

1 L_1 and L_2 Regularization

1.1 Concept Check Questions

1. Consider the following two minimization problems:

$$\arg \min_w \Omega(w) + \frac{\lambda}{n} \sum_{i=1}^n L(f_w(x_i), y_i)$$

and

$$\arg \min_w C\Omega(w) + \frac{1}{n} \sum_{i=1}^n L(f_w(x_i), y_i),$$

where $\Omega(w)$ is the penalty function (for regularization) and L is the loss function. Give sufficient conditions under which these two give the same minimizer.

2. (\star) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a differentiable function. Prove that $\|\nabla f(x)\|_2 \leq L$ if and only if f is Lipschitz with constant L .
3. (\star) Let \hat{w} denote the minimizer for

$$\begin{array}{ll} \text{minimize}_w & \|Xw - y\|_2^2 \\ \text{subject to} & \|w\|_1 \leq r. \end{array}$$

Prove that $f(x) = \hat{w}^T x$ is Lipschitz with constant r .

4. Two of the plots in the lecture slides use the fact that $\|\hat{w}\|/\|\tilde{w}\|$ is always between 0 and 1. Here \hat{w} is the parameter vector of the linear model resulting from the regularized least squares problem. Analogously, \tilde{w} is the parameter vector from the unregularized problem. Why is this true that the quotient lies in $[0, 1]$?
5. Explain why feature normalization is important if you are using L_1 or L_2 regularization.