

Quiz 5: Solutions

1 Question 1 (100 points)

Consider the optimization problem shown below when answering the following questions.

$$\begin{aligned} \min_{\mathbf{w}, \epsilon} \quad & \frac{1}{2} \|\mathbf{w}\|_2^2 + \frac{C}{2} \|\epsilon\|_2^2 \\ \text{s.t.} \quad & \epsilon_n = y_n - \mathbf{w}^T \mathbf{x}_n \text{ for } 1 \leq n \leq N \end{aligned}$$

Note that C is a constant as are the y_n 's and \mathbf{x}_n 's. The optimization variables are \mathbf{w} and $\epsilon = [\epsilon_1, \dots, \epsilon_N]$. Both \mathbf{w} and the \mathbf{x}_n 's are column vectors.

Explain what model this optimization problem is learning and give an alternate unconstrained optimization problem that results in the same optimal parameters. Show your work and/or explain your answer.

Solution: Since this is an equality constrained problem, we can move the constraint directly into the objective function letting $\mathbf{y} = [y_1, \dots, y_N]$ and $\mathbf{X} = [\mathbf{x}_1, \dots, \mathbf{x}_N]$ we have:

$$\min_{\mathbf{w}, \epsilon} \frac{1}{2} \|\mathbf{w}\|_2^2 + \frac{C}{2} \|\mathbf{y} - \mathbf{w}^T \mathbf{X}\|_2^2 \tag{1}$$

Expanding the second norm, we have:

$$\min_{\mathbf{w}, \epsilon} \frac{1}{2} \|\mathbf{w}\|_2^2 + \frac{C}{2} \sum_{n=1}^N (y_n - \mathbf{w}^T \mathbf{x}_n)^2 \tag{2}$$

So, this optimization problem is learning a linear regression model $f(\mathbf{x}) = \mathbf{w}^T \mathbf{x}$.

Rubric:

- 50 points: An explanation indicating that a linear regression model is being learned.
- 25 points: A valid unconstrained form of the problem as in (1) or (2) or similar.
- 25 points: Some valid work or a brief explanation of how the equivalent form was obtained.