

# Machine Learning Homework 4

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**Problem.** Given a matrix  $A = (a_{ij})_{n \times m}$ , show that  $\min_{i \leq n} \max_{j \leq m} a_{ij} \geq \max_{i \leq m} \min_{j \leq n} a_{ji}$ .

*Proof.* Denote  $\min_{i \leq n} \max_{j \leq m} a_{ij} = a_{pq}$ ,  $\max_{i \leq m} \min_{j \leq n} a_{ji} = a_{rt}$ . Then

$$a_{pq} \geq a_{rq} \geq a_{rt}$$

Therefore,

$$\min_{i \leq n} \max_{j \leq m} a_{ij} \geq \max_{i \leq m} \min_{j \leq n} a_{ji}$$

□

**Problem.** Show that KKT conditions are necessary and if  $f, g_i$  are convex and each  $h_i$  is linear then it's also suffice for  $(X^*, \lambda^*, \mu^*)$  to be the optima of primal and dual programmings.

*Proof.* First we show that KKT conditions are necessary. Denote  $(x^*, \lambda^*, \mu^*)$  the solution of primal and dual problems. Obviously, it satisfies (1), (2) and (3) of KKT conditions. For (4), consider the primal problem, if  $g_i(x^*) < 0$ , then  $\lambda_i(x^*)$  should equal to 0 in order to max the primal function. Therefore,  $\lambda_i g_i(x^*) = 0$ .

Then we show that with the additional condition that  $f(x), g_i(x)$  are convex and  $h_i(x)$  is linear, KKT condition is sufficient.

First, a  $x_0$  satisfies the above conditions also satisfies the constraints in primal and dual problems. Denote  $x_1$  the solution of primal problem and  $x_2$  the solution of dual problem, denote  $L(x, \mu, \lambda) = f(x) + \sum_i \mu_i h_i(x) + \sum_i \lambda_i g_i(x)$ .

With the additional condition,  $L(x, \mu, \lambda)$  is a convex function w.r.t  $x$ , then  $x_0 = \arg \min_x L(x, \mu, \lambda)$ . Therefore,

$$L(x_0, \mu, \lambda) \leq L(x_2, \mu, \lambda) \leq L(x_1, \mu, \lambda) \quad (1)$$

Note that  $x_0$  satisfies primal constraints and  $x_2$  is the argmin of primal problem. Therefore,

$$L(x_1, \mu, \lambda) \leq L(x_0, \mu, \lambda) \quad (2)$$

With (1) and (2) we have our conclusion that

$$L(x_0, \mu, \lambda) = L(x_2, \mu, \lambda) = L(x_1, \mu, \lambda)$$

□