

Data Analytics – Exercises

(Week 10)

In these exercises, you will learn:

- to perform cluster analyses using k-means clustering,
- to determine the optimal number of clusters (k) using the elbow method,
- to evaluate k-means clustering models,
- to visualize the results of k-means clustering.

In the data analytics process model, these exercises cover part of the steps “Statistical data analysis and/or Modeling” and “Evaluation & Interpretation” (see figure 1). Results of the exercises must be uploaded as separate files (**no .zip files!**) by each student on Moodle. Details on how to submit the results can be found in the tasks below.

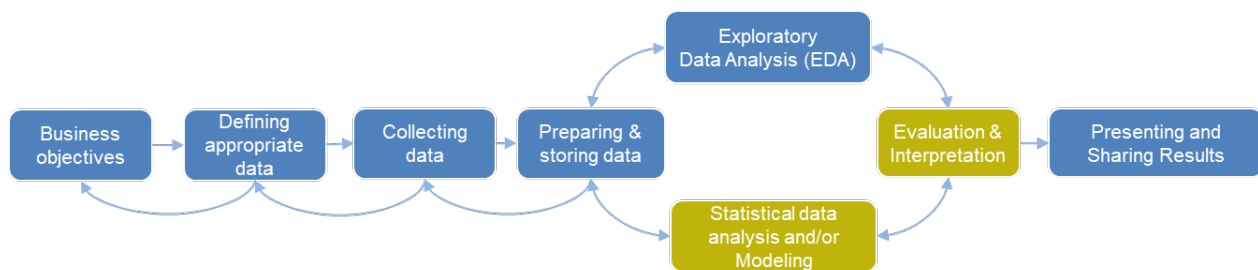


Figure 1: Data analytics process model (see slides of week 01)

Task 1

In this exercise, you will learn to determine the optimal number of clusters for a k-means clustering model by using the elbow method. In addition, you will learn to perform image segmentation and 3D point cloud segmentation based on k-means clustering.

- a) Install the following additional Python libraries from within your Jupyter notebook:

```
!pip install opencv-python
!pip install plyfile
```

- b) Run the Jupyter notebook '[k-means_clustering.ipynb](#)' step by step and try to find out, what the Python code does.
- c) In section '1.) Simple k-means clustering example', go to the subsection with the elbow method. In the Jupyter notebook, based on the elbow chart, state which is the optimal k value for this clustering model.
- d) In the subsection 'Perform Silhouette Analysis', perform a Silhouette Analysis for different values of k . Use $k = 2, 3, 4, 5, 6, 7, 8, 9, 10, 11$. You can implement this in the `for` loop by replacing the code line:

```
for i in [2,3]:
    by
    for i in range(2,12,1):
```

- e) Go to the section '2.) Image segmentation'. In this example, an image with a parrot is used for image segmentation. Find another .jpg image (e.g., by using Google's image search) and run the Jupyter notebook to perform image segmentation for your own image.
- f) Use the elbow method to determine the optimal number of clusters k for the segmentation of your image.
- g) Create 4 different image segmentations for your image based on different numbers of clusters k and plot the images.
- h) Go to the section '3.) 3D point cloud segmentation' and look at the segmentation example based on the airport data.
- i) Open the file '3d_point_cloud_excavator.txt' from Moodle and copy the content to a new cell in your Jupyter notebook. Run the cell, then change the value for k and look what happens.

To be submitted on Moodle:

- The Jupyter notebook as html-file '[k-means_clustering_task01.html](#)' with the changes and short explanations according to c), d), e), f), g), h), and i).

Task 2

In this exercise, you will learn to perform k-means clustering based on the apartments data.

- a) Go to the section '4.) Finding clusters in the apartments data ...'. In the prepared example, the variables `rooms`, `area` and `price_per_m2` are used to form the clusters.
- b) Go to the subsection 'Subset of the apartment data ...' and extend the data frame which is used for k-means clustering (name = 'X3') by including additional numerical variables (e.g. `lat`, `lon`, `pop_dens`, `tax_income`, ...).
- c) Use the elbow method to find the optimal number of clusters k for the extended data frame (X3).
- d) Perform k-means clustering based on the optimal number of clusters. Use the (already included) name '`kmeans_apmts`' as the name for the k-means model.
- e) Use the following Python code to derive the attribute values from '`kmeans_apmts`':

```
print(kmeans_apmts.labels_, '\n')
print(kmeans_apmts.inertia_, '\n')
print(kmeans_apmts.cluster_centers_, '\n')
print(kmeans_apmts.feature_names_in_)
```

- f) In the Jupyter notebook, explain the meaning of the output of the code in e). You can find the required information in the sklearn.cluster documentation on this [Link](#) under "Attributes".
- g) Calculate the Silhouette Score for 'kmeans_apmts'. In the Jupyter notebook, state whether the optimal value for k suggested by the elbow method is consistent with the value for k showing the highest Silhouette Score.

To be submitted on Moodle:

The Jupyter notebook as html-file '[k-means_clustering_task02.html](#)' with the changes and short explanations according to b), c), d), e), f) and g).