## Exercise 3

If we have a linear model

$$\mathbf{y} = \mathbf{w} \cdot \mathbf{x} \tag{1}$$

to show that

$$\mathbf{w} = \langle \mathbf{x} \mathbf{x} \rangle^{-1} \cdot \langle \mathbf{x} \mathbf{y} \rangle \tag{2}$$

we define the mean squared loss as follow

$$\mathcal{L}\left(\mathbf{x}^{\mu}\mathbf{y}^{\mu}\right) = \frac{1}{2} \sum_{\mu=1}^{n} \left(\mathbf{y}^{\mu} - \mathbf{w} \cdot \mathbf{x}^{\mu}\right)^{2}$$
(3)

where

$$\langle \mathbf{x} \mathbf{x} \rangle := \frac{1}{n} \sum_{\mu=1}^{n} \mathbf{x}^{\mu} \mathbf{x}^{\mu} \tag{4}$$

and

$$\langle \mathbf{x} \mathbf{y} \rangle := \frac{1}{n} \sum_{\mu=1}^{n} \mathbf{x} \mathbf{y}^{\mu} \mathbf{y}^{\mu} \tag{5}$$

Solution

Proof.

$$\mathcal{L}(\mathbf{x}^{\mu}\mathbf{y}^{\mu}) = \frac{1}{2} \sum_{\mu=1}^{n} (\mathbf{y}^{\mu} - \mathbf{w} \cdot \mathbf{x}^{\mu})^{2}$$

$$\frac{\partial \mathcal{L}}{\partial \mathbf{w}} = \frac{\partial}{\partial \mathbf{w}} ((\mathbf{y}^{\mu} - \mathbf{w} \cdot \mathbf{x}^{\mu})^{2})$$

$$= \frac{1}{2} \left[ (\mathbf{y} - \mathbf{w} \cdot \mathbf{x})^{2} + (\mathbf{y}^{2} - \mathbf{w} \cdot \mathbf{x}^{2})^{2} + \dots + (\mathbf{y}^{n} - \mathbf{w} \cdot \mathbf{x}^{n})^{2} \right]$$

$$= -\mathbf{x}\mathbf{y} + \mathbf{w}\mathbf{x}^{2} - \mathbf{x}^{2}\mathbf{y}^{2} + \dots + \mathbf{x}^{n}\mathbf{y}^{n} + \mathbf{w}\mathbf{x}^{2n} = 0$$

$$\mathbf{w} (\mathbf{x}^{2} + \mathbf{x}^{4} + \dots + \mathbf{x}^{2n}) = \mathbf{x}\mathbf{y} + \mathbf{x}^{2}\mathbf{y}^{2} + \dots + \mathbf{x}^{n}\mathbf{y}^{n}$$

$$\mathbf{w} = \frac{\mathbf{x}\mathbf{y} + \mathbf{x}^{2}\mathbf{y}^{2} + \dots + \mathbf{x}^{n}\mathbf{y}^{n}}{(\mathbf{x}^{2} + \mathbf{x}^{4} + \dots + \mathbf{x}^{2n})}$$

$$\mathbf{w} = \frac{\langle \mathbf{x}\mathbf{y} \rangle}{\langle \mathbf{x}\mathbf{x} \rangle}$$

$$\therefore \mathbf{w} = \langle \mathbf{x}\mathbf{x} \rangle^{-1} \langle \mathbf{x}\mathbf{y} \rangle$$