

18CS10069_Assignment1_Q10

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NOTE: The answers to the questions can be found in the text of the this document.

```
[13]: import pandas as pd
import wikipedia
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.cluster import KMeans
```

```
articles= ['Linear algebra',
'Data Science',
'Artificial intelligence',
'European Central Bank',
'Financial technology',
'International Monetary Fund',
'Basketball',
'Swimming',
'Cricket']
```

```
wiki_lst, title = [], []
for article in articles:
    wiki_lst.append(wikipedia.page(article).content)
    title.append(article)
```

```
[14]: vectorizer = TfidfVectorizer(stop_words={'english'})
X = vectorizer.fit_transform(wiki_lst)
```

```
[15]: print(f' +++++ Shape of the input data: {X.shape}')
```

```
+++++ Shape of the input data: (9, 8285)
```

0.0.1 Ans 10 (a),(b) with k=4

```
[43]: # kmeans with k=4
kmeans4 = KMeans(n_clusters=4, random_state=0)
kmeans4 = kmeans4.fit(X)
print(f'+++++ Jclust value: {kmeans4.inertia_/X.shape[0]}')
print(f'+++++ Finding Document Association +++++')
for i, val in enumerate(kmeans4.labels_):
    print(f"\t +++++ Article Name: {articles[i]} {'-' * (30 - len(articles[i])) } Cluster: {kmeans4.labels_[i]}")

+++++ Jclust value: 0.20494844342027826
+++++ Finding Document Association +++++
+++++ Article Name: Linear algebra ----- Cluster: 2
+++++ Article Name: Data Science ----- Cluster: 1
+++++ Article Name: Artificial intelligence ----- Cluster: 0
+++++ Article Name: European Central Bank ----- Cluster: 0
+++++ Article Name: Financial technology ----- Cluster: 0
+++++ Article Name: International Monetary Fund --- Cluster: 0
+++++ Article Name: Basketball ----- Cluster: 0
+++++ Article Name: Swimming ----- Cluster: 3
+++++ Article Name: Cricket ----- Cluster: 0
```

0.0.2 Ans 10 (a),(b) with k=8

```
[44]: # kmeans with k=8
kmeans8 = KMeans(n_clusters=8, random_state=0)
kmeans8 = kmeans8.fit(X)
print(f'+++++ Jclust value: {kmeans8.inertia_/X.shape[0]}')
print(f'+++++ Finding Document Association +++++')
for i, val in enumerate(kmeans8.labels_):
    print(f"\t +++++ Article Name: {articles[i]} {'-' * (30 - len(articles[i])) } Cluster: {kmeans8.labels_[i]}")

+++++ Jclust value: 0.025551532806781845
+++++ Finding Document Association +++++
+++++ Article Name: Linear algebra ----- Cluster: 2
+++++ Article Name: Data Science ----- Cluster: 1
+++++ Article Name: Artificial intelligence ----- Cluster: 5
+++++ Article Name: European Central Bank ----- Cluster: 0
+++++ Article Name: Financial technology ----- Cluster: 4
+++++ Article Name: International Monetary Fund --- Cluster: 7
+++++ Article Name: Basketball ----- Cluster: 6
+++++ Article Name: Swimming ----- Cluster: 3
+++++ Article Name: Cricket ----- Cluster: 6
```

0.0.3 Ans 10 (c)

- With the help of only `Jclust` values, from comparison we can say that since it is lower for $k=8$, hence 8 is the optimal number of clusters.