clustering-Text-K-MEANS-NLP

March 23, 2024

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[1]: from sklearn.datasets import fetch_20newsgroups
     from nltk.tokenize import word_tokenize #Used to extract words from documents
     from nltk.stem import WordNetLemmatizer #Used to lemmatize words
     from sklearn.feature_extraction.text import TfidfVectorizer
     from sklearn.pipeline import make_pipeline
     from sklearn.preprocessing import Normalizer
     from sklearn import metrics
     from sklearn.cluster import KMeans
     import sys
     from time import time
     import pandas as pd
     import numpy as np
[2]: # Selected 3 categories from the 20 newsgroups dataset
     categories = [
         'talk.religion.misc',
         'comp.graphics',
         'sci.space',
     ]
     print("Loading 20 newsgroups dataset for categories:")
     print(categories)
    Loading 20 newsgroups dataset for categories:
    ['talk.religion.misc', 'comp.graphics', 'sci.space']
[3]: df = fetch_20newsgroups(subset='all', categories=categories,
                                  shuffle=False, remove=('headers', 'footers', __

¬'quotes'))
[4]: labels = df.target
     true_k = len(np.unique(labels)) ## This should be 3 in this example
     print(true_k)
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[5]: #Perform Lemmatization
 [6]: lemmatizer = WordNetLemmatizer()
      for i in range(len(df.data)):
          word_list = word_tokenize(df.data[i])
          lemmatized doc = ""
          for word in word list:
              lemmatized_doc = lemmatized_doc + " " + lemmatizer.lemmatize(word)
          df.data[i] = lemmatized_doc
 [7]: print(df.data[1])
      In regard to fractal commpression , I have seen 2 fractal compressed `` movie
     '' . They were both fairly impressive . The first one wa a 64 gray scale ``
     movie '' of Casablanca , it wa 1.3MB and had 11 minute of 13 fps video . It wa a
     little grainy but not bad at all . The second one I saw wa only 3 minute but it
     had 8 bit color with 10fps and measured in at 1.2MB . I consider the fractal
     movie a practical thing to explore . But unlike many other format out there,
     you do end up losing resolution . I do n't know what kind of software/hardware
     wa used for creating the `` movie '' I saw but the guy that showed them to me
     said it took 5-15 minute per frame to generate . But a I said above playback wa
     10 or more frame per second . And how else could you put 11 minute on one floppy
     disk ?
 [8]: #We next convert our corpus into tf-idf vectors. We remove common stop words,
       sterms with very low document frequency (many of them are numbers
      #or misspells), accents.
 [9]: vectorizer = TfidfVectorizer(strip_accents='unicode', stop_words='english', u
       ⇒min_df=2) ## Corpus is in English
      X = vectorizer.fit_transform(df.data)
[10]: print(X.shape)
     (2588, 14439)
[11]: #Clustering using standard k-means¶
[12]: #We choose a value that reflects our knowledge about the data
[13]: #We may try several value, possibly in increasing order
[14]: | #we set k=3 |
[15]: km = KMeans(n_clusters=true_k, init='k-means++', max_iter=100)
      t0 = time()
      km.fit(X)
      print("done in %0.3fs" % (time() - t0))
```

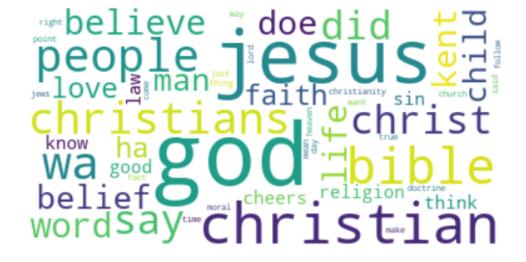
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done in 0.162s
     /home/nmit/anaconda3/lib/python3.11/site-
     packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of
     `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init`
     explicitly to suppress the warning
       warnings.warn(
[16]: #Standard measures of cluster quality¶
[17]: print("Homogeneity: %0.3f" % metrics.homogeneity_score(labels, km.labels_))
      print("Completeness: %0.3f" % metrics.completeness_score(labels, km.labels_))
      print("V-measure: %0.3f" % metrics.v_measure_score(labels, km.labels_))
      print("Adjusted Rand-Index: %.3f"
            % metrics.adjusted_rand_score(labels, km.labels_))
      print("Silhouette Coefficient: %0.3f"
            % metrics.silhouette_score(X, km.labels_, sample_size=1000))
     Homogeneity: 0.293
     Completeness: 0.401
     V-measure: 0.339
     Adjusted Rand-Index: 0.210
     Silhouette Coefficient: 0.009
[18]: #Identify the 10 most relevant terms in each cluster¶
[20]: centroids = km.cluster_centers_.argsort()[:, ::-1] ## Indices of largest_
      ⇔centroids' entries in descending order
      #vectorizer.get_feature_names use instead vectorizer.get_feature_names_out
      terms = vectorizer.get_feature_names_out()
      for i in range(true_k):
          print("Cluster %d:" % i, end='')
          for ind in centroids[i, :10]:
              print(' %s' % terms[ind], end='')
          print()
     Cluster 0: god jesus christian bible wa people christians say christ did
     Cluster 1: wa space just like think ha time know did people
     Cluster 2: file image thanks format program know graphic ftp bit gif
[21]: #Visualization
[22]: from wordcloud import WordCloud
      import matplotlib.pyplot as plt
[23]: def frequencies dict(cluster index):
          if cluster_index > true_k - 1:
          term_frequencies = km.cluster_centers_[cluster_index]
```

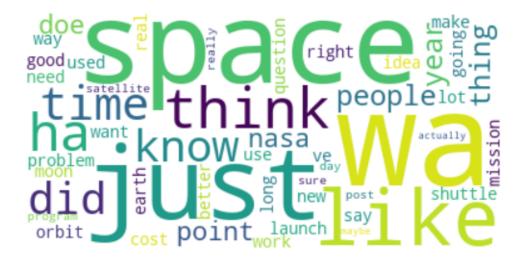
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sorted_terms = centroids[cluster_index]
frequencies = {terms[i]: term_frequencies[i] for i in sorted_terms}
return frequencies
```

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[24]: def makeImage(frequencies):
    wc = WordCloud(background_color="white", max_words=50)
    # generate word cloud
    wc.generate_from_frequencies(frequencies)

# show
    plt.imshow(wc, interpolation="bilinear")
    plt.axis("off")
    plt.show()
[25]: for i in range(true_k):
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[25]: for i in range(true_k):
    freq = frequencies_dict(i)
    makeImage(freq)
    print()
```







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