```
In [1]:
         from collections import Counter
         from scipy.sparse import csr matrix
         from collections import defaultdict
         from sklearn.cluster import KMeans
         from numpy import *
         %matplotlib inline
         import matplotlib.pyplot as plt
         def build matrix 1(docs, labels):
             distinctWordsAndIndex = {}
             indexIter = 0
             nnz = 0
             for idx, doc in enumerate(docs):
                 frequency = doc.split()
                 while frequency:
                     term, freq, *frequency = frequency
                     if term not in labels:
                         continue
                     nnz += 1
                     if term not in distinctWordsAndIndex:
                         distinctWordsAndIndex[term] = indexIter
                         indexIter += 1
             nrows = len(docs)
             ncols = len(distinctWordsAndIndex)
             ind = np.zeros(nnz, dtype=int)
             val = np.zeros(nnz, dtype=int)
             ptr = np.zeros(nrows+1, dtype=int)
             i = 0
             j = 0
             for idx, doc in enumerate(docs):
                 ptr[j] = i
                 i += 1
                 frequency = doc.split()
                 while frequency:
                     term, freq, *frequency = frequency
                     if term not in labels:
                         continue
                     ind[i] = distinctWordsAndIndex[term]
                     val[i] = int(freq)
                     i += 1
             ptr[j] = i
             mat = csr matrix((val, ind, ptr), shape=(nrows, ncols), dtype=np.double)
             mat.sort indices()
             return mat
```

```
In [2]: #pre-processing
#Filter labels with length <= 3 and is present in stop words collection

import nltk
from nltk.corpus import stopwords

def getValidWords():
    stop_words = stopwords.words('english')
    with open("train.clabel", "r", encoding="utf8") as fh:
    labels = {}</pre>
```

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for idx, word in enumerate(fh.readlines()):
    if len(word.rstrip()) < 4 or word.rstrip() in stop_words:
        continue
    labels[str(idx+1)] = word.rstrip()
return labels
# print(labels)</pre>
```

```
In [3]:
        def BisectingKMeans(mat4, k start, k end, step):
            for k in range(k_start, k_end+1, step):
                print('----')
                X = mat4
                k list = []
                sse list = []
                total SSE = 0
                current clusters = 1
                clusterMap = {}
                for idx,row in enumerate(X):
                    clusterMap[idx] = idx
                finalClusterLabels = {}
                while current clusters != k:
                    kmeans = KMeans(n clusters=2, n init=50).fit(X)
                    cluster centers = kmeans.cluster centers
                    sse = [0]*2
                    for point, label in zip(X, kmeans.labels ):
                        sse[label] += np.square(point-cluster centers[label]).sum()
                    chosen cluster = np.argmax(sse, axis=0)
                    total SSE += sse[np.argmin(sse, axis=0)]
                    chosen cluster data = X[kmeans.labels == chosen cluster]
                    newClusterMap = {}
                    clusterIter = 0
                    #to keep track of the index of the clusters
                    for idx, x in enumerate(kmeans.labels ):
                       if(x != chosen cluster):
                           finalClusterLabels[clusterMap[idx]] = current clusters
                        elif current clusters + 1 == k:
                           finalClusterLabels[clusterMap[idx]] = current clusters +
                       else:
                           newClusterMap[clusterIter] = clusterMap[idx]
                           clusterIter += 1
                    clusterMap = newClusterMap
                    current clusters += 1
                    assigned cluster data = X[kmeans.labels != chosen cluster]
                    X = chosen_cluster_data
                k list.append(k)
                sse list.append(kmeans.inertia )
                print_internal_metrics(mat4, finalClusterLabels)
                print('----')
            return finalClusterLabels
```

```
In [4]: #Internal Metrics
from sklearn import metrics

def print_internal_metrics(mat, labels_dict):
    labels = [labels_dict[key] for key in sorted(labels_dict.keys())]
    n_clusters_ = len(set(labels)) - (1 if -1 in labels else 0)

    print('K: %d' % n_clusters_)
    print("Silhouette Coefficient: %0.3f" % metrics.silhouette_score(mat, lake print("Calinski Harabasz Score: %0.3f" % metrics.calinski_harabasz_score()
```

```
import numpy as np
In [5]:
        from sklearn.datasets import make blobs
        from sklearn.decomposition import TruncatedSVD
        from sklearn.feature extraction.text import TfidfTransformer
        from datetime import datetime
         #Read the file
        with open("train.dat", "r", encoding="utf8") as fh:
            rows = fh.readlines()
         #Select valid words
        valid words = getValidWords()
         #build csr matrix from valid words only
        mat1 = build matrix 1(rows, valid words)
         # TF-IDF transform to normalise the matrix
        tfidf transformer=TfidfTransformer(smooth idf=True, use idf=True)
        tf idf vector=tfidf transformer.fit(mat1).transform(mat1)
         #print CST information
        print('Shape before SVD', tf idf vector.shape)
        print("Start Time =", datetime.now().strftime("%H:%M:%S"))
         # components = 200
         # while components < 501:
              print('concepts', components)
              wcss = []
         #
         #
              for i in range(2,23,2):
         #
                  kmeans = KMeans(n clusters=i,init='k-means++',max iter=300,n init=
         #
                  kmeans.fit(tf idf vector)
         #
                  wcss.append(kmeans.inertia )
         #
                  print('k', i)
             plt.plot(range(2,23,2),wcss)
         #
              plt.title('The Elbow Method')
         #
              plt.xlabel('Number of clusters')
         #
         #
              plt.ylabel('WCSS')
         #
              plt.savefig('elbow.png')
         #
              plt.show()
              components += 50
         # This
        components = 4500
        while components < 4501:</pre>
            print('=======')
            print('SVD number of concepts = ', components)
            tsvd = TruncatedSVD(n components=components)
            mat4 = tsvd.fit(tf idf vector).transform(tf idf vector)
            print('Variance ratio sum', tsvd.explained_variance_ratio_.sum())
            finalClusterLabels = BisectingKMeans(mat4, 14, 14, 2)
            components += 50
            print('=======')
        print("End Time =", datetime.now().strftime("%H:%M:%S"))
        Shape before SVD (8580, 26237)
        Start Time = 17:05:21
        SVD number of concepts = 4500
        Variance ratio sum 0.9590014631709839
        K: 14
        Silhouette Coefficient: 0.022
        Calinski Harabasz Score: 38.850
```

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09/10/2021, 17:54
                                                   BisectingKMeans
             End Time = 17:23:38
    In [6]:
              #write the cluster info to output file
             labels = [finalClusterLabels[key] for key in sorted(finalClusterLabels.keys()
             with open("output.dat", "w", encoding="utf8") as file:
                   for item in labels:
                      file.write("%s\n" % str(item))
    In [7]:
              #print the number of elements in each cluster
              count labels = {}
              for label in labels:
                  if label not in count_labels:
                      count labels[label] = 1
                  else:
                      count labels[label] = int(count labels[label]) + 1
              print(count labels)
```

{1: 2914, 14: 1116, 10: 160, 5: 1137, 11: 43, 12: 21, 6: 646, 4: 611, 13: 48, 3: 654, 2: 754, 7: 301, 9: 133, 8: 42}