

# Hands on Machine Learning

## Linear Regression

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Given  $m$  data points, e.g.,

$$(a_{1,1}, \dots, a_{1,n-1}, b_1), \dots, (a_{m,1}, \dots, a_{m,n-1}, b_m)$$

Let  $\mathbf{a}_i = (a_{i,1}, \dots, a_{i,n})^T$ . We search for a linear function  $f$  with  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , such that,

$$\forall i \in \{1, \dots, m\} : f(\mathbf{a}_i) = x_0 + x_1 \cdot a_{i,1} + \dots x_n \cdot a_{i,n-1} = \langle \mathbf{a}_i^T, \mathbf{x} \rangle = b_i$$

In other words, we search for  $\mathbf{x}$ , such that

$$\mathbf{A}\mathbf{x} = \mathbf{b}$$

with

$$\mathbf{A} = \begin{bmatrix} 1 & a_{1,1} & \dots & a_{1,n-1} \\ 1 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ 1 & \dots & \dots & a_{m,n-1} \end{bmatrix}, \mathbf{x} = \begin{bmatrix} x_0 \\ \vdots \\ x_{n-1} \end{bmatrix}, \mathbf{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$

We want to solve the linear equation

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{b} \quad (1)$$

where  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $m > n$  (more rows than columns).

It is very likely that  $\mathbf{A}^{-1}$  does not exist and therefore, there is no solution, i.e., no  $\mathbf{x}$  satisfying the equation.

In general we search for

$$\arg \min_{\hat{\mathbf{x}} \in \mathbb{R}^n} \|\mathbf{A}\hat{\mathbf{x}} - \mathbf{b}\| = \mathbf{e} \quad (2)$$

Since  $\mathbf{e}$  is the minimal Euclidean distance  $\mathbf{A}\hat{\mathbf{x}} = \mathbf{p}$  is the projection of  $\mathbf{b}$  onto  $C(\mathbf{A})$  (column space of  $\mathbf{A}$ )

Since  $\mathbf{b} - \mathbf{A}\hat{\mathbf{x}} = \mathbf{e} \perp C(\mathbf{A})$ , it follows that

$$\mathbf{A}^T \mathbf{p} = \mathbf{A}^T (\mathbf{b} - \mathbf{A}\hat{\mathbf{x}}) = \mathbf{0} \quad (3)$$

Therefore, we solve

$$\mathbf{A}^T \mathbf{A} \hat{\mathbf{x}} = \mathbf{A}^T \mathbf{b} \quad (4)$$

for which a solution exists!

$$\hat{\mathbf{x}} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b} \quad (5)$$

We can also compute the projection matrix

$$\mathbf{P} = \mathbf{A}(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \quad (6)$$