7. Text Clustering using K-Means

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In [1]: from pprint import pprint
In [2]: # Code for the K-means clustering algorithm
        def vectorize(doc,terms):
            docvector=[]
            count=0
            for i in range(len(terms)):
                docvector.append(0)
                for syn in terms[i]:
                    docvector[i]+=doc.lower().split(" ",500).count(syn.lower())
            return docvector
        def vectorAvg(a):
            center=[0 for i in range(len(a[0]))]
            for vec in a:
                center=list(map(sum,zip(center,vec)))
            center[:]=[round(x/n,3) for x in center]
            return center
        def distInitknn(mat,n,k):
            for i in range(n):
                mat.append([0 for j in range(k)])
            return None
        def kMean(docvectors,centroids,k=2,prev=[],n=1):
            print("\niteration: ",n,"\n")
            cluster=[]
            distMat=[]
            print("\ncentroid vectors:\n")
            pprint(centroids)
            print("\ndoc vectors:\n")
            pprint(docvectors)
            veclen=len(docvectors)
            distInitknn(distMat, veclen,k)
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for i in range(k):
                cluster.append([i])
            if prev==[]:
                for i in range(k,len(docvectors)):
                    cluster[i%k].append(i)
            else:
                for i in range(k, veclen):
                    cluster[getMinIndex(i,distMat)].append(i)
            newcentroid=[]
            print("cluster formed:\n",cluster)
            for clus in cluster:
                vec=[docvectors[i] for i in clus]
                center=vectorAvg(vec)
                newcentroid.append(list(center))
            print("\nnew centroids:\n")
            pprint(newcentroid)
            print("*"*50)
            if cluster!=prev:
                kMean(docvectors, newcentroid, k, cluster, n+1)
            return None
        def calcManDist(a,b):
            dist=0
            for x,y in zip(a,b):
                dist+=abs(x-y)
            return dist
        def genDistMat(distMat,docvectors,centroids):
            for i in range(len(docvectors)):
                for j in range(len(centroids)):
                    dist=calcManDist(docvectors[i],centroids[j])
                    distMat[i][j]=dist
            return None
        def getMinIndex(i,mat):
            mindist=float("inf")
            minindex=0
            for j in range(len(mat[i])):
                if mindist>mat[i][j]:
                    mindist=mat[i][j]
                    minindex=j
            return minindex
In [3]: # Code for the clustering of documents in part 1
        terms=[['automotive'],['car','cars'],['motorcycles','motorcycle'],['self-drive'],['IoT
        doc1='Electric automotive maker Tesla Inc. is likely to introduce its products in India
        doc2='Automotive major Mahindra likely to introduce driverless cars'
                                         2
```

genDistMat(distMat,docvectors,centroids)

```
doc3='BMW plans to introduce its own motorcycles in india'
       doc4='Just drive, a self-drive car rental firm uses smart vehicle technology based on
       doc5='Automotive industry going to hire thousands in 2018'
       doc6='Famous cricket player Dhoni brought his priced car Hummer which is an SUV'
       doc7='Dhoni led india to its second world cup victory'
       doc8='IoT in cars will lead to more safety and make driverless vehicle revolution poss
       doc9='Sachin recommended Dhoni for the indian skipper post'
       docvectors=[]
       doclist=[doc1,doc2,doc3,doc4,doc5,doc6,doc7,doc8,doc9]
       for doc in doclist:
           docvectors.append(vectorize(doc,terms))
       k=4
       kMean(docvectors, docvectors[:k],k)
iteration: 1
centroid vectors:
[[1, 0, 0, 0, 0, 0, 0],
[1, 1, 0, 0, 0, 0, 0],
 [0, 0, 1, 0, 0, 0, 0],
 [0, 1, 0, 1, 1, 0, 0]
doc vectors:
[[1, 0, 0, 0, 0, 0, 0],
 [1, 1, 0, 0, 0, 0, 0],
 [0, 0, 1, 0, 0, 0, 0],
 [0, 1, 0, 1, 1, 0, 0],
 [1, 0, 0, 0, 0, 1, 0],
 [0, 1, 0, 0, 0, 0, 1],
 [0, 0, 0, 0, 0, 0, 1],
 [0, 1, 0, 0, 1, 0, 0],
 [0, 0, 0, 0, 0, 0, 1]]
cluster formed:
 [[0, 4, 8], [1, 5], [2, 6], [3, 7]]
new centroids:
[[0.667, 0.0, 0.0, 0.0, 0.0, 0.333, 0.333],
 [0.5, 1.0, 0.0, 0.0, 0.0, 0.0, 0.5],
 [0.0, 0.0, 0.5, 0.0, 0.0, 0.0, 0.5],
 [0.0, 1.0, 0.0, 0.5, 1.0, 0.0, 0.0]
**************
```

iteration: 2

centroid vectors:

```
[[0.667, 0.0, 0.0, 0.0, 0.0, 0.333, 0.333], [0.5, 1.0, 0.0, 0.0, 0.0, 0.0, 0.5], [0.0, 0.0, 0.5, 0.0, 0.0, 0.0, 0.5], [0.0, 1.0, 0.0, 0.5, 1.0, 0.0, 0.0]]
```

doc vectors:

```
[[1, 0, 0, 0, 0, 0, 0],
[1, 1, 0, 0, 0, 0, 0],
[0, 0, 1, 0, 0, 0, 0],
[0, 1, 0, 1, 1, 0, 0],
[1, 0, 0, 0, 0, 1, 0],
[0, 1, 0, 0, 0, 0, 1],
[0, 0, 0, 0, 0, 0, 1],
[0, 1, 0, 0, 1, 0, 0],
[0, 0, 0, 0, 0, 0, 1]]

cluster formed:
[[0, 4], [1, 5], [2, 6, 8], [3, 7]]
```

new centroids:

iteration: 3

centroid vectors:

```
[[1.0, 0.0, 0.0, 0.0, 0.0, 0.5, 0.0],
[0.5, 1.0, 0.0, 0.0, 0.0, 0.0, 0.5],
[0.0, 0.0, 0.333, 0.0, 0.0, 0.0, 0.667],
[0.0, 1.0, 0.0, 0.5, 1.0, 0.0, 0.0]]
```

doc vectors:

```
[[1, 0, 0, 0, 0, 0, 0],
[1, 1, 0, 0, 0, 0, 0],
[0, 0, 1, 0, 0, 0, 0],
[0, 1, 0, 1, 1, 0, 0],
```

```
[1, 0, 0, 0, 0, 1, 0],
 [0, 1, 0, 0, 0, 0, 1],
 [0, 0, 0, 0, 0, 0, 1],
 [0, 1, 0, 0, 1, 0, 0],
 [0, 0, 0, 0, 0, 0, 1]]
cluster formed:
 [[0, 4], [1, 5], [2, 6, 8], [3, 7]]
new centroids:
[[1.0, 0.0, 0.0, 0.0, 0.0, 0.5, 0.0],
 [0.5, 1.0, 0.0, 0.0, 0.0, 0.0, 0.5],
 [0.0, 0.0, 0.333, 0.0, 0.0, 0.0, 0.667],
 [0.0, 1.0, 0.0, 0.5, 1.0, 0.0, 0.0]
**************
In [4]: # Code for text minining from web for part 2
       from bs4 import BeautifulSoup
        import requests
       from string import punctuation
       doclist=['doc1.txt','doc2.txt','doc3.txt','doc4.txt','doc5.txt','doc6.txt','doc7.txt',
       links=['https://www.zigwheels.com/newcars/Tesla',' https://www.financialexpress.com/au
               ' https://en.wikipedia.org/wiki/Toyota_Prius','https://economictimes.indiatimes
               https://indianexpress.com/article/india/india-news-india/demonetisation-hits-
               https://www.livemint.com/Politics/ySbMKTIC4MINsz1btccBJO/How-demonetisation-a
               ' https://inc42.com/buzz/electric-vehicles-this-week-centre-reduces-gst-on-lith
               'https://www.youthkiawaaz.com/2017/12/impact-of-demonetisation-on-the-indian-ec-
               ' https://www.news18.com/news/business/how-gst-will-curb-tax-evasion-1446035.htm
       for i in range(len(doclist)):
           page=requests.get(links[i])
           soup=BeautifulSoup(page.text, 'html.parser')
           p_tags=soup.find_all('p')
           text = (''.join(s.findAll(text=True))for s in soup.findAll('p'))
           f=open(doclist[i],'w')
            #f.write("abc")
           gen=[str(y.lower()) for y in text ]
           count=0
           for t in gen:
               f.write(t)
               count=count+1
               if count>=500:
                   break
           f.close()
In [5]: # Code for text clustering in part 2
```

```
files = ['doc1.txt', 'doc2.txt', 'doc3.txt', 'doc4.txt', 'doc5.txt', 'doc6.txt', 'doc7.
       terms=[['tesla',"tesla's"], ['electric'], ['car','cars','vehicle','vehicles','automobil
       docvectors=[]
       for fname in files:
            file=open(fname,'r')
            doclines=file.read().split('.')
            doc=''
            for line in doclines:
               doc+=" "+str(line)
            docvectors.append(vectorize(doc,terms))
       kMean(docvectors, docvectors[:k],k)
iteration: 1
centroid vectors:
[[5, 4, 3, 0, 0, 0, 0],
 [0, 22, 13, 0, 0, 0, 0],
 [0, 1, 7, 0, 0, 0, 0],
 [0, 3, 3, 0, 0, 0, 0]
doc vectors:
[[5, 4, 3, 0, 0, 0, 0],
 [0, 22, 13, 0, 0, 0, 0],
 [0, 1, 7, 0, 0, 0, 0],
 [0, 3, 3, 0, 0, 0, 0],
 [0, 5, 2, 0, 1, 0, 0],
 [0, 0, 1, 0, 6, 0, 0],
 [0, 0, 4, 0, 0, 14, 0],
 [0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 4, 0]]
cluster formed:
 [[0, 4, 8], [1, 5, 9], [2, 6, 10], [3, 7, 11]]
new centroids:
[[1.667, 3.0, 1.667, 0.0, 0.333, 0.0, 0.0],
 [0.0, 7.333, 4.667, 0.0, 2.0, 0.0, 0.0],
 [0.0, 0.333, 3.667, 0.0, 0.0, 4.667, 0.0],
 [0.0, 1.0, 1.0, 0.0, 0.0, 1.333, 0.0]]
**************
```

iteration: 2

centroid vectors:

```
[[1.667, 3.0, 1.667, 0.0, 0.333, 0.0, 0.0], [0.0, 7.333, 4.667, 0.0, 2.0, 0.0, 0.0], [0.0, 0.333, 3.667, 0.0, 0.0, 4.667, 0.0], [0.0, 1.0, 1.0, 0.0, 0.0, 1.333, 0.0]]
```

doc vectors:

```
[[5, 4, 3, 0, 0, 0, 0],
[0, 22, 13, 0, 0, 0, 0],
[0, 1, 7, 0, 0, 0, 0],
[0, 3, 3, 0, 0, 0, 0],
[0, 5, 2, 0, 1, 0, 0],
[0, 0, 1, 0, 6, 0, 0],
[0, 0, 4, 0, 0, 14, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0]]
```

new centroids:

iteration: 3

centroid vectors:

```
[[2.5, 4.5, 2.5, 0.0, 0.5, 0.0, 0.0],
[0.0, 22.0, 13.0, 0.0, 0.0, 0.0, 0.0],
[0.0, 0.333, 3.667, 0.0, 0.0, 6.0, 0.0],
[0.0, 0.5, 0.667, 0.0, 1.0, 0.0, 0.0]]
```

doc vectors:

```
[[5, 4, 3, 0, 0, 0, 0],
[0, 22, 13, 0, 0, 0, 0],
[0, 1, 7, 0, 0, 0, 0],
[0, 3, 3, 0, 0, 0, 0],
[0, 5, 2, 0, 1, 0, 0],
[0, 0, 1, 0, 6, 0, 0],
[0, 0, 4, 0, 0, 14, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0],
[0, 0, 0, 0, 0, 0, 0]]
```

new centroids: