

HW 2B – ENGR 4399: Machine Learning

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Problem 1 Show that the ridge regression objective function $f(\beta) = \|\mathbf{Y} - \mathbb{X}\beta\|^2 + \lambda\|\beta\|^2$ is convex in β . Find the optimal solution and provide justification for each step. Here

$$\|\beta\| = \|\beta\|_2 = \sqrt{\beta_1^2 + \beta_2^2 + \cdots + \beta_p^2}.$$

You should get the solution

$$\beta_{\text{ridge}} = (\mathbb{X}^T \mathbb{X} + \lambda I_p)^{-1} \mathbb{X}^T \mathbf{Y}.$$

Now, assume that the columns of the data matrix \mathbb{X} are orthonormal, i.e.,

$$\mathbb{X}^T \mathbb{X} = I_p.$$

Show that under this assumption, the ridge solution **shrinks** the OLS solutions towards the all zero vector.

$$\beta_{\text{ridge}} = \frac{\mathbb{X}^T \mathbf{Y}}{1 + \lambda} = \frac{\beta_{\text{ols}}}{1 + \lambda}.$$

Problem 2 Show that the Elastic Net regression function, for $\lambda_1 > 0$ and $\lambda_2 > 0$,

$$f(\beta) = \|\mathbf{Y} - \mathbb{X}\beta\|^2 + \lambda_1 \|\beta\|_2^2 + 2\lambda_2 \|\beta\|_1,$$

is convex in β . Here

$$\|\beta\|_1 = \sqrt{|\beta_1| + |\beta_2| + \cdots + |\beta_p|}.$$

Problem 3 Define the BIC regression function

$$f(\beta) = \|\mathbf{Y} - \mathbb{X}\beta\|^2 + \lambda \|\beta\|_0,$$

where $\|\beta\|_0 = \sum_{i=1}^p \mathbb{I}_{\{\beta_i \neq 0\}}$. Show that the function $f(\beta)$ is **not convex**.

Problem 4 Define the logistic regression function

$$f(\beta) = - \sum_{i=1}^n \left(Y_i \beta^T X_i + \log \frac{1}{1 + e^{\beta^T X_i}} \right).$$

Here $(X_1, Y_1), \dots, (X_n, Y_n)$ is our training data. Show that the function above is convex in β .

Problem 5 Define the support vector machine objective function

$$L(\beta, \beta_0, \lambda) = \frac{1}{2} \|\beta\|^2 + \sum_{i=1}^n \lambda_i (1 - Y_i (X_i^T \beta + \beta_0)).$$

Show that the function is convex in β and β_0 .