Module 12

Nonlinear Optimization and Hard Margin SVM Solutions

Problem 1: 50 points total

Linear programming with box constraints is the following optimization problem:

minimize
$$c^{\top}x$$

subject to $Ax = b$
 $l \le x \le u$,

where A is an $m \times n$ matrix, $c, u, l, x \in \mathbb{R}^n$ and $b \in \mathbb{R}^m$, with $l \leq u$ (which means that $l_i \leq u_i$, for $i = 1, \ldots, n$).

(1) (20 points) Prove that the dual of the above program is the following program:

$$\begin{aligned} \text{maximize} & & -\nu^\top b - \lambda_1^\top u + \lambda_2^\top l \\ \text{subject to} & & A^\top \nu + \lambda_1 - \lambda_2 + c = 0 \\ & & & \lambda_1 \geq 0, \quad \lambda_2 \geq 0. \end{aligned}$$

(2) (10 points) The primal problem in (1) can be reformulated by incorporating the constraints $l \leq x \leq u$ into the objective function by defining

$$f_0(x) = \begin{cases} c^{\top} x & \text{if } l \le x \le u \\ +\infty & \text{otherwise.} \end{cases}$$

The primal is reformulated as

minimize
$$f_0(x)$$

subject to $Ax = b$.

Prove that the new dual function is given by

$$G(\nu) = \inf_{l \le x \le u} (c^{\mathsf{T}} x + \nu^{\mathsf{T}} (Ax - b)).$$

(3) (20 points) Given any real number $s \in \mathbb{R}$, let

$$s^+ = \max\{s, 0\}, \quad s^- = \max\{-s, 0\}.$$

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Prove that for any reals $s, \lambda, \mu \in \mathbb{R}$ with $\lambda \leq \mu$,

$$\inf_{\lambda \le y \le \mu} sy = \lambda s^+ - \mu s^-.$$

Hint. Consider the cases $s \geq 0$ and $s \leq 0$.

We extend the above operators to vectors $z \in \mathbb{R}^n$ componentwise by

$$z^+ = (z_1^+, \dots, z_n^+), \quad z^- = (z_1^-, \dots, z_n^-).$$

For any $w \in \mathbb{R}^n$, prove that

$$\inf_{l \le x \le u} x^\top w = l^\top w^+ - u^\top w^-.$$

Use the above to prove that

$$G(\nu) = -\nu^{\top} b + l^{\top} (A^{\top} \nu + c)^{+} - u^{\top} (A^{\top} \nu + c)^{-}$$

and deduce that the dual program is the unconstained problem

maximize
$$-\nu^{\top}b + l^{\top}(A^{\top}\nu + c)^{+} - u^{\top}(A^{\top}\nu + c)^{-}$$

with respect to ν .

Solution. (1) First we rewrite the inequalities $x \leq u$ and $l \leq x$ as $x-u \leq 0$ and $-x+l \leq 0$. To form the Lagrangian we assign the Lagrange multipliers $\nu \in \mathbb{R}^m$ to the system of equations Ax - b = 0, $\lambda_1 \in \mathbb{R}^n_+$ ($\lambda_1 \geq 0$) to the inequalities $x - u \leq 0$, and $\lambda_2 \in \mathbb{R}^n_+$ ($\lambda_2 \geq 0$) to the inequalities $-x + l \leq 0$. The Lagrangian $L(x, \nu, \lambda_1, \lambda_2)$ is given by

$$L(x, \nu, \lambda_1, \lambda_2) = c^{\mathsf{T}} x + \nu^{\mathsf{T}} (Ax - b) + \lambda_1^{\mathsf{T}} (x - u) + \lambda_2^{\mathsf{T}} (-x + l).$$

The Lagrangian can be rewritten as

$$L(x, \nu, \lambda_1, \lambda_2) = (c^\top + \nu^\top A + \lambda_1^\top - \lambda_2^\top) x - \nu^\top b - \lambda_1^\top u + \lambda_2^\top l.$$

Next we want to contruct the dual function $G(\nu, \lambda_1, \lambda_2)$ obtained by minimizing $L(x, \nu, \lambda_1, \lambda_2)$ over $x \in \mathbb{R}^n$,

$$G(\nu, \lambda_1, \lambda_2) = \inf_{x \in \mathbb{R}^n} L(x, \nu, \lambda_1, \lambda_2).$$

The linear form $L(x, \nu, \lambda_1, \lambda_2)$ (in x, with $\nu, \lambda_1, \lambda_2$ fixed) is unbounded below unless

$$c^{\top} + \nu^{\top} A + \lambda_1^{\top} - \lambda_2^{\top} = 0,$$

which (by transposition) is equivalent to

$$A^{\top}\nu + \lambda_1 - \lambda_2 + c = 0.$$

Therefore the dual function $G(\nu, \lambda_1, \lambda_2)$ is given by

$$G(\nu, \lambda_1, \lambda_2) = \begin{cases} -\nu^{\top} b - \lambda_1^{\top} u + \lambda_2^{\top} l & \text{if } A^{\top} \nu + \lambda_1 - \lambda_2 + c = 0\\ -\infty & \text{otherwise.} \end{cases}$$

The above defintion yields the dual program

maximize
$$-\nu^{\top}b - \lambda_1^{\top}u + \lambda_2^{\top}l$$

subject to $A^{\top}\nu + \lambda_1 - \lambda_2 + c = 0$
 $\lambda_1 \ge 0, \quad \lambda_2 \ge 0.$

(2) This time we only need to assign the Lagrange multipliers $\nu \in \mathbb{R}^m$ to the system of equations Ax - b = 0 and we obtain the Lagrangian

$$L(x, \nu) = f_0(x) + \nu^{\top} (Ax - b).$$

The dual function $G(\nu)$ is obtained by minimizing $L(x,\nu)$ over $x \in \mathbb{R}^n$ (with ν fixed). By definition of

$$f_0(x) = \begin{cases} c^{\top} x & \text{if } l \le x \le u \\ +\infty & \text{otherwise,} \end{cases}$$

this minimization can be restricted to those $x \in \mathbb{R}^n$ such that $l \leq x \leq u$, in which case $f_0(x) = c^{\top}x$, so

$$G(\nu) = \inf_{l \le x \le u} (c^{\mathsf{T}} x + \nu^{\mathsf{T}} (Ax - b)).$$

(3) We have

$$c^{\mathsf{T}}x + \nu^{\mathsf{T}}(Ax - b) = -\nu^{\mathsf{T}}b + (\nu^{\mathsf{T}}A + c^{\mathsf{T}})x = -\nu^{\mathsf{T}}b + x^{\mathsf{T}}(A^{\mathsf{T}}\nu + c),$$

and so

$$G(\nu) = \inf_{l < x < u} (-\nu^\top b + x^\top (A^\top \nu + c)).$$

Given any real number $s \in \mathbb{R}$, let

$$s^+ = \max\{s, 0\}, \quad s^- = \max\{-s, 0\}.$$

Then for any fixed reals $s, \lambda, \mu \in \mathbb{R}$ with $\lambda \leq \mu$, we claim that

$$\inf_{\lambda \le y \le \mu} sy = \lambda s^+ - \mu s^-.$$

Case 1. $s \ge 0$. In this case $s^+ = s$, and since $-s \le 0$, we have $s^- = 0$. The minimum of $y \mapsto sy$ for $\lambda \le y \le \mu$ is $\lambda s = \lambda s^+$.

Case 2. $s \le 0$. In this case $s^+ = 0$ and $s^- = -s \ge 0$. The minimum of $y \mapsto sy$ for $\lambda \le y \le \mu$ is $\mu s = -\mu s^-$.

In summary, we have

$$\inf_{\lambda \le y \le \mu} sy = \lambda s^+ - \mu s^-,$$

as claimed.

For any $w \in \mathbb{R}^n$, the minimum of a linear form $x \mapsto x^\top w = \sum_{i=1}^n x_i w_i$ for $1 \le x \le u$ is obtained by computing the minima componentwise,

$$\inf_{l \le x \le u} \left(\sum_{i=1}^{n} w_i x_i \right) = \sum_{i=1}^{n} \inf_{l_i \le x_i \le u_i} w_i x_i$$

$$= \sum_{i=1}^{n} (l_i w_i^+ - u_i w_i^-)$$

$$= \sum_{i=1}^{n} l_i w_i^+ - \sum_{i=1}^{n} u_i w_i^-$$

$$= l^\top w^+ - u^\top w^-,$$

so we get

$$\inf_{l \le x \le u} x^\top w = l^\top w^+ - u^\top w^-.$$

As a consequence, since

$$G(\nu) = \inf_{l < x < u} (-\nu^{\top} b + x^{\top} (A^{\top} \nu + c)),$$

with $w = A^{\top} \nu + c$ we obtain

$$G(\nu) = -\nu^{\mathsf{T}}b + l^{\mathsf{T}}(A^{\mathsf{T}}\nu + c)^{+} - u^{\mathsf{T}}(A^{\mathsf{T}}\nu + c)^{-}.$$

If follows immediately that the dual program is the unconstrained problem

maximize
$$-\nu^{\mathsf{T}}b + l^{\mathsf{T}}(A^{\mathsf{T}}\nu + c)^{+} - u^{\mathsf{T}}(A^{\mathsf{T}}\nu + c)^{-}$$

with respect to ν .

Total: 50 points