

ECE M146

Introduction to Machine Learning

Instructor: Lara Dolecek

TA: Zehui (Alex) Chen, Ruiyi (John) Wu

Discussion 4

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1. Introduction on optimization problem

- (a) Convex sets.
- (b) Convex functions.
- (c) Optimization problem in standard form.
 - Convex optimization.
- (d) Global and local optimal.
- (e) Duality.
 - Lagrange dual problem.
 - Geometric interpretation.
 - KKT conditions.

2. Find the dual problem of the following Quadratic program

$$\begin{array}{ll}\text{minimize}_x & x^T P x \\ \text{subject to} & Ax \leq b\end{array}$$

Assume $P \in \mathcal{S}_{++}^n$.

3. Quadratic program example Consider the objective function

$$J(x_1, x_2) = 5x_1^2 + 4x_1x_2 + 2x_2^2 + 2x_1 - 4x_2.$$

Find the optimal x that minimize $J(x)$ under the following constrains:

- (a) No constrain.
- (b) $x_1 + x_2 + 2 = 0$.
- (c) $x_1 + x_2 + 2 \leq 0$.
- (d) $x_1 + x_2 + 2 \geq 0$.

4. Relation soft SVM loss and logistic regression

We learned that the loss function for the primal problem of SVM is of the form

$$\frac{1}{2}\|w\|^2 + C \sum_{n=1}^m \xi_n.$$

We have seen that for data points that are on the correct side of the margin boundary, and which therefore satisfy $y_n h(x_n) \geq 1$, we have $\xi_n = 0$. For the remaining points we have $\xi_n = 1 - y_n h(x_n)$. Thus the objective function can be written in the form

$$\sum_{n=1}^N E_{SV}(y_n h(x_n)) + \lambda \|w\|^2.$$

where $\lambda = (2C)^{-1}$, find out what is the error function $E_{SV}(\cdot)$ in terms of $y_n h(x_n)$.