Introduction to Statistical Machine Learning Homework 1

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1.

$$\frac{\partial}{\partial \mathbf{w}} R(\mathbf{w}) = 0$$

$$\frac{\partial}{\partial \mathbf{w}} E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x})^2 \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[2(y - \mathbf{w}^T \mathbf{x})(-\mathbf{x}) \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) \mathbf{x} \right] = 0$$

$$\mathbf{a}^T E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) \mathbf{x} \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) \mathbf{a}^T \mathbf{x} \right] = 0$$

2.

$$E_{p(\mathbf{x},y)} \left[y - \mathbf{w}^T \mathbf{x} \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) E_{p(\mathbf{x})} \left[\mathbf{x} \right] \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) \mathbf{x} \right] + E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) E_{p(\mathbf{x})} \left[\mathbf{x} \right] \right] = 0$$

$$E_{p(\mathbf{x},y)} \left[(y - \mathbf{w}^T \mathbf{x}) (\mathbf{x} - E_{p(\mathbf{x})} \left[\mathbf{x} \right]) \right] = 0$$

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

3.