```
In [83]:

from sklearn.feature_extraction.text import TfidfVectorizer import pandas as pd import matplotlib.pyplot as plt import numpy as np import matplotlib.cm as cm from collections import defaultdict

import nltk from nltk.corpus import stopwords from nltk.stem import WordNetLemmatizer import random
```

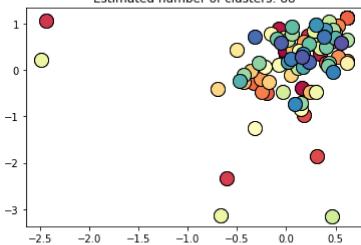
```
In [79]:
           lemmatizer = WordNetLemmatizer()
           def lemma (word):
               word = word. strip()
               safe = ["ns"]
               if word in safe:
                   return word
               noun = lemmatizer. lemmatize(word)
               if noun != word:
                   return noun
               adjective = lemmatizer.lemmatize(word, pos="a")
               if adjective != word:
                   return adjective
               verb = lemmatizer. lemmatize(word, pos="v")
               if verb != word:
                   return verb
               return word
           stopwordDict = defaultdict(int)
           for sw in stopwords.words("english"):
               stopwordDict[sw] += 1
```

```
continue
                   if "Other Languages" in page["labels"]:
                       continue
                     date = page["created date"].split(",")
                     year = date[1][1:5]
           #
                     if year != "2021" and year != "2020":
           #
                         continue
                   unclassified = True
                   if unclassified:
                       unprocessed, processed = page["original post"], []
                       for word in unprocessed. split(" "):
                           lem = lemma(word.strip().lower())
                           try:
                               _{-} = int(lem)
                               continue
                           except:
                               pass
                           if lem not in stopwordDict:
                               flag = False
                               for char in lem:
                                   if char < "a" or char > "z":
                                       flag = True
                               if not flag:
                                   processed. append (lem)
                                   vocab[lem] = 1
                       corpus. append(" ". join(processed))
                       corpus_index. append (page["title"])
                       url_list.append(page["url"])
           print(corpus)
           print(len(corpus))
           from sklearn.feature_extraction.text import CountVectorizer
           vocabList = vocab.keys()
           #cv1 = CountVectorizer(vocabulary=vocabList, analyzer = 'word', ngram_range=(1,2))
           cv1 = CountVectorizer(vocabulary=vocabList, analyzer = 'word')
           corpus vocab count matrix = cv1. transform(corpus)
           from sklearn.feature_extraction.text import TfidfTransformer
           tfidf transformer = TfidfTransformer(smooth idf=True, use idf=True)
           tfidf_transformer.fit(corpus_vocab_count_matrix)
Out[23]: TfidfTransformer()
In [24]:
           df idf = pd. DataFrame(tfidf transformer.idf, index=cv1.get feature names(), columns=
           df idf = df idf. sort values(by=['idf-weights'])
           print(df_idf.head(10))
                      idf-weights
          cloudflare
                         1.428240
                         1.610688
```

```
1.782482
          domain
                         2.051258
          use
                         2.068365
          com
                         2.211908
          record
                         2.227520
          get.
          thank
                         2.256599
          work
                         2.262401
                         2.291928
          server
           articles vocab count matrix = cv1. transform(corpus)
           tfidf_matrix = tfidf_transformer.transform(articles_vocab_count_matrix)
   [26]:
           tfidf array = np. asarray(tfidf matrix. todense())
           df = pd. DataFrame(tfidf_array[0], index=cv1.get_feature_names(), columns=['tfidf'])
           df_descending = df. sort_values(by=['tfidf'], ascending=False)
           print(df_descending. head(10))
                         tfidf
                      0.455808
          cornel1
                      0.312971
          edu
                      0.274808
          query
          confluence 0.265888
                      0.252202
          opt.
                      0.237834
          contegix
                      0.218305
          flag
                      0.196958
          answer.
                      0.149258
          est
                      0.145079
          dec
           from sklearn.metrics import pairwise distances
           from scipy. spatial. distance import cosine
           distance array = pairwise distances(tfidf array, metric='cosine')
   [59]:
           distance array
Out[59]: array([[-27.19565918,
                                  0.76632044,
                                                 0.5809683, ..., -0.87197672,
                    0.31689633,
                                  0.57934432],
                   0. 18277863, -21. 75420146,
                                                 0.03796774, \ldots, -2.41995114,
                   -0.95927878,
                                  0.37833943],
                 [ 0.0375315 ,
                                  0.09069644, -21.2904778, ..., -0.51003695,
                                  0.57934432],
                   -1.3721348 ,
                                               -0.12230912, ..., -23.73957429,
                 -1.85603471,
                                  0.57934432],
                    0.75989932,
                 [ -0.29794515,
                                 -0.60088823.
                                               -0. 94421591, ...,
                                                                    0.32749841,
                  -28.73846948,
                                  0.57934432],
                 0.61148409,
                                  0.93072739.
                                                 1.04191954, ...,
                                                                    0.76756499,
                    1. 29927718, -23. 84607878]])
In [41]:
           #distance_array = StandardScaler().fit_transform(distance_array)
           import numpy as np
           from sklearn.cluster import DBSCAN
           from sklearn import metrics
           from sklearn.datasets import make blobs
           from sklearn.preprocessing import StandardScaler
           clustering = DBSCAN(eps=0.6, min samples=2). fit(distance array)
```

```
for i in range(max(clustering.labels_)):
                print(np. where(clustering. labels_==i)[0])
           [ \quad 120 \quad 1219 \quad 1776 \quad 3116 \quad 4977 \quad 6466 \quad 6701 \quad 7503 \quad 7580 \quad 8313 \quad 8721 \quad 8864
            10101
           [ 2039 9813 11947]
           [2729 5499 6883]
           core_samples_mask = np. zeros_like(clustering. labels_, dtype=bool)
            core samples mask[clustering.core sample indices] = True
            labels = clustering.labels_
            labels
Out[73]: array([-1, -1, -1, ..., -1, -1, -1], dtype=int64)
In [74]:
            n_{clusters} = len(set(labels)) - (1 if -1 in labels else 0)
            n_{noise} = list(labels).count(-1)
In [47]:
            import matplotlib.pyplot as plt
            unique labels = set(labels)
            colors = [plt. cm. Spectral(each) for each in np. linspace(0, 1, len(unique labels))]
            for k, col in zip(unique_labels, colors):
                if k == -1:
                    # Black used for noise.
                    continue
                    co1 = [0, 0, 0, 1]
                class member mask = labels == k
                xy = distance array[class member mask & core samples mask]
                plt.plot(
                    xy[:, 0],
                    xy[:, 1],
                    ″o″,
                    markerfacecolor=tuple(col),
                    markeredgecolor="k",
                    markersize=14,
                )
                xy = distance_array[class_member_mask & ~core_samples_mask]
                plt. plot (
                    xy[:, 0],
                    xy[:, 1],
                    markerfacecolor=tuple(col),
                    markeredgecolor="k",
                    markersize=6,
            plt. title ("Estimated number of clusters: %d" % n_clusters_)
            plt. show()
```





In [48]: metrics.silhouette_score(distance_array, labels)

Out[48]: -0.20606649528416796

In []: