

Problem 1

Define

$$J_1(\mathbf{w}) = \|\mathbf{y} - \mathbf{X}\mathbf{w}\|^2 + \lambda_2 \|\mathbf{w}\|^2 + \lambda_1 \|\mathbf{w}\|_1$$

and

$$J_2(\mathbf{w}) = \|\tilde{\mathbf{y}} - \tilde{\mathbf{X}}\mathbf{w}\|^2 + c\lambda_1 \|\mathbf{w}\|_1$$

where $c = (1 + \lambda_2)^{-\frac{1}{2}}$ and

$$\tilde{\mathbf{X}} = c \begin{pmatrix} \mathbf{X} \\ \sqrt{\lambda_2} \mathbf{I}_d \end{pmatrix}, \quad \tilde{\mathbf{y}} = \begin{pmatrix} \mathbf{y} \\ \mathbf{0}_{d \times 1} \end{pmatrix}$$

Show

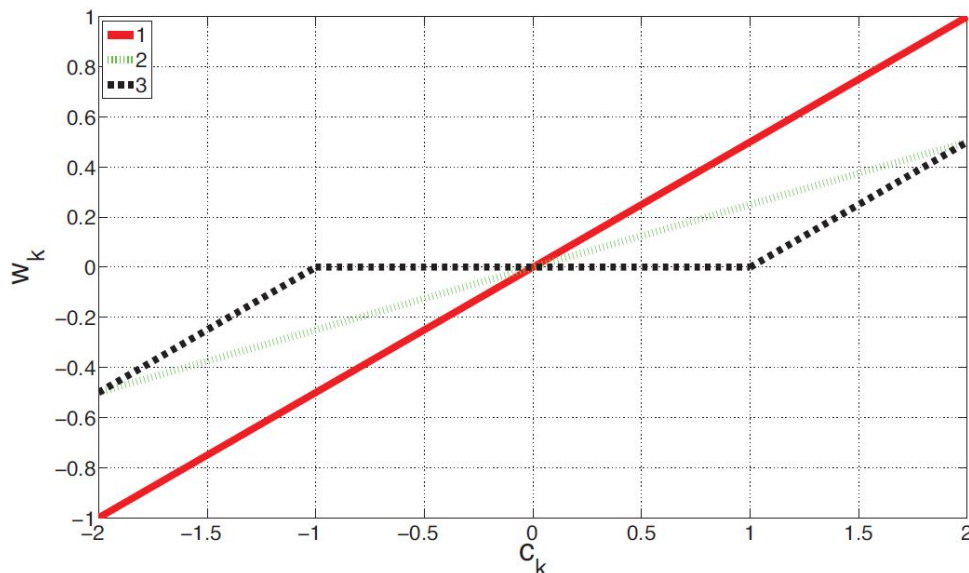
$$\arg \min J_1(\mathbf{w}) = c(\arg \min J_2(\mathbf{w}))$$

i.e. $J_1(c\mathbf{w}) = J_2(\mathbf{w})$ and hence that one can solve an elastic net problem using a lasso solver on modified data.

Problem 2

Consider performing linear regression with an orthonormal design matrix, so $\|\mathbf{x}_{:,k}\|_2^2 = 1$ for each column (feature) k , and $\mathbf{x}_{:,k}^T \mathbf{x}_{:,j} = 0$ so we can estimate each parameter w_k separately.

The following figure plots \tilde{w}_k vs $c_k = 2\mathbf{y}^T \mathbf{x}_{:,k}$, the correlation of feature k with the response, for 3 different estimation methods: ordinary least squares (OLS), ridge regression with parameter λ_2 , and lasso with parameter λ_1 .



Assignment 9

CMPT 727
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1. Unfortunately we forgot to label the plots. Which method does the solid (1), dotted (2) and dashed (3) line correspond to?
2. What is the value of λ_1 ?
3. What is the value of λ_2 ?

Problem 3

Please write one thing from this course so far that you found confusing, a topic you would like to hear more about, or something you found particularly interesting.