## ANALYSIS AND OPTIMIZATION: MIDTERM 2 PRACTICE PROBLEMS SOLUTIONS

## SPRING 2016

## PRACTICE PROBLEM SOLUTIONS

At times, I have only written the final answer or only sketched the solution. Let me know if something is unclear. I will add more explanation.

(1) Write the definition of a convex function. Let  $f(\vec{x})$  and  $g(\vec{x})$  be two convex functions on  $\mathbf{R}^n$ . Using the definition, show that the function  $h(\vec{x})$  defined by

$$h(\vec{x}) = \max(f(\vec{x}), g(\vec{x}))$$

is also convex.

Solution. A function *h* is convex if for every  $\vec{x}$  and  $\vec{y}$  in the domain and  $\lambda$  in [0, 1], we have

$$\lambda h(\vec{x}) + (1 - \lambda)h(\vec{y}) \ge h(\lambda \vec{x} + (1 - \lambda)\vec{y}).$$

Let us show that  $h = \max(f, g)$  is convex by verifying the above inequality. Since  $h(\vec{x}) \ge f(\vec{x})$  and  $h(\vec{y}) \ge f(\vec{y})$ , we have

$$\lambda h(\vec{x}) + (1 - \lambda)h(\vec{y}) \ge \lambda f(\vec{x}) + (1 - \lambda)f(\vec{y}).$$

Since f is convex, we have

$$\lambda f(\vec{x}) + (1 - \lambda)f(\vec{y}) \ge f(\lambda \vec{x} + (1 - \lambda)\vec{y}).$$

Combining the two inequalities gives

$$\lambda h(\vec{x}) + (1 - \lambda)h(\vec{y}) \ge f(\lambda \vec{x} + (1 - \lambda)\vec{y}).$$

Similarly we get

$$\lambda h(\vec{x}) + (1 - \lambda)h(\vec{y}) \ge g(\lambda \vec{x} + (1 - \lambda)\vec{y}).$$

Combining the last two inequalities gives

$$\lambda h(\vec{x}) + (1 - \lambda)h(\vec{y}) \ge \max(f(\lambda \vec{x} + (1 - \lambda)\vec{y}), g(\lambda \vec{x} + (1 - \lambda)\vec{y})),$$

which is the same as

$$\lambda h(\vec{x}) + (1-\lambda)h(\vec{y}) \ge h(\lambda \vec{x} + (1-\lambda)\vec{y}).$$

(2) Use Jensen's inequality to prove that for positive real numbers  $x_1, \ldots, x_n$ , we have

$$\sqrt[3]{\frac{x_1^3 + \dots + x_n^3}{n}} \ge \frac{x_1 + \dots + x_n}{n}.$$

Solution. Use Jensen's inequality for  $f(x) = x^3$  (which is convex for x > 0) with all  $\lambda_i = 1/n$ .

- (3) Find the global minimum and maximum of the function  $f(x, y) = 2x^3 + 4y^3$  on the set  $S = \{x^2 + y^2 \le 1\}$  by the following outline.
  - (a) Show that the maximum and the minimum exists. Solution. f is a continuous function on a compact set S, so the max/min exist by the maximum theorem.  $\square$
  - (b) Using the gradient, find the possible points where the max/min could be achieved on the interior  $\{x^2 + y^2 < 1\}$ . Solution. In the interior, the max/min can only be achieved when the gradient is equal to zero. The gradient is zero only at (x, y) = (0, 0).  $\square$
  - (c) Using Lagrange multipliers, find the possible points where the max/min could be achieved on the boundary  $\{x^2 + y^2 = 1\}$ .

Solution. The Lagrange multiplier problem is

$$6x^{2} = 2\lambda x$$
$$12y^{2} = 2\lambda y$$
$$x^{2} + y^{2} = 1.$$

The solutions are  $(x, y) = (0, \pm 1), (\pm 1, 0), (2/\sqrt{5}, 1/\sqrt{5}), (-2/\sqrt{5}, -1/\sqrt{5}).$ (d) Check all the possibilities.

 $\square$ 

Solution. The max is at (0, 1) and min at (0, -1).

(4) Let *A* be the matrix

$$A = \begin{pmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{pmatrix}$$

(a) Write down the quadratic form Q(x, y, z) associated with A. Solution.

$$Q(x, y, z) = 2x^{2} + 2y^{2} + 2z^{2} - 2xy - 2yz$$

- (b) Show that the function  $f(x, y, z) = e^{Q(x, y, z)}$  is strictly convex. Solution. First, Q is strictly convex because the leading principal minors are positive. Now  $e^{Q(x,y,z)}$  is the composition of a strictly convex function with a strictly increasing strictly convex function.
- (5) Check if the following equation defines *z* as a function z = g(x, y) in a neighborhood of (0,0,1). If it does, find  $\frac{\partial g}{\partial x}$  and  $\frac{\partial g}{\partial y}$  at (0,0,1).

$$x^3 + y^3 + z^3 - xyz - 1 = 0.$$

(6) The same question at (1,0,0) for the equation

$$e^{z} - z^{2} - x^{2} - y^{2} = 0.$$

(7) Consider the system of equations

$$1 + (x + y)u - (2 + u)^{1+\nu} = 0$$
  
2u - (1 + xy)e<sup>u(x-1)</sup> = 0.

Show that it defines u and v as functions of x and y near the point (x, y, u, v) = (1, 1, 1, 0). Find  $\frac{\partial u}{\partial x}, \frac{\partial v}{\partial y}, \frac{\partial u}{\partial y}$  at this point.

*Solution*. The three problems above are applications of the implicit function theorem and the equation

$$\left(\frac{\partial F}{\partial Y}\right)\left(\frac{\partial Y}{\partial X}\right) = -\left(\frac{\partial F}{\partial X}\right),\,$$

where F(X, Y) = 0 is the constraint equation and where the goal is to write *Y* as a function of *X*. By the implicit function theorem, this is possible if the  $m \times m$  matrix  $\left(\frac{\partial F}{\partial Y}\right)$  is invertible.

For example, denote the two equations in the last problem by  $f_1$  and  $f_2$ . Then we get

$$\begin{pmatrix} \frac{\partial(f_1, f_2)}{\partial(u, v)} \end{pmatrix}_{(1,1,1,0)} = \begin{pmatrix} x + y - (1+v)(2+u)^v & -(2+u)^{1+v}\ln(2+u) \\ 2 - (1+xy)(x-1)e^{u(x-1)} & 0 \end{pmatrix}_{(1,1,1,0)} \\ = \begin{pmatrix} 1 & -3\ln 3 \\ 2 & 0 \end{pmatrix},$$

which is invertible. Therefore, it is possible to write u and v as functions of x and y around (1, 1, 1, 0). To find the partials, solve

$$\begin{pmatrix} 1 & -3\ln 3 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} \partial u/\partial x & \partial u/\partial y \\ \partial v/\partial x & \partial v/\partial y \end{pmatrix} = - \begin{pmatrix} u & u \\ -u(1+xy)e^{u(x-1)} - ye^{u(x-1)} & -xe^{u(x-1)} \end{pmatrix}_{(1,1,1,0)}$$
$$\begin{pmatrix} 1 & -3\ln 3 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} \partial u/\partial x & \partial u/\partial y \\ \partial v/\partial x & \partial v/\partial y \end{pmatrix} = - \begin{pmatrix} 1 & 1 \\ -3 & -1 \end{pmatrix}.$$

(8) Write down a function on  $\mathbf{R}^2$  with a critical point at (0,0) which is neither a local minimum nor a local maximum.

*Solution*. The easiest is to write down an indefinite quadratic form like  $x^2 - y^2$ .

(9) Write down a function whose gradient at (0,0) is (1,3) and whose Hessian is  $\begin{pmatrix} 2 & 1 \\ 1 & 8 \end{pmatrix}$ . *Solution.* 

$$x + 3y + \frac{1}{2}(2x^2 + 2xy + 8y^2)$$

(10) Consider the matrix

$$A = \begin{pmatrix} 0 & 2 & 2 \\ 2 & 1 & 0 \\ 2 & 0 & -1 \end{pmatrix}.$$

Find an orthogonal matrix P such that  $P^TAP$  is diagonal. Solution. The columns of P will be the *unit eigenvectors* of A.

(11) State the spectral theorem.

Solution. For every symmetric matrix A, there exists an orthogonal matrix P such that  $P^{T}AP$  is diagonal.

You may also state it using eigenvectors – every symmetric matrix has an orthogonal basis of eigenvectors.  $\hfill \Box$ 

- (12) Let  $f(x, y, z) = \sin(x + 2y)e^{z-y}$ . Find the gradient and the Hessian of f. Write the second order Taylor approximation for f at (0,0,0).
- (13) Consider the function

$$f(x, y, z) = x^{2} + y^{2} + 3z^{2} - xy + 2xz + yz.$$

Find all critical points and use the second derivative test to determine if each one is a local minimum, local maximum, or neither (or say that the test cannot determine the answer).

(14) Suppose a differentiable convex function f on  $\mathbb{R}^n$  has a global maximum at a point  $\vec{p}$ . Show that f must be a constant function.

Solution. Since f is convex, the graph of f lies above the tangent (hyper)plane at any point on the graph. But the tangent (hyper)plane at  $\vec{p}$  is horizontal (since  $\vec{p}$  is a maximum), and the graph of f cannot lie strictly above this hyperplane (since  $\vec{p}$  is a maximum). So the graph must be this (hyper)plane. In other words, f is constant.

A less wordy and more math-y (and rigorous) way to write the above is as follows. Since f is convex, we have the inequality

$$f(\vec{x}) \ge f(\vec{p}) + \nabla f(p) \cdot (\vec{x} - \vec{p}).$$

Since  $\vec{p}$  is a maximum,  $\nabla f(p) = 0$ , so we get

$$f(\vec{x}) \ge f(\vec{p}).$$

But since  $\vec{p}$  is a maximum, we cannot have strict inequality, so we get

$$f(\vec{x}) = f(\vec{p})$$

for all  $\vec{x}$ . So f is constant.