

Convex optimization HW3

KALLEL Mahdi

November 22, 2020

Exercise 1

1. From the previous homework we have :

$$\|u\|_1^* = \begin{cases} 0 & \text{if } \|u\|_\infty \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

(LASSO) can be written as :

$$\begin{aligned} \min_w \quad & \frac{1}{2} \|z\|_2^2 + \lambda \|w\|_1 \\ \text{s.t. } & z = Xw - y \end{aligned} \quad [1.1]$$

The dual is :

$$\begin{aligned} g(v) &= \inf_{w,z} \mathcal{L}(w, z, v) \\ &= y^T v + \inf_z \left(\frac{1}{2} \|z\|_2^2 + v^T z \right) + \inf_w \left(\lambda \|w\|_1 - (X^T v)^T w \right) \end{aligned}$$

$h : z \mapsto \frac{1}{2} \|z\|_2^2 + v^T z$ is convex and differentiable.

$\nabla h(z) = z + v$ $\nabla h(z) = 0$ iff $z = -v$.

\implies The minimum of h is $\frac{1}{2} \|v\|_2^2 - \|v\|_2^2 = -\frac{1}{2} \|v\|_2^2$.

The second term can be reformulated using $\|\cdot\|_1$:

$$\inf_w \lambda \|w\|_1 - (X^T v)^T w = \sup_w \left(\frac{1}{\lambda} X^T v \right)^T w - \|w\|_1 = \left\| \frac{1}{\lambda} X^T v \right\|_1^*$$

Thus,

$$g(v) = y^T v - \frac{1}{2} \|v\|_2^2 + \left\| \frac{1}{\lambda} X^T v \right\|_1^*$$

The dual can be re writte as :

$$\begin{aligned} & \max_v y^T v - \frac{1}{2} \|v\|_2^2 \\ \text{s.t. } & \left\| \frac{1}{\lambda} X^T v \right\|_{\infty} \leq 1 \end{aligned}$$

The constraint can be reformulated

$$\begin{aligned} \left\| \frac{1}{\lambda} X^T v \right\|_{\infty} \leq 1 & \text{ iff } \forall i, -1 \leq \left[\frac{1}{\lambda} X^T v \right]_i \leq 1 \\ & \text{ iff } \forall i, \left[\frac{1}{\lambda} X^T v \right]_i \leq 1 \text{ and } \left[-\frac{1}{\lambda} X^T v \right]_i \leq 1 \\ & \text{ iff } Av \leq \lambda \mathbf{1}_{2d} \text{ where } A = \begin{pmatrix} X^T \\ -X^T \end{pmatrix} \end{aligned}$$

The problem can be re written as :

$\begin{aligned} & \min_v v^T Q v + p^T v \quad \text{with } Q = \frac{1}{2} I_n, p = -y, b = \lambda \cdot \mathbf{1}_{2d} \\ \text{s.t. } & Av \leq b \end{aligned}$
--

2. For questions 2 and 3 please check the code