```
In [1]:
            import nltk
            # python -m nltk.downloader all ##type this on command line to download required
  In [163]:
             import warnings
             warnings.filterwarnings('ignore')
   In [2]:
            import pandas as pd
            import os
            os.chdir("D:/PlayGround/NLP/")
   In [3]: | df_init = pd.read_excel('keyword_grouping.xlsx')

    In [4]: df_init.info()

               <class 'pandas.core.frame.DataFrame'>
               RangeIndex: 6554 entries, 0 to 6553
               Data columns (total 2 columns):
               Keyword
                                   6553 non-null object
               Refined Keyword
                                   3391 non-null object
               dtypes: object(2)
               memory usage: 102.5+ KB
   In [5]:
            df_init[:5]
   Out[5]:
                                      Keyword Refined Keyword
             0 0_4da850ed6c-18e2076de5-99212033
                                                        NaN
             1
                        0_5f41bbf30e-9fad566447-
                                                        NaN
             2 0_161dc7017d-fb01084e85-186735817
                                                        NaN
             3 0_161dc7017d-fb01084e85-186735837
                                                        NaN
             4 0 161dc7017d-fb01084e85-186735845
                                                        NaN
            keywords = df init['Refined Keyword'].value counts()
   In [6]:
```

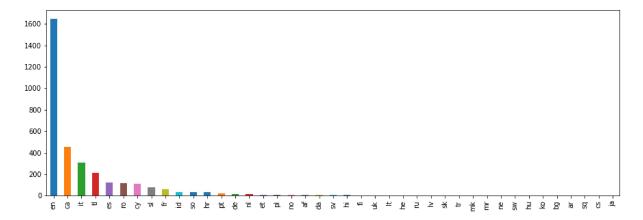
```
In [9]:
              keywords.plot(kind = 'bar', figsize = (15,5))
 Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x1bdf9b42978>
                    500
                    400
                    300
                    200
                    100
                                                                    GL - Poland
                                                                                      High-Tech
                        GL - India
                                          Technical Keyword
                                                                                                                                           Content Engineering
                                                                                                                                              Mobile Application Development
                                    Ukraine - Careers
                                             Cashless Economy
                                                                          Digital Product Engineering
                                                                             Retail / e-Commerce
                                                                                   Customer Focused
                                                                                              Slovakia - Careers
                                                                                                  Big Data & Analytics
                                                                                                       Communications & Media
                                                                                                                Electronics & Industrial
                                                                                                                   Application Development
                                                                                                                               Ideas / POC
                              Ę
                                 India - Careers
                                       GlobalLogic
                                                Corporate
                                                         Poland - Careers
                                                           LatAm - Careers
                                                                                         GL - Israel
                                                                                                          OTT Application Services
                                                                                                                            Experience Design
                                                      Mobile App Development
In [11]: | df_init['Keyword'].nunique()
Out[11]: 6553
In [12]:
              df = df_init.copy()
In [13]:
               # Remove NaN/Empty
               df = df.dropna()
In [14]:
               df.info()
                   <class 'pandas.core.frame.DataFrame'>
                   Int64Index: 3391 entries, 77 to 6553
                   Data columns (total 2 columns):
                   Keyword
                                                  3391 non-null object
                   Refined Keyword
                                                  3391 non-null object
                   dtypes: object(2)
                   memory usage: 79.5+ KB
In [15]:
               # checking if keyword has non-english characters. 0 means no non-english chars.
               df['non_eng'] = 0
In [35]:
               # function to detect Language
               def detect_language(line):
                     maxchar = max(line)
                     # print(maxchar, ord(maxchar))
                     if ord(maxchar)>127:
                            return 1
                     else:
                            return 0
```

```
In [36]:
         for index,row in df.iterrows():
             df['non_eng'][index] = detect_language(row['Keyword'])
In [37]: | df['non_eng'].value_counts()
Out[37]:
         0
              3310
                81
         Name: non_eng, dtype: int64
In [38]:
         df[:5]
Out[38]:
                       Keyword
                                    Refined Keyword non_eng
          77
                   1mobile market Mobile App Development
                                                        0
          78
               2 2 channel ukraine
                                        GL - Ukraine
                                                        0
          79
              5 ETAPAS DE BPMN
                                    Technical Keyword
                                                        0
             5 ETAPAS DEL BPMN
                                    Technical Keyword
                                                        0
          81
                   5 g live mobile Mobile App Development
                                                        0
In [63]:
         # function to split urls
         def split_url(line):
             import re
             non empty = []
             for word in words:
                 if len(word)>0:
                      non empty.append(word)
             return ' '.join(non_empty)
         # converting urls to string of space separated words
In [61]:
         for index, row in df.iterrows():
             df['Keyword'][index] = split url(str(row['Keyword']))
         # split url('<http://www.globallogic.com/>')
         # - *- coding: utf- 8 - *-
In [39]:
In [40]:
         # detecting language in the text.
         # Adding new language column and setting values to english.
         df['lang'] = 'en'
```

```
In [62]: # detecting Language of the keywords
          from langdetect import detect
          for index, row in df.iterrows():
              try:
                   df['lang'][index] = detect(row['Keyword'])
              except:
                  print(row)
             Keyword
                                 career@globallogic
             Refined Keyword
                                             Careers
             non_eng
                                                    0
             lang
                                                   en
             Name: 444, dtype: object
In [65]:
         df['Keyword'][444]='career globallogic'
          df['lang'][444] = detect('career globallogic')
In [66]:
          df[:5]
In [74]:
Out[74]:
                        Keyword
                                      Refined Keyword non eng
           77
                    1mobile market
                                Mobile App Development
                                                            0
                                                                 tr
           78
                2 2 channel ukraine
                                          GL - Ukraine
                                                            0
                                                                de
               5 ETAPAS DE BPMN
                                      Technical Keyword
           79
                                                            0
                                                                de
              5 ETAPAS DEL BPMN
                                      Technical Keyword
           80
                                                            0
                                                                de
           81
                    5 g live mobile Mobile App Development
                                                                 s
In [75]:
          df.info()
             <class 'pandas.core.frame.DataFrame'>
             Int64Index: 3391 entries, 77 to 6553
             Data columns (total 4 columns):
             Keyword
                                 3391 non-null object
             Refined Keyword
                                 3391 non-null object
                                 3391 non-null object
             non eng
             lang
                                 3391 non-null object
             dtypes: object(4)
             memory usage: 292.5+ KB
```

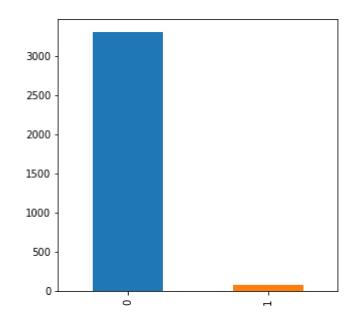
```
In [70]: df['lang'].value_counts().plot(kind = 'bar', figsize=(15, 5))
```

Out[70]: <matplotlib.axes._subplots.AxesSubplot at 0x1bdfc8454a8>



```
In [73]: df['non_eng'].value_counts().plot(kind = 'bar', figsize=(5, 5))
```

Out[73]: <matplotlib.axes._subplots.AxesSubplot at 0x1bdfb693278>



```
In [81]: # separating df with only english characters
df_eng = df[df['non_eng']==0]
```

```
In [88]: | df_eng['lang'].value_counts()[:5]
```

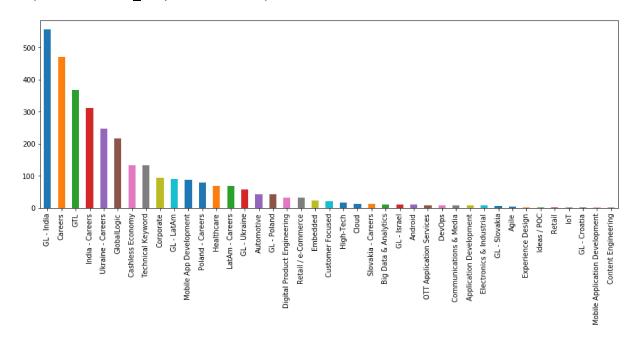
Out[88]: en 1643 ca 446 it 306 tl 215 ro 118

Name: lang, dtype: int64

```
In [82]:
         from sklearn.feature extraction.text import TfidfVectorizer, CountVectorizer
          # create a count vectorizer object
          count vect = CountVectorizer(analyzer='word')
          count vect.fit(df eng['Keyword'])
Out[82]: CountVectorizer(analyzer='word', binary=False, decode_error='strict',
                  dtype=<class 'numpy.int64'>, encoding='utf-8', input='content',
                  lowercase=True, max_df=1.0, max_features=None, min_df=1,
                  ngram range=(1, 1), preprocessor=None, stop words=None,
                  strip_accents=None, token_pattern='(?u)\\b\\w\\w+\\b',
                  tokenizer=None, vocabulary=None)
         # transform the data using count vectorizer object
In [83]:
          keyword_feature_matrix = count_vect.transform(df_eng['Keyword'])
In [84]: keyword_feature_matrix.shape
Out[84]: (3310, 2365)
          features = count_vect.get_feature_names()
In [89]:
         tokenize df = pd.DataFrame(keyword feature matrix.toarray(), columns = features)
In [90]:
In [91]:
         tokenize_df.info()
            <class 'pandas.core.frame.DataFrame'>
            RangeIndex: 3310 entries, 0 to 3309
            Columns: 2365 entries, 003 to zppakode
            dtypes: int64(2365)
            memory usage: 59.7 MB
In [92]:
         tokenize df[:5]
Out[92]:
             003
                 005 0221210211 08 09
                                       0b19a637cc 0h9ms5fxgtkj 10 1042 10698 ... you your you
          0
               0
                   0
                              0
                                 0
                                     0
                                                               0
                                                                                  0
                                                                                       0
          1
               0
                   0
                              0
                                 0
                                     0
                                               0
                                                            0
                                                               0
                                                                    0
                                                                          0
                                                                                  0
                                                                                       0
          2
               0
                   0
                              0
                                 0
                                     0
                                                0
                                                               0
                                                                          0
                                                                                  0
                                                                                       0
          3
                                                                          0 ...
               0
                   0
                              0
                                 0
                                     0
                                                0
                                                            0
                                                               0
                                                                    0
                                                                                  0
                                                                                       0
                              0
                                                0
                                                              0
                                                                          0 ...
                                                                                       0
               0
                   0
                                 0
                                     0
                                                            0
                                                                    0
                                                                                  0
         5 rows × 2365 columns
         tokenize_df['label'] = df_eng['Refined Keyword'].tolist()
In [93]:
         tokenize_df['label'].nunique()
In [94]:
Out[94]: 41
```

```
In [121]: tokenize_df['label'].value_counts().plot(kind = 'bar', figsize = (15,5))
```

Out[121]: <matplotlib.axes._subplots.AxesSubplot at 0x1bdff372630>



```
In [97]: tokenize_df['label'].isna().sum()
```

Out[97]: 0

In [98]: tokenize_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3310 entries, 0 to 3309
Columns: 2366 entries, 003 to label
dtypes: int64(2365), object(1)

memory usage: 59.7+ MB

In [99]: | tokenize_df[:5]

Out[99]:

	003	005	0221210211	08	09	0b19a637cc	0h9ms5fxgtkj	10	1042	10698	 your	youtube
0	0	0	0	0	0	0	0	0	0	0	 0	0
1	0	0	0	0	0	0	0	0	0	0	 0	0
2	0	0	0	0	0	0	0	0	0	0	 0	0
3	0	0	0	0	0	0	0	0	0	0	 0	0
4	0	0	0	0	0	0	0	0	0	0	 0	0

5 rows × 2366 columns

```
In [112]:
           tokenize_df['003'].value_counts()
Out[112]: 0
                 3309
           1
           Name: 003, dtype: int64
In [113]:
           # no of rows where each feature exists
           feat_total = []
           for feat in features:
                feat_sum = tokenize_df[feat].sum()
                feat_total.append([feat, feat_sum])
In [116]:
           feat_total_df = pd.DataFrame(feat_total, columns=['Feature', 'Occurance'])
In [122]:
           feat_total_df.sort_values('Occurance', ascending=False)[:10]
Out[122]:
                    Feature Occurance
                                 1392
             939
                 globallogic
             930
                                  838
                     global
             468
                      com
                                  833
            2344
                                  790
                      www
            1067
                      https
                                  778
            1440
                                  747
                      logic
            1107
                                  506
                        in
            1584
                    nagpur
                                  429
            1115
                                  369
                      india
             893
                   gl_career
                                  280
           feat_total_df.sort_values('Occurance', ascending=False)[:50].plot.bar(y='Occurance')
In [132]:
Out[132]: <matplotlib.axes._subplots.AxesSubplot at 0x1bd8569a908>
              1400
                                                                                           Occurance
              1200
              1000
               800
               600
               400
               200
```

```
Out[134]: <matplotlib.axes._subplots.AxesSubplot at 0x1bd85d0ce10>
                           0.6
                           0.4
                                                                                 globslogic -
globollogic -
globollogic -
globollogic -
grobis -
grobis -
gratis -
gratis -
gratis -
gratis -
helton -
heirarchy -
healthy -
healthy -
healthy -
healthy -
healthy -
harkor -
harkor -
harkor -
harkor -
harkor -
gurgoan -
                                      golang-
golablogic -
goe-
gobellogic -
goblelogic -
gobierno -
gobballe-
goballe-
goballe-
goballe-
goballogical -
glocal -
glocal -
                                                                                                          Feature
In [137]: | # Total features
                      feat_total_df.shape[0]
Out[137]: 2365
In [136]: # no of words occurring only 1 time
                      feat total df[feat total df['Occurance']==1].shape[0]
Out[136]: 1472
In [138]: | # no of words occurring less than or equal to 5 times
                      feat total df[feat total df['Occurance']<=5].shape[0]</pre>
Out[138]: 2062
In [141]: # no of words occuring less than 10 times
                      feat_total_df[feat_total_df['Occurance']<10].shape[0]</pre>
Out[141]: 2155
                     # no of words occuring more than or equal to 10 times
In [142]:
                      feat_total_df[feat_total_df['Occurance']>=10].shape[0]
Out[142]: 210
```

In [134]: feat_total_df.sort_values('Occurance', ascending=False)[-50:].plot.bar(y='Occurance')

Modeling Feature Vectors using various classification algorithms

```
In [185]: # creating train and test dataset randomly
          import numpy as np
          msk = np.random.rand(len(tokenize_df)) <= 0.8</pre>
          train df = tokenize df[msk]
          test df = tokenize df[~msk]
          len(train_df), len(test_df)
Out[185]: (2643, 667)
In [154]: x col = features
                             # x_col : Features
          y_col = ['label'] # y_col : Target Variable
          train_x = train_df[x_col]
          train_y = train_df[y_col]
          test x = test df[x col]
          test_y = test_df[y_col]
In [158]: | train y['label'].nunique(),test y['label'].nunique()
Out[158]: (40, 35)
In [161]:
          # Creating a accuracy function:
           def train_model(classifier, feature_vector_train, label, feature_vector_valid, is
               # fit the training dataset on the classifier
               from sklearn.metrics import accuracy score
               classifier.fit(feature vector train, label)
               # predict the labels on validation dataset
               predictions = classifier.predict(feature_vector_valid)
               if is neural net:
                   predictions = predictions.argmax(axis=-1)
               return accuracy score(predictions, test y)
```

Naive Bayes

```
In [168]: # using sklearn's multinomial naive bayes classifier
    from sklearn.naive_bayes import MultinomialNB
    accuracy_nb = train_model(MultinomialNB(), train_x, train_y, test_x)
    print('NB accuracy : ' , accuracy_nb)
NB accuracy : 0.6341829085457271
```

Linear Classifier

```
In [179]: # Linear Classifier on Count Vectors
    from sklearn import *
    accuracy_lc = train_model(linear_model.LogisticRegression(), train_x, train_y, teleprint('LC accuracy : ' , accuracy_lc)
```

LC accuracy: 0.8185907046476761

SVM

```
In [171]: # SVM
    accuracy_svm = train_model(svm.SVC(), train_x, train_y, test_x)
    print('SVM accuracy : ' , accuracy_svm)
    SVM accuracy : 0.14392803598200898
```

Random Forest

```
In [172]: # Random Forest
accuracy_rf = train_model(ensemble.RandomForestClassifier(), train_x, train_y, temprint('RF accuracy : ' , accuracy_rf)
```

SVM accuracy : 0.7886056971514243

XGBoost

```
In [193]: # Extereme Gradient Boosting on Count Vectors
    import xgboost
    accuracy_xgb = train_model(xgboost.XGBClassifier(), train_x, train_y, test_x)
    print('XGB accuracy : ' , accuracy_xgb)
```

XGB accuracy : 0.7901049475262368

LightGBM

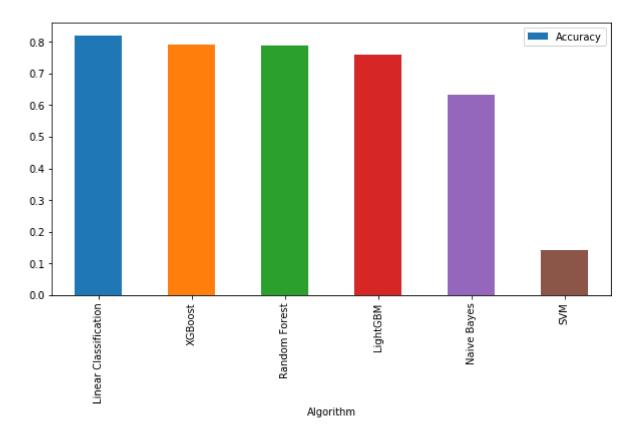
```
In [182]: # using lightgbm

# encoding categorial column label for lightgbm
lgbm_df = tokenize_df.copy()
from sklearn import preprocessing
cat_cols = ['label']
for col in cat_cols:
    lbl = preprocessing.LabelEncoder()
    lbl.fit(list(lgbm_df[col].values.astype('str')))
    lgbm_df[col] = lbl.transform(list(lgbm_df[col].values.astype('str')))
```

```
In [187]: # creating test and train data for Lgbm
          lgbm_train_df = lgbm_df[msk]
          lgbm_test_df = lgbm_df[~msk]
In [188]:
          lgbm train x = lgbm train df[x col]
          lgbm_train_y = lgbm_train_df[y_col]
          lgbm\_test\_x = lgbm\_test\_df[x\_col]
          lgbm_test_y = lgbm_test_df[y_col]
In [190]: # Creating a accuracy function for Lgbm:
          def train_model_lgbm(classifier, feature_vector_train, label, feature_vector_valid
              # fit the training dataset on the classifier
              from sklearn.metrics import accuracy score
              classifier.fit(feature_vector_train, label)
              # predict the labels on validation dataset
              predictions = classifier.predict(feature_vector_valid)
              if is neural net:
                   predictions = predictions.argmax(axis=-1)
              return accuracy_score(predictions, lgbm_test_y)
In [192]:
          import lightgbm as lgb
          accuracy_lgbm = train_model_lgbm(lgb.LGBMClassifier(), lgbm_train_x, lgbm_train_y
          print('LGBM accuracy : ' , accuracy_lgbm)
            LGBM accuracy: 0.760119940029985
In [195]: | accuracy_matrices = [['Naive Bayes', accuracy_nb],
                                ['Linear Classification', accuracy_lc],
                                ['SVM', accuracy svm],
                                ['Random Forest', accuracy_rf],
                                ['XGBoost', accuracy xgb],
                                ['LightGBM', accuracy lgbm],
In [196]: | accuracy df = pd.DataFrame(accuracy matrices, columns=['Algorithm', 'Accuracy'])
```

```
In [204]: accuracy_df.sort_values('Accuracy', ascending=False).plot.bar(y='Accuracy', x='Alg
```

Out[204]: <matplotlib.axes._subplots.AxesSubplot at 0x1bd94280358>



Demo