

## Exercise 2: Discrete & Continuous Optimization

1. Show that the following functions are convex or not convex on their domains using the test of Hessian being positive semidefinite:

(a)  $f(x, y) = x^3 + y^3 - 3xy$  on  $\mathbb{R}^2$

(b)  $g(x, y) = x^2 + 2xy + y^2$  on  $\mathbb{R}^2$

(c)  $h(x, y, z) = x^2 + y^2 + z^2 - 2xyz$  on  $\mathbb{R}^3$

2. Let  $M$  be a convex set. Recall that a function  $f : M \rightarrow \mathbb{R}$  is said to be convex over  $M$  if for all  $x_1, x_2 \in M$  and for all  $\lambda \in [0, 1]$  we have that  $f(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda f(x_1) + (1 - \lambda)f(x_2)$ . Prove that  $f : M \rightarrow \mathbb{R}$  is convex over  $M$  if and only if for any  $x_1, \dots, x_k \in M$  and  $\lambda_1, \dots, \lambda_k \geq 0$ ,  $\sum_{i=1}^k \lambda_i = 1$  we have that

$$f\left(\sum_{i=1}^k \lambda_i x_i\right) \leq \sum_{i=1}^k \lambda_i f(x_i).$$

3. Let  $P$  be a symmetric  $n \times n$  matrix and let  $c, d \in \mathbb{R}^n$ . Prove that  $f(x) = x^T P x + c^T x + d$  is convex if and only if  $P$  is positive semidefinite.
4. Prove that  $f(x) = -\sum_{i=1}^m \ln(b_i - a_i^T x)$ , is convex over  $\{x | a_i^T x < b_i, i = 1, \dots, m\}$ .