```
In [1]: from pyspark.ml.evaluation import RegressionEvaluator
    from pyspark.ml.recommendation import ALS
    from pyspark.sql import Row
    from pyspark.sql import SparkSession
    spark = SparkSession.builder.getOrCreate()
    data = spark.read.format("csv").option("header",True).option("inferSchem a",True).load("RCdata/rating_final.csv")
    data.show()
```

```
+----+
|userID|placeID|rating|food rating|service rating|
       1 | 135085 |
                         2 |
                                        2 |
                                                          2 |
       2 | 135038 |
                         2 |
                                        2 |
                                                          1 |
       3 | 132825 |
                         2
                                        2
                                                          2
       4 | 135060 |
                                        2 |
                                                          2 |
                         1 |
       5 | 135104 |
                         1
                                        1 |
                                                          2
                         0 |
                                        0 |
       6 132740
                                                          0
       7 | 132663 |
                         1 |
                                        1 |
                                                          1 |
       8 | 132732 |
                         0 |
                                        0 |
                                                          0
      9 | 132630 |
                         1
                                        1 |
                                                          1
     10 | 132584 |
                         2
                                        2 |
                                                          2
     11 | 132733 |
                         1 |
                                        1 |
                                                          1
                         1 |
                                        2 |
     12 | 132732 |
                                                          2
     13 | 132630 |
                         1 |
                                        0 |
                                                          1 |
     14 | 135104 |
                         0 |
                                        0 |
                                                          0
     15 | 132560 |
                         1 |
                                        0 |
                                                          0
                                        2 |
     16 | 132584 |
                         1 |
                                                          1
                                        0 |
     17 | 132732 |
                         0
                                                          2
     18 | 132630 |
                         1 |
                                        2
                                                          0 |
     19 | 132613 |
                         2 |
                                        2
                                                          2
     20 | 132667 |
                         1 |
                                        2 |
                                                          2 |
```

only showing top 20 rows

```
predictions = model.transform(test)
evaluator = RegressionEvaluator(metricName="rmse", labelCol="rating",
                                predictionCol="prediction")
#rmse = evaluator.evaluate(predictions)
#print("Root-mean-square error = " + str(rmse))
userRecs = model.recommendForAllUsers(10)
movieRecs = model.recommendForAllItems(10)
userRecs.show(5,False)
|userID|recommendations
463 [[132862,0.48100093], [135060,0.36393434], [135032,0.25766575],
[132754,0.18146174], [135051,0.17885107], [135072,0.1247983], [132723,
0.12071887], [135057,0.104022026], [132872,0.102057114], [135058,0.0910
4104]]
       | [[135038,0.09330819], [132825,0.090056315], [135039,0.0592330
833
1], [135057,0.057779774], [135079,0.049921643], [135045,0.048617616],
 [132921,0.04774813], [135075,0.035188816], [135058,0.031516954], [1350
59,0.0306528111
      [[132560,0.0], [132630,0.0], [132660,0.0], [132740,0.0], [13283
0,0.0], [132870,0.0], [135000,0.0], [135030,0.0], [135040,0.0], [13505
0,0.011
       [[135051, 0.051854], [132862, 0.05014171], [135026, 0.03888851],
 [132572,0.038523607], [135060,0.035370983], [135030,0.031751085], [135
072,0.03148463], [135038,0.030215243], [135076,0.028989723], [134976,0.
0251938]]
      [[135075,0.39391476], [135076,0.23548545], [135030,0.18134171],
1088
[135057,0.15644243], [135066,0.15440075], [135041,0.13692331], [132754,
0.1352996], [135047, 0.13422821], [132723, 0.121487096], [135062, 0.116676
5311
```

only showing top 5 rows

```
In [2]: from pyspark.ml.evaluation import RegressionEvaluator
    from pyspark.ml.recommendation import ALS
    from pyspark.sql import Row
    from pyspark.sql import SparkSession
    spark = SparkSession.builder.getOrCreate()
    data = spark.read.format("csv").option("header",True).option("inferSchem
    a",True).load("ml-20m/ratings.csv")
    data.show()
```

```
+----+
|userId|movieId|rating| timestamp|
       1 |
                2 |
                      3.5 | 1112486027 |
       1 |
               29
                      3.5 | 1112484676 |
       1 |
               32
                      3.5 | 1112484819 |
                      3.5 | 1112484727 |
       1 |
               47
       1 |
               50
                      3.5 | 1112484580 |
                      3.5 | 1094785740 |
       1 |
              112
       1 |
              151
                      4.0 | 1094785734 |
                      4.0 | 1112485573 |
       1 |
              223
       1 |
              253
                      4.0 | 1112484940 |
       1 |
              260
                      4.0 | 1112484826 |
       1 |
                      4.0 | 1112484703 |
              293
       1
              296
                      4.0 | 1112484767 |
       1 |
              318
                      4.0 | 1112484798 |
                      3.5 | 1094785709 |
       1 |
              337
       1 |
              367
                      3.5 | 1112485980 |
                      4.0 | 1112484603 |
       1 |
              541
       1 |
              589
                      3.5 | 1112485557 |
              593
                      3.5 | 1112484661 |
       1 |
              653
                      3.0 | 1094785691 |
       1 |
              919|
                      3.5 | 1094785621 |
       1 |
```

only showing top 20 rows

```
predictions = model.transform(test)
evaluator = RegressionEvaluator(metricName="rmse", labelCol="rating",
                                predictionCol="prediction")
#rmse = evaluator.evaluate(predictions)
#print("Root-mean-square error = " + str(rmse))
userRecs = model.recommendForAllUsers(10)
movieRecs = model.recommendForAllItems(10)
userRecs.show(5,False)
|userId|recommendations
       [[17,0.7760857], [539,0.76959646], [1307,0.71515596], [62,0.698
13776], [597,0.6962341], [1035,0.6932603], [1393,0.6854005], [357,0.673
3391], [11,0.66974914], [708,0.6490281]]
      [[590,0.9840783], [150,0.9808516], [457,0.92620337], [454,0.912
4953], [296,0.91238725], [339,0.911694], [356,0.9104003], [380,0.904721
5], [597,0.9004352], [592,0.8948771]]
      |[[1721,1.1610202], [2028,1.1132988], [1610,1.1045119], [1961,1.
0844686], [1580,1.0815619], [2396,1.0779964], [1307,1.0614403], [2268,
1.0577246], [1270,1.0558075], [3578,1.0534228]]
       [[1196,0.90157634], [1197,0.8774836], [1270,0.87396485], [1198,
496
0.87006056], [1097,0.8494378], [1210,0.8274293], [260,0.825099], [919,
0.8202874], [1136,0.81852674], [1214,0.8003226]]
       [[592,0.9677862], [590,0.9576383], [380,0.9545622], [150,0.9510
537], [457,0.94851166], [480,0.9285179], [349,0.9152355], [165,0.910314
44], [356,0.9014194], [153,0.88558245]]
```

only showing top 5 rows

```
In [ ]: import nltk
    nltk.download('punkt')
    import csv
```

Read a dataset that contains news\_headlines for clustering

```
In [50]: from pyspark.sql import Row
    from pyspark.sql import SparkSession
    from pyspark.ml.feature import HashingTF, IDF, Tokenizer
    spark = SparkSession.builder.getOrCreate()
    data = spark.read.format("csv").option("header",True).option("inferSchem
    a",True).load("abcnews-date-text.csv")
    data.show()
    data = data.limit(500000)
```

```
|publish date|
                  headline text
+----+
    20030219 aba decides again...
    20030219 act fire witnesse...
    20030219 a g calls for inf...
    20030219 air nz staff in a...
    20030219 air nz strike to ...
    20030219 ambitious olsson ...
    20030219 antic delighted w...
    20030219 aussie qualifier ...
    20030219 aust addresses un...
    20030219 australia is lock...
    20030219 australia to cont...
    20030219 barca take record...
    20030219 bathhouse plans m...
    20030219 big hopes for lau...
    20030219 big plan to boost...
    20030219 blizzard buries u...
    20030219 brigadier dismiss...
    20030219 british combat tr...
    20030219|bryant leads lake...|
    20030219 bushfire victims ...
```

+----+

only showing top 20 rows

```
In [51]: tokenizer = Tokenizer(inputCol="headline_text", outputCol="words")
    wordsData = tokenizer.transform(data)
    wordsData.show()
    wordsData.count()
```

```
|publish date|
                    headline text
                                                 words
     20030219 aba decides again... [aba, decides, ag...]
     20030219 act fire witnesse... | [act, fire, witne... |
     20030219 a g calls for inf... [a, g, calls, for...]
     20030219 air nz staff in a... [air, nz, staff, ...
     20030219 air nz strike to ... [air, nz, strike,...]
     20030219 ambitious olsson ... [ambitious, olsso...
     20030219 antic delighted w... [antic, delighted...
     20030219 aussie qualifier ... [aussie, qualifie...]
     20030219 aust addresses un... [aust, addresses,...
     20030219 australia is lock... [australia, is, l...]
     20030219 australia to cont... [australia, to, c...
     20030219 barca take record... | [barca, take, rec... |
     20030219 bathhouse plans m... | [bathhouse, plans...
     20030219|big hopes for lau...|[big, hopes, for,...|
     20030219 big plan to boost... | [big, plan, to, b...
     20030219 blizzard buries u... | [blizzard, buries... |
     20030219|brigadier dismiss...|[brigadier, dismi...|
     20030219 british combat tr... | [british, combat,...
     20030219|bryant leads lake...|[bryant, leads, l...|
     20030219|bushfire victims ...|[bushfire, victim...|
    ----+
only showing top 20 rows
```

Out[51]: 500000

#### TF-IDF

In [53]: featurizedData.show()

```
|publish date|
                     headline_text
                                                   words
                                                                   rawFea
tures
     20030219|aba decides again...|[aba, decides, ag...|(300,[42,57,12
     20030219 act fire witnesse... [act, fire, witne... | (300,[23,43,72,
11...
     20030219 \mid a g calls for inf... \mid [a, g, calls, for... \mid (300, [46,66,14])
4,1...
     20030219 air nz staff in a... [air, nz, staff, ... (300,[83,116,11
     20030219|air nz strike to ...|[air, nz, strike,...|(300,[45,88,10]
     20030219 ambitious olsson ... ambitious, olsso... (300, [18,110,19]
     20030219 antic delighted w... [antic, delighted... | (300, [50,81,10]
5,1...
     20030219 aussie qualifier ... [aussie, qualifie... (300,[0,39,51,1
     20030219 aust addresses un... [aust, addresses,... (300,[79,83,11]
1,1...
     20030219|australia is lock...|[australia, is, l...|(300,[2,62,100,
18...
     20030219 australia to cont... [australia, to, c... (300,[2,42,88,1
     20030219|barca take record...|[barca, take, rec...|(300,[81,95,13
8,1...
     20030219|bathhouse plans m...|[bathhouse, plans...|(300,[107,139,1
     20030219|big hopes for lau...|[big, hopes, for,...|(300,[44,168,22
5,...
     20030219|big plan to boost...|[big, plan, to, b...|(300,[88,123,20
     20030219|blizzard buries u...|[blizzard, buries...|(300,[11,105,15]
     20030219|brigadier dismiss...|[brigadier, dismi...|(300,[35,129,19
     20030219|british combat tr...|[british, combat,...|(300,[35,95,11]
7,1...
     20030219|bryant leads lake...|[bryant, leads, l...|(300,[84,88,20]
0,2...
     20030219 bushfire victims ... | [bushfire, victim... | (300, [15,88,15]
only showing top 20 rows
```

## IDF features

```
In [54]: idf = IDF(inputCol="rawFeatures", outputCol="features")
    idfModel = idf.fit(featurizedData)
    rescaledData = idfModel.transform(featurizedData)

rescaledData.select("headline_text", "features").show()
    rescaledData = rescaledData.select("features")
```

```
headline_text
                                 features
+-----+
aba decides again... (300,[42,57,122,1...
act fire witnesse... | (300, [23, 43, 72, 11... |
a g calls for inf... | (300, [46,66,144,1... |
air nz staff in a... (300,[83,116,117,...
air nz strike to ... (300,[45,88,109,1...
ambitious olsson ... (300,[18,110,192,...
antic delighted w... (300, [50, 81, 105, 1...
aussie qualifier ... (300,[0,39,51,110...)
aust addresses un... | (300,[79,83,111,1... |
australia is lock... (300,[2,62,100,18...
australia to cont... | (300, [2, 42, 88, 111...
barca take record... (300,[81,95,138,1...
bathhouse plans m... (300, [107, 139, 197...]
big hopes for lau... (300,[44,168,225,...
big plan to boost... | (300,[88,123,204,... |
blizzard buries u... (300,[11,105,154,...
brigadier dismiss... (300,[35,129,195,...
british combat tr... (300,[35,95,117,1...
bryant leads lake... (300, [84, 88, 200, 2...
|bushfire victims ...|(300,[15,88,158,1...|
+----+
only showing top 20 rows
```

## I only use the headline of news and that maybe the reason why the Error is so huge

```
In [59]: from pyspark.ml.clustering import KMeans
    kmeans = KMeans().setK(300).setSeed(1)
    model = kmeans.fit(rescaledData)

wssse = model.computeCost(rescaledData)
    print("Within Set Sum of Squared Errors = " + str(wssse))

centers = model.clusterCenters()
```

Within Set Sum of Squared Errors = 33112470.9293

```
In [94]:
         import wikipedia
         import nltk
         from nltk.corpus import stopwords
         from nltk.tokenize import word_tokenize
         t = wikipedia.search("google", results=30)
         print t
         plist = []
         for p in t:
             print p
             page = wikipedia.page(p, auto_suggest=False)
             plist.append(page)
         tokenized docs list = []
         from nltk.tokenize import word tokenize
         for i in plist:
             tokenized docs = word_tokenize(i.content.encode('ascii','ignore'))
             tokenized docs list.append(tokenized docs)
```

[u'Google', u'Google Search', u'Google+', u'Google Play', u'Google Doc s, Sheets, and Slides', u'Google Books', u'Google Translate', u'.googl e', u'Google hacking', u'Google Account', u'Google Chrome', u'Google Ma ps', u'Gmail', u'Google Glass', u'Google Hangouts', u'Thuppakki', u'G S uite', u'AdWords', u'Google Classroom', u'Google Doodle', u'Google Eart h', u'Google Analytics', u'List of Google products', u'Google Drive', u'Googleplex', u'Vikas Gupta', u'Google Brain', u'Google Traffic', u'G oogle Dashboard', u'Motorola'] Google Google Search Google+ Google Play Google Docs, Sheets, and Slides Google Books Google Translate .google Google hacking Google Account Google Chrome Google Maps Gmail Google Glass Google Hangouts Thuppakki G Suite AdWords Google Classroom Google Doodle Google Earth Google Analytics List of Google products Google Drive Googleplex Vikas Gupta Google Brain Google Traffic Google Dashboard Motorola

```
In [106]: print len(tokenized_docs_list)
29
```

## Remove the stopwords

```
In [107]: from nltk.corpus import stopwords
          nltk.download('stopwords')
          print "and" in stopwords.words()
          t_nsw_list = []
          for p in tokenized_docs_list:
              tokenized_nsw = []
              for i in p:
                  if not i in stopwords.words():
                      tokenized_nsw.append(i)
              t_nsw_list.append(tokenized_nsw)
          print len(t_nsw_list)
          [nltk data] Downloading package stopwords to /Users/han/nltk data...
                        Package stopwords is already up-to-date!
          [nltk_data]
          True
          29
```

#### Stem the words

```
In [101]: from nltk.stem.porter import PorterStemmer
    from nltk.stem.snowball import SnowballStemmer
    from nltk.stem.wordnet import WordNetLemmatizer

porter = PorterStemmer()
    #snowball = SnowballStemmer("english")
    wordnet = WordNetLemmatizer()
    after_stem_list = []
    for p in t_nsw_list:
        words = []
        for i in p:
            words.append(porter.stem(i))
        after_stem_list.append(words)
```

[u'unilev', '(', ')', 'dutch-british', u'transnat', u'consum', u'good', u'compani', u'co-headquart', 'rotterdam', ',', u'netherland', 'london', ',', u'unit', 'kingdom', '.', u'it', u'product', u'includ', 'food', ,', u'beverag', ',', u'clean', u'agent', u'person', u'product', '.', 'It', 'world', "'s", 'largest', u'consum', u'good', u'compani', u'meas ur', '2012', u'revenu', ',', 'world', "'s", 'largest', u'produc', 'foo d', u'spread', ',', u'margarin', '.', 'It', u'europ', '7th-most', u'val uabl', u'compani', '.', u'unilev', 'one', 'oldest', u'multin', u'compan i', ';', u'product', u'avail', 'around', '190', u'countri', '.', u'unil ev', u'own', '400', u'brand', ',', u'focus', 'thirteen', u'brand', u'sa le', 'one', 'billion', u'euro', ':', 'axe/lynx', ',', 'dove', ',', 'om ,', 'becel/flora', ',', 'heartbrand', 'ice', u'cream', ',', 'hellm ann', "'s", ',', 'knorr', ',', 'lipton', ',', 'lux', ',', 'magnum', ',', 'rama', ',', 'rexona', ',', 'sunsilk', 'surf', '.', 'It', u'duallist', u'compani', u'consist', u'unilev', 'n.v.', ',', u'base', 'rotter dam', ',', u'unilev', 'plc', ',', u'base', 'london', '.', 'the', 'two', u'compani', u'oper', u'singl', u'busi', ',', 'common', 'board', u'direc tor', '.', u'unilev', u'organis', 'four', 'main', u'divis', u'food', ',', u'refresh', '(', u'beverag', 'ice', 'cream', ')', ',', 'home', 'c are', ',', u'person', 'care', '.', 'It', 'research', u'develop', u'faci l', u'unit', 'kingdom', '(', 'two', ')', ',', u'netherland', ',', 'chin a', ',', 'india', u'unit', u'state', '.', u'unilev', u'found', '1930', 'merger', 'dutch', u'margarin', u'produc', u'margarin', u'uni', 'briti sh', u'soapmak', 'lever', u'brother', '.', u'dure', 'second', 'half', '20th', u'centuri', u'compani', u'increasingli', u'diversifi', 'make r', u'product', 'made', u'oil', u'fat', ',', u'expand', u'oper', u'worl dwid', '.', 'It', 'made', u'numer', u'corpor', u'acquisit', ',', u'incl ud', 'lipton', '(', '1971', ')', ',', u'brook', 'bond', '(', '1984', ')', ',', u'chesebrough-pond', '(', '1987', ')', ',', 'best', u'food', '(', '2000', ')', ',', 'ben', '&', u'jerri', "'s", '(', '2000', ')', ,', u'alberto-culv', '(', '2010', ')', ',', 'dollar', 'shave', 'clu b', '(', '2016', ')', '.', u'unilev', u'divest', u'special', u'chemic', u'busi', 'ici', '1997', '.', 'In', '2015', ',', 'leadership', 'paul', 'polman', ',', u'compani', u'gradual', u'shift', u'focu', u'toward', 'health', u'beauti', u'brand', 'away', 'food', u'brand', u'show', 'slo w', 'growth', '.', u'unilev', 'n.v.', u'primari', u'list', 'euronext', 'amsterdam', u'constitu', 'aex', 'index', '.', u'unilev', 'plc', u'pri mari', u'list', 'london', 'stock', u'exchang', u'constitu', 'ftse', '10 0', 'index', '.', 'the', u'compani', 'lux', 'euro', 'stoxx', '50', 'sto ck', 'market', 'index', '.', '==', u'histori', '==', '===', u'1870s191
0', '===', 'In', '1872', ',', 'antoon', u'jurgen', ',', u'found', 'firs t', u'margarin', u'factori', 'world', u'oss', ',', u'netherland', '.', 'then', ',', '1888', ',', 'samuel', 'bergh', ',', u'oss', ',', u'ope n', u'margarin', u'factori', 'kleve', '.', 'these', 'two', u'compani', u'merg', '1927', 'form', u'margarin', u'uni', '.', '===', u'1910s192
)', '===', 'the', u'initi', u'harvest', 'palm', 'oil', 'british', 'wes t', 'africa', ',', u'news', u'report', 'seen', 'back', 'england', u'sho w', u'worker', 'abroad', u'favour', u'condit', '.', 'In', '1911', ',', u'compani', u'receiv', u'concess', '750,000', u'hectar', 'forest', lgian', 'congo', ',', u'mostli', 'south', 'bandundu', ',', 'system',
u'forc', 'labour', u'oper', '.', '===', u'1920s1930', '===', 'In', '19 22', 'lever', u'brother', u'acquir', 'mac', u'fisheri', ',', 'owner', 'T.', 'wall', '&', u'son', '.', 'In', u'septemb', '1929', ', v', u'form', 'merger', u'oper', 'dutch', u'margarin', u'uni', 'britis h', u'soapmak', 'lever', u'brother', ',', 'name', u'result', u'compan i', 'portmanteau', 'name', u'compani', '.', '===', u'1930s1940', '===', 'In', u'1930', u'busi', 'grew', 'new', u'ventur', u'launch', 'africa',

'latin', 'america', '.', 'the', 'nazi', u'occup', u'europ', 'second', 'world', 'war', 'meant', u'unilev', u'unabl', 'reinvest', u'capit', u'europ', ',', 'instead', u'acquir', 'new', u'busi', 'UK', 'US', '.', 'In', '1943', u'acquir', 'T.', 'J.', 'lipton', ',', u'major', 'stake', u'frost', u'food', '(', 'owner', u'bird', 'eye', 'brand', ')', u'batche lor', u'pea', ',', 'one', 'largest', u'veget', u'canner', 'UK', '.', 'I n', '1944', ',', u'pepsod', u'acquir', '.', 'after', '1945', u'unilev', "'s", u'success', 'US', u'busi', '(', 'lever', u'brother', 't.j.', 'lip ton', ')', 'began', u'declin', '.', 'As', 'result', ',', u'unilev', 'be gan', u'oper', '``', u'hand', "''", u'polici', u'toward', u'subsidiar i', ',', 'left', 'american', u'manag', u'devic', '.', '===', u'1950s196 '===', 'sunsilk', 'first', u'launch', 'UK', '1954', '.', 'dove', 'f irst', u'launch', 'US', '1957', '.', u'unilev', 'took', 'full', 'owners
hip', u'frost', u'food', '1957', ',', u'renam', u'bird', 'eye', '.', 't
he', u'us-bas', 'good', 'humor', 'ice', 'cream', u'busi', u'acquir', '1 961', '.', 'By', u'mid-1960', u'laundri', 'soap', u'edibl', u'fat', 'st ill', u'contribut', 'around', 'half', u'unilev', "'s", u'corpor', u'pro fit', '.', u'howev', 'stagnant', 'market', 'yellow', u'fat', u'increa s', u'competit', u'deterg', u'soap', 'procter', '&', u'gambl', u'forc', u'unilev', u'diversifi', '.', 'In', '1971', ',', u'unilev', u'acquir', u'british-bas', 'lipton', 'ltd', u'alli', u'supplier', '.', 'In', '197 8', ',', u'nation', 'starch', u'acquir', '\$', '487', 'million', ',', u'mark', 'largest', 'ever', u'foreign-acquisit', 'US', u'compani', 'po int', '.', '===', u'1970s1980', '===', 'By', u'1970', ',', u'acquisit', ',', u'unilev', u'gain', '30', 'cent', 'western', 'european', 'ice', 'c ream', 'market', '.', 'In', '1982', u'unilev', u'manag', u'decid', u're posit', u'unwieldi', u'conglomer', u'concentr', 'fmcg', u'compani', '.', 'In', '1984', u'unilev', u'acquir', u'brook', 'bond', '(', 'make r', 'PG', u'tip', 'tea', ')', '390', 'million', u'compani', "'s", 'firs t', u'success', u'hostil', u'takeov', '.', 'In', '1986', u'unilev', u's trengthen', u'posit', 'world', 'skin', 'market', u'acquir', u'chesebrou gh-pond', '(', u'merg', 'chesebrough', u'manufactur', 'pond', "'s", u'c ream', ')', ',', 'maker', 'rag', ',', 'pond', "'s", ',', 'aqua-net', ',', 'cutex', ',', u'vaselin', u'anoth', u'hostil', u'takeov', '.', 'I '1989', ',', u'unilev', 'bought', 'calvin', 'klein', u'cosmet', ',', 'faberg', ',', 'elizabeth', 'arden', ',', 'latter', 'later', 'sol d', '(', '2000', ')', 'ffi', u'fragranc', '.', '===', u'1990', '===', 'In', '1993', u'unilev', u'acquir', u'breyer', 'kraft', ',', 'made', u'compani', 'largest', 'ice', 'cream', u'manufactur', u'unit', u'stat e', '.', 'In', '1996', u'unilev', u'merg', 'elida', u'gibb', 'lever', u'brother', 'UK', u'oper', '.', 'It', u'purchas', u'helen', u'curti', ',', u'significantli', u'expand', u'presenc', u'unit', u'state', 'sham poo', u'deodor', 'market', '.', 'the', u'purchas', 'brought', u'unile v', u'suav', u'finess', u'hair-car', 'product', u'brand', u'degre', u'd eodor', 'brand', '.', 'In', '1997', u'unilev', 'sold', u'special', u'ch emic', u'divis', ',', u'includ', u'nation', 'starch', '&', u'chemic', ',', 'quest', ',', 'unichema', 'crosfield', u'imperi', u'chemic', u'in dustri', '4.9', 'billion', '.', u'unilev', u'establish', u'sustain', u'agricultur', u'programm', '1998', '.', '===', u'2000', '===', 'In', 'april', '2000', u'unilev', 'bought', 'ben', '&', u'jerri', "'s", 'sli m', 'fast', '1.63', 'billion', '.', 'later', 'year', ',', u'compani', u'acquir', 'best', u'food', '13.4', 'billion', '.', 'the', u'bestfood', u'acquisit', u'increas', u'unilev', "'s", 'scale', u'food', 'americ a', ',', u'ad', u'brand', 'knorr', 'hellmann', "'s", 'portfolio', '.', 'the', u'transact', 'second', 'largest', 'cash', u'acquisit', 'world', u'busi', u'histori', '.', 'In', u'exchang', 'european', u'regulatori', u'approv', 'deal', ',', u'unilev', u'divest', 'well-known', u'brand',

'oxo', ',', 'royco', u'batchelor', '.', 'the', 'year', u'compani', 'bo ught', u'worldwid', 'mustard', u'product', 'firm', u'maill', '.', u'mai ll', 'three', u'boutiqu', u'europ', ',', 'sell', 'mustard', 'pump', u't radit', u'maill', 'fashion', '.', u'pari', ',', 'dijon', ',', u'franc', 'london', ',', 'UK', '.', 'the', u'merg', 'best', u'food', u'unilev', u'approv', u'isra', 'anti', 'trust', u'agenc', '.', 'In', '2001', u'un ilev', 'split', 'two', u'divis', ':', 'one', u'food', 'one', 'home', u'person', 'care', '.', 'In', 'UK', u'merg', 'lever', u'brother', 'eli da', u'faberg', u'busi', 'lever', u'faberg', u'januari', '2001', '.', 'In', u'septemb', '2002', ',', u'compani', 'sold', u'specialti', u'oi l', u'fat', u'divis', ',', u'loder', 'croklaan', ',', 'rm814', 'millio n', '(', '218.5', 'million', ')', 'ioi', u'corpor', ' ,', 'kuala', 'lump ur', ',', u'malaysia-bas', 'oil', 'palm', u'compani', '.', 'As', 'par t', 'deal', ',', u'loder', 'croklaan', 'brand', u'maintain', '.', 'als o', '2002', u'unilev', 'sold', 'mazola', ',', 'argo', '&', u'kingsfor d', ',', 'karo', ',', 'golden', u'griddl', ',', 'henri', "'s", u'bran d', ',', 'along', u'sever', 'canadian', u'brand', ',', 'ach', 'food', u'compani', ',', 'american', u'subsidiari', u'associ', 'british', u'fo od', '.', 'In', '2004', u'unilev', 'sold', 'share', 'rushdi', 'food', u'industri', 'bashir', u'famili', u'start', u'use', u'barack', 'bran d', 'name', '.', 'As', '2014', 'roshadi', 'food', u'industri', 'one', 'three', 'largest', 'tahini', u'produc', 'israel', 'one', 'largest', u'produc', 'tahini', u'worldwid', '.', 'In', 'may', '2007', u'unilev', u'becam', 'first', u'large-scal', u'compani', 'commit', u'sourc', 'te a', u'sustain', 'manner', ',', u'employ', 'rainforest', u'allianc', ',', u'intern', u'environment', 'ngo', ',', u'certifi', 'tea', u'esta t', 'east', 'africa', ',', 'well', u'third-parti', u'supplier', 'afric a', u'part', 'world', '.', 'It', u'declar', 'aim', 'lipton', 'yellow', 'label', 'PG', u'tip', 'tea', u'bag', 'sold', 'western', u'europ', u'c ertifi', '2010', ',', u'follow', 'lipton', 'tea', u'bag', u'global', '2 015', '.', 'In', u'septemb', '2009', u'unilev', u'agre', u'acquir', u'p erson', u'busi', 'sara', 'lee', u'corpor', ',', u'includ', u'brand', 'r adox', ',', u'badeda', u'duschda', ',', u'strengthen', u'categori', 'le adership', 'skin', u'cleans', u'deodor', '.', 'the', 'sara', 'lee', u'a cquisit', u'complet', '6', u'decemb', '2010', '.', '===', '20102014', '===', 'On', '9', 'august', '2010', u'unilev', u'sign', 'asset', u'pur chas', 'agreement', 'norwegian', u'dairi', 'group', 'tine', ',', u'acqu ir', u'activ', u'diplom-i', 'denmark', '.', 'On', '24', u'septemb', '20 10', u'unilev', u'announc', u'enter', u'definit', 'agreement', 'sell', u'consum', 'tomato', u'product', u'busi', 'brazil', u'cargil', '.', 'O n', '27', u'septemb', '2010', u'unilev', u'purchas', u'alberto-culv', ,', 'maker', u'person', 'household', u'product', u'includ', u'simpl', 'vo5', ',', u'nexxu', ',', 'tresemm', ',', u'mr', '.', 'dash', ',', 'US', '\$', '3.7', 'billion', '.', 'On', '28', u'septemb', '2010', u'unilev', 'evga', u'announc', u'sign', 'agreement', u'unilev', 'woul d', u'acquir', 'evga', "'s", 'ice', 'cream', u'brand', '(', 'amongst', u'other', ',', 'scandal', ',', u'variet', 'karabola', ')', u'distribu
t', 'network', u'greec', ',', u'undisclos', 'amount', '.', 'In', u'febr uari', '2011', u'unilev', u'announc', 'switch', '100', '%', u'cage-fr e', u'egg', u'product', u'produc', u'worldwid', '.', 'In', 'march', '20 11', u'announc', u'unilev', u'enter', u'bind', 'agreement', 'sell', 'sa nex', 'brand', u'colgate-palmol', '672', 'million', ',', u'unilev', 'wo uld', u'acquir', u'colgate-palmol', "'s", u'laundri', u'deterg', u'bran d', 'colombia', '(', 'fab', ',', u'lavomat', 'vel', ')', 'US', '\$', '21 5', 'million', '.', 'In', 'april', '2011', u'unilev', u'fine', '104', 'million', 'european', u'commiss', u'establish', u'price-fix', 'carte l', u'europ', 'along', 'P', '&', 'G', ',', u'fine', '211.2', 'million',

',', 'henkel', '(', u'fine', ')', '.', 'though', 'fine', 'set', 'highe r', 'first', ',', u'discount', '10', '%', u'unilev', 'P', '&', 'G', u'a dmit', u'run', 'cartel', '.', 'As', u'provid', 'tip-off', u'lead', u'in vestig', ',', 'henkel', u'fine', '.', 'On', '24', 'august', '2011', u'a nnounc', u'unilev', u'agre', 'sell', 'alberto', 'vo5', 'brand', u'uni t', u'state', 'puerto', 'rico', ',', 'rave', 'brand', u'global', ',', 'brynwood', u'partner', 'VI', 'l.p.', 'On', '14', u'octob', '2011', u'announc', u'unilev', u'agre', u'acquir', '82', '%', u'russia-bas', u'beauti', u'compani', 'kalina', '.', 'On', '27', u'decemb', '2012', u'announc', u'unilev', 'would', 'phase', 'use', u'microplast', 'form', u'microbead', u'person', u'product', '2015', '.', 'In', u'januari', '20 13', ',', u'unilev', u'agre', 'sell', u'skippi', 'peanut', 'butter', 'b ',', u'togeth', u'relat', u'manufactur', u'facil', u'littl', 'ro ck', ',', u'arkansa', ',', u'unit', u'state', 'weifang', ',', 'shandon g', ',', 'china', ',', 'hormel', u'food', u'approxim', '\$', '700', 'mil lion', '(', '433', 'million', ',', u'approxim', '540', 'million', ')', 'cash', '.', 'In', u'juli', '2013', u'unilev', u'increas', 'stake', 'i ndian', 'unit', ',', 'hindustan', u'unilev', ',', '67', '%', 'around', '2.45', 'billion', '.', 'On', '12', 'august', '2013', u'unilev', u'ann ounc', u'sign', 'agreement', u'wish-bon', 'western', u'dress', u'bran d', u'pinnacl', u'food', 'inc.', 'total', 'cash', u'consider', u'approx im', 'US', '\$', '580', 'million', ',', 'subject', u'regulatori', u'appr ov', '.', 'On', '6', u'septemb', '2013', u'unilev', u'enter', u'defini t', 'agreement', u'acquir', 'premium', 'australian', 'tea', 'brand', 'T 2', '.', 'On', '21', u'februari', '2014', u'unilev', u'sign', u'defini t', 'agreement', 'meat', u'snack', u'busi', ',', u'includ', 'peperami', '(', 'uk/ireland', ')', 'bifi', '(', u'continent', u'europ', ')', 'jac k', 'link', "'s", ',', u'undisclos', 'amount', '.', 'In', 'march', '201 4', u'unilev', u'agre', u'acquir', u'major', 'stake', u'china-bas', 'wa ter', u'purif', u'compani', 'qinyuan', ',', u'make', 'water', u'purif i', ',', u'drink', 'water', u'equip', 'water', 'treatment', u'membran', ',', u'undisclos', 'price', '.', 'On', '22', 'may', '2014', u'compani', u'announc', 'sold', 'north', 'america', 'pasta', u'sauc', u'busi', u'in clud', 'rag', 'bertolli', u'brand', u'japanes', u'compani', 'mizkan', , 'worth', '\$', '2.15', 'billion', '.', 'On', '10', u'juli', '20 14', '.', u'unilev', u'announc', 'sold', 'slim-fast', 'brand', u'kain o', u'capit', ',', 'yet', u'retain', u'minor', 'stake', u'busi', '.', 'On', '2', u'decemb', '2014', ',', u'unilev', u'announc', u'acquir', 'talenti', 'gelato', '&', 'sorbetto', ':', u'minneapolis-bas', 'talent i', ',', u'found', '2003', ',', 'grown', u'best-sel', u'packag', 'gelat o', u'unit', u'state', '.', 'On', '22', u'decemb', '2014', ',', u'unile v', u'announc', u'purchas', 'camay', 'brand', u'global', 'zest', 'bran d', u'outsid', 'north', 'america', 'caribbean', 'procter', '&', u'gamb l', '.', '====', 'hampton', 'creek', 'lawsuit', '====', 'In', u'novem b', '2014', ',', u'unilev', 'subject', 'media', 'backlash', 'due', 'law suit', 'rival', 'hampton', 'creek', '.', 'In', 'suit', ',', u'unilev', u'reveal', 'hampton', 'creek', '``', u'seiz', 'market', 'share', "''" u'loss', u'caus', u'unilev', '``', u'irrepar', 'harm', '.', "''", u'uni lev', u'use', 'standard', u'ident', u'regul', u'claim', 'hampton', 'cre ek', "'s", '``', 'just', 'mayo', "''", u'product', u'fals', u'adverti s', "n't", 'contain', u'egg', '.', 'the', 'washington', 'post', u'headl in', 'suit', 'read', ',', '``', 'big', 'food', "'s", 'weird', 'war', 'o
ver', 'the', u'mean', u'mayonnais', '.', "''", 'the', u'lo', u'angel', u'time', 'began', u'stori', '``', 'big', 'tobacco', ',', 'big', 'oil',
',', 'big', 'mayo', '?', "''", 'A', 'wall', 'street', 'journal', 'write r', u'describ', '``', 'giant', u'corpor', u'gener', 'huge', u'quantit i', 'free', u'advertis', 'brand', u'equiti', u'tini', 'rival', u'su',

'.', "''", 'eat', 'drink', u'polit', u'headlin', u'controversi', '``', u'unilev', "'s", u'bulli', u'backfir', ',', u'boost', 'hampton', 'cree k', "''", '.', u'neg', 'media', u'coverag', 'big', 'mayo', 'lawsuit', u'goe', 'viral', 'case', u'studi', 'PR', 'blunder', "''", '.', '===', '2015present', '===', 'In', 'march', '2015', ',', u'unilev', u'confir m', u'reach', 'agreement', u'acquir', 'ren', u'skincar', ',', 'britis h', u'nich', u'skincar', 'brand', '.', u'thi', u'follow', 'may', '201 5', u'acquisit', u'prestig', u'skincar', 'brand', 'kate', u'somervil', u'skincar', 'llc', '.', 'In', u'juli', '2015', ',', u'compani', u'sepa r', u'spread', u'busi', ',', u'includ', 'flora', 'I', 'Ca', "n't", u'be liev', 'It', "'s", 'not', 'butter', '!', u'brand', ',', u'standalon', u'entiti', u'name', u'unilev', u'bake', ',', u'cook', u'spread', '.', 'the', u'separ', 'first', u'announc', u'decemb', '2014', 'made', u'res pons', u'declin', u'worldwid', u'sale', 'product', u'categori', '.', 'I n', u'octob', '2015', ',', u'unilev', u'agre', u'acquir', 'italian', 'p remium', 'ice', 'cream', 'maker', 'grom', u'undisclos', 'fee', '.', 'I n', u'juli', '2016', ',', u'unilev', 'bought', 'US', 'start-up', 'dolla r', 'shave', 'club', u'report', '\$', 'lbn', '(', '764m', ')', 'cash',
u'compet', 'male', u'groom', 'market', '.', 'In', u'septemb', u'unile v', u'acquir', 'seventh', u'gener', 'inc.', '\$', '700', 'million', '.', 'On', u'februari', '17', ',', '2017', ',', u'significantli', 'smaller', 'kraft', 'heinz', 'made', '\$', '143', 'billion', 'bid', 'food', u'consu m', u'product', 'giant', u'unilev', 'the', 'deal', u'declin', u'unile v', u'abandon', u'februari', '19', 'UK', 'prime', u'minist', 'theresa', 'may', u'order', u'scrutini', 'deal', '.', '==', u'oper', '==', u'unile v', u'organis', 'four', 'main', u'divis', ':', u'person', 'care', '(', u'product', 'skin', 'hair', u'product', ',', u'deodor', 'oral', u'prod uct', ')', ';', u'food', '(', u'product', u'soup', ',', u'bouillon', ,', u'sauc', ',', u'snack', ',', u'mayonnais', ',', 'salad', u'dres s', ',', u'margarin', u'spread', ')', ';', u'refresh', '(', u'product', 'ice', 'cream', ',', u'tea-bas', u'beverag', ',', u'weight-manag', u'pr oduct', u'nutrit', u'enhanc', u'stapl', 'sold', u'develop', u'market', ')', ';', 'home', 'care', '(', u'product', 'home', u'product', u'inclu d', u'powder', ',', u'liquid', u'capsul', ',', 'soap', u'bar', u'clea n', u'product', ')', '.', 'In', u'financi', 'year', u'end', '31', u'dec emb', '2013', ',', u'unilev', 'total', u'turnov', '49.797', 'billion', '36', '%', u'person', 'care', ',', '27', '%', u'food', ',', '19', '%', u'refresh', '18', '%', 'home', 'care', '.', u'unilev', u'invest', 'tota l', '1.04', 'billion', 'research', u'develop', '2013', '.', u'unilev', 'one', 'largest', 'media', u'buyer', 'world', ',', u'invest', 'aroun d', '6', 'billion', '(', 'US', '\$', '8', 'billion', ')', u'advertis', u'promot', '2010', '.', u'unilev', "'s", 'largest', u'intern', u'compe titor', 'nestl', 'procter', '&', u'gambl', '.', 'It', u'face', u'compet it', 'local', u'market', u'specif', 'product', u'rang', u'numer', u'com pani', ',', u'includ', 'beiersdorf', ',', 'conagra', ',', u'danon', ',', 'henkel', ',', u'mar', ',', 'pepsico', ',', 'reckitt', u'bencki s', 'S.', 'C.', 'johnson', '&', 'son', '.', u'unilev', u'fine', 'autori t', u'concurr', u'franc', '2016', u'price-fix', u'person', u'hygien', u'product', '.', '==', u'product', '==', u'unilev', "'s", u'product', u'includ', u'food', ',', u'beverag', ',', u'clean', u'agent', u'perso n', u'product', '.', 'the', u'compani', u'own', '400', u'brand', ',', u'organis', 'four', 'main', u'categori', '-', u'food', ',', u'refres h', ',', 'home', 'care', ',', u'person', 'care', '.', u'unilev', "'s", 'current', u'largest-sel', u'brand', u'includ', ':', 'axe/lynx', 'ben', '&', u'jerri', "'s", ';', 'dove', ';', 'flora/becel', ';', 'hea rtbrand', ';', "hellmann's/best", u'food', ';', 'knorr', ';', 'lipton', ';', 'lux/radox', ';', 'omo/surf', ';', u'rexona/sur', ';', 'sunsilk',

';', 'tresemm', ';', 'magnum', ';', u'vaselin', 'vo5', '.', u'unilev', "'s", 'standard', u'industri', u'classif', u'code', '10890', ':', u'man  $\,$ ufactur', 'food', u'product', 'n.e.c.', ',', '10410', ':', u'manufactu r', u'oil', u'fat', ',', '10420', ':', u'manufactur', u'margarin', 'sim ilar', u'edibl', u'fat', '.', '==', u'corpor', u'affair', '==', '===', 'legal', u'structur', '===', u'unilev', 'two', u'hold', u'compani', ':', u'unilev', 'n.v.', ',', u'regist', 'head', u'offic', 'rotterdam', ',', u'netherland', u'unilev', 'plc', ',', u'regist', u'offic', 'port', 'sunlight', u'merseysid', ',', u'unit', 'kingdom', 'head', u'offic', u'unilev', u'hous', 'london', ',', u'unit', 'kingdom', '.', u'unilev', 'plc', u'unilev', 'n.v.', u'subsidiari', u'compani', u'oper', u'nearl i', u'practic', u'singl', u'econom', u'entiti', ',', 'whilst', u'remai n', u'separ', 'legal', u'entiti', u'differ', u'sharehold', u'separ', 's tock', u'exchang', u'list', '.', 'there', u'seri', 'legal', u'agreemen t', 'parent', u'compani', ',', u'togeth', 'special', u'provis', u'respe ct', u'articl', u'associ', ',', 'known', u'foundat', u'agreement', '.', 'A', 'key', u'requir', u'agreement', u'peopl', u'board', 'two', 'paren t', u'compani', '.', 'An', u'equalis', 'agreement', u'regul', 'mutual', u'right', u'sharehold', u'unilev', 'plc', u'unilev', 'n.v.', u'object', u'ensur', ',', u'principl', ',', 'make', u'financi', u'differ', 'hold', u'share', u'unilev', 'plc', 'rather', u'unilev', 'n.v.', '(', 'vice',
 'versa', ')', '.', '===', 'senior', u'manag', '===', u'unilev', "'s", 'highest', u'execut', u'bodi', u'unilev', 'leadership', u'execut', ',', 'led', 'chief', u'execut', '(', u'current', 'paul', 'polman', ')', '.', '==', 'logo', '==', 'In', '1930', ',', 'logo', u'unilev', u'use', 'helvetica', 'font', u'cap', '.', 'In', '1969', ',', u'typefa c', 'logo', u'chang', ',', 'basic', 'design', u'remain', '.', 'the', 'c urrent', u'unilev', u'corpor', 'logo', u'introduc', '2004', u'design', 'wolff', u'olin', ',', 'brand', u'consult', u'agenc', '.', "'U", "'", 'shape', 'made', '25', 'distinct', u'symbol', ',', 'icon', u'repres', 'one', u'compani', "'s", u'sub-brand', u'corpor', u'valu', '.', 'the', 'brand', u'ident', u'develop', 'around', 'idea', '``', u'a d', u'vital', 'life', '.', "''", '==', u'advertis', '==', 'dove', ':', 'dove', u'describ', u'dedic', '``', 'help', '...', 'women', 'develop', u'posit', 'relationship', 'way', 'look', u'help', u'rais', 'self-estee m', u'realiz', 'full', u'potenti', "''", '.', '(', 'dove', ',', '``', 'our', 'vision', "''", ')', 'dove', u'employ', 'use', u'advertis', u'p roduct', 'display', u'messag', u'posit', 'self-esteem', '.', 'In', u'se ptemb', '2004', 'dove', u'creat', 'real', u'beauti', 'campaign', ',', u'focus', u'predomin', 'women', u'shape', 'colour', '.', 'later', 7', 'campaign', u'further', u'includ', 'women', u'age', '.', u'thi', 'c ampaign', u'consist', u'mostli', u'advertis', ',', 'shown', u'televis', u'popularis', 'internet', '.', 'dove', 'fell', u'scrutini', u'gener', 'public', 'felt', 'dove', u'advertis', u'describ', 'opinion', u'cellul it', 'still', u'unsightli', ',', 'women', "'s", u'age', u'process', u's ometh', u'asham', '.', u'lynx/ax', ':', 'axe', ',', 'known', 'lynx', u'unit', 'kingdom', ',', u'republ', 'ireland', ',', 'australia', 'ne w', 'zealand', ',', u'toiletri', 'brand', u'market', u'toward', 'youn g', 'age', '16', '24', '.', u'it', u'market', '``', 'tongue-in-cheek', 'take', u"'mate", 'game', "'", "''", ',', u'suggest', 'women', u'instantli', 'drawn', u'use', u'product', '.', u'unlik', 'dove', "'s", 'lon g', u'run', u'beauti', 'campaign', 'lynx', u'advertis', 'often', u'crea t', 'mini', u'seri', "'", u'advertis', u'base', 'around', 'singular', 'product', 'rather', u'commun', u'overarch', 'idea', '.', u'thi', u'ad vertis', 'campaign', u'thrive', u'controversi', '.', u'use', u'imag', u'compani', u'know', u'receiv', u'complaint', u'garner', 'brand', 'fre e', u'public', u'notorieti', '.', 'A', 'wide', u'varieti', u'advert',

u'ban', u'countri', 'around', 'world', '.', 'In', '2012', 'lynx', "'s", "'clean", u'ball', "'", 'advert', u'ban', '.', u'thi', 'advert', u'design', u'televis', ',', u'show', u'attract', 'young', 'woman', u'cl ean', u'variou', 'sport', u'ball', '.', 'In', '2011', 'UK', 'lynx', "'s", 'shower', 'gel', 'campaign', u'ban', '.', 'the', 'poster', 'lyn x', 'shower', 'gel', u'show', 'woman', u'undon', 'bikini', 'shower', 'b each', ',', u'headlin', ':', '``', 'the', 'cleaner', 'dirtier', 'get', '.', "''", 'both', u'advertis', u'campaign', 'make', 'stark', u'compar ison', 'women', u'sexual', u'portray', u'advertis', u'sale', u'effici', '.', 'lynx', u'commonli', u'portray', 'women', 'visual', u'advertis', 'hyper', 'sexual', ',', u'flawless', u'stereotyp', u'attract', u'arou s', ',', u'age', u'statur', ',', 'use', 'lynx', 'product', '.', r', 'target', u'audienc', u'age', '16-24', u'singl', '.', '==', u'envir onment', 'record', '==', u'unilev', u'declar', 'goal', u'decoupl', u'en vironment', 'impact', 'growth', ',', ':', u'halv', u'environment', 'foo tprint', u'product', 'next', '10', u'year', ';', u'help', '1', 'billio n', u'peopl', u'improv', 'health', u'well-b', ';', u'sourc', u'agricult ur', 'raw', u'materi', u'sustain', '.', '===', 'palm', 'oil', '===',
u'unilev', u'criticis', u'greenpeac', u'caus', u'deforest', ',', u'uni lev', u'target', '2008', u'greenpeac', 'UK', ',', u'criticis', u'compan i', u'buy', 'palm', 'oil', u'supplier', u'damag', 'indonesia', "'s", u'rainforest', '.', 'By', '2008', ',', 'indonesia', u'lose', '2', '%', u'remain', 'rainforest', 'year', ',', 'fastest', u'deforest', 'rate', u'countri', '.', 'the', u'unit', u'nation', u'environment', u'program m', u'state', 'palm', 'oil', u'plantat', u'lead', u'caus', u'deforest', 'indonesia', '.', u'furthermor', ',', 'indonesia', 'fourteenth', 'large st', u'emitt', u'greenhous', u'gase', u'larg', 'due', u'destruct', u'ra inforest', 'palm', 'oil', u'industri', ',', u'contribut', '4', '%', 'gl obal', 'green', u'hous', u'ga', u'emiss', '.', u'accord', u'greenpeac', ',', 'palm', 'oil', u'expans', u'take', 'place', u'littl', 'oversight', 'central', 'local', u'govern', u'procedur', u'environment', 'impact', u'assess', ',', u'land-us', u'plan', u'ensur', 'proper', u'process', u'develop', u'concess', u'neglect', '.', u'plantat', u'off-limit', ',', 'law', ',', 'palm', 'oil', u'plantat', u'establish', 'well', u'il leg', 'use', 'fire', 'clear', 'forest', u'area', u'commonplac', '.', u'unilev', ',', u'found', 'member', u'roundtabl', u'sustain', 'palm', 'oil', '(', 'rspo', ')', ',', u'respond', u'publicis', 'plan', 'obtai n', 'palm', 'oil', u'sourc', u'certifi', u'sustain', '2015', '.', 'It', u'claim', 'goal', '2012', u'encourag', 'rest', u'industri', u'becom', '100', '%', u'sustain', '2020', '.', 'In', 'cte', u"d'ivoir", '(', u'i vori', 'coast', ')', ',', 'one', u'unilev', "'s", 'palm', 'oil', u'supp lier', u'accus', u'clear', 'forest', u'plantat', ',', u'activ', u'threa ten', u'primat', u'speci', ',', u'miss', 'waldron', "'s", 'red', u'colo bu', '.', u'unilev', u'interven', 'halt', u'clearanc', u'pend', u'resul t', u'environment', u'assess', '.', u'accord', u'amnesti', u'intern', '2016', u'unilev', 'palm', 'oil', u'provid', 'wilmar', u'intern', u'pr ofit', '8', '14', 'year', 'old', 'child', 'labor', u'forc', 'labor', 'some', u'worker', u'extort', ',', u'threaten', 'paid', 'work', '.', 'some', u'worker', u'suffer', u'sever', u'injuri', 'toxic', u'ba n', u'chemic', '.', 'In', '2016', u'singapore-bas', 'wilmar', u'inter n', u'world', 'biggest', 'palm', 'oil', 'grower', '.', '===', 'paper', 'use', '===', 'for', u'year', ',', u'unilev', u'purchas', 'paper', u'p ackag', 'asia', 'pulp', '&', 'paper', ',', 'third', 'largest', 'paper', u'produc', 'world', ',', u'label', '``', 'forest', u'crimin', "''", u'd estroy', '``', u'preciou', 'habitat', "''", 'indonesia', "'s", 'rainfor .', 'In', '2011', ',', u'unilev', u'cancel', 'contract', 'asia', 'pulp', '&', 'paper', ',', u'greenpeac', u'execut', 'director', 'phil',

'radford', u'commend', u'compani', u'effort', 'made', u'toward', 'fores t', u'protect', ',', '``', u'take', 'rainforest', u'conserv', u'seriou s', '.', "''", '===', 'rainforest', u'allianc', '===', u'unilev', u'cer tifi', 'tea', u'product', 'rainforest', u'allianc', 'scheme', '.', 'th e', u'compani', u'state', 'least', '50', '%', 'tea', u'product', u'orig in', u'certifi', u'farm', ',', u'compar', u'allianc', "'s", '30', '%', 'minimum', u'entri', 'point', '.', u'unilev', u'decid', 'scheme', u'fa irtrad', ',', u'accord', u'compani', "'s", u'analysi', ',', u'fairtra d', 'might', '``', 'lack', 'scale', u'organiz', u'flexibl', u'certifi', u'industri', 'tea', u'estat', "''", '.', '===', u'critic', '===', 'th e', 'rainforest', u'allianc', u'certif', 'scheme', u'criticis', u'offe r', u'produc', 'minimum', u'guarante', 'price', ',', u'therefor', u'lea v', u'vulner', 'market', 'price', u'variat', '.', 'the', u'altern', u'c ertif', ',', u'fairtrad', ',', u'howev', u'receiv', 'similar', u'criti c', 'well', '.', 'the', 'rainforest', u'allianc', u'certif', u'furtherm or', u'criticis', u'allow', 'use', 'seal', u'product', 'contain', 'mini mum', '30', '%', u'certifi', 'content', ',', u'accord', u'endang', u'in tegr', u'certif', '.', '===', u'juli', '-', u'septemb', '2016', 'salmon ella', 'affair', '===', '====', 'salmonella', 'affair', u'cereal', 'isr ael', '====', 'In', u'juli', '2016', ',', u'rumour', 'salmonella', u'co ntamin', u'cereal', 'spread', 'among', u'isra', u'consum', '.', u'init
i', ',', u'unilev', u'provid', 'public', u'inform', 'subject', u'quer , 'matter', u'initi', u'rebuf', u'compani', u'non-stori', u'nonsens', '.', 'On', 'night', '26', u'juli', '2016', ',', u'unilev', u'stop', u't ransfer', u'cornflak', u'retail', u'chain', '.', 'On', '28', u'juli', ',', 'yediot', 'ahronot', u'report', u'ten', u'thousand', u'box', 'bre akfast', 'cereal', u'destroy', '.', 'By', '28', u'juli', ',', u'despi t', u'compani', "'s", u'assur', u'noth', u'contamin', u'releas', u'cons umpt', ',', u'mani', u'custom', u'stop', u'buy', u'unilev', u'product', u'start', 'throw', 'away', u'cornflak', 'made', u'unilev', '.', 'the', u'compani', 'withheld', u'inform', u'affect', u'product', u'date', '.', 'On', '2', 'august', '2016', ',', u'globe', u'report', u'compan i', u'publish', u'inform', 'telma', u'cereal', u'handl', u'packag', 'li ne', u'contamin', u'discov', 'telma', u'announc', 'made', ':', '``', e', u'stress', 'telma', u'product', u'store', u'home', 'safe', 'eat', '.', u'accord', u'compani', "'s", 'strict', u'procedur', ',', u'ever i', u'product', 'batch', u'check', 'put', 'hold', '.', 'these', u'produ ct', u'market', 'test', u'result', u'product', u'seri', u'return', ',', u'confirm', 'well', '.', 'If', 'flaw', u'discov', ',', 'batch', u'marke t', u'store', ',', 'case', '.', "''", 'In', u'follow', u'day', 'healt h', u'minist', ',', 'yakov', 'litzman', ',', u'threaten', 'pull', u'uni lev', "'s", u'licenc', 'israel', '.', 'He', u'accus', u'unilev', u'li e', u'ministri', u'regard', u'salmonella-infect', 'breakfast', u'cerea l', '.', 'On', '7', 'august', ',', u'globe', u'report', u'contamin', 'm ay', u'sourc', 'pigeon', u'faec', ',', 'health', u'ministri', 'said', 'might', u'sourc', u'contamin', 'pigeon', u'faec', u'possibl', u'sour c', '.', u'globe', 'said', u'product', 'line', u'automat', '(', ' $\tilde{}$ ', 'without', 'human', u'hand', "''", ')', u'possibl', u'sourc', 'human', 'slim', u'chanc', '.', 'On', '8', 'august', '2016', ',', u'isra', 'heal th', u'minist', u'suspend', u'manufactur', u'licenc', u'unilev', u'carr i', 'number', u'correct', ';', 'action', 'came', u'inspect', 'arad', 'p lant', ',', u'state', '``', u'thi', u'seri', u'neglig', u'mistak', ' u'incid', u'malici', 'intent', 'firm', "'s", u'manag', u'qualiti', trol', u'procedur', '.', "''", 'An', u'investig', 'led', 'prof.', 'itam r', 'grutto', 'eli', 'gordon', u'conclud', 'event', u'caus', u'neglig', '.', 'On', '23', u'septemb', u'report', u'cereal', u'produc', '18th', '20th', 'arad', 'plant', u'trace', 'salmonella', '.', '=====', u'clas

s', u'action', '=====', 'A', u'file', u'class', 'action', 'must', 'firs t', u'approv', u'isra', 'court', ',', u'approv', 'case', 'held', '.', 'for', 'sum', '1.2', 'million', u'ni', '(', '~', '\$', '329k', 'usd', ')', u'unilev', u'hide', u'contamin', u'mislead', 'public', 'for', 'su m', '76', 'million', u'ni', '(', '~', '\$', '23m', 'usd', ')', u'unile v', '15-year-old', 'teen', u'hospitalis', u'salmonellosi', u'allegedl i', u'contract', u'unilev', u'product', '====', 'salmonella', 'affair',
'tehnia', '====', 'On', '31', 'august', ',', u'unilev', u'state', 'tehn ia', u'product', u'produc', 'rjm', u'contamin', 'salmonella', '.', '== =', u'kodaikan', '===', 'In', '2015', ',', 'indian', 'rapper', 'sofia', 'ashraf', u'releas', 'music', 'video', '``', u'kodaikan', 'Wo', "n't", ',', "''", 'set', 'beat', 'nicki', 'minaj', "'s", '``', 'anaconda', ,', "''", u'criticis', u'unilev', u'dump', u'mercuri', u'wast', 'grou nd', 'indian', 'town', u'kodaikan', '.', u'unilev', u'acknowledg', 'too k', u'thermomet', u'factori', u'acquisit', u'chesebrough-pond', '.', u'accord', u'unilev', "'s", 'statement', ',', u'factori', 'sold', u'me rcury-contamin', 'scrap', u'glass', 'local', 'dealer', ',', u'prompt', 'hindustan', u'unilev', u'immedi', 'close', u'factori', ',', 'plan', 'clean-up', u'affect', u'site', 'monitor', 'health', u'worker', '.', u'unilev', "'s", u'websit', u'state', u'wait', u'sinc', '2010', 'loca l', u'govern', 'tamil', 'nadu', u'pollut', 'control', 'board', 'give', u'permiss', 'clean-up', '.', '==', 'see', '==', 'list', 'food', u'comp ani', '==', u'refer', '==', '==', u'extern', u'link', '==', u'offici', u'websit'l

## Put the cleaned word into a string

## Find a different sort of wiki pages

```
In [98]: t = wikipedia.search('asia',results=30)
        plist = []
        for p in t:
            print p
            if p =="XXX": continue
            page = wikipedia.page(p, auto_suggest=False)
            plist.append(page)
        tokenized_docs_list = []
        for i in plist:
            tokenized docs = word_tokenize(i.content.encode('ascii','ignore'))
            tokenized_docs_list.append(tokenized_docs)
        #----#
        t_nsw_list = []
        for p in tokenized_docs_list:
            tokenized nsw = []
            for i in p:
                if not i in stopwords.words():
                    tokenized nsw.append(i)
            t nsw list.append(tokenized nsw)
        #----#
        after_stem_list = []
        for p in t_nsw_list:
            words = []
            for i in p:
                words.append(porter.stem(i))
            after_stem_list.append(words)
        #----#
        w str list2 = []
        for p in after_stem_list:
            word_str = ''
            for i in p:
                if (i!=p[0]):
                   word_str += ' '
                word str += i
            w str list2.append(word str)
         #----#
```

```
Asia
Unilever
Eurasia
Eric Hoffer
Shamanism
Asian giant hornet
List of Asian pornographic actors
Georgia (country)
Calligraphy
Hornet
Asia Argento
Time (magazine)
Pacific War
Boundaries between the continents of Earth
Demographics of Russia
List of most common surnames in Asia
Humid subtropical climate
Aedes albopictus
Buddhism
Lists of World Heritage Sites
XXX
Shanghai SIPG F.C.
Alcohol flush reaction
Asian Boyz
KLM
XXX (Asia album)
Pre-Indo-European languages
Tatars
Diospyros kaki
Indian subcontinent
```

```
In [110]: f= open("wiki_result.csv","a")
    writer = csv.writer(f, delimiter='|')
    for i in w_str_list2:
        p = ['1',i]
        writer.writerow(p)
    f.close()
```

#### In [96]:

[u'Asia', u'Unilever', u'Eurasia', u'Eric Hoffer', u'Shamanism', u'Asia n giant hornet', u'List of Asian pornographic actors', u'Georgia (count ry)', u'Calligraphy', u'Hornet', u'Asia Argento', u'Time (magazine)', u'Pacific War', u'Boundaries between the continents of Earth', u'Demog raphics of Russia', u'List of most common surnames in Asia', u'Humid su btropical climate', u'Aedes albopictus', u'Buddhism', u'Lists of World Heritage Sites', u'XXX', u'Shanghai SIPG F.C.', u'Alcohol flush reacti on', u'Asian Boyz', u'KLM', u'XXX (Asia album)', u'Pre-Indo-European la nguages', u'Tatars', u'Diospyros kaki', u'Indian subcontinent']

## Open the file that recorded clean wiki pages

```
In [40]: from pyspark.sql import Row
    from pyspark.sql import SparkSession
    import csv
    spark = SparkSession.builder.getOrCreate()
    f = open("wiki_result.csv")
    reader = csv.reader(f,delimiter='|')
    ww = []
    for w in reader:
        ww.append(w)
    ww= map(lambda p: Row(label=int(p[0]), text=str(p[1])),ww)
    wikies = spark.createDataFrame(ww)
    wikies.show()
    f.close
```

```
+----+
label
    0|asia ( ) earth 's...|
    0 unilev ( ) dutch-...
    0 eurasia combin co...
    0 eric hoffer ( jul...
    0 shaman ( shah-men...
    0 the asian giant h...
    0 list asian pornog...
    0 georgia (; georg...
    0 calligraphi (gre...
    0 hornet (insect g...
    0|asia argento ( it...
    0 time american wee...
    0 the pacif war , s...
    0 the boundari cont...
    0 the demograph rus...
    0 thi list common s...
    0 A humid subtrop c...
    0 aed albopictu ( s...
    0|buddhism ( ) reli...
    0 thi list list wor...
   ---+----+
only showing top 20 rows
```

Out[40]: <function close>

### Do tokenize

```
In [41]: from pyspark.ml.feature import HashingTF, IDF, Tokenizer
    tokenizer = Tokenizer(inputCol="text", outputCol="words")
    wordsData = tokenizer.transform(wikies)
    wordsData.show()
```

```
label
                      text
                                          words
     0|asia ( ) earth 's...|[asia, (, ), eart...|
    0 | unilev ( ) dutch-... | [unilev, (, ), du... |
     0 eurasia combin co... [eurasia, combin,...]
    0|eric hoffer ( jul...|[eric, hoffer, (,...
     0|shaman ( shah-men...|[shaman, (, shah-...
     0 the asian giant h... [the, asian, gian...
     0 list asian pornog... [list, asian, por...
     0 georgia (; georg... georgia, (, ;, g...
     0 calligraphi ( gre... calligraphi, (, ...
     0|hornet ( insect g...|[hornet, (, insec...
    0 asia argento (it... asia, argento, (...
    0 time american wee... [time, american, ...]
     0 the pacif war , s... [the, pacif, war,...
     0 the boundari cont... [the, boundari, c...
    0|the demograph rus...|[the, demograph, ...
    0 thi list common s... [thi, list, commo...
     0 A humid subtrop c... [a, humid, subtro...
     0 aed albopictu ( s... [aed, albopictu, ...
    0 | buddhism ( ) reli... | [buddhism, (, ), ... |
    0 thi list list wor... [thi, list, list,...]
        -----+
only showing top 20 rows
```

Do tokenize

```
In [42]: hashingTF = HashingTF(inputCol="words", outputCol="rawFeatures",numFeatures=300)
    featurizedData = hashingTF.transform(wordsData)
    featurizedData.show()
```

```
label
                       text
                                           words
                                                          rawFeatures
    0|asia ( ) earth 's...|[asia, (, ), eart...|(300,[0,1,2,3,4,5...
    0 unilev ( ) dutch-... [unilev, (, ), du... (300,[0,1,2,3,4,6...
    0 | eurasia combin co... | [eurasia, combin,... | (300,[0,1,2,3,4,5...
    0|eric hoffer ( jul...|[eric, hoffer, (,...|(300,[0,1,2,3,4,5...
    0|shaman (shah-men...|[shaman, (, shah-...|(300,[0,1,2,3,4,5...
    0 \mid \text{the asian giant h...} \mid \text{[the, asian, gian...} \mid (300, [0, 1, 2, 3, 4, 5...]
    0|list asian pornog...|[list, asian, por...|(300,[1,4,8,10,12...
    0|georgia (; georg...|[georgia, (,;, g...|(300,[0,1,2,3,4,5...
    0|calligraphi ( gre...|[calligraphi, (, ...|(300,[0,1,2,3,4,5...
    0|hornet (insect g...|[hornet, (, insec...|(300,[0,1,3,4,5,6...
    0|asia argento ( it...|[asia, argento, (...|(300,[0,1,2,3,4,5...
    0 time american wee... [time, american, ... (300,[0,1,2,3,4,5...
    0 the pacif war, s... [the, pacif, war,... (300,[0,1,2,3,4,5...
    0|the boundari cont...|[the, boundari, c...|(300,[0,1,2,3,4,5...
    0 the demograph rus... [the, demograph, ... (300, [0,1,2,3,4,5...
    0 thi list common s... [thi, list, commo... (300,[1,4,5,7,9,1...
    0 | A humid subtrop c... | [a, humid, subtro... | (300, [0,1,2,3,4,5...
    0|aed albopictu ( s...|[aed, albopictu, ...|(300,[0,1,2,3,4,5...
    0 | buddhism ( ) reli... | [buddhism, (, ), ... | (300,[0,1,2,3,4,5...
    0 thi list list wor... [thi, list, list,... (300,[1,2,3,5,6,1...
        ._____+__+
```

only showing top 20 rows

```
In [43]: idf = IDF(inputCol="rawFeatures", outputCol="features")
    idfModel = idf.fit(featurizedData)
    rescaledData = idfModel.transform(featurizedData)

rescaledData.select("text", "features").show()
    rescaledData = rescaledData.select("label", "features")
```

```
text
                                  features
asia ( ) earth 's... | (300, [0,1,2,3,4,5... |
unilev ( ) dutch-... | (300,[0,1,2,3,4,6...
eurasia combin co... (300,[0,1,2,3,4,5...)
eric hoffer ( jul... | (300,[0,1,2,3,4,5...
shaman ( shah-men... |(300,[0,1,2,3,4,5...]
the asian giant h... |(300,[0,1,2,3,4,5...]
list asian pornog... (300,[1,4,8,10,12...
georgia (; georg...|(300,[0,1,2,3,4,5...
calligraphi ( gre... | (300, [0,1,2,3,4,5...
hornet (insect g... | (300, [0,1,3,4,5,6...
asia argento ( it... | (300,[0,1,2,3,4,5...
time american wee... (300,[0,1,2,3,4,5...]
the pacif war, s... |(300,[0,1,2,3,4,5...]
the boundari cont... (300,[0,1,2,3,4,5...
the demograph rus... |(300,[0,1,2,3,4,5...]
thi list common s... (300,[1,4,5,7,9,1...
A humid subtrop c... (300, [0, 1, 2, 3, 4, 5...]
aed albopictu ( s... | (300,[0,1,2,3,4,5...
buddhism ( ) reli... | (300, [0,1,2,3,4,5...
thi list list wor... (300,[1,2,3,5,6,1...
+_____
only showing top 20 rows
```

```
In [44]: from pyspark.ml.clustering import KMeans
    kmeans = KMeans().setK(2).setSeed(1)
    model = kmeans.fit(rescaledData)

wssse = model.computeCost(rescaledData)
print("Within Set Sum of Squared Errors = " + str(wssse))

centers = model.clusterCenters()
print("Cluster Centers: ")
for center in centers:
    print(center)
```

Within Set Sum of Squared Errors = 53447.3276304 Cluster Centers:

```
3.64005606
                               6.9126512
                                             2.73938205
                                                           3.9037373
   2.33138898
                 7.42197374
                               4.86614262
                                             7.00360714
                                                           3.52862577
                                                                         0.
   3.47475518
                 5.25497411
                               2.3896737
                                             2.76852441
                                                           0.71731246
   4.4073333
                 1.77148904
                               4.51706615
                                             5.45735621
                                                           6.33847393
                                                                         0.
   6.18741141
                 2.22842045
                               2.76852441
                                             2.78838379
                                                           2.8098329
   2.76852441
                 2.42906735
                               7.18551901
                                             6.67360095
                                                           1.72702476
   4.41136294
                 5.95761387
                               6.32143761
                                             4.320407
                                                           1.47998832
   3.52862577
                 3.04577306
                               4.95709856
                                             3.20289532
                                                           6.69696294
   3.60291263
                 1.39883339
                               3.1250631
                                             4.17147429
                                                           1.76914732
   3.32222929
                 3.06722217
                               0.58626499
                                             6.32674656
                                                           5.42294066
   4.95709856
                 1.93145832
                               4.42006807
                                             2.89562932
                                                           0.
                                                                         3.4
9406762
                 5.77570199
                               3.95477433
                                                           2.65968915
   7.41290886
                                             3.14214947
   4.07533015
                 5.25497411
                                             4.7132242
                                                           2.71023969
   7.55923641
                 3.00861776
                               1.63013206
                                             2.25215614
                                                           2.56452788
                 2.42374899
                                             3.98585034
                                                           1.79722902
   1.58723385
                               0.
   1.58661625
                 4.1172993
                               2.9343309
                                             2.33138898
                                                           6.57438696
   5.25503099
                 3.68371544
                               2.9796245
                                             3.98585034
                                                           3.46719941
   5.1844884
                 2.9254495
                               4.42872259
                                             5.52584906
                                                           1.77315422
   6.98857382
                 2.46743116
                               2.31674054
                                             4.43996496
                                                           2.97147433
                                                                         0.
   1.51049451
                 2.3896737
                               2.62094938
                                             7.25944877
                                                           7.03758856
   2.31650346
                 4.03243194
                               1.58661625
                                             5.3486538
                                                           3.73807437
   2.6743269
                 2.18780882
                               4.36545513
                                             3.23147834
                                                           1.30357746
                 6.64308389
   3.96863931
                               8.95156643
                                             3.64279528
                                                           4.4004967
                 3.90059934
                               2.35940167
                                                           4.19720749
   5.82117996
                                             2.76852441
   3.61365292
                 4.72970872
                               2.78008864
                                             4.80832739
                                                           9.36358492
   3.49148234
                 4.90293265
                               1.80172491
                                            10.0223733
                                                           1.73120604
                                                                         0.
   3.78850709
                 6.16580924
                               3.50180357
                                             2.09825008
                                                           2.85273111
   5.63926809
                 4.58596375
                               8.45129857
                                             1.51641199
                                                           2.04158162
                                                                         4.2
   6.06759897
                 1.89551498
                               4.4572115
                                             2.38085077
                                                           0.
2199838
   3.25049945
                11.96018419
                               0.
                                             3.00861776
                                                           2.63968013
   3.32222929
                 4.01149035
                               2.63824005
                                             3.16337328
                                                           1.73359971
   2.89718747
                                             2.91058998
                                                                         0.
                 3.63823748
                               3.15719148
                                                           2.08782637
   3.95477433
                 3.34606055
                               4.13967747
                                             3.32222929
                                                           3.1959168
   3.15719148
                 5.82117996
                               2.23070703
                                             4.42963906
                                                           2.50954541
   1.64277965
                 3.46794111
                               1.89612468
                                             4.72970872
                                                           3.84679182
   3.15719148
                 3.03683835
                               4.83763213
                                             3.81764945
                                                           3.8257732
  14.93461634
                 3.67193764
                               2.57389273
                                             3.3198917
                                                           4.73176703
   3.08898738
                                             5.05150637
                                                           5.75733937
                 3.12004805
                              10.24932943
   1.96860175
                 1.89261996
                               3.93082607
                                             3.38005205
                                                           6.33847393
   4.7315499
                 2.27360524
                               2.50954541
                                             4.60449323
                                                           5.11087003
  11.24895182
                 2.80817035
                                                           3.26026412
                               4.55525753
                                             3.97280748
                                             8.14055635
   2.36053134
                 1.01389358
                               6.39001403
                                                           2.01621597
   3.26394457
                 3.45404953
                               4.36545513
                                             2.4143229
                                                           2.70258736
   0.92858573
                 3.19433491
                               2.6743269
                                             6.63978339
                                                           5.02272221
   2.68109733
                 5.53227373
                               1.36196262
                                             2.42374899
                                                           6.57981643
                 3.46794111
   1.65533645
                               4.63875278
                                             3.8593154
                                                           4.31127032
   2.54676623
                               2.59396609
                                                           1.92339591
                 3.52862577
                                             1.66223369
   4.60487422
                 4.38639081
                               0.92231156
                                             2.13814292
                                                           2.19146232
   3.79604794
                 2.21845457
                               0.39314241
                                             5.79125864
                                                           5.38168537
   2.95997664
                 4.38292464
                               2.91058998
                                             8.23151229
                                                           5.07077103
   4.43690915
                 1.13114658
                               1.15825173
                                             1.7776841
                                                          11.70448114
   7.83929544
                               2.55543502
                                             2.42374899
                                                           1.95253827
                 2.87418021
                               3.15719148
                                             5.50283418
   2.04650858
                                                           4.61155781
```

```
3.84679182
                 5.67785988
                              2.23249543
                                            8.41457293
                                                          1.76914732
   3.90006006
                 3.95477433
                              4.06312431
                                            7.31362376
                                                          7.74702369
   2.8559515
                 3.96336126
                              3.72578218
                                            3.00154592
                                                          2.23249543
                              4.35416853
                                            3.38005205
                                                          3.00861776
   4.36545513
                 3.4563256
   0.67592905
                 2.08056329
                              8.98398012
                                            3.64005606
                                                          3.00861776
                 3.04702389
   1.41812603
                              1.84462312]
[ 0.8589418
              0.
                           0.86218648
                                        0.53427664 0.92052414
                                                                 0.7771296
  1.35437477
              0.78638987
                           1.02325429
                                        0.71965394
                                                                 0.6792216
9
                                        0.30175404
  0.97063525
              1.00784003
                           0.60713255
                                                    1.22569617
                                                                 0.7776626
  0.54034797
              2.42549165
                           1.04964045
                                                     1.04640046
                                        0.
                                                                 0.4547796
8
  0.38856483
              0.63006749
                           0.33514228
                                        0.78320099
                                                     0.97408409
                                                                 1.1085254
8
  1.40854751
              0.31591916
                           1.14642379
                                        1.43066109
                                                     1.18432209
                                                                 1.1180000
6
  0.40217074
              0.97501502
                           0.49601058
                                        0.99483056
                                                     1.44069628
                                                                 0.8795829
9
  0.95180037
              0.52213399
                           0.65105481
                                        0.6207551
                                                     0.46510321
                                                                 0.7042737
5
  0.74178158
              0.17099396
                           1.27853003
                                        1.17620859
                                                     0.84323733
                                                                 1.9190771
7
  0.51846037
              1.26907211
                           0.
                                        1.28904404
                                                     1.07062717
                                                                 1.0137797
1
  0.85777069
              0.80133841
                           0.53622765
                                        0.57197616
                                                     0.99320817
  1.68168201
              0.74070171
                           0.8943541
                                        0.82025073
                                                     0.7149702
                                                                 0.6345360
5
              0.49601058
  0.55249062
                           0.67028456
                                        0.
                                                     0.83038549
                                                                 0.4124500
2
  0.38027307
              1.09478627
                           0.46429286
                                        0.6678458
                                                     1.06013537
                                                                 0.9652097
7
  0.82428818
              1.10607273
                           0.83038549
                                        0.53046345
                                                     1.44013567
                                                                 0.8351977
R
  0.79084332
              2.79904119
                           1.16654883
                                        1.70425492
                                                    0.94901198
                                                                 0.5528585
4
  0.94286695
              0.97501502
                                        0.43395105
                                                     0.65570315
                                                                 0.5690219
  1.0609269
              1.17293143
                           0.46026207
                                        0.64347318
                                                     0.21646313
                                                                 0.7660832
3
  0.59818219
              0.50298394
                           0.69262738
                                        0.84356621
                                                     0.79703608
                                                                 0.3118125
1
  0.97910509
              0.98855415
                           1.13751752
                                        0.6192752
                                                     0.78320099
                                                                 0.9758814
1
  0.58689573
              0.55410191
                           0.93498412
                                        0.76608323
                                                     0.97141207
                                                                 0.8527119
1
                                                     0.72739215
  1.00333586
              0.7776626
                           1.79257819
                                        1.33871109
                                                                 0.7641244
                                                     0.85120358
  0.82391132
              0.2816371
                           0.
                                        1.14140919
                                                                 0.9095593
7
  0.74070171
              0.77306153
                           0.75796614
                                        2.51742719
                                                     1.21893729
                                                                 0.4387766
2
  0.31324943
              0.68847384
                           0.41830037
                                        0.68096287
                                                     0.65241031
                                                                 0.
  1.36487016
              0.48531763
                           1.1297793
                                                     0.5881043
                                                                 1.1817716
9
                           0.58091329
                                        0.59313249
  0.83177159
              1.0214443
                                                     0.76747904
                                                                 1.3387110
9
  0.85271191 0.61131894
                          0.57794918
                                        0.67221682
                                                    0.
                                                                 0.6207551
```

			Q2 Wha com	mae		
1	0.75518728	0.78646722	1.05641063	0.57644473	0.42560179	1.4211865
	0.56303903	0.62534652	0.70156451	0.3890486	0.81962894	0.4966040
8	0.91903395	0.73463038	1.0214443	0.93583126	1.86389692	0.5646332
7	1.08335002	1.89293651	0.84998557	0.91605557	0.63479664	1.0384379
9	0.84830146	1.45478431	1.04127704	1.00596787	0.68539754	0.4642928
6	0.66726986	0.77342643	0.58036608	1.5010987	0.7582612	0.6881588
2	0.50494771	1.14140919	0.6230628	0.6192752	0.62891315	0.9094698
2	0.74432663	0.74178158	0.51606266	0.40952334	0.8962891	1.0327288
7	0.51388483	0.49177736	0.92728125	1.29171076	0.65774822	0.5719761
6	0.93632394	0.68096287	1.02918252	1.20327125	1.03123524	0.8317715
9	0.66726986	0.2720415	0.54963334	0.71175899	0.21122783	1.1899797
9	0.66322037	0.76448188	0.65241031	0.58742376	0.88215644	2.9788431
7	0.33480329	0.54034797	0.82391132	0.69625251	0.37089079	0.5704588
3				0.16237241		
6	0.6422718	1.12036137	0.5060557		1.59527726	0.8969475
1	1.27800924	1.15299395	0.72006783	1.07062717	0.78927231	1.2695270
	0.27876326 0.81904968	0.29939377 0.65687887	0.63748917 0.41892146	0.94901198 0.72006783	1.11712389 0.43345069	0. 1.0137060
9	1.37381363	0.91605557	0.80141496	1.09189613	0.44755215	1.0412770
4	0.62013762	0.92084751	0.82391132	0.81262486	1.21893729	0.9254894
3	0.57070459	1.01998268	0.68995966	0.85271191	0.26911632	0.9094698
2	0.88113564	0.95627265	0.71965394	0.70417751	0.20116936	0.5004791
4	1.1994457	1.11430287	1.16847038	0.33054505	0.73901699	0.7417815
0		1.1143020/	1.1004/030	0.33034303	0./3501033	0.741/013

## **Summary**

8]

At first I took 10 pages for each category. The Error is around 2000. When I finally used 30 pages for each category. The Error increased to 50000. So the model is not effective to cluster.

# next we perform the classification to the wikipages. Question3-wiki:

Test set accuracy = 0.382978723404

In [47]: from pyspark.ml.classification import DecisionTreeClassifier
 from pyspark.ml.evaluation import MulticlassClassificationEvaluator

dt = DecisionTreeClassifier()

model = dt.fit(train)

predictions = model.transform(test)

evaluator = MulticlassClassificationEvaluator(labelCol="label", predictionCol="prediction", metricName="accuracy")
 accuracy = evaluator.evaluate(predictions)
 print("Test set accuracy = " + str(accuracy))

Test set accuracy = 0.382978723404

In [48]: from pyspark.ml.classification import RandomForestClassifier
 from pyspark.ml.evaluation import MulticlassClassificationEvaluator

 rf = RandomForestClassifier()

 model = rf.fit(train)

 predictions = model.transform(test)

 evaluator = MulticlassClassificationEvaluator(labelCol="label",predictionCol="prediction",metricName="accuracy")
 accuracy = evaluator.evaluate(predictions)
 print("Test set accuracy of RandomForest= " + str(accuracy))

Test set accuracy of RandomForest= 0.382978723404

#### PIMA DATASET

already preprocessed the dataset.

```
In [3]: from pyspark.ml.feature import HashingTF, IDF, Tokenizer
    from pyspark.sql import SparkSession
    from pyspark.sql.functions import monotonically_increasing_id
    spark = SparkSession.builder.getOrCreate()

data = spark.read.format("csv").option("header",True).option("inferSchem
a",True).\
    load("pima/pima-indians-diabetes.data")
```

#### **Vertorize the Data into feature**

```
In [4]: from pyspark.ml.feature import VectorAssembler
label = ["label"]
assembler = VectorAssembler(
                inputCols=[x for x in data.columns if x not in label],
                outputCol='features')
data = assembler.transform(data)
```

### Split the Data

```
In [5]: splits = data.select("label", "features").randomSplit([0.8, 0.2], 1234)
    train = splits[1]
    test = splits[0]
```

### Use NaiveBayes method to build a model

Test set accuracy = 0.61648177496

#### Use DecisionTree method to build a model

```
In [7]: from pyspark.ml.classification import DecisionTreeClassifier
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator

    dt = DecisionTreeClassifier()

    model = dt.fit(train)

    predictions = model.transform(test)

    evaluator = MulticlassClassificationEvaluator(labelCol="label", predictionCol="prediction", metricName="accuracy")
    accuracy = evaluator.evaluate(predictions)
    print("Test set accuracy = " + str(accuracy))
```

Test set accuracy = 0.698890649762

#### Use RandomForest method to build a model

```
In [10]: from pyspark.ml.classification import RandomForestClassifier
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator

rf = RandomForestClassifier()

model = rf.fit(train)

predictions = model.transform(test)

evaluator = MulticlassClassificationEvaluator(labelCol="label",predictionCol="prediction",metricName="accuracy")
    accuracy = evaluator.evaluate(predictions)
    print("Test set accuracy of RandomForest= " + str(accuracy))
```

Test set accuracy of RandomForest= 0.765451664025

## **Statlog DATASET**

already preprocessed the dataset.

```
In [13]: from pyspark.ml.feature import HashingTF, IDF, Tokenizer
         from pyspark.sql import SparkSession
         from pyspark.sql.functions import monotonically_increasing_id
         from pyspark.sql import Row
         import csv
         spark = SparkSession.builder.getOrCreate()
         f = open("pima/australian.dat")
         reader = csv.reader(f,delimiter=' ')
         ww = []
         for w in reader:
             ww.append(w)
         data = map(lambda p: Row(label=int(p[0]),
         a_1=float(p[1]),a_2=float(p[2]),
                                         label_2=int(p[3]),label_3=int(p[4]),label
         _4=int(p[5]),
                                         a_3=float(p[6]),label_5 =int(p[7]),label_
         6=int(p[8]),
                                         a_4=float(p[9]),label_7=int(p[10]),label_
         8=int(p[11]),
         a_5=float(p[12]),a_6=float(p[13]),label_9=int(p[14]))
         data = spark.createDataFrame(data)
         f.close()
```

In [14]: data.show()

a\_2| a\_3| a\_4| a\_5| a\_6|label\_2|label\_3|label\_4|labe 1\_5|label\_6|label\_7|label\_8|label\_9| |22.08|11.46|1.585| 0.0|100.0|1213.0| 1 | 2 | 4 | 0 0 1 2 |22.67| 7.0|0.165| 0.0|160.0| 0 | 2 | 8 4 1.0 0 0 2 0 | |29.58| 1.75| 1.25| 0.0|280.0| 0 | 1 | 4 | 4 | 1.0 0 0 1 2 0 |21.67| 11.5| 0.0|11.0| 0.0| 0 | 1 | 5 | 3 | 1.0 1 1 | 2 |20.17| 8.17| 1.96|14.0| 60.0| 159.0| 2 1 6 4 1 | 0 | 2 | |15.83|0.585| 1.5| 2.0|100.0| 0 | 2 | 8 | 8 | 1.0 1 0 | 2 |17.42| 6.5|0.125| 0.0| 60.0| 101.0| 1 | 2 | 3 | 4 | 0 | 0 | 2 | |58.67 | 4.46 | 3.04 | 6.0 | 43.0 | 561.0 | 0 | 2 | 8 | 11 0 | 2 | 1 |27.83| 1.0| 3.0| 0.0|176.0| 538.0| 1 | 1 | 2 | 8 | 0 0 2 |55.75| 7.08| 6.75| 3.0|100.0| 51.0| 0 | 2 4 8 1 2 1 | 33.5 | 1.75 | 4.5 | 4.0 | 253.0 | 858.0 | 1 | 2 | 14 8 | 1 | 2 | 1 | |41.42| 5.0| 5.0| 6.0|470.0| 1.0 1 | 2 | 11| 8 | 1 | 1 2 |20.67| 1.25|1.375| 3.0|140.0| 211.0| 1 | 1 | 8 | 8 | 1 | 1 | 2 | |34.92| 5.0| 7.5| 6.0| 0.0|1001.0| 2 | 1 | 14| 8 | 1 1 2 | |58.58| 2.71|2.415| 0.0|320.0| 1 | 2 | 8 | 4 | 1.0 0 | 1 | 2 | |48.08| 6.04| 0.04| 0.0| 0.0|2691.0| 1 | 2 | 4 | 4 | 0 | 0 | 2 | |29.58| 4.5| 7.5| 2.0|330.0| 1 | 2 | 9 | 4 | 1 2 1 | 1 | |18.92| 9.0| 0.75| 2.0| 88.0| 592.0| 0 | 2 | 6 4 | 1 0 2 1 | 20.0 | 1.25 | 0.125 | 0.0 | 140.0 | 1 | 1 | 4 | 4 | 5.0 0 0 0 2 |22.42|5.665|2.585| 7.0|129.0|3258.0| 0 | 2 | 11| 4 | 1 0 | 2 | ---+----+

only showing top 20 rows

Vertorize the Data into feature

```
In [15]: from pyspark.ml.feature import VectorAssembler
label = ["label"]
assembler = VectorAssembler(
          inputCols=[x for x in data.columns if x not in label],
          outputCol='features')
data = assembler.transform(data)
data.show()
```

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```
a_2| a_3| a_4| a_5|
                              a_6|label|label_2|label_3|label_4|labe
                                   features
1_5|label_6|label_7|label_8|label_9|
|22.08|11.46|1.585| 0.0|100.0|1213.0| 1|
                                                           4 |
                       2
             1 |
                              0 | [22.08,11.46,1.58... |
        7.0 | 0.165 | 0.0 | 160.0 |
                                     0 | 2 |
22.67
                              1.0
         0 |
                0 | 2 |
                             0|[22.67,7.0,0.165,...|
|29.58| 1.75| 1.25| 0.0|280.0|
                             1.0 0 1
                                                           4 |
                       2 |
         0
             1 |
                             0|[29.58,1.75,1.25,...|
                                                           3 |
|21.67| 11.5| 0.0|11.0| 0.0|
                              1.0 0 1
         1
                1 |
                       2
                             1 [21.67,11.5,0.0,1...]
|20.17| 8.17| 1.96|14.0| 60.0| 159.0| 1|
                                            2
                                                           4
               0 | 2 |
                              1 | [20.17,8.17,1.96,... |
       1 |
|15.83|0.585| 1.5| 2.0|100.0|
                              1.0 0
                                        2
                                                           8 |
                      2
                             1 | [15.83,0.585,1.5,... |
                0 |
        6.5|0.125| 0.0| 60.0| 101.0| 1| 2|
                0 2
                               0 | [17.42,6.5,0.125,... |
|58.67| 4.46| 3.04| 6.0| 43.0| 561.0| 0| 2|
                                                           8
        1 |
                0 2
                               1 | [58.67, 4.46, 3.04, ... |
|27.83| 1.0| 3.0| 0.0|176.0| 538.0| 1| 1|
                                                           8 |
        0
                0 2
                               0 | [27.83,1.0,3.0,0....
|55.75| 7.08| 6.75| 3.0|100.0| 51.0| 0|
                                                           8 |
             1 | 2 |
        1
                              0 | [55.75,7.08,6.75,... |
33.5 1.75
            4.5 | 4.0 | 253.0 | 858.0 | 1 |
                                        2
                                                           8 |
             1 2 |
         1 |
                               1 | [33.5, 1.75, 4.5, 4.... |
            5.0 | 6.0 | 470.0 |
41.42
        5.0
                              1.0 | 1 | 2 |
                                                           8 |
             1 2
                             1 | [41.42,5.0,5.0,6.... |
        1 |
|20.67| 1.25|1.375| 3.0|140.0| 211.0| 1| 1|
                                                           8 |
                       2 |
                               0 | [20.67,1.25,1.375...]
               1 |
|34.92| 5.0| 7.5| 6.0| 0.0|1001.0| 1| 2|
                                                           8 |
                1 |
         1 |
                       2 |
                               1 | [34.92,5.0,7.5,6.... |
|58.58| 2.71|2.415| 0.0|320.0|
                              1.0 | 1 |
                                         2 |
                                                           4 |
                       2 |
         0 |
                             0|[58.58,2.71,2.415...|
            1 |
|48.08| 6.04| 0.04| 0.0| 0.0|2691.0|
                                     1 2
                       2 |
         0 |
                0 |
                               1 | [48.08,6.04,0.04,...|
                              1.0 | 1 | 2 |
29.58
        4.5 | 7.5 | 2.0 | 330.0 |
                1 |
                       2 |
                               1 | [29.58, 4.5, 7.5, 2.... |
        1 |
        9.0 | 0.75 | 2.0 | 88.0 | 592.0 | 0 | 2 |
                                                           4 |
                0 | 2 |
                             1 | [18.92,9.0,0.75,2... |
20.0 | 1.25 | 0.125 | 0.0 | 140.0 |
                              5.0 1
                                            1 |
                0 |
                       2
                               0 | [20.0, 1.25, 0.125, ... |
|22.42|5.665|2.585| 7.0|129.0|3258.0| 0|
                                            2 |
                       2 |
                               1 | [22.42,5.665,2.58... |
only showing top 20 rows
```

#### Split the Data

```
In [17]: splits = data.select("label", "features").randomSplit([0.8, 0.2], 1234)
    train = splits[1]
    test = splits[0]
```

#### Use NaiveBayes method to build a model

Test set accuracy = 0.388791593695

#### Use DecisionTree method to build a model

```
In [19]: from pyspark.ml.classification import DecisionTreeClassifier
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator

dt = DecisionTreeClassifier()

model = dt.fit(train)

predictions = model.transform(test)

evaluator = MulticlassClassificationEvaluator(labelCol="label", predictionCol="prediction", metricName="accuracy")
    accuracy = evaluator.evaluate(predictions)
    print("Test set accuracy = " + str(accuracy))
```

Test set accuracy = 0.647985989492

#### Use RandomForest method to build a model

```
In [20]: from pyspark.ml.classification import RandomForestClassifier
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator

rf = RandomForestClassifier()

model = rf.fit(train)

predictions = model.transform(test)

evaluator = MulticlassClassificationEvaluator(labelCol="label",predictionCol="prediction",metricName="accuracy")
    accuracy = evaluator.evaluate(predictions)
    print("Test set accuracy of RandomForest= " + str(accuracy))
```

Test set accuracy of RandomForest= 0.647985989492

## **Summary**

Naive Bayes model fails when the input data is very independent. And the classifiers used in this experiment give out poor performance on continuous data.