

Problem 1 (5 points)

Find the solution  $(x^*, y^*)$  to the following problem.

optimize  $xy$

subject to  $2x + 2y = 20$

Make sure the constraint is in standard form.

Standard form

optimize  $xy$

subject to  $2x + 2y - 20 = 0$

Lagrangian:  $L(x, y, \beta) = xy + \beta(2x + 2y - 20)$

Partial Derivatives:

$$\nabla_x L(x, y, \beta) = y + 2\beta = 0$$

$$\nabla_y L(x, y, \beta) = x + 2\beta = 0$$

$$\nabla_\beta L(x, y, \beta) = 2x + 2y - 20 = 0$$

Solve for  $x, y$

$$x^* = 5, y^* = 5$$

Problem 2 (5 points)

Define the Lagrangian function for the following optimization problem.

$$\begin{aligned} &\text{minimize } 7x_1^3 + 8x_2^2 + 9x_1 \\ &\text{subject to } 6x_1 + 11x_2^2 = 13x_2 \\ &\quad 5x_2^2 \leq 4x_1 - 7 \\ &\quad 20x_1 + 5 \geq x_2 \end{aligned}$$

Do not solve for the variables, i.e., do not try to compute the partial derivatives.

Make sure the constraints are in standard form.

Standard form

$$\begin{aligned} &\text{minimize } 7x_1^3 + 8x_2^2 + 9x_1 \\ &\text{subject to } 6x_1 + 11x_2^2 - 13x_2 = 0 \\ &\quad 5x_2^2 - 4x_1 + 7 \leq 0 \\ &\quad -20x_1 - 5 + x_2 \leq 0 \end{aligned}$$

$$\begin{aligned} L(x_1, x_2, \alpha_1, \alpha_2, \beta_1) = &7x_1^3 + 8x_2^2 + 9x_1 + \\ &\alpha_1(5x_2^2 - 4x_1 + 7) + \\ &\alpha_2(-20x_1 - 5 + x_2) + \\ &\beta_1(6x_1 + 11x_2^2 - 13x_2) \end{aligned}$$

Bonus Problem (5 points)

The SVM optimization can be defined by the primal form:

$$\min_{\mathbf{w}} \frac{1}{2} \|\mathbf{w}\|^2$$
$$\text{subject to } y_i(\mathbf{w}^T \mathbf{x}_i + b) \geq 1, \quad i = 1, \dots, N$$

Or by its the dual form:

$$\max_{\alpha} J(\alpha) = \sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (\mathbf{x}_i^T \mathbf{x}_j)$$
$$\text{subject to } \alpha_i \geq 0, i = 1, \dots, N \text{ and } \sum_{i=1}^N \alpha_i y_i = 0$$

Show the objective function  $J(\alpha)$  is the Lagrangian function  $L(\mathbf{w}, b, \alpha)$  evaluated at  $\mathbf{w}$  that minimizes that function.

Hints:

1. Write the primal problem in standard form
2. Form the Lagrangian function  $L(\mathbf{w}, b, \alpha)$
3. Find  $\mathbf{w}$  and  $b$  that minimize  $L(\mathbf{w}, b, \alpha)$
4. Plug the results back into  $L(\mathbf{w}, b, \alpha)$

Note  $\|\mathbf{z}\|^2$  is the inner product  $\langle \mathbf{z}_i, \mathbf{z}_j \rangle$  meaning  $\mathbf{z}^T \mathbf{z}$  between points in the vector space. If  $\mathbf{z} = \sum_i \mathbf{x}_i$  then  $\mathbf{z}^T \mathbf{z} = \sum_i \mathbf{x}_i^T \sum_j \mathbf{x}_j$

(1). Write the primal problem in standard form

$$\min_{\mathbf{w}} \frac{1}{2} \|\mathbf{w}\|^2 \text{ subject to } g_i(\mathbf{w}) = -y_i(\mathbf{w}^T \mathbf{x}_i + b) + 1 \leq 0$$

(2). Form the Lagrangian function

$$L(\mathbf{w}, b, \alpha) = \frac{1}{2} \|\mathbf{w}\|^2 + \sum_{i=1}^N \alpha_i [-y_i(\mathbf{w}^T \mathbf{x}_i + b) + 1] \rightarrow (\text{Eq. 1})$$

- Note there is no  $\beta_i$  because there are no equality constraints

(3). Find  $\mathbf{w}$  and  $b$  that minimize  $L(\mathbf{w}, b, \alpha)$

$$\nabla_{\mathbf{w}} L(\mathbf{w}, b, \alpha) = \mathbf{w} - \sum_{i=1}^N \alpha_i y_i \mathbf{x}_i = 0 \rightarrow \mathbf{w} = \sum_{i=1}^N \alpha_i y_i \mathbf{x}_i \rightarrow (\text{Eq. 2})$$

$$\nabla_b L(\mathbf{w}, b, \alpha) = \sum_{i=1}^N \alpha_i y_i = 0 \rightarrow (\text{Eq. 3})$$

(4). Plug (Eq. 2) and (Eq. 3) back into (Eq. 1). Notice index  $i$  in each equation is only relevant for that equation.

$$L(\mathbf{w}, b, \alpha) = \frac{1}{2} \mathbf{w}^T \mathbf{w} + \sum_{i=1}^N \alpha_i [-y_i(\mathbf{w}^T \mathbf{x}_i + b) + 1]$$

$$L(\mathbf{w}, b, \alpha) = \frac{1}{2} \sum_{i=1}^N \alpha_i y_i \mathbf{x}_i^T \cdot \sum_{j=1}^N \alpha_j y_j \mathbf{x}_j + \sum_{i=1}^N \alpha_i \left[ -y_i \left( \left( \sum_{j=1}^N \alpha_j y_j \mathbf{x}_j^T \right) \mathbf{x}_i + b \right) + 1 \right]$$

$$L(\mathbf{w}, b, \alpha) = \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (\mathbf{x}_i^T \mathbf{x}_j) - \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (\mathbf{x}_j^T \mathbf{x}_i) - \sum_{i=1}^N \alpha_i y_i b + \sum_{i=1}^N \alpha_i$$

But

$$\sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (\mathbf{x}_j^T \mathbf{x}_i) = \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (\mathbf{x}_i^T \mathbf{x}_j)$$

And from (Eq. 3)

$$\sum_{i=1}^N \alpha_i y_i b = 0$$

Thus,

$$J(\alpha) = L(w, b, \alpha) = \sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y_i y_j (x_i^T x_j)$$