## ingenieur wissenschaften htw saar

Hochschule für Technik und Wirtschaft des Saarlandes University of

**Applied Sciences** 

Machine Learning Summer 2023 Prof. Dr.-Ing. Klaus Berberich

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## Programming Assignment 3

The programming assignment will be discussed on **July 10th**, **2023**. To obtain bonus points, you have to submit your solution as a Zip-file via Moodle by **July 7th**, **2023 23:59**. Teams of up to **three** students are allowed. Please indicate your names and your student numbers in your submission.

## 3.1 Clustering News Articles (5.0)

For this programming assignment, we are going to work with a collection of about 40K news articles from CNN. More details about the dataset are available on Kaggle at:

https://www.kaggle.com/datasets/hadasu92/cnn-articles-after-basic-cleaning

You can download the dataset from there or from Moodle.

- (a) As a first step, we need to import the dataset and transform it into a suitable representation.
  - (i) Import the dataset into a pandas DataFrame and transform the column ArticleText using the TfidfVectorizer with its default settings.
  - (ii) To get an idea for a reasonable number of clusters to aim for, determine how often each value of the two columns Category and Section occurs in the dataset.
- (b) Now, we apply **k-Means** to the dataset.
  - (i) Compute two clusterings using k-Means for values of  $k \in \{6, 37\}$ .
  - (ii) For both clusterings, determine for each cluster the data point that is closest to the centroid. Output the text of these articles as a description of the clusters.
  - (iii) Visualize the two determined clusterings in two-dimensional scatter plots. For each cluster, data points should be drawn in a separate color. To map the data points onto two dimensions, please make use of t-SNE (available in sklearn.manifold.TSNE) as a dimensionality reduction technique.



- (c) Now, we apply **DBScan** to the dataset.
  - (i) Determine two choices of the hyperparameters  $\epsilon$  and minPoints, so that the resulting number of clusters is in [5, 10] and [30, 40], respectively.
  - (ii) Visualize the two determined clusterings in two-dimensional scatter plots. For each cluster, data points should be drawn in a separate color. To map the data points onto two dimensions, please make use of t-SNE (available in sklearn.manifold.TSNE) as a dimensionality reduction technique.
- (d) Finally, we want to quantitatively evaluate the four clusterings computed in (b) and (c).
  - (i) For each of the four clusterings, determine its Homogeneity score (as implemented in sklearn.metrics.homogeneity\_score). Homogeneity is another external quality measure. For the smaller (larger) clusterings use Section and Category as a ground-truth labeling (idealized clustering).
  - (ii) For each of the four clusterings, determine its Silhouette score (as implemented in sklearn.metrics.silhouette\_score). Silhouette is another internal quality measure.