
$$

Then substituting the vector $\pi$ from the first equation into the second one we have:

$$
A^T (d_R - c_R) + (d_S - c_S) = 0, \quad \text{(KKT.DE)}
$$

where $d_R$ is the subvector of reduced costs of auxiliary variables (rows), $d_S$ is the subvector of reduced costs of structural variables (columns), $c_R$ and $c_S$ are subvectors of objective coefficients at, respectively, auxiliary and structural variables, $A^T$ is a matrix transposed to the constraint matrix of the original LP problem. In case of exact arithmetic this condition would be satisfied for any basic solution; however, in case of inexact (floating-point) arithmetic, this condition shows how accurate the dual basic solution is, that depends on accuracy of a representation of the basis matrix used by the simplex method routines.

The last, fourth condition checked by the routine is (KKT.DB):

$$
d_k = 0, \quad \text{if } x_k \text{ is basic or free non-basic variable}
$$

$$
0 \leq d_k < +\infty, \quad \text{if } x_k \text{ is non-basic on its lower (minimization) or upper (maximization) bound}
$$

$$
-\infty < d_k \leq 0, \quad \text{if } x_k \text{ is non-basic on its upper (minimization) or lower (maximization) bound}
$$

$$
-\infty < d_k < +\infty, \quad \text{if } x_k \text{ is non-basic fixed variable}
$$

for all $k = 1, \ldots, m+n$, where $d_k$ is a reduced cost (Lagrange multiplier) of auxiliary ($1 \leq k \leq m$) or structural ($m+1 \leq k \leq m+n$) variable $x_k$. Geometrically this condition expresses the requirement that constraints of the original problem must "hold" the point preventing its movement along the anti-gradient (in case of minimization) or the gradient (in case of maximization) of the objective function. Since in case of basic solution reduced
costs of all basic variables are placed on their (zero) bounds, actually the condition (KKT.DB) needs to be checked for non-basic variables only. If the dual basic solution has sufficient accuracy, this condition shows dual feasibility of the solution.

Should note that the complete set of Karush-Kuhn-Tucker optimality conditions also includes the fifth, so called complementary slackness condition, which expresses the requirement that at least either a primal variable \( x_k \) or its dual counterpart \( d_k \) must be on its bound for all \( k = 1, \ldots, m + n \). However, being always satisfied by definition for any basic solution that condition is not checked by the routine.

To check the first condition (KKT.PE) the routine computes a vector of residuals:

\[
g = x_R - Ax_S,
\]
determines component of this vector that correspond to largest absolute and relative errors:

\[
\begin{align*}
pe_{ae\_max} &= \max_{1 \leq i \leq m} |g_i|, \\
pe_{re\_max} &= \max_{1 \leq i \leq m} \frac{|g_i|}{1 + |(x_R)_i|},
\end{align*}
\]
and stores these quantities and corresponding row indices to the structure LPXKKT.

To check the second condition (KKT.PB) the routine computes a vector of residuals:

\[
h_k = \begin{cases} 
0, & \text{if } l_k \leq x_k \leq u_k \\
x_k - l_k, & \text{if } x_k < l_k \\
x_k - u_k, & \text{if } x_k > u_k
\end{cases}
\]
for all \( k = 1, \ldots, m + n \), determines components of this vector that correspond to largest absolute and relative errors:

\[
\begin{align*}
pb_{ae\_max} &= \max_{1 \leq k \leq m+n} |h_k|, \\
pb_{re\_max} &= \max_{1 \leq k \leq m+n} \frac{|h_k|}{1 + |x_k|},
\end{align*}
\]
and stores these quantities and corresponding variable indices to the structure LPXKKT.

To check the third condition (KKT.DE) the routine computes a vector of residuals:

\[
u = A^T(d_R - c_R) + (d_S - c_S),
\]
determines components of this vector that correspond to largest absolute and relative errors:

\[
de_{\text{ae\_max}} = \max_{1 \leq j \leq n} |u_j|,
\]

\[
de_{\text{re\_max}} = \max_{1 \leq j \leq n} \frac{|u_j|}{1 + |(dS)_j - (cS)_j|},
\]

and stores these quantities and corresponding column indices to the structure \text{LPXKKT}.

To check the fourth condition (KKT.DB) the routine computes a vector of residuals:

\[
v_k = \begin{cases} 
0, & \text{if } d_k \text{ has correct sign} \\
d_k, & \text{if } d_k \text{ has wrong sign} 
\end{cases}
\]

for all \( k = 1, \ldots, m + n \), determines components of this vector that correspond to largest absolute and relative errors:

\[
\text{db}_{\text{ae\_max}} = \max_{1 \leq k \leq m+n} |v_k|,
\]

\[
\text{db}_{\text{re\_max}} = \max_{1 \leq k \leq m+n} \frac{|v_k|}{1 + |d_k - c_k|},
\]

and stores these quantities and corresponding variable indices to the structure \text{LPXKKT}.

Using the relative errors for all the four conditions listed above the routine \text{lpx\_check\_kkt} also estimates a "quality" of the basic solution from the standpoint of these conditions and stores corresponding quality indicators to the structure \text{LPXKKT}:

- \text{pe\_quality}—quality of primal solution;
- \text{pb\_quality}—quality of primal feasibility;
- \text{de\_quality}—quality of dual solution;
- \text{db\_quality}—quality of dual feasibility.

Each of these indicators is assigned to one of the following four values:

'\text{H}' means high quality,

'\text{M}' means medium quality,

'\text{L}' means low quality, or

'?\text{}' means wrong or infeasible solution.

If all the indicators show high or medium quality (for an internally scaled LP problem, i.e. when the parameter \text{scaled} in a call to the routine \text{lpx\_check\_kkt} is non-zero), the user can be sure that the obtained basic solution is quite accurate.
If some of the indicators show low quality, the solution can still be considered as relevant, though an additional analysis is needed depending on which indicator shows low quality.

If the indicator pe_quality is assigned to '?', the primal solution is wrong. If the indicator de_quality is assigned to '?', the dual solution is wrong.

If the indicator db_quality is assigned to '?' while other indicators show a good quality, this means that the current basic solution being primal feasible is not dual feasible. Similarly, if the indicator pb_quality is assigned to '?' while other indicators are not, this means that the current basic solution being dual feasible is not primal feasible.
Chapter 3

Utility API routines

3.1 Problem data reading/writing routines

3.1.1 glp_read_mps—read problem data in MPS format

Synopsis

```
int glp_read_mps(glp_prob *lp, int fmt, const void *parm, 
                         const char *fname);
```

Description

The routine `glp_read_mps` reads problem data in MPS format from a text file. (The MPS format is described in Appendix B, page 191.)

The parameter `fmt` specifies the MPS format version as follows:
- `GLP_MPS_DECK` fixed (ancient) MPS format;
- `GLP_MPS_FILE` free (modern) MPS format.

The parameter `parm` is reserved for use in the future and must be specified as `NULL`.

The character string `fname` specifies a name of the text file to be read in. (If the file name ends with suffix `.gz`, the file is assumed to be compressed, in which case the routine `glp_read_mps` decompresses it “on the fly”.)

Note that before reading data the current content of the problem object is completely erased with the routine `glp_erase_prob`.

Returns

If the operation was successful, the routine `glp_read_mps` returns zero. Otherwise, it prints an error message and returns non-zero.
3.1.2  glp_write_mps—write problem data in MPS format

Synopsis

int glp_write_mps(glp_prob *lp, int fmt, const void *parm,
                  const char *fname);

Description

The routine glp_write_mps writes problem data in MPS format to a text file. (The MPS format is described in Appendix B, page 191.)

The parameter fmt specifies the MPS format version as follows:
- GLP_MPS_DECK fixed (ancient) MPS format;
- GLP_MPS_FILE free (modern) MPS format.

The parameter parm is reserved for use in the future and must be specified as NULL.

The character string fname specifies a name of the text file to be written out. (If the file name ends with suffix ‘.gz’, the file is assumed to be compressed, in which case the routine glp_write_mps performs automatic compression on writing it.)

Returns

If the operation was successful, the routine glp_write_mps returns zero. Otherwise, it prints an error message and returns non-zero.

3.1.3  glp_read_lp—read problem data in CPLEX LP format

Synopsis

int glp_read_lp(glp_prob *lp, const void *parm,
                const char *fname);

Description

The routine glp_read_lp reads problem data in CPLEX LP format from a text file. (The CPLEX LP format is described in Appendix C, page 204.)

The parameter parm is reserved for use in the future and must be specified as NULL.

The character string fname specifies a name of the text file to be read in. (If the file name ends with suffix ‘.gz’, the file is assumed to be compressed, in which case the routine glp_read_lp decompresses it “on the fly”.)
Note that before reading data the current content of the problem object is completely erased with the routine `glp_erase_prob`.

**Returns**

If the operation was successful, the routine `glp_read_lp` returns zero. Otherwise, it prints an error message and returns non-zero.

### 3.1.4 `glp_write_lp`—write problem data in CPLEX LP format

**Synopsis**

```c
int glp_write_lp(glp_prob *lp, const void *parm, const char *fname);
```

**Description**

The routine `glp_write_lp` writes problem data in CPLEX LP format to a text file. (The CPLEX LP format is described in Appendix C, page 204.)

The parameter `parm` is reserved for use in the future and must be specified as `NULL`.

The character string `fname` specifies a name of the text file to be written out. (If the file name ends with suffix `.gz`, the file is assumed to be compressed, in which case the routine `glp_write_lp` performs automatic compression on writing it.)

**Returns**

If the operation was successful, the routine `glp_write_lp` returns zero. Otherwise, it prints an error message and returns non-zero.

### 3.1.5 `glp_read_prob`—read problem data in GLPK format

**Synopsis**

```c
int glp_read_prob(glp_prob *P, int flags, const char *fname);
```

**Description**

The routine `glp_read_prob` reads problem data in the GLPK LP/MIP format from a text file. (For description of the GLPK LP/MIP format see below.)
The parameter **flags** is reserved for use in the future and should be specified as zero.

The character string **fname** specifies a name of the text file to be read in. (If the file name ends with suffix `.gz`, the file is assumed to be compressed, in which case the routine `glp_read_prob` decompresses it “on the fly”.)

Note that before reading data the current content of the problem object is completely erased with the routine `glp_erase_prob`.

**Returns**

If the operation was successful, the routine `glp_read_prob` returns zero. Otherwise, it prints an error message and returns non-zero.

**GLPK LP/MIP format**

The GLPK LP/MIP format is a DIMACS-like format.\(^1\) The file in this format is a plain ASCII text file containing lines of several types described below. A line is terminated with the end-of-line character. Fields in each line are separated by at least one blank space. Each line begins with a one-character designator to identify the line type.

The first line of the data file must be the problem line (except optional comment lines, which may precede the problem line). The last line of the data file must be the end line. Other lines may follow in arbitrary order, however, duplicate lines are not allowed.

**Comment lines.** Comment lines give human-readable information about the data file and are ignored by GLPK routines. Comment lines can appear anywhere in the data file. Each comment line begins with the lower-case character `c`.

```
c This is an example of comment line
```

**Problem line.** There must be exactly one problem line in the data file. This line must appear before any other lines except comment lines and has the following format:

```
p CLASS DIR ROWS COLS NONZ
```

\(^1\)The DIMACS formats were developed by the Center for Discrete Mathematics and Theoretical Computer Science (DIMACS) to facilitate exchange of problem data. For details see: <http://dimacs.rutgers.edu/Challenges/>.
The lower-case letter \( p \) specifies that this is the problem line.

The \texttt{CLASS} field defines the problem class and can contain either the keyword \texttt{lp} (that means linear programming problem) or \texttt{mip} (that means mixed integer programming problem).

The \texttt{DIR} field defines the optimization direction (that is, the objective function sense) and can contain either the keyword \texttt{min} (that means minimization) or \texttt{max} (that means maximization).

The \texttt{ROWS}, \texttt{COLS}, and \texttt{NONZ} fields contain non-negative integer values specifying, respectively, the number of rows (constraints), columns (variables), and non-zero constraint coefficients in the problem instance. Note that \texttt{NONZ} value does not account objective coefficients.

**Row descriptors.** There must be at most one row descriptor line in the data file for each row (constraint). This line has one of the following formats:

\begin{verbatim}
  i ROW f
  i ROW l RHS
  i ROW u RHS
  i ROW d RHS1 RHS2
  i ROW s RHS
\end{verbatim}

The lower-case letter \( i \) specifies that this is the row descriptor line. The \texttt{ROW} field specifies the row ordinal number, an integer between 1 and \( m \), where \( m \) is the number of rows in the problem instance.

The next lower-case letter specifies the row type as follows:

- \( f \) — free (unbounded) row: \(-\infty < \sum a_j x_j < +\infty\);
- \( l \) — inequality constraint of ‘\( \geq \)’ type: \( \sum a_j x_j \geq b \);
- \( u \) — inequality constraint of ‘\( \leq \)’ type: \( \sum a_j x_j \leq b \);
- \( d \) — double-sided inequality constraint: \( b_1 \leq \sum a_j x_j \leq b_2 \);
- \( s \) — equality constraint: \( \sum a_j x_j = b \).

The \texttt{RHS} field contains a floating-point value specifying the row right-hand side. The \texttt{RHS1} and \texttt{RHS2} fields contain floating-point values specifying, respectively, the lower and upper right-hand sides for the double-sided row.

If for some row its descriptor line does not appear in the data file, by default that row is assumed to be an equality constraint with zero right-hand side.

**Column descriptors.** There must be at most one column descriptor line in the data file for each column (variable). This line has one of the following formats depending on the problem class specified in the problem line:

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LP class | MIP class
---|---
j COL f | j COL KIND f
j COL l BND | j COL KIND l BND
j COL u BND | j COL KIND u BND
j COL d BND1 BND2 | j COL KIND d BND1 BND2
j COL s BND | j COL KIND s BND

The lower-case letter \( j \) specifies that this is the column descriptor line.
The \( \text{COL} \) field specifies the column ordinal number, an integer between 1 and \( n \), where \( n \) is the number of columns in the problem instance.
The \( \text{KIND} \) field is used only for MIP problems and specifies the column kind as follows:
c — continuous column;
i — integer column;
b — binary column (in this case all remaining fields must be omitted).
The next lower-case letter specifies the column type as follows:
f — free (unbounded) column: \(-\infty < x < +\infty\);
l — column with lower bound: \( x \geq l \);
u — column with upper bound: \( x \leq u \);
d — double-bounded column: \( l \leq x \leq u \);
s — fixed column: \( x = s \).

The \( \text{BND} \) field contains a floating-point value that specifies the column bound. The \( \text{BND1} \) and \( \text{BND2} \) fields contain floating-point values specifying, respectively, the lower and upper bounds for the double-bounded column.

If for some column its descriptor line does not appear in the file, by default that column is assumed to be non-negative (in case of LP class) or binary (in case of MIP class).

**Coefficient descriptors.** There must be exactly one coefficient descriptor line in the data file for each non-zero objective or constraint coefficient. This line has the following format:

\[ a \ ROW \ COL \ VAL \]

The lower-case letter \( a \) specifies that this is the coefficient descriptor line.
For objective coefficients the \( \text{ROW} \) field must contain 0. For constraint coefficients the \( \text{ROW} \) field specifies the row ordinal number, an integer between 1 and \( m \), where \( m \) is the number of rows in the problem instance.
The \( \text{COL} \) field specifies the column ordinal number, an integer between 1 and \( n \), where \( n \) is the number of columns in the problem instance.
If both the \texttt{ROW} and \texttt{COL} fields contain 0, the line specifies the constant term ("shift") of the objective function rather than objective coefficient.

The \texttt{VAL} field contains a floating-point coefficient value (it is allowed to specify zero value in this field).

The number of constraint coefficient descriptor lines must be exactly the same as specified in the field \texttt{NONZ} of the problem line.

**Symbolic name descriptors.** There must be at most one symbolic name descriptor line for the problem instance, objective function, each row (constraint), and each column (variable). This line has one of the following formats:

\begin{verbatim}
 n  p  NAME
 n  z  NAME
 n  i  ROW  NAME
 n  j  COL  NAME
\end{verbatim}

The lower-case letter \texttt{n} specifies that this is the symbolic name descriptor line.

The next lower-case letter specifies which object should be assigned a symbolic name:

- \texttt{p} — problem instance;
- \texttt{z} — objective function;
- \texttt{i} — row (constraint);
- \texttt{j} — column (variable).

The \texttt{ROW} field specifies the row ordinal number, an integer between 1 and \(m\), where \(m\) is the number of rows in the problem instance.

The \texttt{COL} field specifies the column ordinal number, an integer between 1 and \(n\), where \(n\) is the number of columns in the problem instance.

The \texttt{NAME} field contains the symbolic name, a sequence from 1 to 255 arbitrary graphic ASCII characters, assigned to corresponding object.

**End line.** There must be exactly one end line in the data file. This line must appear last in the file and has the following format:

\begin{verbatim}
 e
\end{verbatim}

The lower-case letter \texttt{e} specifies that this is the end line. Anything that follows the end line is ignored by GLPK routines.
Example of data file in GLPK LP/MIP format

The following example of a data file in GLPK LP/MIP format specifies the same LP problem as in Subsection “Example of MPS file”.

```
p lp min 8 7 48
n p PLAN
n z VALUE
i 1 f
n i 1 VALUE
i 2 s 2000
n i 2 YIELD
i 3 u 60
n i 3 FE
n i 4 u 100
n i 4 CU
i 5 u 40
n i 5 MN
i 6 u 30
n i 6 MG
i 7 1 1500
n i 7 AL
i 8 d 250 300
n i 8 SI
j 1 d 0 200
n j 1 BIN1
j 2 d 0 2500
n j 2 BIN2
j 3 d 400 800
n j 3 BIN3
j 4 d 100 700
n j 4 BIN4
j 5 d 0 1500
n j 5 BIN5
n j 6 ALUM
j 7 7 SILICON
a 0 1 0.03
a 0 2 0.08
a 0 3 0.17
a 0 4 0.12
a 0 5 0.15
a 0 6 0.21
a 0 7 0.38
a 1 1 0.03
a 1 2 0.08
a 1 3 0.17
a 1 4 0.12
a 1 5 0.15
a 1 6 0.21
```
3.1.6 glp_write_prob—write problem data in GLPK format

Synopsis

int glp_write_prob(glp_prob *P, int flags, const char *fname);

Description

The routine glp_write_prob writes problem data in the GLPK LP/MIP format to a text file. (For description of the GLPK LP/MIP format see Subsection “Read problem data in GLPK format”.)

The parameter flags is reserved for use in the future and should be specified as zero.

The character string fname specifies a name of the text file to be written out. (If the file name ends with suffix ‘.gz’, the file is assumed to be compressed, in which case the routine glp_write_prob performs automatic compression on writing it.)

Returns

If the operation was successful, the routine glp_read_prob returns zero. Otherwise, it prints an error message and returns non-zero.
3.2 Routines for processing MathProg models

3.2.1 Introduction

GLPK supports the GNU MathProg modeling language. As a rule, models written in MathProg are solved with the GLPK LP/MIP stand-alone solver `glpsol` (see Appendix D) and do not need any programming with API routines. However, for various reasons the user may need to process MathProg models directly in his/her application program, in which case he/she may use API routines described in this section. These routines provide an interface to the MathProg translator, a component of GLPK, which translates MathProg models into an internal code and then interprets (executes) this code.

The processing of a model written in GNU MathProg includes several steps, which should be performed in the following order:

1. Allocating the workspace. The translator allocates the workspace, an internal data structure used on all subsequent steps.

2. Reading model section. The translator reads model section and, optionally, data section from a specified text file and translates them into the internal code. If necessary, on this step data section may be ignored.

3. Reading data section(s). The translator reads one or more data sections from specified text file(s) and translates them into the internal code.

4. Generating the model. The translator executes the internal code to evaluate the content of the model objects such as sets, parameters, variables, constraints, and objectives. On this step the execution is suspended at the solve statement.

5. Building the problem object. The translator obtains all necessary information from the workspace and builds the standard problem object (that is, the program object of type `glp_prob`).

6. Solving the problem. On this step the problem object built on the previous step is passed to a solver, which solves the problem instance and stores its solution back to the problem object.

2The GNU MathProg modeling language is a subset of the AMPL language. For its detailed description see the document “Modeling Language GNU MathProg: Language Reference” included in the GLPK distribution.
7. **Postsolving the model.** The translator copies the solution from the problem object to the workspace and then executes the internal code from the solve statement to the end of the model. (If model has no solve statement, the translator does nothing on this step.)

8. **Freeing the workspace.** The translator frees all the memory allocated to the workspace.

Note that the MathProg translator performs no error correction, so if any of steps 2 to 7 fails (due to errors in the model), the application program should terminate processing and go to step 8.

**Example 1**

In this example the program reads model and data sections from input file `egypt.mod` and writes the model to output file `egypt.mps` in free MPS format (see Appendix B). No solution is performed.

```c
/* mplsamp1.c */

#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
    glp_prob *lp;
    glp_tran *tran;
    int ret;
    lp = glp_create_prob();
    tran = glp_mpl_alloc_wksp();
    ret = glp_mpl_read_model(tran, "egypt.mod", 0);
    if (ret != 0)
    {
        fprintf(stderr, "Error on translating model\n");
        goto skip;
    }
    ret = glp_mpl_generate(tran, NULL);
    if (ret != 0)
    {
        fprintf(stderr, "Error on generating model\n");
        goto skip;
    }
    glp_mpl_build_prob(tran, lp);
    ret = glp_write_mps(lp, GLP_MPS_FILE, NULL, "egypt.mps");
```

This is an example model included in the GLPK distribution.
if (ret != 0)
    fprintf(stderr, "Error on writing MPS file\n");
skip: glp_mpl_free_wksp(tran);
glp_delete_prob(lp);
return 0;
}

/* eof */

Example 2
In this example the program reads model section from file sudoku.mod\(^4\)
ignoring data section in this file, reads alternative data section from file
sudoku.dat, solves the problem instance and passes the solution found back
to the model.

/* mplsamp2.c */

#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{   glp_prob *mip;
    glp_tran *tran;
    int ret;
    mip = glp_create_prob();
    tran = glp_mpl_alloc_wksp();
    ret = glp_mpl_read_model(tran, "sudoku.mod", 1);
    if (ret != 0)
    {
        fprintf(stderr, "Error on translating model\n");
        goto skip;
    }
    ret = glp_mpl_read_data(tran, "sudoku.dat");
    if (ret != 0)
    {
        fprintf(stderr, "Error on translating data\n");
        goto skip;
    }
    ret = glp_mpl_generate(tran, NULL);
    if (ret != 0)
    {
        fprintf(stderr, "Error on generating model\n");
        goto skip;
    }
}\(^4\)This is an example model which is included in the GLPK distribution along with
alternative data file sudoku.dat.

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glp_mpl_build_prob(tran, mip);
glp_simplex(mip, NULL);
glp_intopt(mip, NULL);
ret = glp_mpl_postsolve(tran, mip, GLP_MIP);
if (ret != 0)
    fprintf(stderr, "Error on postsolving model\n");
skip: glp_mpl_free_wksp(tran);
glp_delete_prob(mip);
return 0;
}

/* eof */

3.2.2 glp_mpl_alloc_wksp—allocate the translator workspace

Synopsis

glp_tran *glp_mpl_alloc_wksp(void);

Description

The routine glp_mpl_alloc_wksp allocates the MathProg translator workspace. (Note that multiple instances of the workspace may be allocated, if necessary.)

Returns

The routine returns a pointer to the workspace, which should be used in all subsequent operations.

3.2.3 glp_mpl_read_model—read and translate model section

Synopsis

int glp_mpl_read_model(glp_tran *tran, const char *fname, int skip);

Description

The routine glp_mpl_read_model reads model section and, optionally, data section, which may follow the model section, from a text file, whose name is the character string fname, performs translation of model statements and data blocks, and stores all the information in the workspace.
The parameter skip is a flag. If the input file contains the data section and this flag is non-zero, the data section is not read as if there were no data section and a warning message is printed. This allows reading data section(s) from other file(s).

**Returns**

If the operation is successful, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.

### 3.2.4 glp_mpl_read_data—read and translate data section

**Synopsis**

```c
int glp_mpl_read_data(glp_tran *tran, const char *fname);
```

**Description**

The routine `glp_mpl_read_data` reads data section from a text file, whose name is the character string `fname`, performs translation of data blocks, and stores the data read in the translator workspace. If necessary, this routine may be called more than once.

**Returns**

If the operation is successful, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.

### 3.2.5 glp_mpl_generate—generate the model

**Synopsis**

```c
int glp_mpl_generate(glp_tran *tran, const char *fname);
```

**Description**

The routine `glp_mpl_generate` generates the model using its description stored in the translator workspace. This operation means generating all variables, constraints, and objectives, executing check and display statements, which precede the solve statement (if it is presented).

The character string `fname` specifies the name of an output text file, to which output produced by display statements should be written. If `fname` is NULL, the output is sent to the terminal.
Returns  
If the operation is successful, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.

3.2.6 glp_mpl_build_prob—build problem instance from the model

Synopsis

void glp_mpl_build_prob(glp_tran *tran, glp_prob *prob);

Description

The routine glp_mpl_build_prob obtains all necessary information from the translator workspace and stores it in the specified problem object prob. Note that before building the current content of the problem object is erased with the routine glp_erase_prob.

3.2.7 glp_mpl_postsolve—postsolve the model

Synopsis

int glp_mpl_postsolve(glp_tran *tran, glp_prob *prob,  
                      int sol);

Description

The routine glp_mpl_postsolve copies the solution from the specified problem object prob to the translator workspace and then executes all the remaining model statements, which follow the solve statement.

  The parameter sol specifies which solution should be copied from the problem object to the workspace as follows:

  GLP_SOL  basic solution;
  GLP_IPT  interior-point solution;
  GLP_MIP  mixed integer solution.

Returns

If the operation is successful, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.
3.2.8  glp_mpl_free_wksp—free the translator workspace

Synopsis

void glp_mpl_free_wksp(glp_tran *tran);

Description

The routine `glp_mpl_free_wksp` frees all the memory allocated to the translator workspace. It also frees all other resources, which are still used by the translator.
3.3 Problem solution reading/writing routines

3.3.1 glp_print_sol—write basic solution in printable format

Synopsis

```c
int glp_print_sol(glp_prob *lp, const char *fname);
```

Description

The routine `glp_print_sol` writes the current basic solution of an LP problem, which is specified by the pointer `lp`, to a text file, whose name is the character string `fname`, in printable format.

Information reported by the routine `glp_print_sol` is intended mainly for visual analysis.

Returns

If no errors occurred, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.

3.3.2 glp_read_sol—read basic solution from text file

Synopsis

```c
int glp_read_sol(glp_prob *lp, const char *fname);
```

Description

The routine `glp_read_sol` reads basic solution from a text file whose name is specified by the parameter `fname` into the problem object.

For the file format see description of the routine `glp_write_sol`.

Returns

On success the routine returns zero, otherwise non-zero.
3.3.3 glp_write_sol—write basic solution to text file

Synopsis

```c
int glp_write_sol(glp_prob *lp, const char *fname);
```

Description

The routine `glp_write_sol` writes the current basic solution to a text file whose name is specified by the parameter `fname`. This file can be read back with the routine `glp_read_sol`.

Returns

On success the routine returns zero, otherwise non-zero.

File format

The file created by the routine `glp_write_sol` is a plain text file, which contains the following information:

```plaintext
m n
p_stat d_stat obj_val
... 
r_stat[m] r_prim[m] r_dual[m]
... 
c_stat[n] c_prim[n] c_dual[n]
```

where:
- `m` is the number of rows (auxiliary variables);
- `n` is the number of columns (structural variables);
- `p_stat` is the primal status of the basic solution (`GLP_UNDEF` = 1, `GLP_FEAS` = 2, `GLP_INFEAS` = 3, or `GLP_NOFEAS` = 4);
- `d_stat` is the dual status of the basic solution (`GLP_UNDEF` = 1, `GLP_FEAS` = 2, `GLP_INFEAS` = 3, or `GLP_NOFEAS` = 4);
- `obj_val` is the objective value;
- `r_stat[i]`, `i = 1, ..., m`, is the status of `i`-th row (`GLP_BS` = 1, `GLP_NL` = 2, `GLP_NU` = 3, `GLP_NF` = 4, or `GLP_NS` = 5);
- `r_prim[i]`, `i = 1, ..., m`, is the primal value of `i`-th row;
- `r_dual[i]`, `i = 1, ..., m`, is the dual value of `i`-th row;
c_stat[j], j = 1, ..., n, is the status of j-th column (GLP_BS = 1, GLP_NL = 2, GLP_NU = 3, GLP_NF = 4, or GLP_NS = 5);
c_prim[j], j = 1, ..., n, is the primal value of j-th column;
c_dual[j], j = 1, ..., n, is the dual value of j-th column.

3.3.4 glp_print_ipt—write interior-point solution in printable format

Synopsis

int glp_print_ipt(glp_prob *lp, const char *fname);

Description

The routine glp_print_ipt writes the current interior point solution of an LP problem, which the parameter lp points to, to a text file, whose name is the character string fname, in printable format.

Information reported by the routine glp_print_ipt is intended mainly for visual analysis.

Returns

If no errors occurred, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.

3.3.5 glp_read_ipt—read interior-point solution from text file

Synopsis

int glp_read_ipt(glp_prob *lp, const char *fname);

Description

The routine glp_read_ipt reads interior-point solution from a text file whose name is specified by the parameter fname into the problem object.

For the file format see description of the routine glp_write_ipt.

Returns

On success the routine returns zero, otherwise non-zero.
3.3.6  \texttt{glp_write_ipt}—write interior-point solution to text file

Synopsis

\begin{verbatim}
int glp_write_ipt(glp_prob *lp, const char *fname);
\end{verbatim}

Description

The routine \texttt{glp_write_ipt} writes the current interior-point solution to a text file whose name is specified by the parameter \texttt{fname}. This file can be read back with the routine \texttt{glp_read_ipt}.

Returns

On success the routine returns zero, otherwise non-zero.

File format

The file created by the routine \texttt{glp_write_ipt} is a plain text file, which contains the following information:

\begin{verbatim}
m n
  stat obj_val
  r_prim[1] r_dual[1]
  . . .
  r_prim[m] r_dual[m]
  c_prim[1] c_dual[1]
  . . .
  c_prim[n] c_dual[n]
\end{verbatim}

where:

\begin{itemize}
  \item \texttt{m} is the number of rows (auxiliary variables);
  \item \texttt{n} is the number of columns (structural variables);
  \item \texttt{stat} is the solution status (\texttt{GLP_UNDEF} = 1 or \texttt{GLP_OPT} = 5);
  \item \texttt{obj_val} is the objective value;
  \item \texttt{r_prim[i]}, \texttt{i} = 1, \ldots, \texttt{m}, is the primal value of \texttt{i}-th row;
  \item \texttt{r_dual[i]}, \texttt{i} = 1, \ldots, \texttt{m}, is the dual value of \texttt{i}-th row;
  \item \texttt{c_prim[j]}, \texttt{j} = 1, \ldots, \texttt{n}, is the primal value of \texttt{j}-th column;
  \item \texttt{c_dual[j]}, \texttt{j} = 1, \ldots, \texttt{n}, is the dual value of \texttt{j}-th column.
\end{itemize}
3.3.7 glp_print_mip—write MIP solution in printable format

Synopsis

int glp_print_mip(glp_prob *lp, const char *fname);

Description

The routine glp_print_mip writes a best known integer solution of a MIP problem, which is specified by the pointer lp, to a text file, whose name is the character string fname, in printable format.

Information reported by the routine glp_print_mip is intended mainly for visual analysis.

Returns

If no errors occurred, the routine returns zero. Otherwise the routine prints an error message and returns non-zero.
3.3.8  glp_read_mip—read MIP solution from text file

Synopsis

int glp_read_mip(glp_prob *mip, const char *fname);

Description

The routine glp_read_mip reads MIP solution from a text file whose name is specified by the parameter fname into the problem object.

For the file format see description of the routine glp_write_mip.

Returns

On success the routine returns zero, otherwise non-zero.

3.3.9  glp_write_mip—write MIP solution to text file

Synopsis

int glp_write_mip(glp_prob *mip, const char *fname);

Description

The routine glp_write_mip writes the current MIP solution to a text file whose name is specified by the parameter fname. This file can be read back with the routine glp_read_mip.

Returns

On success the routine returns zero, otherwise non-zero.

File format

The file created by the routine glp_write_sol is a plain text file, which contains the following information:

```
m n
stat obj_val
r_val[1]
. . .
r_val[m]
c_val[1]
```
where:
\( m \) is the number of rows (auxiliary variables);
\( n \) is the number of columns (structural variables);
\texttt{stat} is the solution status (\texttt{GLP_UNDEF} = 1, \texttt{GLP_FEAS} = 2, \texttt{GLP_NOFEAS} = 4, or \texttt{GLP_OPT} = 5);
\texttt{obj_val} is the objective value;
\texttt{r_val}[i], \( i = 1, \ldots, m \), is the value of \( i \)-th row;
\texttt{c_val}[j], \( j = 1, \ldots, n \), is the value of \( j \)-th column.
3.4 Post-optimal analysis routines

3.4.1 glp_print_ranges—print sensitivity analysis report

Synopsis

```c
int glp_print_ranges(glp_prob *P, int len, const int list[],
                     int flags, const char *fname);
```

Description

The routine `glp_print_ranges` performs sensitivity analysis of current optimal basic solution and writes the analysis report in human-readable format to a text file, whose name is the character string `fname`. (Detailed description of the report structure is given below.)

The parameter `len` specifies the length of the row/column list.

The array `list` specifies ordinal number of rows and columns to be analyzed. The ordinal numbers should be passed in locations `list[1]`, `list[2]`, ..., `list[len]`. Ordinal numbers from 1 to `m` refer to rows, and ordinal numbers from `m + 1` to `m + n` refer to columns, where `m` and `n` are, resp., the total number of rows and columns in the problem object. Rows and columns appear in the analysis report in the same order as they follow in the array list.

It is allowed to specify `len = 0`, in which case the array `list` is not used (so it can be specified as `NULL`), and the routine performs analysis for all rows and columns of the problem object.

The parameter `flags` is reserved for use in the future and must be specified as zero.

On entry to the routine `glp_print_ranges` the current basic solution must be optimal and the basis factorization must exist. The application program can check that with the routine `glp_bf_exists`, and if the factorization does not exist, compute it with the routine `glp_factorize`. Note that if the LP preprocessor is not used, on normal exit from the simplex solver routine `glp_simplex` the basis factorization always exists.

Returns

If the operation was successful, the routine `glp_print_ranges` returns zero. Otherwise, it prints an error message and returns non-zero.
**Analysis report example**

An example of the sensitivity analysis report is shown on the next two pages. This example corresponds to the example of LP problem described in Subsection “Example of MPS file”.

**Structure of the analysis report**

For each row and column specified in the array list the routine prints two lines containing generic information and analysis information, which depends on the status of corresponding row or column.

Note that analysis of a row is analysis of its auxiliary variable, which is equal to the row linear form $\sum a_j x_j$, and analysis of a column is analysis of corresponding structural variable. Therefore, formally, on performing the sensitivity analysis there is no difference between rows and columns.

*Generic information*

**No.** is the row or column ordinal number in the problem object. Rows are numbered from 1 to $m$, and columns are numbered from 1 to $n$, where $m$ and $n$ are, resp., the total number of rows and columns in the problem object.

**Row name** is the symbolic name assigned to the row. If the row has no name assigned, this field contains blanks.

**Column name** is the symbolic name assigned to the column. If the column has no name assigned, this field contains blanks.

**St** is the status of the row or column in the optimal solution:

- **BS** — non-active constraint (row), basic column;
- **NL** — inequality constraint having its lower right-hand side active (row), non-basic column having its lower bound active;
- **NU** — inequality constraint having its upper right-hand side active (row), non-basic column having its upper bound active;
- **NS** — active equality constraint (row), non-basic fixed column.
- **NF** — active free row, non-basic free (unbounded) column. (This case means that the optimal solution is dual degenerate.)

**Activity** is the (primal) value of the auxiliary variable (row) or structural variable (column) in the optimal solution.

**Slack** is the (primal) value of the row slack variable.

**Obj coef** is the objective coefficient of the column (structural variable).
Problem: PLAN
Objective: VALUE = 296.2166065 (MINimum)

<table>
<thead>
<tr>
<th>No.</th>
<th>Row name</th>
<th>St</th>
<th>Activity</th>
<th>Slack</th>
<th>Lower bound</th>
<th>Activity range</th>
<th>Obj coef</th>
<th>Obj value at Limiting break point</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>VALUE</td>
<td>BS</td>
<td>296.21661</td>
<td>-296.21661</td>
<td>Inf</td>
<td>296.21661</td>
<td>-1.00000</td>
<td>. MN 296.21661 +Inf +Inf</td>
</tr>
<tr>
<td>3</td>
<td>FE</td>
<td>NU</td>
<td>60.00000</td>
<td>-2.56823</td>
<td>60.00000</td>
<td>55.89016</td>
<td>-Inf</td>
<td>306.77162 BIN4 +Inf</td>
</tr>
<tr>
<td>4</td>
<td>CU</td>
<td>BS</td>
<td>83.96751</td>
<td>16.03249</td>
<td>-Inf</td>
<td>93.88467</td>
<td>-Inf</td>
<td>270.51157 MN 2.56823 289.28294 BIN3</td>
</tr>
<tr>
<td>5</td>
<td>MN</td>
<td>NU</td>
<td>40.00000</td>
<td>-.544440</td>
<td>40.00000</td>
<td>34.42336</td>
<td>-Inf</td>
<td>299.25255 BIN4 -.54440 295.29825 BIN3</td>
</tr>
<tr>
<td>6</td>
<td>MG</td>
<td>BS</td>
<td>19.96029</td>
<td>10.03971</td>
<td>30.00000</td>
<td>24.74427</td>
<td>-1.79618</td>
<td>260.36433 BIN1 +Inf</td>
</tr>
<tr>
<td>7</td>
<td>AL</td>
<td>NL</td>
<td>1500.00000</td>
<td>.</td>
<td>1500.00000</td>
<td>1485.78425</td>
<td>-.25199</td>
<td>292.63444 CU 1504.92126 +Inf</td>
</tr>
<tr>
<td>8</td>
<td>SI</td>
<td>NL</td>
<td>250.00000</td>
<td>50.00000</td>
<td>300.00000</td>
<td>235.32871</td>
<td>-.48520</td>
<td>289.09812 CU 255.06073 +Inf</td>
</tr>
</tbody>
</table>
Problem: PLAN  
Objective: VALUE = 296.2166065 (MINimum)

<table>
<thead>
<tr>
<th>No.</th>
<th>Column name</th>
<th>St</th>
<th>Activity</th>
<th>Obj coef</th>
<th>Lower bound</th>
<th>Activity range</th>
<th>Obj coef</th>
<th>Obj value at Limiting break point variable</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>BIN1</td>
<td>NL</td>
<td>.</td>
<td>.03000</td>
<td>.</td>
<td>.</td>
<td>.</td>
<td>.</td>
</tr>
<tr>
<td>2</td>
<td>BIN2</td>
<td>BS</td>
<td>665.34296</td>
<td>.08000</td>
<td>2500.00000</td>
<td>313.43066</td>
<td>.01722</td>
<td>254.44822 BIN1</td>
</tr>
<tr>
<td>3</td>
<td>BIN3</td>
<td>BS</td>
<td>490.25271</td>
<td>.17000</td>
<td>800.00000</td>
<td>788.61314</td>
<td>.15982</td>
<td>291.22807 MN</td>
</tr>
<tr>
<td>4</td>
<td>BIN4</td>
<td>BS</td>
<td>424.18773</td>
<td>.12000</td>
<td>700.00000</td>
<td>710.52632</td>
<td>.10899</td>
<td>291.54745 MN</td>
</tr>
<tr>
<td>5</td>
<td>BIN5</td>
<td>NL</td>
<td>.</td>
<td>.15000</td>
<td>1500.00000</td>
<td>58.79586</td>
<td>.13544</td>
<td>293.27940 BIN3</td>
</tr>
<tr>
<td>6</td>
<td>ALUM</td>
<td>BS</td>
<td>299.63899</td>
<td>.21000</td>
<td>.</td>
<td>358.26772</td>
<td>.18885</td>
<td>289.87879 AL</td>
</tr>
<tr>
<td>7</td>
<td>SILICON</td>
<td>BS</td>
<td>120.57762</td>
<td>.38000</td>
<td>.</td>
<td>124.27093</td>
<td>.14828</td>
<td>268.27586 BIN5</td>
</tr>
</tbody>
</table>

End of report
Marginal is the reduced cost (dual activity) of the auxiliary variable (row) or structural variable (column).

Lower bound is the lower right-hand side (row) or lower bound (column). If the row or column has no lower bound, this field contains -Inf.

Upper bound is the upper right-hand side (row) or upper bound (column). If the row or column has no upper bound, this field contains +Inf.

Sensitivity analysis of active bounds

The sensitivity analysis of active bounds is performed only for rows, which are active constraints, and only for non-basic columns, because inactive constraints and basic columns have no active bounds.

For every auxiliary (row) or structural (column) non-basic variable the routine starts changing its active bound in both direction. The first of the two lines in the report corresponds to decreasing, and the second line corresponds to increasing of the active bound. Since the variable being analyzed is non-basic, its activity, which is equal to its active bound, also starts changing. This changing leads to changing of basic (auxiliary and structural) variables, which depend on the non-basic variable. The current basis remains primal feasible and therefore optimal while values of all basic variables are primal feasible, i.e. are within their bounds. Therefore, if some basic variable called the limiting variable reaches its (lower or upper) bound first, before any other basic variables, it thereby limits further changing of the non-basic variable, because otherwise the current basis would become primal infeasible. The point, at which this happens, is called the break point. Note that there are two break points: the lower break point, which corresponds to decreasing of the non-basic variable, and the upper break point, which corresponds to increasing of the non-basic variable.

In the analysis report values of the non-basic variable (i.e. of its active bound) being analyzed at both lower and upper break points are printed in the field ‘Activity range’. Corresponding values of the objective function are printed in the field ‘Obj value at break point’, and symbolic names of corresponding limiting basic variables are printed in the field ‘Limiting variable’. If the active bound can decrease or/and increase unlimitedly, the field ‘Activity range’ contains -Inf or/and +Inf, resp.

For example (see the example report above), row SI is a double-sided constraint, which is active on its lower bound (right-hand side), and its activity in the optimal solution being equal to the lower bound is 250. The activity range for this row is [235.32871, 255.06073]. This means that the basis remains optimal while the lower bound is increasing up to 255.06073,
and further increasing is limited by (structural) variable BIN3. If the lower bound reaches this upper break point, the objective value becomes equal to 298.67206.

Note that if the basis does not change, the objective function depends on the non-basic variable linearly, and the per-unit change of the objective function is the reduced cost (marginal value) of the non-basic variable.

**Sensitivity analysis of objective coefficients at non-basic variables**

The sensitivity analysis of the objective coefficient at a non-basic variable is quite simple, because in this case change in the objective coefficient leads to equivalent change in the reduced cost (marginal value).

For every auxiliary (row) or structural (column) non-basic variable the routine starts changing its objective coefficient in both direction. (Note that auxiliary variables are not included in the objective function and therefore always have zero objective coefficients.) The first of the two lines in the report corresponds to decreasing, and the second line corresponds to increasing of the objective coefficient. This changing leads to changing of the reduced cost of the non-basic variable to be analyzed and does affect reduced costs of all other non-basic variables. The current basis remains dual feasible and therefore optimal while the reduced cost keeps its sign. Therefore, if the reduced cost reaches zero, it limits further changing of the objective coefficient (if only the non-basic variable is non-fixed).

In the analysis report minimal and maximal values of the objective coefficient, on which the basis remains optimal, are printed in the field ‘Obj coef range’. If the objective coefficient can decrease or/and increase unlimitedly, this field contains -Inf or/and +Inf, resp.

For example (see the example report above), column BIN5 is non-basic having its lower bound active. Its objective coefficient is 0.15, and reduced cost in the optimal solution 0.01456. The column lower bound remains active while the column reduced cost remains non-negative, thus, minimal value of the objective coefficient, on which the current basis still remains optimal, is 0.15 – 0.01456 = 0.13644, that is indicated in the field ‘Obj coef range’.

**Sensitivity analysis of objective coefficients at basic variables**

To perform sensitivity analysis for every auxiliary (row) or structural (column) variable the routine starts changing its objective coefficient in both direction. (Note that auxiliary variables are not included in the objective function and therefore always have zero objective coefficients.) The first of the two lines in the report corresponds to decreasing, and the second line
corresponds to increasing of the objective coefficient. This changing leads to changing of reduced costs of non-basic variables. The current basis remains dual feasible and therefore optimal while reduced costs of all non-basic variables (except fixed variables) keep their signs. Therefore, if the reduced cost of some non-basic non-fixed variable called the limiting variable reaches zero first, before reduced cost of any other non-basic non-fixed variable, it thereby limits further changing of the objective coefficient, because otherwise the current basis would become dual infeasible (non-optimal). The point, at which this happens, is called the break point. Note that there are two break points: the lower break point, which corresponds to decreasing of the objective coefficient, and the upper break point, which corresponds to increasing of the objective coefficient. Let the objective coefficient reach its limit value and continue changing a bit further in the same direction that makes the current basis dual infeasible (non-optimal). Then the reduced cost of the non-basic limiting variable becomes “a bit” dual infeasible that forces the limiting variable to enter the basis replacing there some basic variable, which leaves the basis to keep its primal feasibility. It should be understood that if we change the current basis in this way exactly at the break point, both the current and adjacent bases will be optimal with the same objective value, because at the break point the limiting variable has zero reduced cost. On the other hand, in the adjacent basis the value of the limiting variable changes, because there it becomes basic, that leads to changing of the value of the basic variable being analyzed. Note that on determining the adjacent basis the bounds of the analyzed basic variable are ignored as if it were a free (unbounded) variable, so it cannot leave the current basis.

In the analysis report lower and upper limits of the objective coefficient at the basic variable being analyzed, when the basis remains optimal, are printed in the field ‘Obj coef range’. Corresponding values of the objective function at both lower and upper break points are printed in the field ‘Obj value at break point’, symbolic names of corresponding non-basic limiting variables are printed in the field ‘Limiting variable’, and values of the basic variable, which it would take on in the adjacent bases (as was explained above) are printed in the field ‘Activity range’. If the objective coefficient can increase or/and decrease unlimitedly, the field ‘Obj coef range’ contains -Inf and/or +Inf, resp. It also may happen that no dual feasible adjacent basis exists (i.e. on entering the basis the limiting variable can increase or decrease unlimitedly), in which case the field ‘Activity range’ contains -Inf and/or +Inf.
For example (see the example report above), structural variable (column) BIN3 is basic, its optimal value is 490.25271, and its objective coefficient is 0.17. The objective coefficient range for this column is \([0.15982, 0.17948]\). This means that the basis remains optimal while the objective coefficient is decreasing down to 0.15982, and further decreasing is limited by (auxiliary) variable MN. If we make the objective coefficient a bit less than 0.15982, the limiting variable MN will enter the basis, and in that adjacent basis the structural variable BIN3 will take on new optimal value 788.61314. At the lower break point, where the objective coefficient is exactly 0.15982, the objective function takes on the value 291.22807 in both the current and adjacent bases.

Note that if the basis does not change, the objective function depends on the objective coefficient at the basic variable linearly, and the per-unit change of the objective function is the value of the basic variable.
4.1 Background

Using vector and matrix notations LP problem (1.1)—(1.3) (see Section 1.1, page 12) can be stated as follows:

minimize (or maximize) \[ z = c^T x_S + c_0 \] (3.1)

subject to linear constraints

\[ x_R = A x_S \] (3.2)

and bounds of variables

\[ l_R \leq x_R \leq u_R \]
\[ l_S \leq x_S \leq u_S \] (3.3)

where:
\( x_R = (x_1, \ldots, x_m) \) is the vector of auxiliary variables;
\( x_S = (x_{m+1}, \ldots, x_{m+n}) \) is the vector of structural variables;
\( z \) is the objective function;
\( c = (c_1, \ldots, c_n) \) is the vector of objective coefficients;
\( c_0 \) is the constant term (“shift”) of the objective function;
\( A = (a_{11}, \ldots, a_{mn}) \) is the constraint matrix;
\( l_R = (l_1, \ldots, l_m) \) is the vector of lower bounds of auxiliary variables;
\( u_R = (u_1, \ldots, u_m) \) is the vector of upper bounds of auxiliary variables;
\( l_S = (l_{m+1}, \ldots, l_{m+n}) \) is the vector of lower bounds of structural variables;
\( u_S = (u_{m+1}, \ldots, u_{m+n}) \) is the vector of upper bounds of structural variables.
From the simplex method’s standpoint there is no difference between auxiliary and structural variables. This allows combining all these variables into one vector that leads to the following problem statement:

minimize (or maximize)

\[ z = (0 \mid c)^T x + c_0 \]  

subject to linear constraints

\[ (I \mid -A)x = 0 \]  

and bounds of variables

\[ l \leq x \leq u \]

where:

- \( x = (x_R \mid x_S) \) is the \((m + n)\)-vector of (all) variables;
- \((0 \mid c)\) is the \((m + n)\)-vector of objective coefficients;\(^1\)
- \((I \mid -A)\) is the augmented constraint \(m \times (m + n)\)-matrix;\(^2\)
- \(l = (l_R \mid l_S)\) is the \((m + n)\)-vector of lower bounds of (all) variables;
- \(u = (u_R \mid u_S)\) is the \((m + n)\)-vector of upper bounds of (all) variables.

By definition an LP basic solution geometrically is a point in the space of all variables, which is the intersection of planes corresponding to active constraints.\(^3\) The space of all variables has the dimension \(m+n\), therefore, to define some basic solution we have to define \(m+n\) active constraints. Note that \(m\) constraints (3.5) being linearly independent equalities are always active, so remaining \(n\) active constraints can be chosen only from bound constraints (3.6).

A variable is called non-basic, if its (lower or upper) bound is active, otherwise it is called basic. Since, as was said above, exactly \(n\) bound constraints must be active, in any basic solution there are always \(n\) non-basic variables and \(m\) basic variables. (Note that a free variable also can be non-basic. Although such variable has no bounds, we can think it as the difference between two non-negative variables, which both are non-basic in this case.)

---

\(^1\)Subvector 0 corresponds to objective coefficients at auxiliary variables.

\(^2\)Note that due to auxiliary variables matrix \((I \mid -A)\) contains the unity submatrix and therefore has full rank. This means, in particular, that the system (3.5) has no linearly dependent constraints.

\(^3\)A constraint is called active if in a given point it is satisfied as equality, otherwise it is called inactive.
Now consider how to determine numeric values of all variables for a given basic solution.

Let $\Pi$ be an appropriate permutation matrix of the order $(m+n)$. Then we can write:

$$
\begin{pmatrix}
  x_B \\
  x_N
\end{pmatrix} = \Pi
\begin{pmatrix}
  x_R \\
  x_S
\end{pmatrix} = \Pi x,
$$

where $x_B$ is the vector of basic variables, $x_N$ is the vector of non-basic variables, $x = (x_R \mid x_S)$ is the vector of all variables in the original order. In this case the system of linear constraints (3.5) can be rewritten as follows:

$$(I \mid -A)\Pi^T \Pi x = 0 \Rightarrow (B \mid N) \begin{pmatrix} x_B \\ x_N \end{pmatrix} = 0,$$

where

$$(B \mid N) = (I \mid -A)\Pi^T. \tag{3.9}$$

Matrix $B$ is a square non-singular $m \times m$-matrix, which is composed from columns of the augmented constraint matrix corresponding to basic variables. It is called the basis matrix or simply the basis. Matrix $N$ is a rectangular $m \times n$-matrix, which is composed from columns of the augmented constraint matrix corresponding to non-basic variables.

From (3.8) it follows that:

$$B x_B + N x_N = 0, \tag{3.10}$$

therefore,

$$x_B = -B^{-1} N x_N. \tag{3.11}$$

Thus, the formula (3.11) shows how to determine numeric values of basic variables $x_B$ assuming that non-basic variables $x_N$ are fixed on their active bounds.

The $m \times n$-matrix

$$\Xi = -B^{-1} N, \tag{3.12}$$

which appears in (3.11), is called the simplex tableau.\footnote{This definition corresponds to the GLPK implementation.} It shows how basic variables depend on non-basic variables:

$$x_B = \Xi x_N. \tag{3.13}$$

The system (3.13) is equivalent to the system (3.5) in the sense that they both define the same set of points in the space of (primal) variables, which
satisfy to these systems. If, moreover, values of all basic variables satisfy to their bound constraints (3.3), the corresponding basic solution is called (primal) feasible, otherwise (primal) infeasible. It is understood that any (primal) feasible basic solution satisfy to all constraints (3.2) and (3.3).

The LP theory says that if LP has optimal solution, it has (at least one) basic feasible solution, which corresponds to the optimum. And the most natural way to determine whether a given basic solution is optimal or not is to use the Karush—Kuhn—Tucker optimality conditions.

For the problem statement (3.4)—(3.6) the optimality conditions are the following:

\[(I \mid -A)x = 0\] (3.14)
\[(I \mid -A)^T \pi + \lambda_l + \lambda_u = \nabla z = (0 \mid c)^T\] (3.15)
\[l \leq x \leq u\] (3.16)
\[\lambda_l \geq 0, \lambda_u \leq 0 \text{ (minimization)}\] (3.17)
\[\lambda_l \leq 0, \lambda_u \geq 0 \text{ (maximization)}\] (3.18)
\[(\lambda_l)_k(x_k - l_k) = 0, (\lambda_u)_k(x_k - u_k) = 0, \quad k = 1, 2, \ldots, m + n\] (3.19)

where: \(\pi = (\pi_1, \pi_2, \ldots, \pi_m)\) is a \(m\)-vector of Lagrange multipliers for equality constraints (3.5); \(\lambda_l = [(\lambda_l)_1, (\lambda_l)_2, \ldots, (\lambda_l)_n]\) is a \(n\)-vector of Lagrange multipliers for lower bound constraints (3.6); \(\lambda_u = [(\lambda_u)_1, (\lambda_u)_2, \ldots, (\lambda_u)_n]\) is a \(n\)-vector of Lagrange multipliers for upper bound constraints (3.6).

Condition (3.14) is the primal (original) system of equality constraints (3.5).

Condition (3.15) is the dual system of equality constraints. It requires the gradient of the objective function to be a linear combination of normals to the planes defined by constraints of the original problem.

Condition (3.16) is the primal (original) system of bound constraints (3.6).

Condition (3.17) (or (3.18) in case of maximization) is the dual system of bound constraints.

Condition (3.19) is the complementary slackness condition. It requires, for each original (auxiliary or structural) variable \(x_k\), that either its (lower or upper) bound must be active, or zero bound of the corresponding Lagrange multiplier \((\lambda_l)_k\) or \((\lambda_u)_k\) must be active.

In GLPK two multipliers \((\lambda_l)_k\) and \((\lambda_u)_k\) for each primal (original) variable \(x_k\), \(k = 1, 2, \ldots, m + n\), are combined into one multiplier:

\[\lambda_k = (\lambda_l)_k + (\lambda_u)_k,\] (3.20)

\[5\] These conditions can be applied to any solution, not only to a basic solution.
which is called a dual variable for \( x_k \). This cannot lead to the ambiguity, because both lower and upper bounds of \( x_k \) cannot be active at the same time,\(^6\) so at least one of \((\lambda_l)_k\) and \((\lambda_u)_k\) must be equal to zero, and because these multipliers have different signs, the combined multiplier, which is their sum, uniquely defines each of them.

Using dual variables \( \lambda_k \) the dual system of bound constraints (3.17) and (3.18) can be written in the form of so called “rule of signs” as follows:

<table>
<thead>
<tr>
<th>Original bound constraint</th>
<th>Minimization</th>
<th>Maximization</th>
</tr>
</thead>
<tbody>
<tr>
<td>(-\infty &lt; x_k &lt; +\infty)</td>
<td>((\lambda_l)_k = 0) (\lambda_k = 0) ((\lambda_l)_k + (\lambda_u)_k = 0)</td>
<td>((\lambda_l)_k = 0) (\lambda_k = 0)</td>
</tr>
<tr>
<td>(x_k \geq l_k)</td>
<td>(\lambda_k \geq 0)</td>
<td>(\lambda_k \leq 0)</td>
</tr>
<tr>
<td>(x_k \leq u_k)</td>
<td>(\lambda_k \leq 0)</td>
<td>(\lambda_k \geq 0)</td>
</tr>
<tr>
<td>(l_k \leq x_k \leq u_k)</td>
<td>(\lambda_k \geq 0) (-\infty &lt; \lambda_k &lt; +\infty)</td>
<td>(\lambda_k \leq 0) (-\infty &lt; \lambda_k &lt; +\infty)</td>
</tr>
<tr>
<td>(x_k = l_k = u_k)</td>
<td>(\lambda_k \geq 0) (-\infty &lt; \lambda_k &lt; +\infty)</td>
<td>(\lambda_k \leq 0) (-\infty &lt; \lambda_k &lt; +\infty)</td>
</tr>
</tbody>
</table>

May note that each primal variable \( x_k \) has its dual counterpart \( \lambda_k \) and vice versa. This allows applying the same partition for the vector of dual variables as (3.7):

\[
\begin{pmatrix} \lambda_B \\ \lambda_N \end{pmatrix} = \Pi \lambda, \tag{3.21}
\]

where \( \lambda_B \) is a vector of dual variables for basic variables \( x_B \), \( \lambda_N \) is a vector of dual variables for non-basic variables \( x_N \).

By definition, bounds of basic variables are inactive constraints, so in any basic solution \( \lambda_B = 0 \). Corresponding values of dual variables \( \lambda_N \) for non-basic variables \( x_N \) can be determined in the following way. From the dual system (3.15) we have:

\[
(I \mid -A)^T \pi + \lambda = (0 \mid c)^T, \tag{3.22}
\]

so multiplying both sides of (3.22) by matrix \( \Pi \) gives:

\[
\Pi(I \mid -A)^T \pi + \Pi \lambda = \Pi(0 \mid c)^T. \tag{3.23}
\]

From (3.9) it follows that

\[
\Pi(I \mid -A)^T = [(I \mid -A)\Pi]^T = (B \mid N)^T. \tag{3.24}
\]

\(^6\)If \( x_k \) is a fixed variable, we can think it as double-bounded variable \( l_k \leq x_k \leq u_k \), where \( l_k = u_k \).
Further, we can apply the partition (3.7) also to the vector of objective coefficients (see (3.4)):

\[
\begin{pmatrix}
c_B \\
c_N
\end{pmatrix} = \Pi \begin{pmatrix} 0 \\ c \end{pmatrix},
\]

(3.25)

where \(c_B\) is a vector of objective coefficients at basic variables, \(c_N\) is a vector of objective coefficients at non-basic variables. Now, substituting (3.24), (3.21), and (3.25) into (3.23), leads to:

\[
(B | N)^T \pi + (\lambda_B | \lambda_N)^T = (c_B | c_N)^T,
\]

(3.26)

and transposing both sides of (3.26) gives the system:

\[
\begin{pmatrix}
B^T \\
N^T
\end{pmatrix} \pi + \begin{pmatrix} \lambda_B \\ \lambda_N 
\end{pmatrix} = \begin{pmatrix} c_B \\ c_T \end{pmatrix},
\]

(3.27)

which can be written as follows:

\[
\begin{cases}
B^T \pi + \lambda_B = c_B \\
N^T \pi + \lambda_N = c_N
\end{cases}
\]

(3.28)

Lagrange multipliers \(\pi = (\pi_i)\) correspond to equality constraints (3.5) and therefore can have any sign. This allows resolving the first subsystem of (3.28) as follows:\footnote{\(B^{-T}\) means \((B^T)^{-1} = (B^{-1})^T\).}

\[
\pi = B^{-T} (c_B - \lambda_B) = -B^{-T} \lambda_B + B^{-T} c_B,
\]

(3.29)

and substitution of \(\pi\) from (3.29) into the second subsystem of (3.28) gives:

\[
\lambda_N = -N^T \pi + c_N = N^T B^{-T} \lambda_B + (c_N - N^T B^{-T} c_B).
\]

(3.30)

The latter system can be written in the following final form:

\[
\lambda_N = -\Xi^T \lambda_B + d,
\]

(3.31)

where \(\Xi\) is the simplex tableau (see (3.12)), and

\[
d = c_N - N^T B^{-T} c_B = c_N + \Xi^T c_B
\]

(3.32)

is the vector of so called \textit{reduced costs} of non-basic variables.
Above it was said that in any basic solution $\lambda_B = 0$, so $\lambda_N = d$ as it follows from (3.31).

The system (3.31) is equivalent to the system (3.15) in the sense that they both define the same set of points in the space of dual variables $\lambda$, which satisfy to these systems. If, moreover, values of all dual variables $\lambda_N$ (i.e. reduced costs $d$) satisfy to their bound constraints (i.e. to the “rule of signs”; see the table above), the corresponding basic solution is called dual feasible, otherwise dual infeasible. It is understood that any dual feasible solution satisfy to all constraints (3.15) and (3.17) (or (3.18) in case of maximization).

It can be easily shown that the complementary slackness condition (3.19) is always satisfied for any basic solution. Therefore, a basic solution is optimal if and only if it is primal and dual feasible, because in this case it satisfy to all the optimality conditions (3.14)—(3.19).

The meaning of reduced costs $d = (d_j)$ of non-basic variables can be explained in the following way. From (3.4), (3.7), and (3.25) it follows that:

$$z = c_B^T x_B + c_N^T x_N + c_0.$$ (3.33)

Substituting $x_B$ from (3.11) into (3.33) we can eliminate basic variables and express the objective only through non-basic variables:

$$z = c_B^T (-B^{-1} N x_N) + c_N^T x_N + c_0 =$$
$$= (c_N^T - c_B^T B^{-1} N) x_N + c_0 =$$
$$= (c_N - N^T B^{-T} c_B)^T x_N + c_0 =$$
$$= d^T x_N + c_0.$$ (3.34)

From (3.34) it is seen that reduced cost $d_j$ shows how the objective function $z$ depends on non-basic variable $(x_N)_j$ in the neighborhood of the current basic solution, i.e. while the current basis remains unchanged.

---

*It is assumed that a complete basic solution has the form $(x, \lambda)$, i.e. it includes primal as well as dual variables.*
4.2 LP basis routines

4.2.1 glp_bf_exists—check if the basis factorization exists

Synopsis

```c
int glp_bf_exists(glp_prob *lp);
```

Returns

If the basis factorization for the current basis associated with the specified problem object exists and therefore is available for computations, the routine `glp_bf_exists` returns non-zero. Otherwise the routine returns zero.

Comments

Let the problem object have \( m \) rows and \( n \) columns. In GLPK the basis matrix \( B \) is a square non-singular matrix of the order \( m \), whose columns correspond to basic (auxiliary and/or structural) variables. It is defined by the following main equality:

\[
(B \mid N) = (I \mid -A)\Pi^T,
\]

where \( I \) is the unity matrix of the order \( m \), whose columns correspond to auxiliary variables; \( A \) is the original constraint \( m \times n \)-matrix, whose columns correspond to structural variables; \( (I \mid -A) \) is the augmented constraint \( m \times (m + n) \)-matrix, whose columns correspond to all (auxiliary and structural) variables following in the original order; \( \Pi \) is a permutation matrix of the order \( m + n \); and \( N \) is a rectangular \( m \times n \)-matrix, whose columns correspond to non-basic (auxiliary and/or structural) variables.

For various reasons it may be necessary to solve linear systems with matrix \( B \). To provide this possibility the GLPK implementation maintains an invertable form of \( B \) (that is, some representation of \( B^{-1} \)) called the basis factorization, which is an internal component of the problem object. Typically, the basis factorization is computed by the simplex solver, which keeps it in the problem object to be available for other computations.

Should note that any changes in the problem object, which affects the basis matrix (e.g. changing the status of a row or column, changing a basic column of the constraint matrix, removing an active constraint, etc.), invalidates the basis factorization. So before calling any API routine, which uses the basis factorization, the application program must make sure (using the

---

\(^9\)For more details see Subsection 4.1, page 115.
routine `glp_bf_exists` that the factorization exists and therefore available for computations.

4.2.2 `glp_factorize`—compute the basis factorization

Synopsis

```
int glp_factorize(glp_prob *lp);
```

Description

The routine `glp_factorize` computes the basis factorization for the current basis associated with the specified problem object.\(^{10}\)

The basis factorization is computed from “scratch” even if it exists, so the application program may use the routine `glp_bf_exists`, and, if the basis factorization already exists, not to call the routine `glp_factorize` to prevent an extra work.

The routine `glp_factorize` does not compute components of the basic solution (i.e. primal and dual values).

Returns

<table>
<thead>
<tr>
<th>Code</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>The basis factorization has been successfully computed.</td>
</tr>
<tr>
<td>GLP_EBADB</td>
<td>The basis matrix is invalid, because the number of basic (auxiliary and structural) variables is not the same as the number of rows in the problem object.</td>
</tr>
<tr>
<td>GLP_ESING</td>
<td>The basis matrix is singular within the working precision.</td>
</tr>
<tr>
<td>GLP_ECOND</td>
<td>The basis matrix is ill-conditioned, i.e. its condition number is too large.</td>
</tr>
</tbody>
</table>

\(^{10}\)The current basis is defined by the current statuses of rows (auxiliary variables) and columns (structural variables).
4.2.3  glp_bf_updated—check if the basis factorization has been updated

Synopsis

int glp_bf_updated(glp_prob *lp);

Returns

If the basis factorization has been just computed from “scratch”, the routine glp_bf_updated returns zero. Otherwise, if the factorization has been updated at least once, the routine returns non-zero.

Comments

Updating the basis factorization means recomputing it to reflect changes in the basis matrix. For example, on every iteration of the simplex method some column of the current basis matrix is replaced by a new column that gives a new basis matrix corresponding to the adjacent basis. In this case computing the basis factorization for the adjacent basis from “scratch” (as the routine glp_factorize does) would be too time-consuming.

On the other hand, since the basis factorization update is a numeric computational procedure, applying it many times may lead to accumulating round-off errors. Therefore the basis is periodically refactorized (reinverted) from “scratch” (with the routine glp_factorize) that allows improving its numerical properties.

The routine glp_bf_updated allows determining if the basis factorization has been updated at least once since it was computed from “scratch”.

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4.2.4  glp_get_bfcp—retrieve basis factorization control parameters

Synopsis

void glp_get_bfcp(glp_prob *lp, glp_bfcp *parm);

Description

The routine glp_get_bfcp retrieves control parameters, which are used on computing and updating the basis factorization associated with the specified problem object.

Current values of the control parameters are stored in a glp_bfcp structure, which the parameter parm points to. For a detailed description of the structure glp_bfcp see comments to the routine glp_set_bfcp in the next subsection.

Comments

The purpose of the routine glp_get_bfcp is two-fold. First, it allows the application program obtaining current values of control parameters used by internal GLPK routines, which compute and update the basis factorization.

The second purpose of this routine is to provide proper values for all fields of the structure glp_bfcp in the case when the application program needs to change some control parameters.

4.2.5  glp_set_bfcp—change basis factorization control parameters

Synopsis

void glp_set_bfcp(glp_prob *lp, const glp_bfcp *parm);

Description

The routine glp_set_bfcp changes control parameters, which are used by internal GLPK routines on computing and updating the basis factorization associated with the specified problem object.

New values of the control parameters should be passed in a structure glp_bfcp, which the parameter parm points to. For a detailed description of the structure glp_bfcp see paragraph “Control parameters” below.

The parameter parm can be specified as NULL, in which case all control parameters are reset to their default values.
Comments

Before changing some control parameters with the routine `glp_set_bfcp` the application program should retrieve current values of all control parameters with the routine `glp_get_bfcp`. This is needed for backward compatibility, because in the future there may appear new members in the structure `glp_bfcp`.

Note that new values of control parameters come into effect on a next computation of the basis factorization, not immediately.

Example

```c
glp_prob *lp;
glp_bfcp parm;
...
/* retrieve current values of control parameters */
glp_get_bfcp(lp, &parm);
/* change the threshold pivoting tolerance */
parm.piv_tol = 0.05;
/* set new values of control parameters */
glp_set_bfcp(lp, &parm);
...
```

Control parameters

This paragraph describes all basis factorization control parameters currently used in the package. Symbolic names of control parameters are names of corresponding members in the structure `glp_bfcp`.

`int type` (default: `GLP_BF_FT`)
Basis factorization type:
- `GLP_BF_FT`—`LU` + Forrest–Tomlin update;
- `GLP_BF_BG`—`LU` + Schur complement + Bartels–Golub update;
- `GLP_BF_GR`—`LU` + Schur complement + Givens rotation update.

In case of `GLP_BF_FT` the update is applied to matrix $U$, while in cases of `GLP_BF_BG` and `GLP_BF_GR` the update is applied to the Schur complement.

`int lu_size` (default: 0)
The initial size of the Sparse Vector Area, in non-zeros, used on computing `LU`-factorization of the basis matrix for the first time. If this parameter is set to 0, the initial SVA size is determined automatically.
double piv_tol (default: 0.10)
Threshold pivoting (Markowitz) tolerance, $0 < \text{piv\_tol} < 1$, used on computing $LU$-factorization of the basis matrix. Element $u_{ij}$ of the active submatrix of factor $U$ fits to be pivot if it satisfies to the stability criterion $|u_{ij}| \geq \text{piv\_tol} \cdot \max |u_{is}|$, i.e. if it is not very small in the magnitude among other elements in the same row. Decreasing this parameter may lead to better sparsity at the expense of numerical accuracy, and vice versa.

int piv_lim (default: 4)
This parameter is used on computing $LU$-factorization of the basis matrix and specifies how many pivot candidates needs to be considered on choosing a pivot element, $\text{piv\_lim} \geq 1$. If $\text{piv\_lim}$ candidates have been considered, the pivoting routine prematurely terminates the search with the best candidate found.

int suhl (default: GLP_ON)
This parameter is used on computing $LU$-factorization of the basis matrix. Being set to GLP_ON it enables applying the following heuristic proposed by Uwe Suhl: if a column of the active submatrix has no eligible pivot candidates, it is no more considered until it becomes a column singleton. In many cases this allows reducing the time needed for pivot searching. To disable this heuristic the parameter suhl should be set to GLP_OFF.

double eps_tol (default: 1e-15)
Epsilon tolerance, $\text{eps\_tol} \geq 0$, used on computing $LU$-factorization of the basis matrix. If an element of the active submatrix of factor $U$ is less than $\text{eps\_tol}$ in the magnitude, it is replaced by exact zero.

double max_gro (default: 1e+10)
Maximal growth of elements of factor $U$, $\text{max\_gro} \geq 1$, allowable on computing $LU$-factorization of the basis matrix. If on some elimination step the ratio $u_{\text{big}}/b_{\text{max}}$ (where $u_{\text{big}}$ is the largest magnitude of elements of factor $U$ appeared in its active submatrix during all the factorization process, $b_{\text{max}}$ is the largest magnitude of elements of the basis matrix to be factorized), the basis matrix is considered as ill-conditioned.
int nfs_max (default: 100)
Maximal number of additional row-like factors (entries of the eta file),
\(nfs_{\text{max}} \geq 1\), which can be added to \(LU\)-factorization of the basis matrix
on updating it with the Forrest–Tomlin technique. This parameter is
used only once, before \(LU\)-factorization is computed for the first time,
to allocate working arrays. As a rule, each update adds one new factor
(however, some updates may need no addition), so this parameter limits
the number of updates between refactorizations.

double upd_tol (default: 1e-6)
Update tolerance, \(0 < \text{upd\_tol} < 1\), used on updating \(LU\)-factorization
of the basis matrix with the Forrest–Tomlin technique. If after updat-
ing the magnitude of some diagonal element \(u_{kk}\) of factor \(U\) becomes
less than \(\text{upd\_tol} \cdot \max(|u_{k*}|, |u_{*k}|)\), the factorization is considered as
inaccurate.

int nrs_max (default: 100)
Maximal number of additional rows and columns, \(nrs_{\text{max}} \geq 1\), which
can be added to \(LU\)-factorization of the basis matrix on updating it with
the Schur complement technique. This parameter is used only once, be-
fore \(LU\)-factorization is computed for the first time, to allocate working
arrays. As a rule, each update adds one new row and column (how-
ever, some updates may need no addition), so this parameter limits the
number of updates between refactorizations.

int rs_size (default: 0)
The initial size of the Sparse Vector Area, in non-zeros, used to store
non-zero elements of additional rows and columns introduced on up-
dating \(LU\)-factorization of the basis matrix with the Schur complement
technique. If this parameter is set to 0, the initial SVA size is determined
automatically.
4.2.6  glp_get_bhead—retrieve the basis header information

Synopsis

int glp_get_bhead(glp_prob *lp, int k);

Description

The routine glp_get_bhead returns the basis header information for the current basis associated with the specified problem object.

Returns

If basic variable \((x_B)_k\), \(1 \leq k \leq m\), is \(i\)-th auxiliary variable \((1 \leq i \leq m)\), the routine returns \(i\). Otherwise, if \((x_B)_k\) is \(j\)-th structural variable \((1 \leq j \leq n)\), the routine returns \(m+j\). Here \(m\) is the number of rows and \(n\) is the number of columns in the problem object.

Comments

Sometimes the application program may need to know which original (auxiliary and structural) variable correspond to a given basic variable, or, that is the same, which column of the augmented constraint matrix \((I \mid -A)\) correspond to a given column of the basis matrix \(B\).

The correspondence is defined as follows:\(^{11}\)

\[
\begin{pmatrix}
  x_B \\
  x_N
\end{pmatrix} = \Pi \begin{pmatrix}
  x_R \\
  x_S
\end{pmatrix} \Leftrightarrow \begin{pmatrix}
  x_R \\
  x_S
\end{pmatrix} = \Pi^T \begin{pmatrix}
  x_B \\
  x_N
\end{pmatrix},
\]

where \(x_B\) is the vector of basic variables, \(x_N\) is the vector of non-basic variables, \(x_R\) is the vector of auxiliary variables following in their original order,\(^{12}\) \(x_S\) is the vector of structural variables following in their original order, \(\Pi\) is a permutation matrix (which is a component of the basis factorization).

Thus, if \((x_B)_k = (x_R)_i\) is \(i\)-th auxiliary variable, the routine returns \(i\), and if \((x_B)_k = (x_S)_j\) is \(j\)-th structural variable, the routine returns \(m+j\), where \(m\) is the number of rows in the problem object.

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\(^{11}\)For more details see Subsection 4.1, page 115.

\(^{12}\)The original order of auxiliary and structural variables is defined by the ordinal numbers of corresponding rows and columns in the problem object.
4.2.7  glp_get_row_bind—retrieve row index in the basis header

Synopsis

int glp_get_row_bind(glp_prob *lp, int i);

Returns

The routine glp_get_row_bind returns the index \( k \) of basic variable \((xB)k\), \( 1 \leq k \leq m \), which is \( i \)-th auxiliary variable (that is, the auxiliary variable corresponding to \( i \)-th row), \( 1 \leq i \leq m \), in the current basis associated with the specified problem object, where \( m \) is the number of rows. However, if \( i \)-th auxiliary variable is non-basic, the routine returns zero.

Comments

The routine glp_get_row_bind is an inverse to the routine glp_get_bhead: if glp_get_bhead\((lp,k)\) returns \( i \), glp_get_row_bind\((lp,i)\) returns \( k \), and vice versa.

4.2.8  glp_get_col_bind—retrieve column index in the basis header

Synopsis

int glp_get_col_bind(glp_prob *lp, int j);

Returns

The routine glp_get_col_bind returns the index \( k \) of basic variable \((xB)k\), \( 1 \leq k \leq m \), which is \( j \)-th structural variable (that is, the structural variable corresponding to \( j \)-th column), \( 1 \leq j \leq n \), in the current basis associated with the specified problem object, where \( m \) is the number of rows, \( n \) is the number of columns. However, if \( j \)-th structural variable is non-basic, the routine returns zero.

Comments

The routine glp_get_col_bind is an inverse to the routine glp_get_bhead: if glp_get_bhead\((lp,k)\) returns \( m+j \), glp_get_col_bind\((lp,j)\) returns \( k \), and vice versa.
### 4.2.9 glp_ftran—perform forward transformation

**Synopsis**

```c
void glp_ftran(glp_prob *lp, double x[]);
```

**Description**

The routine `glp_ftran` performs forward transformation (FTRAN), i.e. it solves the system \( Bx = b \), where \( B \) is the basis matrix associated with the specified problem object, \( x \) is the vector of unknowns to be computed, \( b \) is the vector of right-hand sides.

On entry to the routine elements of the vector \( b \) should be stored in locations \( x[1], \ldots, x[m] \), where \( m \) is the number of rows. On exit the routine stores elements of the vector \( x \) in the same locations.

### 4.2.10 glp_btran—perform backward transformation

**Synopsis**

```c
void glp_btran(glp_prob *lp, double x[]);
```

**Description**

The routine `glp_btran` performs backward transformation (BTRAN), i.e. it solves the system \( B^T x = b \), where \( B^T \) is a matrix transposed to the basis matrix \( B \) associated with the specified problem object, \( x \) is the vector of unknowns to be computed, \( b \) is the vector of right-hand sides.

On entry to the routine elements of the vector \( b \) should be stored in locations \( x[1], \ldots, x[m] \), where \( m \) is the number of rows. On exit the routine stores elements of the vector \( x \) in the same locations.
4.2.11  glp_warm_up—“warm up” LP basis

Synopsis

int glp_warm_up(glp_prob *P);

Description

The routine glp_warm_up “warms up” the LP basis for the specified problem object using current statuses assigned to rows and columns (that is, to auxiliary and structural variables).

This operation includes computing factorization of the basis matrix (if it does not exist), computing primal and dual components of basic solution, and determining the solution status.

Returns

0  The operation has been successfully performed.
GLP_EBADB  The basis matrix is invalid, because the number of basic (auxiliary and structural) variables is not the same as the number of rows in the problem object.
GLP_ESING  The basis matrix is singular within the working precision.
GLP_ECOND  The basis matrix is ill-conditioned, i.e. its condition number is too large.
4.3 Simplex tableau routines

4.3.1 glp_eval_tab_row—compute row of the tableau

Synopsis

int glp_eval_tab_row(glp_prob *lp, int k, int ind[],
                     double val[]);

Description

The routine glp_eval_tab_row computes a row of the current simplex tableau (see Subsection 3.1.1, formula (3.12)), which (row) corresponds to some basic variable specified by the parameter \( k \) as follows: if \( 1 \leq k \leq m \), the basic variable is \( k \)-th auxiliary variable, and if \( m + 1 \leq k \leq m + n \), the basic variable is \((k - m)\)-th structural variable, where \( m \) is the number of rows and \( n \) is the number of columns in the specified problem object. The basis factorization must exist.

The computed row shows how the specified basic variable depends on non-basic variables:

\[
x_k = (x_B)_i = \xi_{i1}(x_N)_1 + \xi_{i2}(x_N)_2 + \ldots + \xi_{in}(x_N)_n,
\]

where \( \xi_{i1}, \xi_{i2}, \ldots, \xi_{in} \) are elements of the simplex table row, \((x_N)_1, (x_N)_2, \ldots, (x_N)_n\) are non-basic (auxiliary and structural) variables.

The routine stores column indices and corresponding numeric values of non-zero elements of the computed row in unordered sparse format in locations ind[1], \ldots, ind[len] and val[1], \ldots, val[len], respectively, where \( 0 \leq \text{len} \leq n \) is the number of non-zero elements in the row returned on exit.

Element indices stored in the array ind have the same sense as index \( k \), i.e. indices 1 to \( m \) denote auxiliary variables while indices \( m + 1 \) to \( m + n \) denote structural variables (all these variables are obviously non-basic by definition).

Returns

The routine glp_eval_tab_row returns \( \text{len} \), which is the number of non-zero elements in the simplex table row stored in the arrays ind and val.

Comments

A row of the simplex table is computed as follows. At first, the routine checks that the specified variable \( x_k \) is basic and uses the permutation matrix \( \Pi \).
(3.7) to determine index $i$ of basic variable $(x_B)_i$, which corresponds to $x_k$.

The row to be computed is $i$-th row of the matrix $\Xi$ (3.12), therefore:

$$\xi_i = e_i^T \Xi = -e_i^T B^{-1} N = -(B^{-T}e_i)^T N,$$

where $e_i$ is $i$-th unity vector. So the routine performs BTRAN to obtain $i$-th row of the inverse $B^{-1}$:

$$g_i = B^{-T}e_i,$$

and then computes elements of the simplex table row as inner products:

$$\xi_{ij} = -g_i^T N_j, \quad j = 1, 2, \ldots, n,$$

where $N_j$ is $j$-th column of matrix $N$ (3.9), which (column) corresponds to non-basic variable $(x_N)_j$. The permutation matrix $\Pi$ is used again to convert indices $j$ of non-basic columns to original ordinal numbers of auxiliary and structural variables.

4.3.2  \texttt{glp\_eval\_tab\_col}—compute column of the tableau

Synopsis

```c
int glp_eval_tab_col(glp_prob *lp, int k, int ind[],
    double val[]);
```

Description

The routine \texttt{glp\_eval\_tab\_col} computes a column of the current simplex tableau (see Subsection 3.1.1, formula (3.12)), which (column) corresponds to some non-basic variable specified by the parameter $k$: if $1 \leq k \leq m$, the non-basic variable is $k$-th auxiliary variable, and if $m + 1 \leq k \leq m + n$, the non-basic variable is $(k - m)$-th structural variable, where $m$ is the number of rows and $n$ is the number of columns in the specified problem object. The basis factorization must exist.

The computed column shows how basic variables depends on the specified non-basic variable $x_k = (x_N)_j$:

$$(x_B)_1 = \ldots + \xi_1j(x_N)_j + \ldots$$
$$(x_B)_2 = \ldots + \xi_2j(x_N)_j + \ldots$$

$$(x_B)_m = \ldots + \xi_{mj}(x_N)_j + \ldots$$
where $\xi_{1j}, \xi_{2j}, \ldots, \xi_{mj}$ are elements of the simplex table column, $(x_B)_1,$
$(x_B)_2, \ldots, (x_B)_m$ are basic (auxiliary and structural) variables.

The routine stores row indices and corresponding numeric values of non-zero elements of the computed column in unordered sparse format in locations $\text{ind}[1], \ldots, \text{ind}[\text{len}]$ and $\text{val}[1], \ldots, \text{val}[\text{len}]$, respectively, where $0 \leq \text{len} \leq m$ is the number of non-zero elements in the column returned on exit.

Element indices stored in the array $\text{ind}$ have the same sense as index $k$, i.e. indices 1 to $m$ denote auxiliary variables while indices $m + 1$ to $m + n$ denote structural variables (all these variables are obviously basic by definition).

**Returns**

The routine `glp_eval_tab_col` returns $\text{len}$, which is the number of non-zero elements in the simplex table column stored in the arrays $\text{ind}$ and $\text{val}$.

**Comments**

A column of the simplex table is computed as follows. At first, the routine checks that the specified variable $x_k$ is non-basic and uses the permutation matrix $\Pi$ (3.7) to determine index $j$ of non-basic variable $(x_N)_j$, which corresponds to $x_k$.

The column to be computed is $j$-th column of the matrix $\Xi$ (3.12), therefore:

$$\Xi_j = \Xi e_j = -B^{-1}N e_j = -B^{-1}N_j,$$

where $e_j$ is $j$-th unity vector, $N_j$ is $j$-th column of matrix $N$ (3.9). So the routine performs FTRAN to transform $N_j$ to the simplex table column $\Xi_j = (\xi_{ij})$ and uses the permutation matrix $\Pi$ to convert row indices $i$ to original ordinal numbers of auxiliary and structural variables.
### 4.3.3 `glp_transform_row`—transform explicitly specified row

**Synopsis**

```c
int glp_transform_row(glp_prob *P, int len, int ind[],
                     double val[]);
```

**Description**

The routine `glp_transform_row` performs the same operation as the routine `glp_eval_tab_row` with exception that the row to be transformed is specified explicitly as a sparse vector.

The explicitly specified row may be thought as a linear form:

\[ x = a_1 x_{m+1} + a_2 x_{m+2} + \ldots + a_n x_{m+n}, \]

where \( x \) is an auxiliary variable for this row, \( a_j \) are coefficients of the linear form, \( x_{m+j} \) are structural variables.

On entry column indices and numerical values of non-zero coefficients \( a_j \) of the specified row should be placed in locations `ind[1]`, \ldots, `ind[len]` and `val[1]`, \ldots, `val[len]`, where `len` is number of non-zero coefficients.

This routine uses the system of equality constraints and the current basis in order to express the auxiliary variable \( x \) through the current non-basic variables (as if the transformed row were added to the problem object and the auxiliary variable \( x \) were basic), i.e. the resultant row has the form:

\[ x = \xi_1 (x_{N})_1 + \xi_2 (x_{N})_2 + \ldots + \xi_n (x_{N})_n, \]

where \( \xi_j \) are influence coefficients, \( (x_{N})_j \) are non-basic (auxiliary and structural) variables, \( n \) is the number of columns in the problem object.

On exit the routine stores indices and numerical values of non-zero coefficients \( \xi_j \) of the resultant row in locations `ind[1]`, \ldots, `ind[len']` and `val[1]`, \ldots, `val[len']`, where \( 0 \leq \text{len}' \leq n \) is the number of non-zero coefficients in the resultant row returned by the routine. Note that indices of non-basic variables stored in the array `ind` correspond to original ordinal numbers of variables: indices 1 to \( m \) mean auxiliary variables and indices \( m + 1 \) to \( m + n \) mean structural ones.

**Returns**

The routine `glp_transform_row` returns `len'`, the number of non-zero coefficients in the resultant row stored in the arrays `ind` and `val`. 
4.3.4 glp_transform_col—transform explicitly specified column

Synopsis

```c
int glp_transform_col(glp_prob *P, int len, int ind[],
                     double val[]);
```

Description

The routine `glp_transform_col` performs the same operation as the routine `glp_eval_tab_col` with exception that the column to be transformed is specified explicitly as a sparse vector.

The explicitly specified column may be thought as it were added to the original system of equality constraints:

\[
\begin{align*}
    x_1 & = a_{11}x_{m+1} + \ldots + a_{1n}x_{m+n} + a_1x \\
    x_2 & = a_{21}x_{m+1} + \ldots + a_{2n}x_{m+n} + a_2x \\
    \quad \quad \quad \quad \quad \quad \quad \quad \vdots \\
    x_m & = a_{m1}x_{m+1} + \ldots + a_{mn}x_{m+n} + a_mx
\end{align*}
\]

where \(x_i\) are auxiliary variables, \(x_{m+j}\) are structural variables (presented in the problem object), \(x\) is a structural variable for the explicitly specified column, \(a_i\) are constraint coefficients at \(x\).

On entry row indices and numerical values of non-zero coefficients \(a_i\) of the specified column should be placed in locations `ind[1]`, ..., `ind[len]` and `val[1]`, ..., `val[len]`, where `len` is number of non-zero coefficients.

This routine uses the system of equality constraints and the current basis in order to express the current basic variables through the structural variable \(x\) (as if the transformed column were added to the problem object and the variable \(x\) were non-basic):

\[
\begin{align*}
    (x_B)_1 & = \ldots + \xi_1x \\
    (x_B)_2 & = \ldots + \xi_2x \\
    \quad \quad \quad \quad \quad \quad \vdots \\
    (x_B)_m & = \ldots + \xi_mx
\end{align*}
\]

where \(\xi_i\) are influence coefficients, \(x_B\) are basic (auxiliary and structural) variables, \(m\) is the number of rows in the problem object.

On exit the routine stores indices and numerical values of non-zero coefficients \(\xi_i\) of the resultant column in locations `ind[1]`, ..., `ind[len']` and `val[1]`, ..., `val[len']`, where \(0 \leq len' \leq m\) is the number of non-zero
coefficients in the resultant column returned by the routine. Note that indices of basic variables stored in the array \texttt{ind} correspond to original ordinal numbers of variables, i.e. indices 1 to \(m\) mean auxiliary variables, indices \(m + 1\) to \(m + n\) mean structural ones.

**Returns**

The routine \texttt{glp_transform_col} returns \texttt{len'}, the number of non-zero coefficients in the resultant column stored in the arrays \texttt{ind} and \texttt{val}.

### 4.3.5 \texttt{glp_prim_rtest}—perform primal ratio test

**Synopsis**

```c
int glp_prim_rtest(glp_prob *P, int len, const int ind[],
                  const double val[], int dir, double eps);
```

**Description**

The routine \texttt{glp_prim_rtest} performs the primal ratio test using an explicitly specified column of the simplex table.

The current basic solution associated with the LP problem object must be primal feasible.

The explicitly specified column of the simplex table shows how the basic variables \(x_B\) depend on some non-basic variable \(x\) (which is not necessarily presented in the problem object):

\[
    (x_B)_{1} = \ldots + \xi_1x \\
    (x_B)_{2} = \ldots + \xi_2x \\
    \ldots \ldots \ldots \\
    (x_B)_{m} = \ldots + \xi_mx
\]

The column is specified on entry to the routine in sparse format. Ordinal numbers of basic variables \((x_B)_i\) should be placed in locations \texttt{ind[1]}, \ldots, \texttt{ind[len]}, where ordinal number 1 to \(m\) denote auxiliary variables, and ordinal numbers \(m + 1\) to \(m + n\) denote structural variables. The corresponding non-zero coefficients \(\xi_i\) should be placed in locations \texttt{val[1]}, \ldots, \texttt{val:len}.

The arrays \texttt{ind} and \texttt{val} are not changed by the routine.

The parameter \texttt{dir} specifies direction in which the variable \(x\) changes on entering the basis: +1 means increasing, −1 means decreasing.

The parameter \texttt{eps} is an absolute tolerance (small positive number, say, \(10^{-9}\)) used by the routine to skip \(\xi_i\)'s whose magnitude is less than \texttt{eps}.  

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The routine determines which basic variable (among those specified in \texttt{ind[1]}, \ldots, \texttt{ind[len]}) reaches its (lower or upper) bound first before any other basic variables do, and which, therefore, should leave the basis in order to keep primal feasibility.

**Returns**

The routine \texttt{glp_prim_rtest} returns the index, \texttt{piv}, in the arrays \texttt{ind} and \texttt{val} corresponding to the pivot element chosen, \(1 \leq \texttt{piv} \leq \texttt{len} \). If the adjacent basic solution is primal unbounded, and therefore the choice cannot be made, the routine returns zero.

**Comments**

If the non-basic variable \(x\) is presented in the LP problem object, the input column can be computed with the routine \texttt{glp_eval_tab_col}; otherwise, it can be computed with the routine \texttt{glp_transform_col}.

### 4.3.6 \texttt{glp_dual_rtest}—perform dual ratio test

**Synopsis**

\begin{verbatim}
int glp_dual_rtest(glp_prob *P, int len, const int ind[],
                 const double val[], int dir, double eps);
\end{verbatim}

**Description**

The routine \texttt{glp_dual_rtest} performs the dual ratio test using an explicitly specified row of the simplex table.

The current basic solution associated with the LP problem object must be dual feasible.

The explicitly specified row of the simplex table is a linear form that shows how some basic variable \(x\) (which is not necessarily presented in the problem object) depends on non-basic variables \(x_N\):

\[
x = \xi_1(x_N)_1 + \xi_2(x_N)_2 + \ldots + \xi_n(x_N)_n.
\]

The row is specified on entry to the routine in sparse format. Ordinal numbers of non-basic variables \((x_N)_j\) should be placed in locations \texttt{ind[1]}, \ldots, \texttt{ind[len]}, where ordinal numbers 1 to \(m\) denote auxiliary variables, and ordinal numbers \(m + 1\) to \(m + n\) denote structural variables. The
corresponding non-zero coefficients $\xi_j$ should be placed in locations val[1], ..., val[len]. The arrays ind and val are not changed by the routine.

The parameter dir specifies direction in which the variable $x$ changes on leaving the basis: +1 means that $x$ goes on its lower bound, so its reduced cost (dual variable) is increasing (minimization) or decreasing (maximization); −1 means that $x$ goes on its upper bound, so its reduced cost is decreasing (minimization) or increasing (maximization).

The parameter eps is an absolute tolerance (small positive number, say, $10^{-9}$) used by the routine to skip $\xi_j$’s whose magnitude is less than eps.

The routine determines which non-basic variable (among those specified in ind[1], ..., ind[len]) should enter the basis in order to keep dual feasibility, because its reduced cost reaches the (zero) bound first before this occurs for any other non-basic variables.

Returns

The routine glp_dual_rtest returns the index, piv, in the arrays ind and val corresponding to the pivot element chosen, $1 \leq \text{piv} \leq \text{len}$. If the adjacent basic solution is dual unbounded, and therefore the choice cannot be made, the routine returns zero.

Comments

If the basic variable $x$ is presented in the LP problem object, the input row can be computed with the routine glp_eval_tab_row; otherwise, it can be computed with the routine glp_transform_row.
4.4 Post-optimal analysis routines

4.4.1 glp_analyze_bound—analyze active bound of non-basic variable

Synopsis

void glp_analyze_bound(glp_prob *P, int k, double *limit1, 
                      int *var1, double *limit2, int *var2);

Description

The routine glp_analyze_bound analyzes the effect of varying the active bound of specified non-basic variable.

The non-basic variable is specified by the parameter $k$, where $1 \leq k \leq m$ means auxiliary variable of corresponding row, and $m + 1 \leq k \leq m + n$ means structural variable (column).

Note that the current basic solution must be optimal, and the basis factorization must exist.

Results of the analysis have the following meaning.

$\text{value1}$ is the minimal value of the active bound, at which the basis still remains primal feasible and thus optimal. -$\text{DBL_MAX}$ means that the active bound has no lower limit.

$\text{var1}$ is the ordinal number of an auxiliary (1 to $m$) or structural ($m + 1$ to $m + n$) basic variable, which reaches its bound first and thereby limits further decreasing the active bound being analyzed. if $\text{value1} = -$DBL_MAX, $\text{var1}$ is set to 0.

$\text{value2}$ is the maximal value of the active bound, at which the basis still remains primal feasible and thus optimal. +$\text{DBL_MAX}$ means that the active bound has no upper limit.

$\text{var2}$ is the ordinal number of an auxiliary (1 to $m$) or structural ($m + 1$ to $m + n$) basic variable, which reaches its bound first and thereby limits further increasing the active bound being analyzed. if $\text{value2} = +\text{DBL_MAX}$, $\text{var2}$ is set to 0.

The parameters $\text{value1}$, $\text{var1}$, $\text{value2}$, $\text{var2}$ can be specified as NULL, in which case corresponding information is not stored.
4.4.2 \texttt{glp\_analyze\_coef}—analyze objective coefficient at basic variable

Synopsis

\begin{verbatim}
void glp_analyze_coef(glp_prob *P, int k, double *coef1,
    int *var1, double *value1, double *coef2, int *var2,
    double *value2);
\end{verbatim}

Description

The routine \texttt{glp\_analyze\_coef} analyzes the effect of varying the objective coefficient at specified basic variable.

The basic variable is specified by the parameter \( k \), where \( 1 \leq k \leq m \) means auxiliary variable of corresponding row, and \( m + 1 \leq k \leq m + n \) means structural variable (column).

Note that the current basic solution must be optimal, and the basis factorization must exist.

Results of the analysis have the following meaning.

\( \text{coef1} \) is the minimal value of the objective coefficient, at which the basis still remains dual feasible and thus optimal. \( -\text{DBL\_MAX} \) means that the objective coefficient has no lower limit.

\( \text{var1} \) is the ordinal number of an auxiliary (1 to \( m \)) or structural (\( m + 1 \) to \( m + n \)) non-basic variable, whose reduced cost reaches its zero bound first and thereby limits further decreasing the objective coefficient being analyzed. If \( \text{coef1} = -\text{DBL\_MAX} \), \( \text{var1} \) is set to 0.

\( \text{value1} \) is value of the basic variable being analyzed in an adjacent basis, which is defined as follows. Let the objective coefficient reaches its minimal value (\( \text{coef1} \)) and continues decreasing. Then the reduced cost of the limiting non-basic variable (\( \text{var1} \)) becomes dual infeasible and the current basis becomes non-optimal that forces the limiting non-basic variable to enter the basis replacing there some basic variable that leaves the basis to keep primal feasibility. Should note that on determining the adjacent basis current bounds of the basic variable being analyzed are ignored as if it were free (unbounded) variable, so it cannot leave the basis. It may happen that no dual feasible adjacent basis exists, in which case \( \text{value1} \) is set to \( -\text{DBL\_MAX} \) or \( +\text{DBL\_MAX} \).

\( \text{coef2} \) is the maximal value of the objective coefficient, at which the basis still remains dual feasible and thus optimal. \( +\text{DBL\_MAX} \) means that the objective coefficient has no upper limit.
var2 is the ordinal number of an auxiliary (1 to m) or structural (m + 1 to m + n) non-basic variable, whose reduced cost reaches its zero bound first and thereby limits further increasing the objective coefficient being analyzed. If coef2 = +DBL_MAX, var2 is set to 0.

value2 is value of the basic variable being analyzed in an adjacent basis, which is defined exactly in the same way as value1 above with exception that now the objective coefficient is increasing.

The parameters coef1, var1, value1, coef2, var2, value2 can be specified as NULL, in which case corresponding information is not stored.
Chapter 5

Branch-and-Cut API Routines

5.1 Introduction

5.1.1 Using the callback routine

The GLPK MIP solver based on the branch-and-cut method allows the application program to control the solution process. This is attained by means of the user-defined callback routine, which is called by the solver at various points of the branch-and-cut algorithm.

The callback routine passed to the MIP solver should be written by the user and has the following specification:

\[
\text{void foo_bar(glp_tree *tree, void *info);}\]

where \text{tree} is a pointer to the data structure \text{glp_tree}, which should be used on subsequent calls to branch-and-cut interface routines, and \text{info} is a transit pointer passed to the routine \text{glp_intopt}, which may be used by the application program to pass some external data to the callback routine.

The callback routine is passed to the MIP solver through the control parameter structure \text{glp_iocp} (see Chapter “Basic API Routines”, Section “Mixed integer programming routines”, Subsection “Solve MIP problem with the branch-and-cut method”) as follows:

\[\text{1The name foo_bar used here is a placeholder for the callback routine name.}\]
To determine why it is being called by the MIP solver the callback routine should use the routine `glp_ios_reason` (described in this section below), which returns a code indicating the reason for calling. Depending on the reason the callback routine may perform necessary actions to control the solution process.

The reason codes, which correspond to various points of the branch-and-cut algorithm implemented in the MIP solver, are described in Subsection “Reasons for calling the callback routine” below.

To ignore calls for reasons, which are not processed by the callback routine, it should just return to the MIP solver doing nothing. For example:

```c
void foo_bar(glp_tree *tree, void *info)
{
    . . .
    switch (glp_ios_reason(tree))
    {
    case GLP_IBRANCH:
        . . .
        break;
    case GLP_ISELECT:
        . . .
        break;
    default:
        /* ignore call for other reasons */
        break;
    }
    return;
}
```

To control the solution process as well as to obtain necessary information the callback routine may use the branch-and-cut API routines described in this chapter. Names of all these routines begin with `glp_ios_`.

---

```
glp_prob *mip;
glp_ioocp parm;
. . .
glp_init_ioocp(&parm);
. . .
parm.cb_func = foo_bar;
parm.cb_info = ... ;
ret = glp_intopt(mip, &parm);
. . .
```
5.1.2 Branch-and-cut algorithm

This section gives a schematic description of the branch-and-cut algorithm as it is implemented in the GLPK MIP solver.

1. Initialization
   Set \( L := \{P_0\} \), where \( L \) is the active list (i.e. the list of active subproblems), \( P_0 \) is the original MIP problem to be solved.
   Set \( z^{best} := +\infty \) (in case of minimization) or \( z^{best} := -\infty \) (in case of maximization), where \( z^{best} \) is incumbent value, i.e. an upper (minimization) or lower (maximization) global bound for \( z^{opt} \), the optimal objective value for \( P^0 \).

2. Subproblem selection
   If \( L = \emptyset \) then GO TO 9.
   Select \( P \in L \), i.e. make active subproblem \( P \) current.

3. Solving LP relaxation
   Solve \( P^{LP} \), which is LP relaxation of \( P \).
   If \( P^{LP} \) has no primal feasible solution then GO TO 8.
   Let \( z^{LP} \) be the optimal objective value for \( P^{LP} \).
   If \( z^{LP} \geq z^{best} \) (in case of minimization) or \( z^{LP} \leq z^{best} \) (in case of maximization) then GO TO 8.

4. Adding “lazy” constraints
   Let \( x^{LP} \) be the optimal solution to \( P^{LP} \).
   If there are “lazy” constraints (i.e. essential constraints not included in the original MIP problem \( P_0 \)), which are violated at the optimal point \( x^{LP} \), add them to \( P \), and GO TO 3.

5. Check for integrality
   Let \( x_j \) be a variable, which is required to be integer, and let \( x_j^{LP} \in x^{LP} \) be its value in the optimal solution to \( P^{LP} \).
   If \( x_j^{LP} \) are integral for all integer variables, then a better integer feasible solution is found. Store its components, set \( z^{best} := z^{LP} \), and GO TO 8.

6. Adding cutting planes
   If there are cutting planes (i.e. valid constraints for \( P \)), which are violated at the optimal point \( x^{LP} \), add them to \( P \), and GO TO 3.

7. Branching
   Select branching variable \( x_j \), i.e. a variable, which is required to be integer, and whose value \( x_j^{LP} \in x^{LP} \) is fractional in the optimal solution to \( P^{LP} \).
Create new subproblem $P^D$ (so called down branch), which is identical to the current subproblem $P$ with exception that the upper bound of $x_j$ is replaced by $\lfloor x_j^{LP} \rfloor$. (For example, if $x_j^{LP} = 3.14$, the new upper bound of $x_j$ in the down branch will be $\lfloor 3.14 \rfloor = 3$.)

Create new subproblem $P^U$ (so called up branch), which is identical to the current subproblem $P$ with exception that the lower bound of $x_j$ is replaced by $\lceil x_j^{LP} \rceil$. (For example, if $x_j^{LP} = 3.14$, the new lower bound of $x_j$ in the up branch will be $\lceil 3.14 \rceil = 4$.)

Set $L := (L \setminus \{P\}) \cup \{P^D, P^U\}$, i.e. remove the current subproblem $P$ from the active list $L$ and add two new subproblems $P^D$ and $P^U$ to it. Then GO TO 2.

8. Pruning

Remove from the active list $L$ all subproblems (including the current one), whose local bound $\tilde{z}$ is not better than the global bound $z_{\text{best}}$, i.e. set $L := L \setminus \{P\}$ for all $P$, where $\tilde{z} \geq z_{\text{best}}$ (in case of minimization) or $\tilde{z} \leq z_{\text{best}}$ (in case of maximization), and then GO TO 2.

The local bound $\tilde{z}$ for subproblem $P$ is an lower (minimization) or upper (maximization) bound for integer optimal solution to this subproblem (not to the original problem). This bound is local in the sense that only subproblems in the subtree rooted at node $P$ cannot have better integer feasible solutions. Note that the local bound is not necessarily the optimal objective value to LP relaxation $P^{LP}$.

9. Termination

If $z_{\text{best}} = +\infty$ (in case of minimization) or $z_{\text{best}} = -\infty$ (in case of maximization), the original problem $P_0$ has no integer feasible solution. Otherwise, the last integer feasible solution stored on step 5 is the integer optimal solution to the original problem $P_0$ with $z_{\text{opt}} = z_{\text{best}}$. STOP.

5.1.3 The search tree

On the branching step of the branch-and-cut algorithm the current subproblem is divided into two new subproblems, so the set of all subproblems can be represented in the form of a rooted tree, which is called the search or branch-and-bound tree. An example of the search tree is shown on Fig. 1. Each node of the search tree corresponds to a subproblem, so the terms ‘node’ and ‘subproblem’ may be used synonymously.

---

\(^2\)In more general cases the current subproblem may be divided into more than two subproblems. However, currently such feature is not used in GLPK.
In GLPK each node may have one of the following four statuses:
- **current node** is the active node currently being processed;
- **active node** is a leaf node, which still has to be processed;
- **non-active node** is a node, which has been processed, but not fathomed;
- **fathomed node** is a node, which has been processed and fathomed.

In the data structure representing the search tree GLPK keeps only current, active, and non-active nodes. Once a node has been fathomed, it is removed from the tree data structure.

Being created each node of the search tree is assigned a distinct positive integer called the *subproblem reference number*, which may be used by the application program to specify a particular node of the tree. The root node corresponding to the original problem to be solved is always assigned the reference number 1.

### 5.1.4 Current subproblem

The current subproblem is a MIP problem corresponding to the current node of the search tree. It is represented as the GLPK problem object (*glp_prob*) that allows the application program using API routines to access its content in the standard way. If the MIP presolver is not used, it is the original
problem object passed to the routine `glp_intopt`; otherwise, it is an internal problem object built by the MIP presolver.

Note that the problem object is used by the MIP solver itself during the solution process for various purposes (to solve LP relaxations, to perform branching, etc.), and even if the MIP presolver is not used, the current content of the problem object may differ from its original content. For example, it may have additional rows, bounds of some rows and columns may be changed, etc. In particular, LP segment of the problem object corresponds to LP relaxation of the current subproblem. However, on exit from the MIP solver the content of the problem object is restored to its original state.

To obtain information from the problem object the application program may use any API routines, which do not change the object. Using API routines, which change the problem object, is restricted to stipulated cases.

5.1.5 The cut pool

The cut pool is a set of cutting plane constraints maintained by the MIP solver. It is used by the GLPK cut generation routines and may be used by the application program in the same way, i.e. rather than to add cutting plane constraints directly to the problem object the application program may store them to the cut pool. In the latter case the solver looks through the cut pool, selects efficient constraints, and adds them to the problem object.

5.1.6 Reasons for calling the callback routine

The callback routine may be called by the MIP solver for the following reasons.

Request for subproblem selection

The callback routine is called with the reason code `GLP_ISELECT` if the current subproblem has been fathomed and therefore there is no current subproblem.

In response the callback routine may select some subproblem from the active list and pass its reference number to the solver using the routine `glp_ios_select_node`, in which case the solver continues the search from the specified active subproblem. If no selection is made by the callback routine, the solver uses a backtracking technique specified by the control parameter `bt_tech`. 
To explore the active list (i.e. active nodes of the branch-and-bound tree) the callback routine may use the routines \texttt{glp_ios_next_node} and \texttt{glp_ios_prev_node}.

**Request for preprocessing**

The callback routine is called with the reason code \texttt{GLP_IPREPRO} if the current subproblem has just been selected from the active list and its LP relaxation is not solved yet.

In response the callback routine may perform some preprocessing of the current subproblem like tightening bounds of some variables or removing bounds of some redundant constraints.

**Request for row generation**

The callback routine is called with the reason code \texttt{GLP_IROWGEN} if LP relaxation of the current subproblem has just been solved to optimality and its objective value is better than the best known integer feasible solution.

In response the callback routine may add one or more “lazy” constraints (rows), which are violated by the current optimal solution of LP relaxation, using API routines \texttt{glp_add_rows}, \texttt{glp_set_row_name}, \texttt{glp_set_row_bnds}, and \texttt{glp_set_mat_row}, in which case the solver will perform re-optimization of LP relaxation. If there are no violated constraints, the callback routine should just return.

Optimal solution components for LP relaxation can be obtained with API routines \texttt{glp_get_obj_val}, \texttt{glp_get_row_prim}, \texttt{glp_get_row_dual}, \texttt{glp_get_col_prim}, and \texttt{glp_get_col_dual}.

**Request for heuristic solution**

The callback routine is called with the reason code \texttt{GLP_IHEUR} if LP relaxation of the current subproblem being solved to optimality is integer infeasible (i.e. values of some structural variables of integer kind are fractional), though its objective value is better than the best known integer feasible solution.

In response the callback routine may try applying a primal heuristic to find an integer feasible solution,\(^3\) which is better than the best known one. In case of success the callback routine may store such better solution in the problem object using the routine \texttt{glp_ios_heur_sol}.

\(^3\)Integer feasible to the original MIP problem, not to the current subproblem.
Request for cut generation

The callback routine is called with the reason code GLP_ICUTGEN if LP relaxation of the current subproblem being solved to optimality is integer infeasible (i.e. values of some structural variables of integer kind are fractional), though its objective value is better than the best known integer feasible solution.

In response the callback routine may reformulate the current subproblem (before it will be splitted up due to branching) by adding to the problem object one or more cutting plane constraints, which cut off the fractional optimal point from the MIP polytope.4

Adding cutting plane constraints may be performed in two ways. One way is the same as for the reason code GLP_IROWGEN (see above), in which case the callback routine adds new rows corresponding to cutting plane constraints directly to the current subproblem.

The other way is to add cutting plane constraints to the cut pool, a set of cutting plane constraints maintained by the solver, rather than directly to the current subproblem. In this case after return from the callback routine the solver looks through the cut pool, selects efficient cutting plane constraints, adds them to the current subproblem, drops other constraints, and then performs re-optimization.

Request for branching

The callback routine is called with the reason code GLP_IBRANCH if LP relaxation of the current subproblem being solved to optimality is integer infeasible (i.e. values of some structural variables of integer kind are fractional), though its objective value is better than the best known integer feasible solution.

In response the callback routine may choose some variable suitable for branching (i.e. integer variable, whose value in optimal solution to LP relaxation of the current subproblem is fractional) and pass its ordinal number to the solver using the routine glp_ios_branch_upon, in which case the solver splits the current subproblem in two new subproblems and continues the search. If no choice is made by the callback routine, the solver uses a branching technique specified by the control parameter br_tech.

4Since these constraints are added to the current subproblem, they may be globally as well as locally valid.
Better integer solution found

The callback routine is called with the reason code GLP_IBINGO if LP relaxation of the current subproblem being solved to optimality is integer feasible (i.e. values of all structural variables of integer kind are integral within the working precision) and its objective value is better than the best known integer feasible solution.

Optimal solution components for LP relaxation can be obtained in the same way as for the reason code GLP_IROWGEN (see above).

Components of the new MIP solution can be obtained with API routines glp_mip_obj_val, glp_mip_row_val, and glp_mip_col_val. Note, however, that due to row/cut generation there may be additional rows in the problem object.

The difference between optimal solution to LP relaxation and corresponding MIP solution is that in the former case some structural variables of integer kind (namely, basic variables) may have values, which are close to nearest integers within the working precision, while in the latter case all such variables have exact integral values.

The reason GLP_IBINGO is intended only for informational purposes, so the callback routine should not modify the problem object in this case.
5.2 Basic routines

5.2.1 glp_ios_reason—determine reason for calling the callback routine

Synopsis

int glp_ios_reason(glp_tree *tree);

Returns

The routine glp_ios_reason returns a code, which indicates why the user-defined callback routine is being called:
- GLP_ISELECT—request for subproblem selection;
- GLP_IPREPRO—request for preprocessing;
- GLP_IROWGEN—request for row generation;
- GLP_IHEUR—request for heuristic solution;
- GLP_ICUTGEN—request for cut generation;
- GLP_IBRANCH—request for branching;
- GLP_IBINGO—better integer solution found.

5.2.2 glp_ios_get_prob—access the problem object

Synopsis

glp_prob *glp_ios_get_prob(glp_tree *tree);

Description

The routine glp_ios_get_prob can be called from the user-defined callback routine to access the problem object, which is used by the MIP solver. It is the original problem object passed to the routine glp_intopt if the MIP presolver is not used; otherwise it is an internal problem object built by the presolver.

Returns

The routine glp_ios_get_prob returns a pointer to the problem object used by the MIP solver.
Comments

To obtain various information about the problem instance the callback routine can access the problem object (i.e. the object of type `glp_prob`) using the routine `glp_ios_get_prob`. It is the original problem object passed to the routine `glp_intopt` if the MIP presolver is not used; otherwise it is an internal problem object built by the presolver.

5.2.3 `glp_ios_row_attr`—determine additional row attributes

Synopsis

```c
void glp_ios_row_attr(glp_tree *tree, int i, glp_attr *attr);
```

Description

The routine `glp_ios_row_attr` retrieves additional attributes of `i`-th row of the current subproblem and stores them in the structure `glp_attr`, which the parameter `attr` points to.

The structure `glp_attr` has the following fields:

- **int level**
  Subproblem level at which the row was created. (If `level = 0`, the row was added either to the original problem object passed to the routine `glp_intopt` or to the root subproblem on generating “lazy” or/and cutting plane constraints.)

- **int origin**
  The row origin flag:
  - `GLP_RF_REG`—regular constraint;
  - `GLP_RF_LAZY`—“lazy” constraint;
  - `GLP_RF_CUT`—cutting plane constraint.

- **int klass**
  The row class descriptor, which is a number passed to the routine `glp_ios_add_row` as its third parameter. If the row is a cutting plane constraint generated by the solver, its class may be the following:
  - `GLP_RF_GMI`—Gomory’s mixed integer cut;
  - `GLP_RF_MIR`—mixed integer rounding cut;
  - `GLP_RF_COV`—mixed cover cut;
  - `GLP_RF_CLQ`—clique cut.
5.2.4  glp_ios_mip_gap—compute relative MIP gap

Synopsis

double glp_ios_mip_gap(glp_tree *tree);

Description

The routine glp_ios_mip_gap computes the relative MIP gap (also called duality gap) with the following formula:

\[
gap = \frac{|best\_mip - best\_bnd|}{|best\_mip| + DBL\_EPSILON}
\]

where best_mip is the best integer feasible solution found so far, best_bnd is the best (global) bound. If no integer feasible solution has been found yet, gap is set to DBL_MAX.

Returns

The routine glp_ios_mip_gap returns the relative MIP gap.

Comments

The relative MIP gap is used to measure the quality of the best integer feasible solution found so far, because the optimal solution value \( z^* \) for the original MIP problem always lies in the range

for minimization, or in the range

for maximization.

To express the relative MIP gap in percents the value returned by the routine glp_ios_mip_gap should be multiplied by 100%.
5.2.5  glp_ios_node_data—access application-specific data

Synopsis

void *glp_ios_node_data(glp_tree *tree, int p);

Description

The routine glp_ios_node_data allows the application accessing a memory
block allocated for the subproblem (which may be active or inactive), whose
reference number is p.

The size of the block is defined by the control parameter cb_size passed
to the routine glp_intopt. The block is initialized by binary zeros on creat-
ing corresponding subproblem, and its contents is kept until the subproblem
will be removed from the tree.

The application may use these memory blocks to store specific data for
each subproblem.

Returns

The routine glp_ios_node_data returns a pointer to the memory block for
the specified subproblem. Note that if cb_size = 0, the routine returns a
null pointer.

5.2.6  glp_ios_select_node—select subproblem to continue the
search

Synopsis

void glp_ios_select_node(glp_tree *tree, int p);

Description

The routine glp_ios_select_node can be called from the user-defined call-
back routine in response to the reason GLP_ISELECT to select an active sub-
problem, whose reference number is p. The search will be continued from
the subproblem selected.
5.2.7 glp_ios_heur_sol—provide solution found by heuristic

Synopsis

int glp_ios_heur_sol(glp_tree *tree, const double x[]);

Description

The routine glp_ios_heur_sol can be called from the user-defined callback routine in response to the reason GLP_IHEUR to provide an integer feasible solution found by a primal heuristic.

Primal values of all variables (columns) found by the heuristic should be placed in locations $x[1], \ldots, x[n]$, where $n$ is the number of columns in the original problem object. Note that the routine glp_ios_heur_sol does not check primal feasibility of the solution provided.

Using the solution passed in the array $x$ the routine computes value of the objective function. If the objective value is better than the best known integer feasible solution, the routine computes values of auxiliary variables (rows) and stores all solution components in the problem object.

Returns

If the provided solution is accepted, the routine glp_ios_heur_sol returns zero. Otherwise, if the provided solution is rejected, the routine returns non-zero.

5.2.8 glp_ios_can_branch—check if can branch upon specified variable

Synopsis

int glp_ios_can_branch(glp_tree *tree, int j);

Returns

If $j$-th variable (column) can be used to branch upon, the routine returns non-zero, otherwise zero.
5.2.9  glp_ios_branch_upon—choose variable to branch upon

Synopsis

void glp_ios_branch_upon(glp_tree *tree, int j, int sel);

Description

The routine glp_ios_branch_upon can be called from the user-defined call-
back routine in response to the reason GLP_IBRANCH to choose a branching
variable, whose ordinal number is $j$. Should note that only variables, for
which the routine glp_ios_can_branch returns non-zero, can be used to
branch upon.

The parameter sel is a flag that indicates which branch (subproblem)
should be selected next to continue the search:

- GLP_DN_BRNCH—select down-branch;
- GLP_UP_BRNCH—select up-branch;
- GLP_NO_BRNCH—use general selection technique.

Comments

On branching the solver removes the current active subproblem from the
active list and creates two new subproblems (down- and up-branches), which
are added to the end of the active list. Note that the down-branch is created
before the up-branch, so the last active subproblem will be the up-branch.

The down- and up-branches are identical to the current subproblem with
exception that in the down-branch the upper bound of $x_j$, the variable
chosen to branch upon, is replaced by $\lfloor x^*_j \rfloor$, while in the up-branch the
lower bound of $x_j$ is replaced by $\lceil x^*_j \rceil$, where $x^*_j$ is the value of $x_j$ in optimal
solution to LP relaxation of the current subproblem. For example, if $x^*_j = 3.14$, the new upper bound of $x_j$ in the down-branch is $\lfloor 3.14 \rfloor = 3$, and the
new lower bound in the up-branch is $\lceil 3.14 \rceil = 4$.

Additionally the callback routine may select either down- or up-branch,
from which the solver will continue the search. If none of the branches is
selected, a general selection technique will be used.
5.2.10  glp_ios_terminate—terminate the solution process

Synopsis
void glp_ios_terminate(glp_tree *tree);

Description
The routine glp_ios_terminate sets a flag indicating that the MIP solver should prematurely terminate the search.
5.3 The search tree exploring routines

5.3.1 glp_ios_tree_size—determine size of the search tree

Synopsis

void glp_ios_tree_size(glp_tree *tree, int *a_cnt, int *n_cnt,
int *t_cnt);

Description

The routine glp_ios_tree_size stores the following three counts which characterize the current size of the search tree:

- \texttt{a\_cnt} is the current number of active nodes, i.e. the current size of the active list;
- \texttt{n\_cnt} is the current number of all (active and inactive) nodes;
- \texttt{t\_cnt} is the total number of nodes including those which have been already removed from the tree. This count is increased whenever a new node appears in the tree and never decreased.

If some of the parameters \texttt{a\_cnt}, \texttt{n\_cnt}, \texttt{t\_cnt} is a null pointer, the corresponding count is not stored.

5.3.2 glp_ios_curr_node—determine current active subproblem

Synopsis

int glp_ios_curr_node(glp_tree *tree);

Returns

The routine glp_ios_curr_node returns the reference number of the current active subproblem. However, if the current subproblem does not exist, the routine returns zero.
5.3.3  glp_ios_next_node—determine next active subproblem

Synopsis

```c
int glp_ios_next_node(glp_tree *tree, int p);
```

Returns

If the parameter `p` is zero, the routine `glp_ios_next_node` returns the reference number of the first active subproblem. However, if the tree is empty, zero is returned.

If the parameter `p` is not zero, it must specify the reference number of some active subproblem, in which case the routine returns the reference number of the next active subproblem. However, if there is no next active subproblem in the list, zero is returned.

All subproblems in the active list are ordered chronologically, i.e. subproblem `A` precedes subproblem `B` if `A` was created before `B`.

5.3.4  glp_ios_prev_node—determine previous active subproblem

Synopsis

```c
int glp_ios_prev_node(glp_tree *tree, int p);
```

Returns

If the parameter `p` is zero, the routine `glp_ios_prev_node` returns the reference number of the last active subproblem. However, if the tree is empty, zero is returned.

If the parameter `p` is not zero, it must specify the reference number of some active subproblem, in which case the routine returns the reference number of the previous active subproblem. However, if there is no previous active subproblem in the list, zero is returned.

All subproblems in the active list are ordered chronologically, i.e. subproblem `A` precedes subproblem `B` if `A` was created before `B`. 
5.3.5 glp_ios_up_node—determine parent subproblem

Synopsis

int glp_ios_up_node(glp_tree *tree, int p);

Returns

The parameter p must specify the reference number of some (active or inactive) subproblem, in which case the routine iet_get_up_node returns the reference number of its parent subproblem. However, if the specified subproblem is the root of the tree and, therefore, has no parent, the routine returns zero.

5.3.6 glp_ios_node_level—determine subproblem level

Synopsis

int glp_ios_node_level(glp_tree *tree, int p);

Returns

The routine glp_ios_node_level returns the level of the subproblem, whose reference number is p, in the branch-and-bound tree. (The root subproblem has level 0, and the level of any other subproblem is the level of its parent plus one.)

5.3.7 glp_ios_node_bound—determine subproblem local bound

Synopsis

double glp_ios_node_bound(glp_tree *tree, int p);

Returns

The routine glp_ios_node_bound returns the local bound for (active or inactive) subproblem, whose reference number is p.

Comments

The local bound for subproblem p is an lower (minimization) or upper (maximization) bound for integer optimal solution to this subproblem (not to the
original problem). This bound is local in the sense that only subproblems in
the subtree rooted at node \( p \) cannot have better integer feasible solutions.

On creating a subproblem (due to the branching step) its local bound
is inherited from its parent and then may get only stronger (never weaker).
For the root subproblem its local bound is initially set to \(-DBL\_MAX\) (min-
imization) or \(+DBL\_MAX\) (maximization) and then improved as the root LP
relaxation has been solved.

Note that the local bound is not necessarily the optimal objective value
to corresponding LP relaxation.

5.3.8 glp_ios_best_node—find active subproblem with best
local bound

Synopsis

\[
\text{int glp_ios_best_node(glp_tree *tree);} \\
\]

Returns

The routine \texttt{glp_ios_best_node} returns the reference number of the active
subproblem, whose local bound is best (i.e. smallest in case of minimization
or largest in case of maximization). However, if the tree is empty, the routine
returns zero.

Comments

The best local bound is an lower (minimization) or upper (maximization)
bound for integer optimal solution to the original MIP problem.
5.4 The cut pool routines

5.4.1 glp_ios_pool_size—determine current size of the cut pool

Synopsis

\begin{verbatim}
int glp_ios_pool_size(glp_tree *tree);
\end{verbatim}

Returns

The routine \texttt{glp_ios_pool_size} returns the current size of the cut pool, that is, the number of cutting plane constraints currently added to it.

5.4.2 glp_ios_add_row—add constraint to the cut pool

Synopsis

\begin{verbatim}
int glp_ios_add_row(glp_tree *tree, const char *name, int klass, int flags, int len, const int ind[], const double val[], int type, double rhs);
\end{verbatim}

Description

The routine \texttt{glp_ios_add_row} adds specified row (cutting plane constraint) to the cut pool.

The cutting plane constraint should have the following format:

\[
\sum_{j \in J} a_j x_j \begin{cases} \geq \\ \leq \end{cases} b,
\]

where \( J \) is a set of indices (ordinal numbers) of structural variables, \( a_j \) are constraint coefficients, \( x_j \) are structural variables, \( b \) is the right-hand side.

The parameter \texttt{name} specifies a symbolic name assigned to the constraint (1 up to 255 characters). If it is \texttt{NULL} or an empty string, no name is assigned.

The parameter \texttt{klass} specifies the constraint class, which must be either zero or a number in the range from 101 to 200. The application may use this attribute to distinguish between cutting plane constraints of different classes.\(^5\)

The parameter \texttt{flags} currently is not used and must be zero.

\(^5\)Constraint classes numbered from 1 to 100 are reserved for GLPK cutting plane generators.
Ordinal numbers of structural variables (i.e. column indices) $j \in J$ and numerical values of corresponding constraint coefficients $a_j$ must be placed in locations $\text{ind}[1], \ldots, \text{ind}[\text{len}]$ and $\text{val}[1], \ldots, \text{val}[\text{len}]$, respectively, where $\text{len} = |J|$ is the number of constraint coefficients, $0 \leq \text{len} \leq n$, and $n$ is the number of columns in the problem object. Coefficients with identical column indices are not allowed. Zero coefficients are allowed, however, they are ignored.

The parameter $\text{type}$ specifies the constraint type as follows:
- $\text{GLP_L0}$ means inequality constraint $\sum a_j x_j \geq b$;
- $\text{GLP_UP}$ means inequality constraint $\sum a_j x_j \leq b$;

The parameter $\text{rhs}$ specifies the right-hand side $b$.

All cutting plane constraints in the cut pool are identified by their ordinal numbers $1, 2, \ldots, \text{size}$, where $\text{size}$ is the current size of the cut pool. New constraints are always added to the end of the cut pool, thus, ordinal numbers of previously added constraints are not changed.

**Returns**

The routine $\text{glp_ios_add_row}$ returns the ordinal number of the cutting plane constraint added, which is the new size of the cut pool.

**Example**

/* generate triangle cutting plane:
   $x[i] + x[j] + x[k] \leq 1$ */

   . . .
/* add the constraint to the cut pool */
\text{ind}[1] = i, \text{val}[1] = 1.0;
\text{ind}[2] = j, \text{val}[2] = 1.0;
\text{ind}[3] = k, \text{val}[3] = 1.0;
\text{glp_ios_add_row}(\text{tree}, \text{NULL}, \text{TRIANGLE_CUT}, 0, 3, \text{ind}, \text{val},
   \text{GLP_UP}, 1.0);

**Comments**

Cutting plane constraints added to the cut pool are intended to be then added only to the current subproblem, so these constraints can be globally as well as locally valid. However, adding a constraint to the cut pool does not mean that it will be added to the current subproblem—it depends on
the solver’s decision: if the constraint seems to be efficient, it is moved from the pool to the current subproblem, otherwise it is simply dropped.\(^6\)

Normally, every time the callback routine is called for cut generation, the cut pool is empty. On the other hand, the solver itself can generate cutting plane constraints (like Gomory’s or mixed integer rounding cuts), in which case the cut pool may be non-empty.

5.4.3 glp_ios_del_row—remove constraint from the cut pool

Synopsis

```c
void glp_ios_del_row(glp_tree *tree, int i);
```

Description

The routine `glp_ios_del_row` deletes \(i\)-th row (cutting plane constraint) from the cut pool, where \(1 \leq i \leq size\) is the ordinal number of the constraint in the pool, `size` is the current size of the cut pool.

Note that deleting a constraint from the cut pool leads to changing ordinal numbers of other constraints remaining in the pool. New ordinal numbers of the remaining constraints are assigned under assumption that the original order of constraints is not changed. Let, for example, there be four constraints \(a, b, c\) and \(d\) in the cut pool, which have ordinal numbers 1, 2, 3 and 4, respectively, and let constraint \(b\) have been deleted. Then after deletion the remaining constraint \(a, c\) and \(d\) are assigned new ordinal numbers 1, 2 and 3, respectively.

To find the constraint to be deleted the routine `glp_ios_del_row` uses “smart” linear search, so it is recommended to remove constraints in a natural or reverse order and avoid removing them in a random order.

Example

```c
/* keep first 10 constraints in the cut pool and remove other constraints */
while (glp_ios_pool_size(tree) > 10)
    glp_ios_del_row(tree, glp_ios_pool_size(tree));
```

\(^6\)Globally valid constraints could be saved and then re-used for other subproblems, but currently such feature is not implemented.
5.4.4  glp_ios_clear_pool—remove all constraints from the cut pool

Synopsis

void glp_ios_clear_pool(glp_tree *tree);

Description

The routine glp_ios_clear_pool makes the cut pool empty deleting all existing rows (cutting plane constraints) from it.
Chapter 6

Miscellaneous API Routines

6.1 GLPK environment routines

6.1.1 glp_long—64-bit integer data type

Some GLPK API routines use 64-bit integer data type, which is declared in the header glpk.h as follows:

typedef struct { int lo, hi; } glp_long;

where \textit{lo} contains low 32 bits, and \textit{hi} contains high 32 bits of 64-bit integer value.\footnote{GLPK conforms to ILP32, LLP64, and LP64 programming models, where the built-in type \textit{int} corresponds to 32-bit integers.}

6.1.2 glp_init_env—initialize GLPK environment

Synopsis

\texttt{int glp_init_env(void);} 

Description

The routine \texttt{glp_init_env} initializes the GLPK environment. Normally the application program does not need to call this routine, because it is called automatically on the first call to any API routine.
Returns

The routine glp_init_env returns one of the following codes:
0 — initialization successful;
1 — environment is already initialized;
2 — initialization failed (insufficient memory);
3 — initialization failed (unsupported programming model).

6.1.3 glp_version—determine library version

Synopsis

const char *glp_version(void);

Returns

The routine glp_version returns a pointer to a null-terminated character string, which specifies the version of the GLPK library in the form "X.Y", where ‘X’ is the major version number, and ‘Y’ is the minor version number, for example, "4.16".

Example

printf("GLPK version is %s\n", glp_version());

6.1.4 glp_free_env—free GLPK environment

Synopsis

int glp_free_env(void);

Description

The routine glp_free_env frees all resources used by GLPK routines (memory blocks, etc.) which are currently still in use.

Normally the application program does not need to call this routine, because GLPK routines always free all unused resources. However, if the application program even has deleted all problem objects, there will be several memory blocks still allocated for the internal library needs. For some reasons the application program may want GLPK to free this memory, in which case it should call glp_free_env.

Note that a call to glp_free_env invalidates all problem objects which still exist.
Returns

The routine `glp_free_env` returns one of the following codes:
0 — termination successful;
1 — environment is inactive (was not initialized).

### 6.1.5 glp_printf—write formatted output to terminal

**Synopsis**

```c
void glp_printf(const char *fmt, ...);
```

**Description**

The routine `glp_printf` uses the format control string `fmt` to format its parameters and writes the formatted output to the terminal.

This routine is a replacement of the standard C function `printf` and used by all GLPK routines to perform terminal output. The application program may use `glp_printf` for the same purpose that allows controlling its terminal output with the routines `glp_term_out` and `glp_term_hook`.

### 6.1.6 glp_vprintf—write formatted output to terminal

**Synopsis**

```c
void glp_vprintf(const char *fmt, va_list arg);
```

**Description**

The routine `glp_vprintf` uses the format control string `fmt` to format its parameters specified by the list `arg` and writes the formatted output to the terminal.

This routine is a replacement of the standard C function `vprintf` and used by all GLPK routines to perform terminal output. The application program may use `glp_vprintf` for the same purpose that allows controlling its terminal output with the routines `glp_term_out` and `glp_term_hook`. 
6.1.7  glp_term_out—enable/disable terminal output

Synopsis

int glp_term_out(int flag);

Description

Depending on the parameter flag the routine glp_term_out enables or dis-
ables terminal output performed by glpk routines:

   GLP_ON — enable terminal output;
   GLP_OFF — disable terminal output.

Returns

The routine glp_term_out returns the previous value of the terminal output
flag (GLP_ON or GLP_OFF).

6.1.8  glp_term_hook—intercept terminal output

Synopsis

void glp_term_hook(int (*func)(void *info, const char *s),
                   void *info);

Description

The routine glp_term_hook installs the user-defined hook routine to inter-
cept all terminal output performed by GLPK routines.

   The parameter func specifies the user-defined hook routine. It is called
   from an internal printing routine, which passes to it two parameters: info
   and s. The parameter info is a transit pointer specified in corresponding
   call to the routine glp_term_hook: it may be used to pass some additional
   information to the hook routine. The parameter s is a pointer to the null
   terminated character string, which is intended to be written to the terminal.
   If the hook routine returns zero, the printing routine writes the string s to
   the terminal in a usual way; otherwise, if the hook routine returns non-zero,
   no terminal output is performed.

   To uninstall the hook routine both parameters func and info should be
   specified as NULL.
Example

```c
static int hook(void *info, const char *s)
{  FILE *foo = info;
   fputs(s, foo);
   return 1;
}

int main(void)
{  FILE *foo;
   ...
   /* redirect terminal output */
   glp_term_hook(hook, foo);
   ...
   /* resume terminal output */
   glp_term_hook(NULL, NULL);
   ...
}
```

6.1.9 glp_open_tee—start copying terminal output

Synopsis

```c
int glp_open_tee(const char *fname);
```

Description

The routine `glp_open_tee` starts copying all the terminal output to an output text file, whose name is specified by the character string `fname`.

Returns

The routine `glp_open_tee` returns one of the following codes:

- 0 — operation successful;
- 1 — copying terminal output is already active;
- 2 — unable to create output file.
6.1.10  glp_close_tee—stop copying terminal output

Synopsis

int glp_close_tee(void);

Description

The routine glp_close_tee stops copying the terminal output to the output
text file previously open by the routine glp_open_tee closing that file.

Returns

The routine glp_close_tee returns one of the following codes:
0 — operation successful;
1 — copying terminal output was not started.

6.1.11  glp_error—display error message and terminate execution

Synopsis

void glp_error(const char *fmt, ...);

Description

The routine glp_error (implemented as a macro) formats its parameters
using the format control string fmt, writes the formatted message to the
terminal, and then abnormally terminates the program.

6.1.12  glp_assert—check logical condition

Synopsis

void glp_assert(int expr);

Description

The routine glp_assert (implemented as a macro) checks a logical condition
specified by the expression expr. If the condition is true (non-zero), the
routine does nothing; otherwise, if the condition is false (zero), the routine
prints an error message and abnormally terminates the program.
This routine is a replacement of the standard C function `assert` and used by all GLPK routines to check program logic. The application program may use `glp_assert` for the same purpose.

### 6.1.13 glp_error_hook—install hook to intercept abnormal termination

**Synopsis**

```c
void glp_error_hook(void (*func)(void *info), void *info);
```

**Description**

The routine `glp_error_hook` installs a user-defined hook routine to intercept abnormal termination.

The parameter `func` specifies the user-defined hook routine. It is called from the routine `glp_error` before the latter calls the abort function to abnormally terminate the application program because of fatal error. The parameter `info` is a transit pointer, specified in the corresponding call to the routine `glp_error_hook`; it may be used to pass some information to the hook routine.

To uninstall the hook routine the parameters `func` and `info` should be specified as `NULL`.

**Usage note**

If the hook routine returns, the application program is abnormally terminated. To prevent abnormal termination the hook routine may perform a global jump using the standard function `longjmp`, in which case the application program must call the routine `glp_free_env`.

### 6.1.14 glp_malloc—allocate memory block

**Synopsis**

```c
void *glp_malloc(int size);
```

**Description**

The routine `glp_malloc` dynamically allocates a memory block of `size` bytes long. Should note that:

1) the parameter `size` must be positive;
2) being allocated the memory block contains arbitrary data, that is, it is not initialized by binary zeros;

3) if the block cannot be allocated due to insufficient memory, the routine prints an error message and abnormally terminates the program.

This routine is a replacement of the standard C function `malloc` and used by all GLPK routines for dynamic memory allocation. The application program may use `glp_malloc` for the same purpose.

**Returns**

The routine `glp_malloc` returns a pointer to the memory block allocated. To free this block the routine `glp_free` (not the standard C function `free`) must be used.

### 6.1.15 glp_calloc—allocate memory block

**Synopsis**

```c
void *glp_calloc(int n, int size);
```

**Description**

The routine `glp_calloc` dynamically allocates a memory block of \( n \times \text{size} \) bytes long. Should note that:

1) both parameters \( n \) and \( \text{size} \) must be positive;

2) being allocated the memory block contains arbitrary data, that is, it is not initialized by binary zeros;

3) if the block cannot be allocated due to insufficient memory, the routine prints an error message and abnormally terminates the program.

This routine is a replacement of the standard C function `calloc` (with exception that the block is not cleaned) and used by all GLPK routines for dynamic memory allocation. The application program may use `glp_calloc` for the same purpose.

**Returns**

The routine `glp_calloc` returns a pointer to the memory block allocated. To free this block the routine `glp_free` (not the standard C function `free`) must be used.
6.1.16 glp_free—free memory block

Synopsis

void glp_free(void *ptr);

Description

The routine glp_free frees (deallocates) a memory block pointed to by ptr, which was previously allocated by the routine glp_malloc or glp_calloc. Note that the pointer ptr must valid and must not be NULL.

This routine is a replacement of the standard C function free and used by all GLPK routines for dynamic memory allocation. The application program may use glp_free for the same purpose.

6.1.17 glp_mem_usage—get memory usage information

Synopsis

void glp_mem_usage(int *count, int *cpeak, glp_long *total, 
 glp_long *tpeak);

Description

The routine glp_mem_usage reports some information about utilization of the memory by the routines glp_malloc, glp_calloc, and glp_free. Information is stored to locations specified by corresponding parameters (see below). Any parameter can be specified as NULL, in which case corresponding information is not stored.

*count is the number of currently allocated memory blocks.
*cpeak is the peak value of *count reached since the initialization of the GLPK library environment.
*total is the total amount, in bytes, of currently allocated memory blocks.
*tpeak is the peak value of *total reached since the initialization of the GLPK library environment.

Example

glp_mem_usage(&count, NULL, NULL, NULL);
printf("%d memory block(s) are still allocated\n", count);

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6.1.18  glp_mem_limit—set memory usage limit

Synopsis
void glp_mem_limit(int limit);

Description
The routine glp_mem_limit limits the amount of memory available for dy-
namic allocation (with the routines glp_malloc and glp_calloc) to limit
megabytes.

6.1.19  glp_time—determine current universal time

Synopsis

Returns
The routine glp_time returns the current universal time (UTC), in milli-
seconds, elapsed since 00:00:00 GMT January 1, 1970.

6.1.20  glp_difftime—compute difference between two time
values

Synopsis
double glp_difftime(glp_long t1, glp_long t0);

Returns
The routine glp_difftime returns the difference between two time values
t1 and t0, expressed in seconds.
6.2 Plain data file reading routines

6.2.1 Introduction

On developing simple applications to solve optimization problems it is often needed to read data from plain text files. To do this the standard C function \texttt{fscanf} may be used, however, it is not convenient; for example, if it scans an integer number according to the format specification '%d', and that number is coded incorrectly, no diagnostics is provided.

This section describes a set of GLPK API routines, which may be used in application programs to simplify reading data from plain text files.

Example 1

The following main program reads ten integer numbers from plain text file \texttt{data.txt} and prints their sum.

```c
/* sdfsamp1.c */
#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>
int main(void)
{
  glp_data *data;
  int j, num, sum;
  /* open plain data file */
  data = glp_sdf_open_file("data.txt");
  if (data == NULL) exit(EXIT_FAILURE);
  sum = 0;
  for (j = 1; j <= 10; j++)
  { /* read next integer number */
    num = glp_sdf_read_int(data);
    sum += num;
  }
  printf("sum = %d\n", sum);
  /* close plain data file */
  glp_sdf_close_file(data);
  return 0;
}
/* eof */
```

The input data are coded in free format. For example, the file \texttt{data.txt} may look like this:

```
123 65 432 890 -12 743 895 -7 111 326
```
or like this:

```
123  65  432  890  -12
743  895  -7   111  326
```

If the input data file contains incorrect data, the routine `glp_sdf_read_int` prints an error message and, if no error handling is provided by the application program, abnormally terminates program execution. For example, the file `data.txt` could contain the following data:

```
123  65  432  890  -12
743  895  =7   111  326
```

in which case the error message would be the following:

```
data.txt:2: cannot convert ‘=7’ to integer
```

**Example 2**

As it was said above, by default any attempt to read incorrect data leads to abnormal termination. However, sometimes it is desirable to catch such errors. This feature is illustrated by the following main program, which does the same job as in the previous example.

```
/* sdfsamp2.c */

#include <setjmp.h>
#include <stdio.h>
#include <stdlib.h>
#include <glpk.h>

int main(void)
{
    glp_data *data;
    jmp_buf jump;
    int j, num, sum, ret;
    /* open plain data file */
    data = glp_sdf_open_file("data.txt");
    if (data == NULL)
    {
        ret = EXIT_FAILURE;
        goto done;
    }
    /* set up error handling */
    if (setjmp(jump))
    {
        ret = EXIT_FAILURE;
        goto done;
    }
    /* set up error handling */
    if (setjmp(jump))
    {
        ret = EXIT_FAILURE;
        goto done;
    }
```

glp_sdf_set_jump(data, jump);
/* read and process data */
sum = 0;
for (j = 1; j <= 10; j++)
{ /* read next integer number */
    num = glp_sdf_read_int(data);
    if (abs(num) > 1000)
        glp_sdf_error(data, "integer %d too big\n", num);
    if (num < 0)
        glp_sdf_warning(data, "integer %d is negative\n", num);
    sum += num;
}
printf("sum = %d\n", sum);
ret = EXIT_SUCCESS;
done: /* close plain data file */
    if (data != NULL) glp_sdf_close_file(data);
    return ret;

/* eof */

6.2.2  glp_sdf_open_file—open plain data file

Synopsis

glp_data *glp_sdf_open_file(const char *fname);

Description

The routine glp_sdf_open_file opens a plain data file, whose name is specified by the character string fname.

Returns

If the operation was successful, the routine glp_sdf_open_file returns a pointer to the opaque program object of the type glp_data associated with the plain data file. Otherwise, if the operation failed, the routine prints an error message and returns NULL.

Note

The application program should use the pointer returned by the routine glp_sdf_open_file to perform all subsequent operations on the data file.

---

^2This data structure is declared in the header file glpk.h.
6.2.3 glp_sdf_set_jump—set up error handling

Synopsis

void glp_sdf_set_jump(glp_data *data, jmp_buf jump);

Description

The routine glp_sdf_set_jump sets up error handling for the plain data file specified by the parameter data.

The parameter jump specifies the environment buffer, which must be initialized with the standard C function setjmp prior to call to the routine glp_sdf_set_jump. Detecting any incorrect data in the corresponding plain data file will cause non-local “go to” by a call to the standard C function longjmp.

The parameter jump can be specified as NULL, in which case the routine glp_sdf_set_jump restores the default behavior, in which case detecting incorrect data leads to abnormal termination.

6.2.4 glp_sdf_error—print error message

Synopsis

void glp_sdf_error(glp_data *data, const char *fmt, ...);

Description

The routine glp_sdf_error prints an error message related to the plain data file specified by the parameter data. If error handling was not previously provided, the routine then abnormally terminates execution of the application program. Otherwise, it signals about the error by a call to the standard C function longjmp.

The character string fmt and optional parameters following it have the same meaning as for the standard C function printf.

The message produced by the routine glp_sdf_error looks like follows:

file:line: message text

where file is the filename passed to the routine glp_sdf_open and line is the current line number.
6.2.5 glp_sdf_warning—print warning message

Synopsis

void glp_sdf_warning(glp_data *data, const char *fmt, ...);

Description

The routine glp_sdf_warning prints a warning message related to the plain data file specified by the parameter data.

The character string fmt and optional parameters following it have the same meaning as for the standard C function printf.

The message produced by the routine glp_sdf_warning looks like follows:

    file:line: warning: message text

where file is the filename passed to the routine glp_sdf_open and line is the current line number.

6.2.6 glp_sdf_read_int—read integer number

Synopsis

int glp_sdf_read_int(glp_data *data);

Description

The routine glp_sdf_read_int skips optional white-space characters and then reads an integer number from the plain data file specified by the parameter data. If the operation failed, the routine glp_sdf_read_int calls the routine glp_sdf_error (see above).

Returns

The routine glp_sdf_read_int returns the integer number read.
6.2.7  glp_sdf_read_num—read floating-point number

Synopsis

double glp_sdf_read_num(glp_data *data);

Description

The routine glp_sdf_read_num skips optional white-space characters and then reads a floating-point number from the plain data file specified by the parameter data. If the operation failed, the routine glp_sdf_num calls the routine glp_sdf_error (see above).

Returns

The routine glp_sdf_read_num returns the floating-point number read.

6.2.8  glp_sdf_read_item—read data item

Synopsis

cconst char *glp_sdf_read_item(glp_data *data);

Description

The routine glp_sdf_read_item skips optional white-space characters and then reads a data item from the plain data file specified by the parameter data. If the operation failed, the routine glp_sdf_read_item calls the routine glp_sdf_error (see above).

Data item is a sequence of 1 to 255 arbitrary graphic characters delimited by white-space characters. Data items may be used to represent symbolic names, identifiers, etc.

Returns

The routine glp_sdf_read_item returns a pointer to the internal buffer, which contains the data item read in the form of a null-terminated character string.
6.2.9  glp_sdf_read_text—read text until end of line

Synopsis

const char *glp_sdf_read_text(glp_data *data);

Description

The routine glp_sdf_read_text reads a text from the plain data file specified by the parameter data.

Reading starts from the current position and extends until end of the current line. Initial and trailing white-space characters as well as the newline character are not included in the text.

Returns

The routine glp_sdf_read_text returns a pointer to the internal buffer, which contains the text read in the form of a null-terminated character string.

6.2.10  glp_sdf_line—determine current line number

Synopsis

int glp_sdf_line(glp_data *data);

Returns

The routine glp_sdf_line returns the current line number for the plain data file specified by the parameter data.

6.2.11  glp_sdf_close_file—close plain data file

Synopsis

void glp_sdf_close_file(glp_data *data);

Description

The routine glp_sdf_close_file closes the plain data file specified by the parameter data and frees all the resources allocated to this program object.
Appendix A

Installing GLPK on Your Computer

A.1 Downloading the distribution tarball

The distribution tarball of the most recent version of the GLPK package can be found on http://ftp.gnu.org/gnu/glpk/ [via http] and ftp://ftp.gnu.org/gnu/glpk/ [via FTP]. It can also be found on one of the FTP mirrors; see http://www.gnu.org/prep/ftp.html. Please use a mirror if possible.

To make sure that the GLPK distribution tarball you have downloaded is intact you need to download the corresponding `.sig` file and run a command like this:

```
gpg --verify glpk-4.38.tar.gz.sig
```

If that command fails because you do not have the required public key, run the following command to import it:

```
gpg --keyserver keys.gnupg.net --recv-keys 5981E818
```

and then re-run the previous command.

A.2 Unpacking the distribution tarball

The GLPK package (like all other GNU software) is distributed in the form of packed archive. This is one file named `glpk-X.Y.tar.gz`, where `X` is the major version number and `Y` is the minor version number.
In order to prepare the distribution for installation you should:

1. Copy the GLPK distribution file to a working subdirectory.
2. Unpack the distribution file with the following command:
   
   ```bash
   gzip -d glpk-X.Y.tar.gz
   ```

   that renames the distribution file to `glpk-X.Y.tar`.

3. Unarchive the distribution file with the following command:
   
   ```bash
   tar -x < glpk-X.Y.tar
   ```

   that automatically creates the subdirectory `glpk-X.Y` containing the GLPK distribution.

### A.3 Configuring the package

After unpacking and unarchiving the GLPK distribution you should configure the package, i.e. automatically tune it for your platform.

Normally, you should just `cd` to the subdirectory `glpk-X.Y` and run the configure script, e.g.

```bash
./configure
```

The `configure` shell script attempts to guess correct values for various system-dependent variables used during compilation. It uses those values to create a `Makefile` in each directory of the package. It also creates file `config.h` containing platform-dependent definitions. Finally, it creates a shell script `config.status` that you can run in the future to recreate the current configuration, a file `config.cache` that saves the results of its tests to speed up reconfiguring, and a file `config.log` containing compiler output (useful mainly for debugging `configure`).

Running `configure` takes about a few minutes. While it is running, it displays some informational messages that tell you what it is doing. If you don’t want to see these messages, run `configure` with its standard output redirected to `dev/null`; for example, `./configure > /dev/null`.

By default both static and shared versions of the GLPK library will be compiled. Compilation of the shared library can be turned off by specifying the `--disable-shared` option to `configure`, e.g.

```bash
./configure --disable-shared
```

If you encounter problems building the library try using the above option, because some platforms do not support shared libraries.
The GLPK package has some optional features listed below. By default all these features are disabled. To enable a feature the corresponding option should be passed to the configure script.

```
--with-gmp    Enable using the GNU MP bignum library

This feature allows the exact simplex solver to use the GNU MP bignum library. If it is disabled, the exact simplex solver uses the GLPK bignum module, which provides the same functionality as GNU MP, however, it is much less efficient.

For details about the GNU MP bignum library see its web page at http://gmplib.org/.
```

```
--with-zlib    Enable using the zlib data compression library

This feature allows GLPK API routines and the stand-alone solver to read and write compressed data files performing compression and decompression “on the fly” (compressed data files are recognized by suffix ‘.gz’ in the file name). It may be useful in case of large MPS files to save the disk space.

For details about the zlib compression library see its web page at http://www.zlib.net/.
```

```
--enable-dl    The same as ‘--enable-dl=ltdl’
--enable-dl=ltdl Enable shared library support (GNU)
--enable-dl=dlfcn Enable shared library support (POSIX)

Currently this feature is only needed to provide dynamic linking to ODBC and MySQL shared libraries (see below).

For details about the GNU shared library support see the manual at http://www.gnu.org/software/libtool/manual/.
```

```
--enable-odbc   Enable using ODBC table driver (libiodbc)
--enable-odbc=unix Enable using ODBC table driver (libodbc)

This feature allows transmitting data between MathProg model objects and relational databases accessed through ODBC.

For more details about this feature see the supplement “Using Data Tables in the GNU MathProg Modeling Language” (doc/tables.pdf).
```

```
--enable-mysql  Enable using MySQL table driver (libmysql)

This feature allows transmitting data between MathProg model objects and MySQL relational databases.
```

For more details about this feature see the supplement “Using Data Tables in the GNU MathProg Modeling Language” (doc/tables.pdf).

### A.4 Compiling the package

Normally, you can compile (build) the package by typing the command:

```
make
```

It reads ‘Makefile’ generated by ‘configure’ and performs all necessary jobs.

If you want, you can override the ‘make’ variables `CFLAGS` and `LDFLAGS` like this:

```
make CFLAGS=-O2 LDFLAGS=-s
```

To compile the package in a different directory from the one containing the source code, you must use a version of `make` that supports ‘VPATH’ variable, such as GNU `make`. ‘cd’ to the directory where you want the object files and executables to go and run the ‘configure’ script. ‘configure’ automatically checks for the source code in the directory that ‘configure’ is in and in ‘..’. If for some reason ‘configure’ is not in the source code directory that you are configuring, then it will report that it can’t find the source code. In that case, run ‘configure’ with the option ‘--srcdir=DIR’, where DIR is the directory that contains the source code.

Some systems require unusual options for compilation or linking that the ‘configure’ script does not know about. You can give ‘configure’ initial values for variables by setting them in the environment. Using a Bourne-compatible shell, you can do that on the command line like this:

```
CC=c89 CFLAGS=-O2 LIBS=-lposix ./configure
```

Or on systems that have the ‘env’ program, you can do it like this:

```
env CPPFLAGS=-I/usr/local/include LDFLAGS=-s ./configure
```

Here are the ‘make’ variables that you might want to override with environment variables when running ‘configure’.

For these variables, any value given in the environment overrides the value that ‘configure’ would choose:
CC       C compiler program. The default is ‘cc’.

INSTALL  Program used to install files. The default value is ‘install’ if
         you have it, otherwise ‘cp’.

         For these variables, any value given in the environment is added to the
         value that ‘configure’ chooses:

DEFS     Configuration options, in the form ‘-Dfoo -Dbar ...’.
LIBS     Libraries to link with, in the form ‘-lfoo -lbar ...’.

A.5 Checking the package

To check the package, i.e. to run some tests included in the package, you
can use the following command:

  make check

A.6 Installing the package

Normally, to install the GLPK package you should type the following com-
mand:

  make install

By default, ‘make install’ will install the package’s files in ‘usr/local/bin’,
‘usr/local/lib’, etc. You can specify an installation prefix other than
‘/usr/local’ by giving ‘configure’ the option ‘--prefix=PATH’. Alter-
ately, you can do so by consistently giving a value for the ‘prefix’ variable
when you run ‘make’, e.g.

  make prefix=/usr/gnu
  make prefix=/usr/gnu install

After installing you can remove the program binaries and object files
from the source directory by typing ‘make clean’. To remove all files
that ‘configure’ created (‘Makefile’, ‘config.status’, etc.), just type
‘make distclean’.

The file ‘configure.ac’ is used to create ‘configure’ by a program
called ‘autoconf’. You only need it if you want to remake ‘configure’
using a newer version of ‘autoconf’.
A.7 Uninstalling the package

To uninstall the GLPK package, i.e. to remove all the package’s files from the system places, you can use the following command:

```
make uninstall
```
Appendix B

MPS Format

B.1 Fixed MPS Format

The MPS format\(^1\) is intended for coding LP/MIP problem data. This format assumes the formulation of LP/MIP problem (1.1)—(1.3) (see Section 1.1, page 12).

*MPS file* is a text file, which contains two types of cards\(^2\): indicator cards and data cards.

Indicator cards determine a kind of succeeding data. Each indicator card has one word in uppercase letters beginning in column 1.

Data cards contain problem data. Each data card is divided into six fixed fields:

<table>
<thead>
<tr>
<th>Field 1</th>
<th>Field 2</th>
<th>Field 3</th>
<th>Field 4</th>
<th>Field 5</th>
<th>Field 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Columns</td>
<td>Code</td>
<td>Name</td>
<td>Name</td>
<td>Number</td>
<td>Number</td>
</tr>
<tr>
<td>2—3</td>
<td>5—12</td>
<td>15—22</td>
<td>25—36</td>
<td>40—47</td>
<td>50—61</td>
</tr>
</tbody>
</table>

On a particular data card some fields may be optional.

Names are used to identify rows, columns, and some vectors (see below). Aligning the indicator code in the field 1 to the left margin is optional.

All names specified in the fields 2, 3, and 5 should contain from 1 up to 8 arbitrary characters (except control characters). If a name is placed in the

\(^1\)The MPS format was developed in 1960’s by IBM as input format for their mathematical programming system MPS/360. Today the MPS format is a most widely used format understood by most mathematical programming packages. This appendix describes only the features of the MPS format, which are implemented in the GLPK package.

\(^2\)In 1960’s MPS file was a deck of 80-column punched cards, so the author decided to keep the word “card”, which may be understood as “line of text file”.

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field 3 or 5, its first character should not be the dollar sign ‘$’. If a name contains spaces, the spaces are ignored.

All numerical values in the fields 4 and 6 should be coded in the form \( sxxE_{syy} \), where \( s \) is the plus ‘+’ or the minus ‘-’ sign, \( xx \) is a real number with optional decimal point, \( yy \) is an integer decimal exponent. Any number should contain up to 12 characters. If the sign \( s \) is omitted, the plus sign is assumed. The exponent part is optional. If a number contains spaces, the spaces are ignored.

If a card has the asterisk ‘*’ in the column 1, this card is considered as a comment and ignored. Besides, if the first character in the field 3 or 5 is the dollar sign ‘$’, all characters from the dollar sign to the end of card are considered as a comment and ignored.

MPS file should contain cards in the following order:
- NAME indicator card;
- ROWS indicator card;
- data cards specifying rows (constraints);
- COLUMNS indicator card;
- data cards specifying columns (structural variables) and constraint coefficients;
- RHS indicator card;
- data cards specifying right-hand sides of constraints;
- RANGES indicator card;
- data cards specifying ranges for double-bounded constraints;
- BOUNDS indicator card;
- data cards specifying types and bounds of structural variables;
- ENDATA indicator card.

Section is a group of cards consisting of an indicator card and data cards succeeding this indicator card. For example, the ROWS section consists of the ROWS indicator card and data cards specifying rows.

The sections RHS, RANGES, and BOUNDS are optional and may be omitted.

### B.2 Free MPS Format

Free MPS format is an improved version of the standard (fixed) MPS format described above.\(^3\) Note that all changes in free MPS format concern only

---

\(^3\)This format was developed in the beginning of 1990’s by IBM as an alternative to the standard fixed MPS format for Optimization Subroutine Library (OSL).
the coding of data while the structure of data is the same for both fixed and free versions of the MPS format.

In free MPS format indicator and data records may have arbitrary length not limited to 80 characters. Fields of data records have no pre-defined positions, i.e. the fields may begin in any position, except position 1, which must be blank, and must be separated from each other by one or more blanks. However, the fields must appear in the same order as in fixed MPS format.

Symbolic names in fields 2, 3, and 5 may be longer than 8 characters and must not contain embedded blanks.

Numeric values in fields 4 and 6 are limited to 12 characters and must not contain embedded blanks.

Only six fields on each data record are used. Any other fields are ignored.

If the first character of any field (not necessarily fields 3 and 5) is the dollar sign ($), all characters from the dollar sign to the end of record are considered as a comment and ignored.

B.3 NAME indicator card

The NAME indicator card should be the first card in the MPS file (except optional comment cards, which may precede the NAME card). This card should contain the word NAME in the columns 1—4 and the problem name in the field 3. The problem name is optional and may be omitted.

B.4 ROWS section

The ROWS section should start with the indicator card, which contains the word ROWS in the columns 1—4.

Each data card in the ROWS section specifies one row (constraint) of the problem. All these data cards have the following format.

‘N’ in the field 1 means that the row is free (unbounded):

\[-\infty < x_i = a_{i1}x_{m+1} + a_{i2}x_{m+2} + \ldots + a_{in}x_{m+n} < +\infty;\]

‘L’ in the field 1 means that the row is of “less than or equal to” type:

\[-\infty < x_i = a_{i1}x_{m+1} + a_{i2}x_{m+2} + \ldots + a_{in}x_{m+n} \leq b_i;\]

---

4 Record in free MPS format has the same meaning as card in fixed MPS format.
5 GLPK allows symbolic names having up to 255 characters.
‘G’ in the field 1 means that the row is of “greater than or equal to” type: 

\[ b_i \leq x_i = a_{i1}x_{m+1} + a_{i2}x_{m+2} + \ldots + a_{in}x_{m+n} < +\infty; \]

‘E’ in the field 1 means that the row is of “equal to” type: 

\[ x_i = a_{i1}x_{m+1} + a_{i2}x_{m+2} + \ldots + a_{in}x_{m+n} \leq b_i; \]

where \( b_i \) is a right-hand side. Note that each constraint has a corresponding implicitly defined auxiliary variable \( (x_i \text{ above}) \), whose value is a value of the corresponding linear form, therefore row bounds can be considered as bounds of such auxiliary variable.

The field 2 specifies a row name (which is considered as the name of the corresponding auxiliary variable).

The fields 3, 4, 5, and 6 are not used and should be empty.

Numerical values of all non-zero right-hand sides \( b_i \) should be specified in the RHS section (see below). All double-bounded (ranged) constraints should be specified in the RANGES section (see below).

### B.5 COLUMNS section

The COLUMNS section should start with the indicator card, which contains the word COLUMNS in the columns 1—7.

Each data card in the COLUMNS section specifies one or two constraint coefficients \( a_{ij} \) and also introduces names of columns, i.e. names of structural variables. All these data cards have the following format.

The field 1 is not used and should be empty.

The field 2 specifies a column name. If this field is empty, the column name from the immediately preceding data card is assumed.

The field 3 specifies a row name defined in the ROWS section.

The field 4 specifies a numerical value of the constraint coefficient \( a_{ij} \), which is placed in the corresponding row and column.

The fields 5 and 6 are optional. If they are used, they should contain a second pair “row name—constraint coefficient” for the same column.

Elements of the constraint matrix (i.e. constraint coefficients) should be enumerated in the column wise manner: all elements for the current column should be specified before elements for the next column. However, the order of rows in the COLUMNS section may differ from the order of rows in the ROWS section.

Constraint coefficients not specified in the COLUMNS section are considered as zeros. Therefore zero coefficients may be omitted, although it is allowed to explicitly specify them.
B.6 RHS section

The RHS section should start with the indicator card, which contains the word RHS in the columns 1—3.

Each data card in the RHS section specifies one or two right-hand sides $b_i$ (see Section B.4, page 193). All these data cards have the following format.

The field 1 is not used and should be empty.

The field 2 specifies a name of the right-hand side (RHS) vector. If this field is empty, the RHS vector name from the immediately preceding data card is assumed.

The field 3 specifies a row name defined in the ROWS section.

The field 4 specifies a right-hand side $b_i$ for the row, whose name is specified in the field 3. Depending on the row type $b_i$ is a lower bound (for the row of $G$ type), an upper bound (for the row of $L$ type), or a fixed value (for the row of $E$ type).

The fields 5 and 6 are optional. If they are used, they should contain a second pair “row name—right-hand side” for the same RHS vector.

All right-hand sides for the current RHS vector should be specified before right-hand sides for the next RHS vector. However, the order of rows in the RHS section may differ from the order of rows in the ROWS section.

Right-hand sides not specified in the RHS section are considered as zeros. Therefore zero right-hand sides may be omitted, although it is allowed to explicitly specify them.

B.7 RANGES section

The RANGES section should start with the indicator card, which contains the word RANGES in the columns 1—6.

Each data card in the RANGES section specifies one or two ranges for double-side constraints, i.e. for constraints that are of the types $L$ and $G$ at the same time:

$$l_i \leq x_i = a_{i1}x_{m+1} + a_{i2}x_{m+2} + \ldots + a_{im}x_{m+n} \leq u_i,$$

where $l_i$ is a lower bound, $u_i$ is an upper bound. All these data cards have the following format.

---

6 This feature allows the user to specify several RHS vectors in the same MPS file. However, before solving the problem a particular RHS vector should be chosen.

7 If the row is of $N$ type, $b_i$ is considered as a constant term of the corresponding linear form. Should note, however, this convention is non-standard.
The field 1 is not used and should be empty.

The field 2 specifies a name of the range vector\(^8\). If this field is empty, the range vector name from the immediately preceding data card is assumed.

The field 3 specifies a row name defined in the ROWS section.

The field 4 specifies a range value \( r_i \) (see the table below) for the row, whose name is specified in the field 3.

The fields 5 and 6 are optional. If they are used, they should contain a second pair “row name—range value” for the same range vector.

All range values for the current range vector should be specified before range values for the next range vector. However, the order of rows in the RANGES section may differ from the order of rows in the ROWS section.

For each double-side constraint specified in the RANGES section its lower and upper bounds are determined as follows:

<table>
<thead>
<tr>
<th>Row type</th>
<th>Sign of ( r_i )</th>
<th>Lower bound</th>
<th>Upper bound</th>
</tr>
</thead>
<tbody>
<tr>
<td>G</td>
<td>+ or −</td>
<td>( b_i )</td>
<td>( b_i +</td>
</tr>
<tr>
<td>L</td>
<td>+ or −</td>
<td>( b_i -</td>
<td>r_i</td>
</tr>
<tr>
<td>E</td>
<td>+</td>
<td>( b_i )</td>
<td>( b_i +</td>
</tr>
<tr>
<td>E</td>
<td>−</td>
<td>( b_i -</td>
<td>r_i</td>
</tr>
</tbody>
</table>

where \( b_i \) is a right-hand side specified in the RHS section (if \( b_i \) is not specified, it is considered as zero), \( r_i \) is a range value specified in the RANGES section.

### B.8 BOUNDS section

The BOUNDS section should start with the indicator card, which contains the word **BOUNDS** in the columns 1—6.

Each data card in the BOUNDS section specifies one (lower or upper) bound for one structural variable (column). All these data cards have the following format.

The indicator in the field 1 specifies the bound type:
- **LO** lower bound;
- **UP** upper bound;
- **FX** fixed variable (lower and upper bounds are equal);
- **FR** free variable (no bounds);
- **MI** no lower bound (lower bound is “minus infinity”);
- **PL** no upper bound (upper bound is “plus infinity”);

---

\(^8\)This feature allows the user to specify several range vectors in the same MPS file. However, before solving the problem a particular range vector should be chosen.
The field 2 specifies a name of the bound vector\(^9\). If this field is empty, the bound vector name from the immediately preceding data card is assumed.

The field 3 specifies a column name defined in the COLUMNS section.

The field 4 specifies a bound value. If the bound type in the field 1 differs from LO, UP, and FX, the value in the field 4 is ignored and may be omitted.

The fields 5 and 6 are not used and should be empty.

All bound values for the current bound vector should be specified before bound values for the next bound vector. However, the order of columns in the BOUNDS section may differ from the order of columns in the COLUMNS section. Specification of a lower bound should precede specification of an upper bound for the same column (if both the lower and upper bounds are explicitly specified).

By default, all columns (structural variables) are non-negative, i.e. have zero lower bound and no upper bound. Lower \(l_j\) and upper \(u_j\) bounds of some column (structural variable \(x_j\)) are set in the following way, where \(s_j\) is a corresponding bound value explicitly specified in the BOUNDS section:

- **LO** sets \(l_j\) to \(s_j\);
- **UP** sets \(u_j\) to \(s_j\);
- **FX** sets both \(l_j\) and \(u_j\) to \(s_j\);
- **FR** sets \(l_j\) to \(-\infty\) and \(u_j\) to \(+\infty\);
- **MI** sets \(l_j\) to \(-\infty\);
- **PL** sets \(u_j\) to \(+\infty\).

### B.9 ENDATA indicator card

The ENDATA indicator card should be the last card of MPS file (except optional comment cards, which may follow the ENDATA card). This card should contain the word ENDATA in the columns 1—6.

### B.10 Specifying objective function

It is impossible to explicitly specify the objective function and optimization direction in the MPS file. However, the following implicit rule is used by default: the first row of \(N\) type is considered as a row of the objective function (i.e. the objective function is the corresponding auxiliary variable), which should be *minimized.*

\(^9\)This feature allows the user to specify several bound vectors in the same MPS file. However, before solving the problem a particular bound vector should be chosen.
GLPK also allows specifying a constant term of the objective function as a right-hand side of the corresponding row in the RHS section.

B.11 Example of MPS file

In order to illustrate what the MPS format is, consider the following example of LP problem:

minimize

\[ value = .03 \text{bin}_1 + .08 \text{bin}_2 + .17 \text{bin}_3 + .12 \text{bin}_4 + .15 \text{bin}_5 + .21 \text{al} + .38 \text{si} \]

subject to linear constraints

\[ \text{yield} = \text{bin}_1 + \text{bin}_2 + \text{bin}_3 + \text{bin}_4 + \text{bin}_5 + \text{al} + \text{si} \]
\[ FE = .15 \text{bin}_1 + .04 \text{bin}_2 + .02 \text{bin}_3 + .04 \text{bin}_4 + .02 \text{bin}_5 + .01 \text{al} + .03 \text{si} \]
\[ CU = .03 \text{bin}_1 + .05 \text{bin}_2 + .08 \text{bin}_3 + .02 \text{bin}_4 + .06 \text{bin}_5 + .01 \text{al} \]
\[ MN = .02 \text{bin}_1 + .04 \text{bin}_2 + .01 \text{bin}_3 + .02 \text{bin}_4 + .02 \text{bin}_5 \]
\[ MG = .02 \text{bin}_1 + .03 \text{bin}_2 + .01 \text{bin}_5 \]
\[ AL = .70 \text{bin}_1 + .75 \text{bin}_2 + .80 \text{bin}_3 + .75 \text{bin}_4 + .80 \text{bin}_5 + .97 \text{al} \]
\[ SI = .02 \text{bin}_1 + .06 \text{bin}_2 + .08 \text{bin}_3 + .12 \text{bin}_4 + .02 \text{bin}_5 + .01 \text{al} + .97 \text{si} \]

and bounds of (auxiliary and structural) variables

\[ \text{yield} = 2000 \quad 0 \leq \text{bin}_1 \leq 200 \]
\[-\infty < FE \leq 60 \quad 0 \leq \text{bin}_2 \leq 2500 \]
\[-\infty < CU \leq 100 \quad 400 \leq \text{bin}_3 \leq 800 \]
\[-\infty < MN \leq 40 \quad 100 \leq \text{bin}_4 \leq 700 \]
\[-\infty < MG \leq 30 \quad 0 \leq \text{bin}_5 \leq 1500 \]
\[ 1500 \leq AL < +\infty \quad 0 \leq \text{al} < +\infty \]
\[ 250 \leq SI < 300 \quad 0 \leq \text{si} < +\infty \]

A complete MPS file which specifies data for this example is shown below (the first two comment lines show card positions).

*00000000111111111122222222223333333333444444444455555555566
*234567890123456789012345678901234567890123456789012345678901
NAME PLAN
ROWS
N VALUE
E YIELD
L FE

198
<table>
<thead>
<tr>
<th>COLUMNS</th>
<th>BIN1</th>
<th>VALUE</th>
<th>.03000</th>
<th>YIELD</th>
<th>1.00000</th>
</tr>
</thead>
<tbody>
<tr>
<td>FE</td>
<td>.15000</td>
<td>CU</td>
<td>.03000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MN</td>
<td>.02000</td>
<td>MG</td>
<td>.02000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>AL</td>
<td>.70000</td>
<td>SI</td>
<td>.02000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>BIN2</td>
<td>VALUE</td>
<td>.08000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.04000</td>
<td>CU</td>
<td>.05000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MN</td>
<td>.04000</td>
<td>MG</td>
<td>.03000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>AL</td>
<td>.75000</td>
<td>SI</td>
<td>.06000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>BIN3</td>
<td>VALUE</td>
<td>.17000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.02000</td>
<td>CU</td>
<td>.08000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MN</td>
<td>.01000</td>
<td>AL</td>
<td>.80000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SI</td>
<td>.08000</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>BIN4</td>
<td>VALUE</td>
<td>.12000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.04000</td>
<td>CU</td>
<td>.02000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MN</td>
<td>.02000</td>
<td>AL</td>
<td>.75000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SI</td>
<td>.12000</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>BIN5</td>
<td>VALUE</td>
<td>.15000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.02000</td>
<td>CU</td>
<td>.06000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MN</td>
<td>.02000</td>
<td>MG</td>
<td>.01000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>AL</td>
<td>.80000</td>
<td>SI</td>
<td>.02000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>ALUM</td>
<td>VALUE</td>
<td>.21000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.01000</td>
<td>CU</td>
<td>.01000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>AL</td>
<td>.97000</td>
<td>SI</td>
<td>.01000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SILICON</td>
<td>VALUE</td>
<td>.38000</td>
<td>YIELD</td>
<td>1.00000</td>
<td></td>
</tr>
<tr>
<td>FE</td>
<td>.03000</td>
<td>SI</td>
<td>.97000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>RHS</td>
<td>RHS1</td>
<td>YIELD</td>
<td>2000.00000</td>
<td>FE</td>
<td>60.00000</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>100.00000</td>
<td>MN</td>
<td>40.00000</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>300.00000</td>
<td>SI</td>
<td>30.00000</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>30.00000</td>
<td>AL</td>
<td>1500.00000</td>
</tr>
<tr>
<td>RANGES</td>
<td>RNG1</td>
<td>SI</td>
<td>50.00000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>BOUNDS</td>
<td>UP</td>
<td>BND1</td>
<td>BIN1</td>
<td>200.00000</td>
<td></td>
</tr>
</tbody>
</table>
### B.12 MIP features

The MPS format provides two ways for introducing integer variables into the problem.

The first way is most general and based on using special marker cards INTORG and INTEND. These marker cards are placed in the COLUMNS section. The INTORG card indicates the start of a group of integer variables (columns), and the card INTEND indicates the end of the group. The MPS file may contain arbitrary number of the marker cards.

The marker cards have the same format as the data cards (see Section B.1, page 191).

- The fields 1, 2, and 6 are not used and should be empty.
- The field 2 should contain a marker name. This name may be arbitrary.
- The field 3 should contain the word 'MARKER' (including apostrophes).
- The field 5 should contain either the word 'INTORG' (including apostrophes) for the marker card, which begins a group of integer columns, or the word 'INTEND' (including apostrophes) for the marker card, which ends the group.

The second way is less general but more convenient in some cases. It allows the user declaring integer columns using three additional types of bounds, which are specified in the field 1 of data cards in the BOUNDS section (see Section B.8, page 196):

- **LI** lower integer. This bound type specifies that the corresponding column (structural variable), whose name is specified in field 3, is of integer kind. In this case an lower bound of the column should be specified in field 4 (like in the case of LO bound type).
- **UI** upper integer. This bound type specifies that the corresponding column (structural variable), whose name is specified in field 3, is of integer kind. In this case an upper bound of the column should be specified in field 4 (like in the case of UP bound type).
BV  binary variable. This bound type specifies that the corresponding
column (structural variable), whose name is specified in the field
3, is of integer kind, its lower bound is zero, and its upper bound
is one (thus, such variable being of integer kind can have only two
values zero and one). In this case a numeric value specified in the
field 4 is ignored and may be omitted.

Consider the following example of MIP problem:

minimize

\[ Z = 3x_1 + 7x_2 - x_3 + x_4 \]

subject to linear constraints

\[ r_1 = 2x_1 - x_2 + x_3 - x_4 \]
\[ r_2 = x_1 - x_2 - 6x_3 + 4x_4 \]
\[ r_3 = 5x_1 + 3x_2 + x_4 \]

and bound of variables

\[ 1 \leq r_1 < +\infty \quad 0 \leq x_1 \leq 4 \quad \text{(continuous)} \]
\[ 8 \leq r_2 < +\infty \quad 2 \leq x_2 \leq 5 \quad \text{(integer)} \]
\[ 5 \leq r_3 < +\infty \quad 0 \leq x_3 \leq 1 \quad \text{(integer)} \]
\[ 3 \leq x_4 \leq 8 \quad \text{(continuous)} \]

The corresponding MPS file may look like the following:

```
NAME SAMP1
ROWS
N  Z
G  R1
G  R2
G  R3
COLUMNS
X1  R1  2.0  R2  1.0
X1  R3  5.0  Z  3.0
MARK0001 ’MARKER’ ’INTORG’
X2  R1  -1.0  R2  -1.0
X2  R3  3.0  Z  7.0
X3  R1  1.0  R2  -6.0
X3  Z  -1.0
MARK0002 ’MARKER’ ’INTEND’
X4  R1  -1.0  R2  4.0
```
The same example may be coded without INTORG/INTEND markers using the bound type UI for the variable \(x_2\) and the bound type BV for the variable \(x_3\):

```
NAME SAMP2
ROWS
 N Z
 G R1
 G R2
 G R3
COLUMNS
 X1 R1 2.0 R2 1.0
 X1 R3 5.0 Z 3.0
 X2 R1 -1.0 R2 -1.0
 X2 R3 3.0 Z 7.0
 X3 R1 1.0 R2 -6.0
 X3 Z -1.0
 X4 R1 -1.0 R2 4.0
 X4 R3 1.0 Z 1.0
RHS
 RHS1 R1 1.0
 RHS1 R2 8.0
 RHS1 R3 5.0
BOUNDS
 UP BND1 X4 4.0
 LO BND1 X2 2.0
 UP BND1 X2 5.0
 UP BND1 X3 1.0
 LO BND1 X4 3.0
 UP BND1 X4 8.0
ENDATA
```
<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>UI BND1</td>
<td>X2</td>
<td>5.0</td>
</tr>
<tr>
<td>BV BND1</td>
<td>X3</td>
<td></td>
</tr>
<tr>
<td>LO BND1</td>
<td>X4</td>
<td>3.0</td>
</tr>
<tr>
<td>UP BND1</td>
<td>X4</td>
<td>8.0</td>
</tr>
<tr>
<td>ENDATA</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Appendix C

CPLEX LP Format

C.1 Prelude

The CPLEX LP format\(^1\) is intended for coding LP/MIP problem data. It is a row-oriented format that assumes the formulation of LP/MIP problem (1.1)—(1.3) (see Section 1.1, page 12).

CPLEX LP file is a plain text file written in CPLEX LP format. Each text line of this file may contain up to 255 characters\(^2\). Blank lines are ignored. If a line contains the backslash character (\), this character and everything that follows it until the end of line are considered as a comment and also ignored.

An LP file is coded by the user using the following elements:

- keywords;
- symbolic names;
- numeric constants;
- delimiters;
- blanks.

\(^1\)The CPLEX LP format was developed in the end of 1980’s by CPLEX Optimization, Inc. as an input format for the CPLEX linear programming system. Although the CPLEX LP format is not as widely used as the MPS format, being row-oriented it is more convenient for coding mathematical programming models by human. This appendix describes only the features of the CPLEX LP format which are implemented in the GLPK package.

\(^2\)GLPK allows text lines of arbitrary length.
Keywords which may be used in the LP file are the following:

- minimize
- maximize
- subject to
- bounds
- general
- integer
- binary
- infinity
- free
end

All the keywords are case insensitive. Keywords given above on the same line are equivalent. Any keyword (except infinity, inf, and free) being used in the LP file must start at the beginning of a text line.

Symbolic names are used to identify the objective function, constraints (rows), and variables (columns). All symbolic names are case sensitive and may contain up to 16 alphanumeric characters (a, ..., z, A, ..., Z, 0, ..., 9) as well as the following characters:

! " # $ % & ( ) / , . ; ? @ \ ' ' { } | ~

with exception that no symbolic name can begin with a digit or a period.

Numeric constants are used to denote constraint and objective coefficients, right-hand sides of constraints, and bounds of variables. They are coded in the standard form $xxEsyy$, where $xx$ is a real number with optional decimal point, $s$ is a sign (+ or -), $yy$ is an integer decimal exponent. Numeric constants may contain arbitrary number of characters. The exponent part is optional. The letter ‘E’ can be coded as ‘e’. If the sign $s$ is omitted, plus is assumed.

Delimiters that may be used in the LP file are the following:

: + - < <= =< > >= => =

GLPK allows symbolic names having up to 255 characters.
Delimiters given above on the same line are equivalent. The meaning of the delimiters will be explained below.

Blanks are non-significant characters. They may be used freely to improve readability of the LP file. Besides, blanks should be used to separate elements from each other if there is no other way to do that (for example, to separate a keyword from a following symbolic name).

The order of an LP file is:
- objective function definition;
- constraints section;
- bounds section;
- general, integer, and binary sections (can appear in arbitrary order);
- end keyword.

These components are discussed in following sections.

C.2 Objective function definition

The objective function definition must appear first in the LP file. It defines the objective function and specifies the optimization direction.

The objective function definition has the following form:

$$\begin{cases}
\text{minimize} \\
\text{maximize}
\end{cases} f : s \ c \ x \ s \ c \ x \ldots \ s \ c \ x$$

where $f$ is a symbolic name of the objective function, $s$ is a sign $+$ or $-$, $c$ is a numeric constant that denotes an objective coefficient, $x$ is a symbolic name of a variable.

If necessary, the objective function definition can be continued on as many text lines as desired.

The name of the objective function is optional and may be omitted (together with the semicolon that follows it). In this case the default name ‘obj’ is assigned to the objective function.

If the very first sign $s$ is omitted, the sign plus is assumed. Other signs cannot be omitted.

If some objective coefficient $c$ is omitted, 1 is assumed.

Symbolic names $x$ used to denote variables are recognized by context and therefore needn’t to be declared somewhere else.

Here is an example of the objective function definition:

Minimize $Z : - x_1 + 2 \ x_2 - 3.5 \ x_3 + 4.997e3x(4) + x_5 + x_6 + x_7 - .01x_8$
C.3 Constraints section

The constraints section must follow the objective function definition. It defines a system of equality and/or inequality constraints.

The constraint section has the following form:

\[
\text{subject to} \\
\text{constraint}_1 \\
\text{constraint}_2 \\
\vdots \\
\text{constraint}_m
\]

where \( \text{constraint}_i \), \( i = 1, \ldots, m \), is a particular constraint definition.

Each constraint definition can be continued on as many text lines as desired. However, each constraint definition must begin on a new line except the very first constraint definition which can begin on the same line as the keyword ‘subject to’.

Constraint definitions have the following form:

\[
\begin{array}{c}
r : s c x s c x \ldots s c x \\ \{ \leq \} \\ \{ \geq \} \\ \{ = \} \\
\end{array} \quad b
\]

where \( r \) is a symbolic name of a constraint, \( s \) is a sign + or −, \( c \) is a numeric constant that denotes a constraint coefficient, \( x \) is a symbolic name of a variable, \( b \) is a right-hand side.

The name \( r \) of a constraint (which is the name of the corresponding auxiliary variable) is optional and may be omitted (together with the semicolon that follows it). In this case the default names like ‘\( r.nnn \)’ are assigned to unnamed constraints.

The linear form \( s c x s c x \ldots s c x \) in the left-hand side of a constraint definition has exactly the same meaning as in the case of the objective function definition (see above).

After the linear form one of the following delimiters that indicate the constraint sense must be specified:

- \( \leq \) means ‘less than or equal to’
- \( \geq \) means ‘greater than or equal to’
- \( = \) means ‘equal to’

The right hand side \( b \) is a numeric constant with an optional sign.

Here is an example of the constraints section:
Subject To
  one: \( y_1 + 3 \ a_1 - a_2 - b \geq 1.5 \)
  \( y_2 + 2 \ a_3 + 2 \ a_4 - b \geq -1.5 \)
  two : \( y_4 + 3 \ a_1 + 4 \ a_5 - b \leq +1 \)
  \( .20 y_5 + 5 \ a_2 - b = 0 \)
  \( 1.7 \ y_6 - a_6 + 5 \ a_7 - b \geq 1 \)

(Should note that it is impossible to express ranged constraints in the CPLEX LP format. Each a ranged constraint can be coded as two constraints with identical linear forms in the left-hand side, one of which specifies a lower bound and other does an upper one of the original ranged constraint.)

C.4 Bounds section

The bounds section is intended to define bounds of variables. This section is optional; if it is specified, it must follow the constraints section. If the bound section is omitted, all variables are assumed to be non-negative (i.e.
that they have zero lower bound and no upper bound).

The bounds section has the following form:

\[
\text{bounds} \\
definition_1 \\
definition_2 \\
\ldots \\
definition_p
\]

where \( \text{definition}_k, k = 1, \ldots, p \), is a particular bound definition.

Each bound definition must begin on a new line\(^4\) except the very first bound definition which can begin on the same line as the keyword ‘bounds’.

Syntactically constraint definitions can have one of the following six forms:

\[
x \geq l \quad \text{specifies a lower bound} \\
l \leq x \quad \text{specifies a lower bound} \\
x \leq u \quad \text{specifies an upper bound} \\
l \leq x \leq u \quad \text{specifies both lower and upper bounds} \\
x = t \quad \text{specifies a fixed value} \\
x \text{free} \quad \text{specifies free variable}
\]

\(^4\)The GLPK implementation allows several bound definitions to be placed on the same line.
where \( x \) is a symbolic name of a variable, \( l \) is a numeric constant with an optional sign that defines a lower bound of the variable or \(-\text{inf}\) that means that the variable has no lower bound, \( u \) is a numeric constant with an optional sign that defines an upper bound of the variable or \(+\text{inf}\) that means that the variable has no upper bound, \( t \) is a numeric constant with an optional sign that defines a fixed value of the variable.

By default all variables are non-negative, i.e. have zero lower bound and no upper bound. Therefore definitions of these default bounds can be omitted in the bounds section.

Here is an example of the bounds section:

```
Bounds
  -inf <= a1 <= 100
  -100 <= a2
  b <= 100
  x2 = +123.456
  x3 free
```

### C.5 General, integer, and binary sections

The general, integer, and binary sections are intended to define some variables as integer or binary. All these sections are optional and needed only in case of MIP problems. If they are specified, they must follow the bounds section or, if the latter is omitted, the constraints section.

All the general, integer, and binary sections have the same form as follows:

```
{general}
{integer}
{binary}

x_1
x_2
...  
x_q
```

where \( x_k \) is a symbolic name of variable, \( k = 1, \ldots, q \).

Each symbolic name must begin on a new line\(^5\) except the very first symbolic name which can begin on the same line as the keyword ‘general’, ‘integer’, or ‘binary’.

\(^5\)The GLPK implementation allows several symbolic names to be placed on the same line.
If a variable appears in the general or the integer section, it is assumed to be general integer variable. If a variable appears in the binary section, it is assumed to be binary variable, i.e. an integer variable whose lower bound is zero and upper bound is one. (Note that if bounds of a variable are specified in the bounds section and then the variable appears in the binary section, its previously specified bounds are ignored.)

Here is an example of the integer section:

```
Integer
  z12
  z22
  z35
```

C.6 End keyword

The keyword ‘end’ is intended to end the LP file. It must begin on a separate line and no other elements (except comments and blank lines) must follow it. Although this keyword is optional, it is strongly recommended to include it in the LP file.

C.7 Example of CPLEX LP file

Here is a complete example of CPLEX LP file that corresponds to the example given in Section B.11, page 198.

```
\* plan.lp *

Minimize
  value: .03 bin1 + .08 bin2 + .17 bin3 + .12 bin4 + .15 bin5 +
         .21 alum + .38 silicon

Subject To
  yield:   bin1 +   bin2 +   bin3 +   bin4 +   bin5 +
          alum +   silicon          = 2000
  fe:      .15 bin1 + .04 bin2 + .02 bin3 + .04 bin4 + .02 bin5 +
          .01 alum + .03 silicon          <=  60
  cu:      .03 bin1 + .05 bin2 + .08 bin3 + .02 bin4 + .06 bin5 +
          .01 alum          <=  100
  mn:      .02 bin1 + .04 bin2 + .01 bin3 + .02 bin4 + .02 bin5    <=  40
```

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mg: .02 bin1 + .03 bin2 + .01 bin5 <= 30
al: .70 bin1 + .75 bin2 + .80 bin3 + .75 bin4 + .80 bin5 + .97 alum >= 1500
si1: .02 bin1 + .06 bin2 + .08 bin3 + .12 bin4 + .02 bin5 + .01 alum + .97 silicon >= 250
si2: .02 bin1 + .06 bin2 + .08 bin3 + .12 bin4 + .02 bin5 + .01 alum + .97 silicon <= 300

Bounds
bin1 <= 200
bin2 <= 2500
400 <= bin3 <= 800
100 <= bin4 <= 700
bin5 <= 1500

End

/* eof */
Appendix D

Stand-alone LP/MIP Solver

The GLPK package includes the program `glpsol`, which is a stand-alone LP/MIP solver. This program can be invoked from the command line of from the shell to read LP/MIP problem data in any format supported by GLPK, solve the problem, and write the problem solution obtained to an output text file.

Usage

`glpsol [options...] [filename]`

General options

|--mps
|  read LP/MIP problem in fixed MPS format
|--freemps
|  read LP/MIP problem in free MPS format (default)
|--lp
|  read LP/MIP problem in CPLEX LP format
|--glp
|  read LP/MIP problem in GLPK format
|--math
|  read LP/MIP model written in GNU MathProg modeling language

|--m filename|--model filename
|  read model section and optional data section from filename (the same as --math)

|--d filename|--data filename
|  read data section from filename (for --math only); if model file also has data section, that section is ignored

|--y filename|--display filename
|  send display output to filename (for --math only); by default the output is sent to stdout

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--seed value initialize pseudo-random number generator used in
MathProg model with specified seed (any integer); if
the seed value is specified as ? (question mark), some
random seed will be used
--mincost read min-cost flow problem in DIMACS format
--maxflow read maximum flow problem in DIMACS format
--simplex use simplex method (default)
--interior use interior point method (for pure LP only)
-r filename, --read filename read solution from filename rather to find it with the
solver
--min minimization
--max maximization
--scale scale problem (default)
--noscale do not scale problem
-o filename, --output filename
write solution to filename in printable format
-w filename, --write filename
write solution to filename in plain text format
--ranges filename
write sensitivity analysis report to filename in printable
format (simplex only)
--tmlim nnn limit solution time to nnn seconds (--tmlim 0 allows
obtaining solution at initial point)
--memlim nnn limit available memory to nnn megabytes
--check do not solve problem, check input data only
--name probname change problem name to probname
--wmps filename write problem to filename in fixed MPS format
--wfreemps filename
write problem to filename in free MPS format
--wlp filename write problem to filename in CPLEX LP format
--wglp filename write problem to filename in GLPK format
--log filename write copy of terminal output to filename
-h, --help display this help information and exit
-v, --version display program version and exit

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LP basis factorization options

--luf  LU + Forrest–Tomlin update
       (faster, less stable; default)
--cbg  LU + Schur complement + Bartels–Golub update
       (slower, more stable)
--cgr  LU + Schur complement + Givens rotation update
       (slower, more stable)

Options specific to the simplex solver

--primal use primal simplex (default)
--dual  use dual simplex
--std   use standard initial basis of all slacks
--adv   use advanced initial basis (default)
--bib   use Bixby’s initial basis
--ini filename use as initial basis previously saved with --w
       (disables LP presolver)
--steep use steepest edge technique (default)
--nosteep use standard “textbook” pricing
--relax use Harris’ two-pass ratio test (default)
--norelax use standard “textbook” ratio test
--presol use LP presolver (default; assumes --scale and --adv)
--nopresol do not use LP presolver
--exact use simplex method based on exact arithmetic
--xcheck check final basis using exact arithmetic

Options specific to the interior-point solver

--nord  use natural (original) ordering
--qmd   use quotient minimum degree ordering
--amd   use approximate minimum degree ordering (default)
--symamd use approximate minimum degree ordering

Options specific to the MIP solver

--nomip consider all integer variables as continuous (allows solving MIP as pure LP)
--first branch on first integer variable
--last  branch on last integer variable
--mostf branch on most fractional variable
--drtom branch using heuristic by Driebeck and Tomlin (default)
--pcost branch using hybrid pseudocost heuristic (may be useful for hard instances)
--dfs backtrack using depth first search
--bfs backtrack using breadth first search
--bestp backtrack using the best projection heuristic (default)
--bestb backtrack using node with best local bound
--intopt use MIP presolver (default)
--nointopt do not use MIP presolver
--binarize replace general integer variables by binary ones (assumes --intopt)
--fpump apply feasibility pump heuristic
--gomory generate Gomory’s mixed integer cuts
--mir generate MIR (mixed integer rounding) cuts
--cover generate mixed cover cuts
--clique generate clique cuts
--cuts generate cuts of all classes above (assumes --intopt)
--mipgap tol set relative mip gap tolerance to tol

For description of the MPS format see Appendix B, page 191.

For description of the CPLEX LP format see Appendix C, page 204.

For description of the modeling language see the document “Modeling Language GNU MathProg: Language Reference” included in the GLPK distribution.

For description of the DIMACS min-cost flow problem format and DIMACS maximum flow problem format see the document “GLPK: Graph and Network Routines” included in the GLPK distribution.
Appendix E

External Software Modules Used In GLPK

In the GLPK package there are used some external software modules listed in this Appendix. Note that these modules are not part of GLPK, but are used with GLPK and included in the distribution.

E.1 AMD

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Description

AMD is a set of routines for pre-ordering sparse matrices prior to Cholesky or LU factorization, using the approximate minimum degree ordering algorithm.

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Availability

http://www.cise.ufl.edu/research/sparse/amd

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E.2 COLAMD/SYMAMD


Description

colamd:  an approximate minimum degree column ordering algorithm, for LU factorization of symmetric or unsymmetric matrices, QR factorization, least squares, interior point methods for linear programming problems, and other related problems.

symamd:  an approximate minimum degree ordering algorithm for Cholesky factorization of symmetric matrices.

Authors

The authors of the code itself are Stefan I. Larimore and Timothy A. Davis (davis at cise.ufl.edu), University of Florida. The algorithm was developed in collaboration with John Gilbert, Xerox PARC, and Esmond Ng, Oak Ridge National Laboratory.

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Availability

http://www.cise.ufl.edu/research/sparse/colamd

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Version 3, 29 June 2007


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