

## Home assignment 1: Distance function, classification, clustering.

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### Exercise 1: Distance functions.

Implemented following functions: chebyshev, manhattan, canberra, euclidean, minkowski, mahalanobis and generic distance function that incorporates all previously mentioned distance functions. All functions work on n-dimensional (n-column) matrixes as inputs.

E.g: euclidean <- function(x1, x2){ return(sqrt(rowSums((x1 - x2) ^ 2))) }

### Exercise 2: Clustering.

Implemented k-means algorithm, represented by the following pseudocode:

```
kmeans_clusters <- function(data, num_clusters, num_iterations, metric):  
  Start iterating until solution.  
  for (1 to num_iterations){  
    1. Initialize random center points for clusters or store new centers.  
    2. Assign each point to a cluster.  
    3. Calculate center of each cluster.  
    4. Check if solution. If found solution then exit function.
```

For cluster analysis implemented following functions:

intra\_cluster\_distances, inter\_cluster\_distances, intra\_to\_inter\_ratio, silhouette\_coefficient, silhouette\_score, cluster\_inertia.

### Exercise 3: Classification.

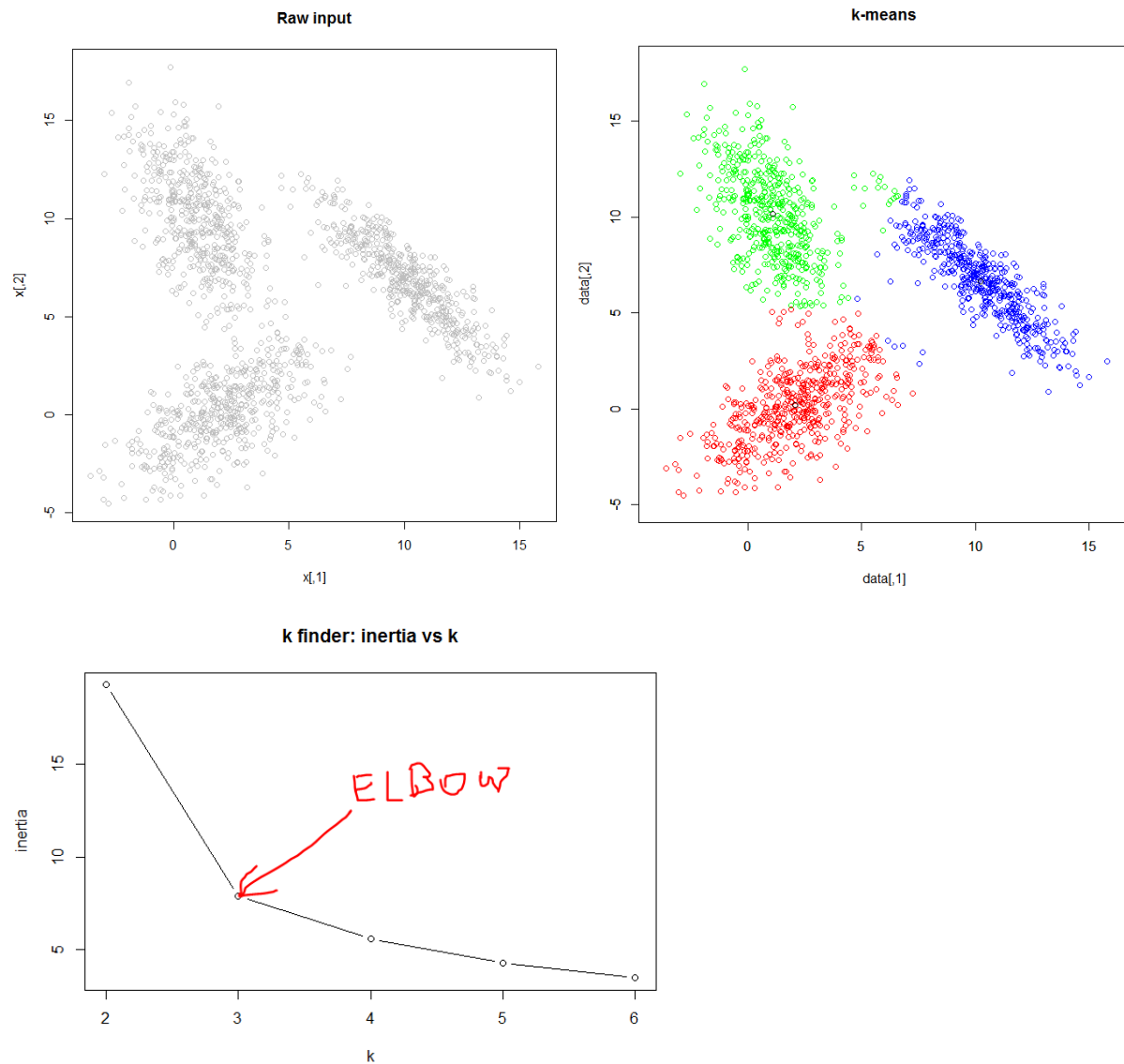
Implemented knneighbors algorithm. For each datapoint we calculate distances to all other datapoints and take most popular label among k closest datapoints.

For classification analysis I used same analytics as for cluster analysis. On top of that I implemented functions: accuracy\_score, confusion\_matrix function.

### Exercise 4: Classification wrapper.

In order to find optimal k for knneighbors algorithm, I implemented an iterative function that finds “best” classification results for different k-values. Best classification result can be determined based on different analytics/scores: accuracy\_score, cluster\_inertia, silhouette\_score etc. I applied the similar analysis for clustering using same techniques in finding optimal k for kmeans algorithm (“elbow” rule, using inertia as clustering score).

**Figure 1: k-means: from raw data to 3 clusters using “elbow” rule.**



## Appendix:

Source: <https://gitlab.cs.ttu.ee/urpits/data-mining-iti8730/tree/master/assignment1>

### List of source files:

cluster\_analysis.R  
clusters\_generator.R  
demo\_cluster\_analysis.R  
demo\_mahalanobis.R  
distances.R