

# Homework (17)

1. [Continued from Homework 16] Write down the dual problems of the following problems and check whether the strong dualities hold.

$$\begin{aligned} \text{a)} \quad & \min_{\mathbf{x}} x_1^2 + x_2^2, \\ & s.t. \quad -x_2 \leq 0, \\ & \quad \quad x_1^3 - x_2 \leq 0, \\ & \quad \quad x_1^3(x_2 - x_1^3) \leq 0. \end{aligned}$$

b) Support vector machine:

$$\begin{aligned} & \min_{\mathbf{w}} \frac{1}{2} \|\mathbf{w}\|^2, \\ & s.t. \quad y_i(\langle \mathbf{w}, \mathbf{x}_i \rangle + \beta) \geq 1, \quad i = 1, \dots, m. \end{aligned}$$

$$\begin{aligned} \text{c)} \quad & \min_x x^2 + 1, \\ & s.t. \quad (x - 2)(x - 4) \leq 0. \end{aligned}$$

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2. Express the dual problem of

$$\begin{aligned} \min_{\mathbf{x}} \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & f(\mathbf{x}) \leq 0, \end{aligned}$$

with  $\mathbf{c} \neq \mathbf{0}$ , in terms of the conjugate  $f^*$ . Explain why the problem you give is convex. We do not assume  $f$  is convex.

3. We consider the convex piecewise-linear minimization problem

$$\min_{\mathbf{x}} \max_{i=1, \dots, m} (\mathbf{a}_i^T \mathbf{x} + b_i) \tag{1}$$

with variable  $\mathbf{x} \in \mathbb{R}^n$ .

(a) Derive a dual problem, based on the Lagrange dual of the equivalent problem

$$\begin{aligned} \min_{\mathbf{x}} \quad & \max_{i=1, \dots, m} y_i \\ \text{s.t.} \quad & \mathbf{a}_i^T \mathbf{x} + b_i = y_i, \quad i = 1, \dots, m, \end{aligned}$$

with variables  $\mathbf{x} \in \mathbb{R}^n$ ,  $\mathbf{y} \in \mathbb{R}^m$ .

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- (b) Formulate the piecewise-linear minimization problem (1) as an LP, and form the dual of the LP. Relate the LP dual to the dual obtained in part (a).
- (c) Suppose we approximate the objective function in (2) by the smooth function

$$f_0(\mathbf{x}) = \log \left( \sum_{i=1}^m \exp(\mathbf{a}_i^T \mathbf{x} + b_i) \right),$$

and solve the unconstrained geometric program

$$\min_{\mathbf{x}} \log \left( \sum_{i=1}^m \exp(\mathbf{a}_i^T \mathbf{x} + b_i) \right). \quad (2)$$

Let  $p_{pwl}^*$  and  $p_{gp}^*$  be the optimal values of (1) and (2), respectively. Show that

$$0 \leq p_{gp}^* - p_{pwl}^* \leq \log m.$$

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- (d) Derive similar bounds for the difference between  $p_{pwl}^*$  and the optimal value of

$$\min_{\mathbf{x}} (1/\gamma) \log \left( \sum_{i=1}^m \exp(\gamma(\mathbf{a}_i^T \mathbf{x} + b_i)) \right),$$

where  $\gamma > 0$  is a parameter. What happens as we increase  $\gamma$ ?