IT 350

Assignment 4

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Batch no.: 7

Link to Colab notebook - https://colab.research.google.com/drive/1r7tHEf3k3quMNWdz1dZiAWKD6wwD5UWL?usp=sharing)

1. Find the clusters in the given dataset based on the content similarity and image similarity using k-means clustering and hierarchical clustering methods.

```
In [2]: import pytesseract
        from glob import glob
        import pandas as pd
        from PIL import Image
        from pprint import pprint
        from google.colab import drive
        drive.mount('/content/drive')
        images = glob("/content/drive/MyDrive/Assignment 4 Clustering/Q2_Clusters/Cluster_1/*")
        pprint(images)
        data = pd.DataFrame({"FileName": images})
        imageText = []
        for img in images:
            text = pytesseract.image_to_string(Image.open(img))
            imageText.append(text.replace("\n", ""))
        data["Text"] = imageText
        pprint(imageText)
```

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Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/
drive", force remount=True).
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In [3]: from sklearn.feature extraction.text import CountVectorizer
        vectorizer = CountVectorizer()
        X = vectorizer.fit transform(data["Text"])
In [4]: # K-means clustering
        from sklearn.cluster import KMeans
```

data["Kmeans"] = KMeans(n clusters=5, random state=0).fit predict(X.toarray())

```
In [5]: data
```

Out[5]:

	FileName	Text	Kmeans
0	/content/drive/MyDrive/Assignment 4 Clustering		0
1	/content/drive/MyDrive/Assignment 4 Clustering		0
2	/content/drive/MyDrive/Assignment 4 Clustering	be thee Wah amahinWO) 2 Unt wod ¢WOW) F2uta wo	0
3	/content/drive/MyDrive/Assignment 4 Clustering	1 a hash. jumdioo hE) tush Saksee	0
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9	/content/drive/MyDrive/Assignment 4 Clustering	Pet Gaatang Cnet neei ? oN a SShtite3 / ate	3
10	/content/drive/MyDrive/Assignment 4 Clustering		0
11	/content/drive/MyDrive/Assignment 4 Clustering		0
12	/content/drive/MyDrive/Assignment 4 Clustering		0

In [6]: # Agglomerative clustering

```
from sklearn.cluster import AgglomerativeClustering
data["Agglomerative"] = AgglomerativeClustering(n_clusters=4).fit_predict(X.toarray())
```

In [7]: data

Out[7]:

	FileName	Text	Kmeans	Agglomerative
0	/content/drive/MyDrive/Assignment 4 Clustering		0	0
1	/content/drive/MyDrive/Assignment 4 Clustering		0	0
2	/content/drive/MyDrive/Assignment 4 Clustering	be thee Wah amahinWO) 2 Unt wod ¢WOW) F2uta wo	0	0
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5	/content/drive/MyDrive/Assignment 4 Clustering	wnee aneweyQorper if te3Fist moth whae diate	4	3
6	/content/drive/MyDrive/Assignment 4 Clustering	@@) Gfveur , Hirer har functions ,AU) = A141 m	1	1
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8	/content/drive/MyDrive/Assignment 4 Clustering	bP	0	0
9	/content/drive/MyDrive/Assignment 4 Clustering	Pet Gaatang Cnet neei ? oN a SShtite3 / ate	3	0
10	/content/drive/MyDrive/Assignment 4 Clustering		0	0
11	/content/drive/MyDrive/Assignment 4 Clustering		0	0
12	/content/drive/MyDrive/Assignment 4 Clustering		0	0

2. Plot t-SNE visualization for derived clusters.

```
In [8]: import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.manifold import TSNE
import numpy as np
X_embedded = TSNE(n_components=2, init='pca').fit_transform(X.toarray().astype("float"))
X_pd = pd.DataFrame(X_embedded)
```

/usr/local/lib/python3.7/dist-packages/sklearn/manifold/_t_sne.py:793: FutureWarning: The default le arning rate in TSNE will change from 200.0 to 'auto' in 1.2.

FutureWarning,

/usr/local/lib/python3.7/dist-packages/sklearn/manifold/_t_sne.py:986: FutureWarning: The PCA initia lization in TSNE will change to have the standard deviation of PC1 equal to 1e-4 in 1.2. This will e nsure better convergence.

FutureWarning,

In [9]: X_pd

Out[9]:

0	-204.667938	-552.070923
1	-204.667938	-552.070923
2	-1438.736450	118.308029
3	-1037.713257	1094.138794
4	-37.757725	650.830383
5	1003.618408	1080.078369
6	840.637329	-1423.135132
7	1159.793213	-102.525795
8	1159.793213	-102.525795
9	-8.582541	1675.383179
10	-204.667938	-552.070923
11	-204.667938	-552.070923
12	-1415.424561	-1086.230591

0

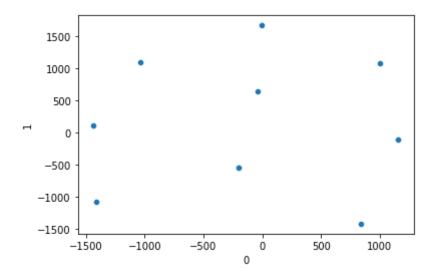
1

```
In [10]: sns.scatterplot(X_pd[0], X_pd[1])
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data `, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Out[10]: <matplotlib.axes. subplots.AxesSubplot at 0x7febadd4c750>



3. Evaluate the clusters that are obtained using appropriate methods.

Performed clustering of answer sheet using pytesseract to recognize OCR text from images, and then used the extracted text to perform clustering. I used 2 types of clustering techniques K-means and Agglomerative clustering. Agglomerative clustering shows a wider distribution of papers into clusters than K-means. Since OCR detection is extremely poor and fails to identify text, I proceeded in checking similarity using image features.