On the Tree Search Problem with Non-Uniform costs

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Input: *n* objects where one is special Output: The special object



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Questions: Is the special object in a subset *S*? Answers: Yes/No. Objective: Minimize number of questions.

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Questions: Is the special object in a subset *S*? Answers: Yes/No. Objective: Minimize number of questions.

Find a treasure quickly!

Online and offline versions

Online: Questions and answers alternate.



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Online: Questions and answers alternate.



Offline: All questions first, then all answers.



We consider only online version.











Input: Tree T, one hidden vertex



Output:

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Output: Decision tree D of minimum depth

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THEOREM (LAM, YUE '98)

An optimal decision tree D can be computed in a polynomial time.

SEARCHING IN TREES WITH COSTS ON EDGES Input: Tree T



Output: Optimal decision tree D



SEARCHING IN TREES WITH COSTS ON EDGES Input: Tree T, cost $c : E(T) \to \mathbb{Z}$



Output: Optimal decision tree D



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Output: Optimal decision tree D



SEARCHING IN TREES WITH COSTS ON EDGES Input: Tree T, cost $c : E(T) \rightarrow \mathbb{Z}$



Output: Optimal decision tree D (cheapest worst case)



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Output: Optimal decision tree D (cheapest worst case)



 $cost(D) := \max_{v \in V(T)} \{cost^D(v)\} \quad OPT(T, c) := \min_D \{cost(D)\}$

Input: Tree T on n vertices, cost c Output: Decision tree D with cost(D) = OPT(T, c)

THEOREM (DERENIOWSKY, '06)

NP-complete if diameter of T is 10 $O(\log(n))$ -approximation algorithm

 $cost(D) = O(log(n)) \cdot OPT(T, c)$

Input: Tree *T* on *n* vertices, cost *c* Output: Decision tree *D* with cost(D) = OPT(T, c)

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THEOREM (CICALESE, JACOBS, LABER, VALENTIN '12) NP-complete for diameter of T is 6 NP-complete for max degree of T is 3

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THEOREM (CICALESE, JACOBS, LABER, VALENTIN '12) *NP*-complete for diameter of *T* is 6 *NP*-complete for max degree of *T* is 3 $O(n^2)$ -time algorithm if *T* is a path $O(n^2)$ -time algorithm

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THEOREM (CICALESE, JACOBS, LABER, VALENTIN '12) *NP*-complete for diameter of *T* is 6 *NP*-complete for max degree of *T* is 3 $O(n^2)$ -time algorithm if *T* is a path $O(n2^n)$ -time algorithm $O\left(\frac{\log n}{\log\log\log n}\right)$ -approximation algorithm

OUR RESULTS

Input: Tree *T* on *n* vertices, cost *c* Output: Decision tree *D* with cost(D) = OPT(T, c)THEOREM (CICALESE, KESZEGH, L., PÁLVÖLGYI, VALLA) *NP-complete for diameter 6 subdivided stars (from Knapsack)* $O\left(\frac{\log n}{\log \log n}\right)$ -approximation algorithm



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PROBLEM Is there $O\left(\frac{\log n}{\log n}\right)$ -approximation algorithm?

LEMMA

There is a 2-approximation algorithm for subdivided stars.



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Find a decision tree for ${\sf S}$

Proof warm-up

LEMMA

There is a 2-approximation algorithm for subdivided stars.



Find a decision tree for S Cost : OPT(T, c)

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Find a decision tree for S and for P_i Cost : OPT(T, c)

LEMMA

There is a 2-approximation algorithm for subdivided stars.



Find a decision tree for S and for P_i and combine them. Cost: OPT(T, c) + OPT(T, c)

Idea: Find a small separator S, small resulting components, recurse



Fix $t = \log(n)$;

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Fix $t = \log(n)$; pick t centroids $S = \{x_1 \dots\}$; (components n/t)

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Fix $t = \log(n)$; pick t centroids $S = \{x_1 \dots\}$; (components n/t) build auxiliary graph Y of size 2t

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Fix $t = \log(n)$; pick t centroids $S = \{x_1 \dots\}$; (components n/t) build auxiliary graph Y of size 2t, solve Y in $O(t2^{2t})$

Cost: OPT(T, c)

Idea: Find a small separator S, small resulting components, recurse



Fix $t = \log(n)$; pick t centroids $S = \{x_1 \dots\}$; (components n/t) build auxiliary graph Y of size 2t, solve Y in $O(t2^{2t})$ solve subdivided stars in S

Cost: OPT(T, c) + 2OPT(T, c)

Idea: Find a small separator S, small resulting components, recurse



Fix $t = \log(n)$; pick t centroids $S = \{x_1 \dots\}$; (components n/t) build auxiliary graph Y of size 2t, solve Y in $O(t2^{2t})$ solve subdivided stars in S, solve neighbors of S

Cost: OPT(T, c) + 2OPT(T, c) + OPT(T, c)

Idea: Find a small separator S, small resulting components, recurse



Fix $t = \log(n)$; pick t centroids $S = \{x_1...\}$; (components n/t) build auxiliary graph Y of size 2t, solve Y in $O(t2^{2t})$ solve subdivided stars in S, solve neighbors of S, recursion $\times k$;

Cost: $(OPT(T, c) + 2OPT(T, c) + OPT(T, c)) \times k$

Idea: Find a small separator S, small resulting components, recurse



Fix $t = \log(n)$; pick t centroids $S = \{x_1...\}$; (components n/t) build auxiliary graph Y of size 2t, solve Y in $O(t2^{2t})$ solve subdivided stars in S, solve neighbors of S, recursion $\times k$; $t^k = n$ hence $k = \log(n)/\log(t)$ Cost: $(OPT(T, c) + 2OPT(T, c) + OPT(T, c)) \times k$ Thank you for your attention!