

Linear Regression

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Abstract—This manual provides an introduction to linear methods in regression.

1 LEAST SQUARES

1.1 Show that

$$\hat{\mathbf{w}} = \min_{\mathbf{w}} \|\mathbf{y} - \mathbf{X}\mathbf{w}\|^2 \quad (1.1)$$

$$= (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y} \quad (1.2)$$

1.2 Using the Gram-Schmidt orthogonalization procedure, show that

$$\mathbf{X} = \mathbf{Q}\mathbf{R} \quad (1.3)$$

where $\mathbf{Q}^T \mathbf{Q} = \mathbf{I}$ and \mathbf{R} is upper triangular.

1.3 Show that

$$\hat{\mathbf{w}} = \mathbf{R}\mathbf{Q}^T \mathbf{y} \quad (1.4)$$

1.4 Find $\hat{\mathbf{y}}$

2 RIDGE REGRESSION

2.1 The ridge problem is defined as

$$\hat{\mathbf{w}} = \min_{\mathbf{w}} \|\mathbf{y} - \mathbf{X}\mathbf{w}\| \quad (2.1)$$

$$\text{s.t. } \|\mathbf{w}\|^2 \leq t \quad (2.2)$$

Using the Lagrangian, show that

$$\hat{\mathbf{w}} = (\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I})^{-1} \mathbf{X}^T \mathbf{y} \quad (2.3)$$

3 THE LASSO

3.1 The Lasso is defined as

$$\hat{\mathbf{w}} = \min_{\mathbf{w}} \|\mathbf{y} - \mathbf{X}\mathbf{w}\| \quad (3.1)$$

$$\text{s.t. } \sum_i |w_i| \leq t \quad (3.2)$$

Obtain the corresponding Lagrangian.

3.2 Show that this is a quadratic programming problem and find a suitable algorithm.

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