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Due at 1:30pm in class, on October 13th, 2016

**Problem 1:** For each function below, tell us if it is “not convex”, “convex but not strictly convex”, or “strictly convex”. Justify your answer.

(a)  $f(x) = x^T Q x + b^T x + c$ , where  $x = (x_1, x_2)^T$ ,  $Q = \begin{pmatrix} 2 & 2 \\ 2 & 3 \end{pmatrix}$ ,  $b = (-1, -3)^T$  and  $c = 5$ ,

(b)  $f(x_1, x_2) = (2x_1 - 3x_2)^2 + 4x_1^3$ ,

(c)  $f(x_1, x_2) = (5x_1 - x_2)^2$ .

**Problem 2:** Prove or disprove the following claim:

The sum of two quasiconvex functions is quasiconvex.

**Problem 3: Minimizers of convex problems.**

Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a convex function and  $\Omega \subseteq \mathbb{R}^n$  be a convex set. Show that the set of minimizers of  $f$  over  $\Omega$  is convex (i.e., a convex set).

**Problem 4:** Consider a homogeneous polynomial  $p(x) = p(x_1, \dots, x_n)$  of degree  $d \geq 2$ . Show that if  $p$  is convex, then it is nonnegative; i.e.,  $p(x) \geq 0$ , for all  $x \in \mathbb{R}^n$ . (Note: A polynomial is homogeneous of degree  $d$  if all of its monomials have degree exactly  $d$ . The degree of a monomial  $x_1^{\alpha_1} \dots x_n^{\alpha_n}$  is equal to  $\alpha_1 + \dots + \alpha_n$ .)

**Problem 5: Regression with different penalties using convex optimization**

We are given 7 data points  $(x_i, y_i)$  in  $\mathbb{R}^2$ :  $\{(0, 0), (1, 3), (2, 7), (3, -1), (4, 0), (5, 5), (6, 10)\}$ .

We would like to fit a cubic polynomial

$$p(x) = c_3 x^3 + c_2 x^2 + c_1 x + c_0$$

to our data to minimize the so-called  $L^2$ -error :

$$\underset{c_0, \dots, c_3}{\text{minimize}} \quad \sum_{i=1}^7 (p(x_i) - y_i)^2, \quad (1)$$

1. (a) Confirm that problem (1) is a least squares problem by writing it as the problem of minimizing  $\|Ac - b\|^2$  over  $c = (c_0, \dots, c_3)^T$ , for some appropriate matrix  $A$  and vector  $b$ . Find the optimal solution and the optimal value using the closed-form solution we derived in class for the least squares problem.
- (b) Solve problem (1) in CVX and compare the answer with what you got in part (a).
2. Now consider two other types of error:

- the  $L^1$ -error:

$$\underset{c_0, \dots, c_3}{\text{minimize}} \quad \sum_{i=1}^7 |p(x_i) - y_i|, \quad (2)$$

- and the  $L^\infty$ -error :

$$\underset{c_0, \dots, c_3}{\text{minimize}} \quad \max_{i=1, \dots, 7} |p(x_i) - y_i|. \quad (3)$$

Although these problems no longer admit a closed-form solution, we can just as easily solve them using convex optimization.

- (a) Show that problems (2) and (3) are convex optimization problems. Solve them using CVX and report their optimal values and optimal solutions.
- (b) Generate a single plot with the seven data points, as well as the three cubic polynomials that are solutions to problems (1), (2), and (3). Let the  $x$  axis run from  $-1$  to  $7$  and label your plots clearly. You may want to use the `hold on` command to display multiple plots on the same figure.