

Name: _____ M.Number: _____

KDDM1 VO (INP.31101UF)

1. *Distance Functions*. Given the dataset “distance-function-dataset.csv” (available in TeachCenter), select or develop a suitable distance function to compare instances (row), based on the values of the features (columns).
 - (a) On what observations from the dataset do you base your decisions? (bullet points)
 - (b) Would you conduct any additional feature engineering steps?
 - (c) What distance function do you choose? (in case of a custom one, please provide a description/pseudocode/...)
 - (d) Would the distance function depend on the succeeding processing, e.g., different function for PCA, DBSCAN, or SVM?

2. *Dimensionality Reduction.* Consider a dataset of 100 dimensions/features (real numbers), and the goal is to derive a 2D visualisation of the dataset.

- (a) What would be a suitable approach if the dependencies in the data are all linear?
- (b) What would be a suitable approach if there are density-based local structures in the data?
- (c) What would be a suitable approach if most of the features are Gaussian noise?
- (d) What types of noise are there and how do they affect the dimensionality reduction?

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3. *Clustering*. Given the dataset “clustering-dataset.csv” (available in TeachCenter), which consists of observations of 5 dimensions, the goal is to find the groups of rows that form clusters.

(a) Which methods did you apply to find the clusters, and why? (bullet points)

- Describe pre-processing steps (if conducted)
- Describe what distance measures you have chosen
- Describe how you determine the number of clusters

(b) What clusters did you find and how would you describe the distribution of the points within each cluster?

- Describe each found cluster, including shape, and amount of points within the cluster.

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4. *Classification.* Select 3 classification algorithms of your choice that should be diverse (i.e., not based on the same underlying principles).
- (a) For each algorithm list the main assumptions (e.g., on the data characteristics, types of dependencies). (bullet points)
 - (b) For each algorithm list 1-2 application scenarios, where these assumptions are being met.