

**STAT 280: OPTIMIZATION**  
**SPRING 2017**  
**PROBLEM SET 3**

In the following, we write  $\mathbb{R}_+ = [0, \infty)$  and  $\mathbb{R}_{++} = (0, \infty)$ . So  $\mathbb{R}_+^n = \{(x_1, \dots, x_n) \in \mathbb{R}^n : x_i \geq 0\}$  and  $\mathbb{R}_{++}^n = \{(x_1, \dots, x_n) \in \mathbb{R}^n : x_i > 0\}$ .

1. (a) Find all stationary points of the cubic polynomial

$$f(x, y) = x^3 + y^3 - 3x - 12y + 20.$$

Indicate which are the local maximizers and local minimizers.

- (b) A rectangular box, open at the top, has a volume of 32 cubic feet. Find the dimensions of the box so that the total surface area is minimized.  
(c) Prove that for any  $x \geq 0, y \geq 0$ , we always have

$$\frac{x^2 + y^2}{4} \leq e^{x+y-2}.$$

- (d) Let  $A \in \mathbb{S}^n$ ,  $\mathbf{b} \in \mathbb{R}^n$ , and  $c \in \mathbb{R}$ . Show that  $\frac{1}{2}\mathbf{x}^\top A\mathbf{x} + \mathbf{b}^\top \mathbf{x} + c$  has a unique global minimizer iff  $A \succ 0$ . What is it?  
(e) Let  $A \in \mathbb{R}^{m \times n}$  and  $\mathbf{b} \in \mathbb{R}^m$ . Show that  $\mathbf{x}_*$  is a global minimizer of  $\|A\mathbf{x} - \mathbf{b}\|^2$  iff  $\mathbf{x}_*$  is a solution to  $A^\top A\mathbf{x} = A^\top \mathbf{b}$ .

2. Compute the Hessians of the following functions and decide if they are convex, concave, or neither on their respective domains.

- (a)  $e, f, g, h : \mathbb{R}_{++}^2 \rightarrow \mathbb{R}$  defined by

$$e(x, y) = xy, \quad f(x, y) = \frac{1}{xy}, \quad g(x, y) = \frac{x}{y}, \quad h(x, y) = x^\alpha y^{1-\alpha},$$

where  $\alpha \in [0, 1]$ .

- (b)  $\varphi : \mathbb{R} \times \mathbb{R}_{++} \rightarrow \mathbb{R}$  defined by  $\varphi(x, y) = x^2/y$ . (Hint: Write  $\nabla^2 \varphi(x, y)$  as a rank-1 matrix).  
(c) Approximate  $e, f, g, \varphi$  by a 3-term Taylor series about the point  $\mathbf{x} = [\frac{2}{3}]$ . Compare the value of each function at  $\mathbf{y} = [\frac{2}{3}, \frac{1}{2}]$  and the value of its corresponding 3-term Taylor series approximation.

3. (a) Find the Hessian of the function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  defined by  $f(\mathbf{x}) = \log(e^{x_1} + \dots + e^{x_n})$ . Show that for any  $\mathbf{v} \in \mathbb{R}^n$ ,

$$\mathbf{v}^\top \nabla^2 f(\mathbf{x}) \mathbf{v} = \frac{1}{(e^{x_1} + \dots + e^{x_n})^2} \left[ \left( \sum_{i=1}^n e^{x_i} \right) \left( \sum_{i=1}^n v_i^2 e^{x_i} \right) - \left( \sum_{i=1}^n v_i e^{x_i} \right)^2 \right].$$

Hence or otherwise, deduce that  $f$  is a convex function.

- (b) Find the Hessian of the function  $g : \mathbb{R}_{++}^n \rightarrow \mathbb{R}$  defined by  $g(\mathbf{x}) = (x_1 \cdots x_n)^{1/n}$ . Show that for any  $\mathbf{v} \in \mathbb{R}^n$ ,

$$\mathbf{v}^\top \nabla^2 g(\mathbf{x}) \mathbf{v} = -\frac{g(\mathbf{x})}{n^2} \left[ n \sum_{i=1}^n \frac{v_i^2}{x_i^2} - \left( \sum_{i=1}^n \frac{v_i}{x_i} \right)^2 \right]$$

Hence or otherwise, deduce that  $g$  is a concave function.

- (c) Find the Hessian of the function  $h : \mathbb{R}_{++}^n \rightarrow \mathbb{R}$  defined by

$$h(\mathbf{x}) = \frac{1}{1/x_1 + \cdots + 1/x_n}.$$

By emulating what we did in the previous two parts or otherwise, decide if  $h$  is convex, concave, or neither on  $\mathbb{R}_{++}^n$ .

- (d) Find the Hessian of the function  $\varphi : \mathbb{R}_{++}^n \rightarrow \mathbb{R}$  defined by  $\varphi(\mathbf{x}) = \log h(\mathbf{x})$ . Decide if  $\varphi$  is convex, concave, or neither on  $\mathbb{R}_{++}^n$ .