

Algorithmic Aspects of Data Analytics and Machine Learning

SS 2025 — Sheet 4

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

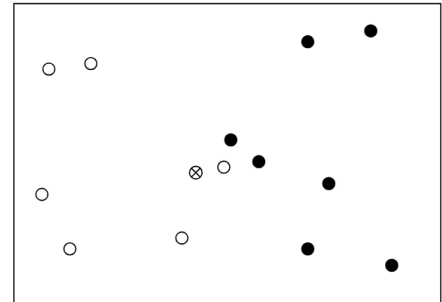
Due: May 23, 2025, 2 p.m.

Task 1

(4 points)

Consider the following dataset $D \subset \mathbb{R}^2 \times \{1, 2\}$, where white points have label 1 and black points have label 2. The label of the point marked with a cross is unknown.

Using visual inspection only, determine the label of the point marked with a cross according to the k-nearest neighbors (k-NN) rule with majority vote for $k = 1, \dots, 7$.



Task 2

(4 points)

Given the following Netflix rating matrix

$$R = \begin{pmatrix} & M_1 & M_2 & M_3 & M_4 \\ U_1 & 5 & 3 & - & 1 \\ U_2 & 4 & - & - & 1 \\ U_3 & 1 & 1 & - & 5 \\ U_4 & 1 & - & - & 4 \\ U_5 & - & 1 & 5 & 4 \end{pmatrix}.$$

With ratings given by users U_1, \dots, U_5 for movies M_1, \dots, M_4 .

Complete the matrix R using the k-nearest neighbors k-NN algorithm with $k = 2$. Use cosine similarity, with respect to the users.

Task 3

(4 points)

Let $A \subset \mathbb{R}^d$ be a finite set, and let $d : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ denote the Euclidean metric. Show that for every finite set $A \subset \mathbb{R}^d$, the following holds:

$$\frac{1}{|A|} \sum_{a \in A} a = \arg \min_{\mu \in \mathbb{R}^d} \sum_{a \in A} d(a, \mu)^2.$$

That is, the mean of the points in A uniquely minimizes the sum of squared Euclidean distances, and is thus consistent with the standard definition of the centroid in linear algebra.

Task 4

(4 points)

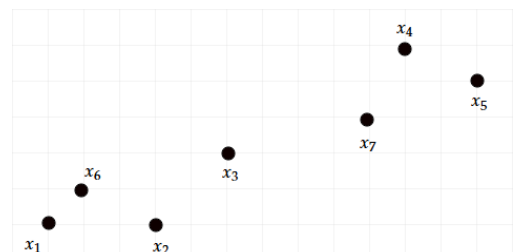
Apply the 2-means algorithm to the given dataset, using the following initialization:

$$\mu_1 = x_7, \quad \mu_2 = x_2$$

Assume the Euclidean metric on \mathbb{R}^2 . For each iteration, indicate the resulting clusterings C_i and the updated centroids μ_i (a rough estimation by eye is sufficient).

Repeat the algorithm with the initialization:

$$\mu_1 = x_1, \quad \mu_2 = x_6.$$



Practical exercise

The following exercise is not mandatory; the points are bonus points that you can collect. Please submit your solutions as a MATLAB or Python file by May 23, 2 p.m., via email to tatjana.stiefken@mathematik.uni-freiburg.de. Please comment your code and your results.

Project

(4* points)

Generate a 2D synthetic dataset with two classes:

- Class 0: points in $B_1(x) := \{x \mid |x| < 1\}$
- Class 1: points on $\partial B_1(x) = \{x \mid |x| = 1\}$.

Split the dataset into training (70%) and test (30%) sets. Train a k-NN classifier for various values of k (e.g., $k = 1, 3, 5, 10, 50$). For each k :

- Plot the decision boundary (use a mesh grid for visualization).
- Compute training and test classification errors.

Plot training and test error vs. k in a line chart. Comment your results in terms of underfitting and overfitting.

Implementation: Use either Python or MATLAB. In MATLAB, you can use `fitcknn` and visualize with `gscatter`. In Python, consider using `sklearn.neighbors.KNeighborsClassifier`.