

Introduction to Statistical Machine Learning

Homework 1

Yota Toyama

October 20, 2016

1.

$$\begin{aligned}
 \frac{\partial}{\partial \mathbf{w}} R(\mathbf{w}) &= 0 \\
 \frac{\partial}{\partial \mathbf{w}} E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})^2] &= 0 \\
 E_{p(\mathbf{x}, y)} [2(y - \mathbf{w}^T \mathbf{x})(-\mathbf{x})] &= 0 \\
 E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})\mathbf{x}] &= 0 \\
 \mathbf{a}^T E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})\mathbf{x}] &= 0 \\
 E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})\mathbf{a}^T \mathbf{x}] &= 0
 \end{aligned}$$

2.

$$\begin{aligned}
 E_{p(\mathbf{x}, y)} [y - \mathbf{w}^T \mathbf{x}] &= 0 \\
 E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})E_{p(\mathbf{x})} [\mathbf{x}]] &= 0 \\
 E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})\mathbf{x}] + E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})E_{p(\mathbf{x})} [\mathbf{x}]] &= 0 \\
 E_{p(\mathbf{x}, y)} [(y - \mathbf{w}^T \mathbf{x})(\mathbf{x} - E_{p(\mathbf{x})} [\mathbf{x}])] &= 0
 \end{aligned}$$

\therefore the correlation between data and prediction errors is 0.

3.

$$\begin{aligned}\hat{\mathbf{w}} &= \underset{\mathbf{w}}{\operatorname{argmin}} \sum_{i=1}^N (y_i - \mathbf{w}^T \mathbf{x}_i)^2 \\ &= (X^T X)^{-1} X^T y\end{aligned}$$

Let $C \in \mathbb{R}^{(d+1) \times (d+1)}$ be a diagonal matrix s.t. $\tilde{X} = XC$

$$\begin{aligned}\hat{\tilde{\mathbf{w}}} &= \underset{\mathbf{w}}{\operatorname{argmin}} \sum_{i=1}^N (y_i - \mathbf{w}^T \tilde{\mathbf{x}}_i)^2 \\ &= (\tilde{X}^T \tilde{X})^{-1} \tilde{X}^T y \\ &= ((XC)^T XC)^{-1} (XC)^T y \\ &= (CX^T XC)^{-1} CX^T y \\ &= C^{-1} (X^T X)^{-1} C^{-1} CX^T y \\ &= C^{-1} (X^T X)^{-1} X^T y \\ \tilde{X} \hat{\tilde{\mathbf{w}}} &= XCC^{-1} (X^T X)^{-1} X^T y \\ &= X(X^T X)^{-1} X^T y \\ &= X\hat{\mathbf{w}} \text{ as required}\end{aligned}$$

4.