

Algorithmic Aspects of Data Analytics and Machine Learning

SS 2025 — Sheet 1

<https://aam.uni-freiburg.de/agba/lehre/ss25/algml/index.html>

Due: May 2, 2025, 2 p.m.

Note: You must register for the exercises between April 23, 12 a.m. and April 25, 2 p.m. via HisInOne.

Task 1

(4 points)

Let $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, and $x, y \in \mathbb{R}^n$. Compute the derivative of the function

$$t \mapsto \|A(x + ty) - b\|_2^2, \quad t \in \mathbb{R},$$

and show the following:

If x minimizes $\|Ax - b\|_2^2$, then it satisfies

$$A^\top Ax = A^\top b.$$

This equation is called the (*Gauss*) *normal equation*.

Task 2

(4 points)

Given the data set $D = \{(1, 2), (2, 2), (2, 3), (3, 3), (4, 3)\} \subseteq \mathbb{R}_x \times \mathbb{R}_y$. Calculate by hand

$$f^* = \arg \min \left\{ \sum_{i=1}^n (f(x_i) - y_i)^2 \mid f : \mathbb{R} \rightarrow \mathbb{R} \text{ affine-linear} \right\}$$

and the mean values

$$\bar{x} = \frac{x_1 + \cdots + x_n}{n} \quad \text{and} \quad \bar{y} = \frac{y_1 + \cdots + y_n}{n}.$$

Verify that $f^*(\bar{x}) = \bar{y}$.

Sketch the data points and the regression line.

Task 3

(4 points)

Find a data set in which not all x -values are the same, such that the optimization problem

$$\arg \min \left\{ \sum_{i=1}^n |f(x_i) - y_i| \mid f : \mathbb{R} \rightarrow \mathbb{R} \text{ affine-linear} \right\}$$

has more than one solution. Sketch the data points and at least two different regression lines.

Then, for your chosen data set, compute the unique solution to

$$\arg \min \left\{ \sum_{i=1}^n (f(x_i) - y_i)^2 \mid f : \mathbb{R} \rightarrow \mathbb{R} \text{ affine-linear} \right\}.$$

Task 4

(4 points)

Show, using a contradiction argument, that the solution to

$$f^* = \arg \min \left\{ \sum_{i=1}^n (f(x_i) - y_i)^2 \mid f : \mathbb{R} \rightarrow \mathbb{R} \text{ affine-linear} \right\}$$

is unique.

Hint: Assume, for contradiction, that there are two minimizers $f_1^*(x) = \lambda_1 x + b_1$ and $f_2^*(x) = \lambda_2 x + b_2$. Show that for $s \in (0, 1)$

$$\sum_{i=1}^n (h(x_i) - y_i)^2 < \sum_{i=1}^n (f_1^*(x_i) - y_i)^2 = \sum_{i=1}^n (f_2^*(x_i) - y_i)^2,$$

where $h(x) := sf_1^*(x) + (1-s)f_2^*(x)$.