DEEP LEARNING CASE STUDY - Quora Question Pairs similarity using S-BERT

> Parameshwari S - CB.SC.I5DAS18026

1. Importing all necessary libraries

```
In [1]:
import numpy as np
import pandas as pd
import pandas_profiling
import string
import random
import math
import time
from sklearn.utils import resample
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches
plt.style.use('fivethirtyeight')
import seaborn as sns
sns.set_style('darkgrid')
import os
from os import listdir
import itertools
import collections
import scipy.stats
import nltk
import torch
import zipfile
from nltk import word tokenize
from nltk.stem import WordNetLemmatizer
from gensim.models.doc2vec import Doc2Vec
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.metrics.pairwise import cosine similarity
nltk.download('punkt')
import gensim
from gensim.models import Word2Vec
from gensim.scripts.glove2word2vec import glove2word2vec
from collections import Counter, defaultdict
from tqdm import tqdm
from sklearn import utils
from sklearn import metrics
!pip install -U sentence-transformers
from sentence_transformers import SentenceTransformer
!pip install transformers
from transformers import AutoTokenizer, AutoModel
import warnings
warnings.filterwarnings("ignore")
[nltk data] Downloading package punkt to /root/nltk data...
[nltk data] Unzipping tokenizers/punkt.zip.
Collecting sentence-transformers
  Downloading sentence-transformers-2.1.0.tar.gz (78 kB)
                                      | 78 kB 3.3 MB/s
Collecting transformers<5.0.0,>=4.6.0
  Downloading transformers-4.12.5-py3-none-any.whl (3.1 MB)
                                      | 3.1 MB 18.9 MB/s
Collecting tokenizers>=0.10.3
  Downloading tokenizers-0.10.3-cp37-cp37m-manylinux 2 5 x86 64.manylinux1 x86 64.manylinux 2 12 x86 64.ma
nylinux2010 x86 64.whl (3.3 MB)
                                      | 3.3 MB 25.3 MB/s
Requirement already satisfied: tqdm in /usr/local/lib/python3.7/dist-packages (from sentence-transformers)
(4.62.3)
Requirement already satisfied: torch>=1.6.0 in /usr/local/lib/python3.7/dist-packages (from sentence-trans
formers) (1.10.0+cull1)
Requirement already satisfied: torchvision in /usr/local/lib/python3.7/dist-packages (from sentence-transf
ormers) (0.11.1+cull1)
Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from sentence-transformers
) (1.19.5)
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.7/dist-packages (from sentence-trans
formers) (1.0.1)
Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from sentence-transformers
```

```
(1.4.1)
Requirement already satisfied: nltk in /usr/local/lib/python3.7/dist-packages (from sentence-transformers)
(3.2.5)
Collecting sentencepiece
    \texttt{Downloading sentencepiece-0.1.96-cp37-cp37m-manylinux\_2\_17\_x86\_64.manylinux2014\_x86\_64.whl (1.2 MB) } \\
                          | 1.2 MB 43.9 MB/s
Collecting huggingface-hub
   Downloading huggingface_hub-0.1.2-py3-none-any.whl (59 kB)
                                                              | 59 kB 6.0 MB/s
Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages (from torch>=1.
6.0->sentence-transformers) (3.10.0.2)
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from transformers<5.0.0
,>=4.6.0->sentence-transformers) (2.23.0)
Collecting pyyaml>=5.1
   Downloading PyYAML-6.0-cp37-cp37m-manylinux_2_5_x86_64.manylinux1_x86_64.manylinux_2_12_x86_64.manylinux
2010_x86_64.whl (596 kB)
                                            | 596 kB 43.5 MB/s
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.7/dist-packages (from transformer
s<5.0.0,>=4.6.0->sentence-transformers) (21.2)
Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from transformers<5.0.0
,>=4.6.0->sentence-transformers) (3.3.2)
Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.7/dist-packages (from transform
ers<5.0.0,>=4.6.0->sentence-transformers) (2019.12.20)
Collecting sacremoses
   Downloading sacremoses-0.0.46-py3-none-any.whl (895 kB)
                   | 895 kB 52.2 MB/s
Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packages (from transfor
mers<5.0.0,>=4.6.0->sentence-transformers) (4.8.2)
\label{eq:continuous_power_relation} Requirement already satisfied: pyparsing < 3, >= 2.0.2 in /usr/local/lib/python 3.7/dist-packages (from packaging the packaging of the packaging that the packaging 
\label{eq:ng} \verb"ng">=20.0-\\ \verb">transformers<5.0.0,>=4.6.0-\\ \verb">sentence-transformers) (2.4.7)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-metadat
a->transformers<5.0.0,>=4.6.0->sentence-transformers) (3.6.0)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from nltk->sentence-transfor
mers) (1.15.0)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-pa
ckages (from requests->transformers<5.0.0,>=4.6.0->sentence-transformers) (1.24.3)
Requirement already satisfied: chardet <4, >=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests-packages)
>transformers<5.0.0,>=4.6.0->sentence-transformers) (3.0.4)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->tran
sformers<5.0.0,>=4.6.0->sentence-transformers) (2.10)
\label{lem:requirement} \textbf{Requirement already satisfied: certifi} \verb| = 2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests) | a continuous continuou
-transformers<5.0.0,>=4.6.0->sentence-transformers) (2021.10.8)
Requirement already satisfied: joblib in /usr/local/lib/python3.7/dist-packages (from sacremoses->transfor
mers<5.0.0,>=4.6.0->sentence-transformers) (1.1.0)
Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from sacremoses->transform
ers<5.0.0,>=4.6.0->sentence-transformers) (7.1.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from scikit
-learn->sentence-transformers) (3.0.0)
Requirement already satisfied: pillow!=8.3.0,>=5.3.0 in /usr/local/lib/python3.7/dist-packages (from torch
vision->sentence-transformers) (7.1.2)
Building wheels for collected packages: sentence-transformers
   Building wheel for sentence-transformers (setup.py) ... done
   Created wheel for sentence-transformers: filename=sentence transformers-2.1.0-py3-none-any.whl size=1210
00 sha256=6430659addd677fd798926a605d7cc0d626988ddb7e1d48fb5344939f0184528
   \texttt{Stored in directory: /root/.cache/pip/wheels/90/f0/bb/ed1add84da70092ea526466eadc2bfb197c4bcb8d4fa5f7bad}
Successfully built sentence-transformers
Installing collected packages: pyyaml, tokenizers, sacremoses, huggingface-hub, transformers, sentencepiec
e, sentence-transformers
   Attempting uninstall: pyyaml
       Found existing installation: PyYAML 3.13
      Uninstalling PyYAML-3.13:
          Successfully uninstalled PyYAML-3.13
Successfully installed huggingface-hub-0.1.2 pyyaml-6.0 sacremoses-0.0.46 sentence-transformers-2.1.0 sent
encepiece-0.1.96 tokenizers-0.10.3 transformers-4.12.5
Requirement already satisfied: transformers in /usr/local/lib/python3.7/dist-packages (4.12.5)
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from transformers) (2.2
3.0)
Requirement already satisfied: sacremoses in /usr/local/lib/python3.7/dist-packages (from transformers) (0
.0.46)
Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from transformers) (3.3
.2)
Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.7/dist-packages (from transform
ers) (2019.12.20)
Requirement already satisfied: huggingface-hub<1.0,>=0.1.0 in /usr/local/lib/python3.7/dist-packages (from
transformers) (0.1.2)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.7/dist-packages (from transformers) (4
.62.3)
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.7/dist-packages (from transformers) (
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.7/dist-packages (from transformer
s) (21.2)
Requirement already satisfied: tokenizers<0.11,>=0.10.1 in /usr/local/lib/python3.7/dist-packages (from tr
ansformers) (0.10.3)
```

 $\texttt{kequirement aiready satisfied: importing-metadata in /usr/local/lip/python3.}//\texttt{dist-packages (from transformers)} \end{mers})$

Requirement already satisfied: pyyaml >= 5.1 in /usr/local/lib/python3.7/dist-packages (from transformers) (6.0)

Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.7/dist-packages (from huggingface-hub<1.0,>=0.1.0->transformers) (3.10.0.2)

Requirement already satisfied: pyparsing<3,>=2.0.2 in /usr/local/lib/python3.7/dist-packages (from packagi ng>=20.0->transformers) (2.4.7)

Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-metadat a->transformers) (3.6.0)

Requirement already satisfied: urllib3!=1.25.0, !=1.25.1, <1.26, >=1.21.1 in /usr/local/lib/python3.7/dist-pa ckages (from requests->transformers) (1.24.3)

Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->transformers) (3.0.4)

Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests ->transformers) (2021.10.8)

Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->tran sformers) (2.10)

Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformer s) (1.15.0)

Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from sacremoses->transform ers) (7.1.2)

Requirement already satisfied: joblib in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformers) (1.1.0)

2. Uploading the data

In [5]:

```
train_df = pd.read_csv('/content/train.csv.zip')
train_df.head(3)
```

Out[5]:

	id	qid1	qid2	question1	question2	is_duplicate
0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0
1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0

In [6]:

```
train_df.shape
```

Out[6]:

(404290, 6)

3. Data preprocessing

In [7]:

```
train_df.isnull().sum()
```

Out[7]:

Tn [8]:

train_df[train_df.isnull().any(1)]

Out[8]:

	id	qid1	qid2	question1	question2	is_duplicate
105780	105780	174363	174364	How can I develop android app?	NaN	0
201841	201841	303951	174364	How can I create an Android app?	NaN	0
363362	363362	493340	493341	NaN	My Chinese name is Haichao Yu. What English na	0

4. Data exploration

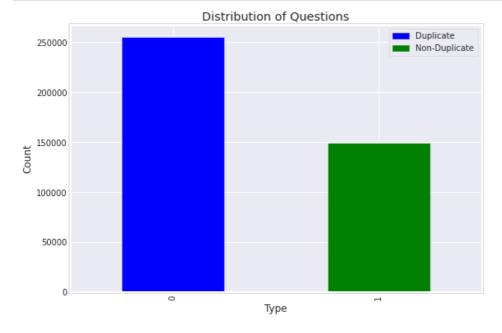
In []:

```
plt.figure(figsize=(8,6))
train_df.is_duplicate.value_counts().plot(kind='bar', color=['b','g'])

D = mpatches.Patch(color='b', label='Duplicate')
ND = mpatches.Patch(color='g', label='Non-Duplicate')

plt.legend(handles=[D,ND], loc='best')

plt.xlabel('Type')
plt.ylabel('Count')
plt.title('Distribution of Questions')
plt.show()
```

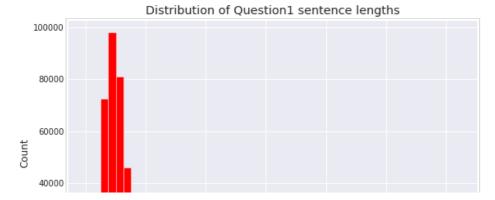


In []:

```
q1_lengths = [len(q1) for q1 in train_df.question1]
print("Mean sentence length for Question1:", np.mean(q1_lengths))

plt.figure(figsize=(8,6))
plt.hist(q1_lengths,bins=50,color='r')
plt.xlabel('Lengths')
plt.ylabel('Count')
plt.title('Distribution of Question1 sentence lengths')
plt.show()
```

Mean sentence length for Question1: 59.53670879813995

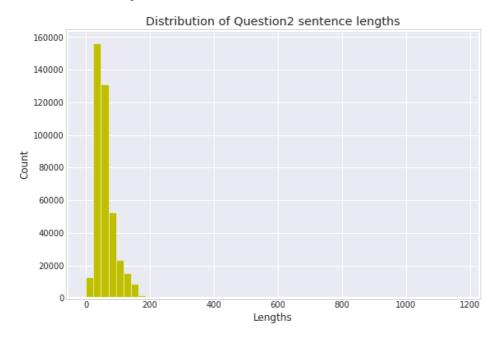


```
20000
0 100 200 300 400 500 600
Lengths
```

```
q2_lengths = [len(q2) for q2 in train_df.question2]
print("Mean sentence length for Question2:", np.mean(q2_lengths))

plt.figure(figsize=(8,6))
plt.hist(q2_lengths,bins=50,color='y')
plt.xlabel('Lengths')
plt.ylabel('Count')
plt.title('Distribution of Question2 sentence lengths')
plt.show()
```

Mean sentence length for Question2: 60.10836528234683



5. S-BERT Embeddings

a) 100000 rows

```
In [ ]:
```

```
st_model = SentenceTransformer('bert-base-nli-mean-tokens')
```

```
In [ ]:
```

```
sbert_df = train_df[:100000]
```

```
In [ ]:
```

```
sentences_question1 = list(sent for sent in sbert_df['question1'].values)
sentences_question2 = list(sent for sent in sbert_df['question2'].values)
```

In []:

```
def generate_sent_embeddings(data):
    return st_model.encode(data)
```

```
In [ ]:
question1_sent_embeddings = generate_sent_embeddings(sentences_question1)
print("shape of question1 sentence embeddings:", question1 sent embeddings.shape)
shape of question1 sentence embeddings: (100000, 768)
In [ ]:
question2_sent_embeddings = generate_sent_embeddings(sentences_question2)
print("shape of question2 sentence embeddings:", question2 sent embeddings.shape)
shape of question2 sentence embeddings: (100000, 768)
In [ ]:
sbert df['question1 sent embeddings'] = pd.DataFrame({'question1 sent embeddings' : list(question1 sent em
beddings) })
sbert df['question2_sent_embeddings'] = pd.DataFrame({'question2_sent_embeddings' : list(question2_sent_em
beddings) })
In [ ]:
\cos \sin = []
spear corr = []
for index, row in sbert_df.iterrows():
  cos_sim.append(cosine_similarity([row['question1_sent_embeddings']],[row['question2_sent_embeddings']]))
  spear corr.append(scipy.stats.spearmanr(row['question1 sent embeddings'],row['question2 sent embeddings
'])[0])
sbert_df['cos_sim'] = cos_sim
sbert df['spear corr'] = spear corr
In [16]:
def similarity to predictions(cos sim, threshold):
    if (cos_sim >= threshold):
         return 1
    else:
         return 0
In [ ]:
sbert_df['pred_res(cos_sim)'] = sbert_df['cos_sim'].apply(similarity_to_predictions, threshold=0.87)
sbert df['pred res(spear corr)'] = sbert df['spear corr'].apply(similarity to predictions, threshold=0.86
In [ ]:
sbert df.head(3)
Out[]:
   id qid1 qid2 question1
                          question2 is_duplicate question1_sent_embeddings question2_sent_embeddings
                                                                                                    cos_sim spear_corr pro
                 What is
                         What is the
                 the step
                            step by
                                               [-0.009722352, -0.32162306,
                                                                          [0.15146354, -0.20154329,
                 by step
                                                                                               [[0.84010166]]
             2
                                            0
                                                                                                              0.810172
0
   0
        1
                          step guide
                                                                            0.9581177, 0.0159406...
                 guide to
                                                     0.9211391, 0.12629...
                         to invest in
                invest in
                               sh...
                    sh...
                 What is
                the story
                         What would
                     of
                           happen if
                                                 [0.27386734, 0.47279105, -
                                                                         [0.19313551, 0.09134984, -
        3
               Kohinoor
                          the Indian
                                            0
                                                                                                 [[0.7469238]]
                                                                                                              0.731909
                                                   0.6623544, 0.1045286...
                                                                            1.0451194, 0.5032031...
                  (Koh-i-
                        government
                   Noor)
                              sto...
                   Dia...
                How can
                           How can
                            Internet
               I increase
                                                \hbox{[-0.20832907, -0.15172529, } \hbox{[0.27955115, 0.0012331137, -}
               the speed
                           speed be
2 2
        5
                                                                                               [[0.89106655]]
                                                                                                              0.881655
                                                    1.1032256, 0.248804...
                                                                              0.03924411, 0.3699...
                  of my
                          increased
                 internet
                                by
                          hacking...
                    CO...
                                                                                                                      ▶
In [ ]:
print("Accuracy for SBERT embeddings using cosine similarity - ", metrics.accuracy score(sbert df['is dupl
```

print("Accuracy for SBERT embeddings using spearman's correlation- ", metrics.accuracy score(sbert df['is

icate'], sbert_df['pred_res(cos_sim)']))

```
duplicate'], sbert df['pred res(spear corr)']))
Accuracy for SBERT embeddings using cosine similarity - 0.72892
Accuracy for SBERT embeddings using spearman's correlation- 0.7286
      b) 500 rows
In [ ]:
sbert df1 = train df[:500]
In [ ]:
sentences_question1 = list(sent for sent in sbert_df1['question1'].values)
sentences question2 = list(sent for sent in sbert df1['question2'].values)
In [ ]:
question1 sent embeddings = generate sent embeddings(sentences question1)
print("shape of question1 sentence embeddings:", question1 sent embeddings.shape)
shape of question1 sentence embeddings: (500, 768)
In [ ]:
question2_sent_embeddings = generate_sent_embeddings(sentences_question2)
print("shape of question2 sentence embeddings:", question2_sent_embeddings.shape)
shape of question2 sentence embeddings: (500, 768)
In [ ]:
sbert df1['question1 sent embeddings'] = pd.DataFrame({'question1 sent embeddings' : list(question1 sent e
sbert df1['question2 sent embeddings'] = pd.DataFrame({'question2 sent embeddings' : list(question2 sent e
mbeddings) })
In [ ]:
\cos \sin = []
spear_corr = []
for index, row in sbert df1.iterrows():
  spear_corr.append(scipy.stats.spearmanr(row['question1_sent_embeddings'],row['question2_sent_embeddings
sbert df1['cos sim'] = cos sim
sbert df1['spear corr'] = spear corr
In [ ]:
sbert df1['pred res(cos sim)'] = sbert df1['cos sim'].apply(similarity to predictions, threshold=0.86)
sbert df1['pred res(spear corr)'] = sbert df1['spear corr'].apply(similarity to predictions, threshold=0.
86)
In [ ]:
sbert dfl.head(3)
Out[]:
  id gid1 gid2 guestion1
                        question2 is_duplicate question1_sent_embeddings question2_sent_embeddings
                                                                                            cos_sim spear_corr pre
                What is
                       What is the
               the step
                          step by
                                                                    [0.15146354, -0.20154329, [[0.84010166]]
                by step
                                           [-0.009722352, -0.32162306,
   0
                                                                                                     0.810172
       1
                        step quide
               guide to
                                                 0.9211391, 0.12629...
                                                                      0.9581177, 0.0159406...
                       to invest in
               invest in
                            sh...
                  sh...
                What is
               the story
                       What would
                    of
                        happen if
                                             [0.27386734, 0.47279105, -
                                                                    [0.19313551, 0.09134984, -
       3
              Kohinoor
                        the Indian
                                         n
                                                                                         [[0.7469238]]
                                                                                                     0.731909
                                               0.6623544. 0.1045286...
                                                                      1.0451194. 0.5032031...
                (Koh-i- government
                 Noor)
                            sto...
                 Dia...
               How can
                         How can
              Lincrease
                          Internet
```

-0.15172529, [0.27955115, 0.0012331137, -embeddings question2 sent embeddings 6, 0.248804... 0.03924411, 0.3699... 2 id qid5 qid2 question1 is_duplicate question | Sent [[0.89**t0665ti]] spæ86165ti** pro internet bv

•

hacking... co...

In []:

print("Accuracy for SBERT embeddings using cosine similarity - ", metrics.accuracy score(sbert df1['is dup licate'], sbert df1['pred res(cos sim)'])) print("Accuracy for SBERT embeddings using spearman's correlation- ", metrics.accuracy score(sbert df1['is _duplicate'], sbert_df1['pred_res(spear corr)']))

Accuracy for SBERT embeddings using cosine similarity - 0.728 Accuracy for SBERT embeddings using spearman's correlation- 0.726

6. BERT Embeddings

In [2]:

!pip install BERTSimilarity

Collecting BERTSimilarity

Downloading BERTSimilarity-0.1.tar.gz (2.7 kB)

Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from BERTSimilarity) (1.19 .5)

Requirement already satisfied: torch in /usr/local/lib/python3.7/dist-packages (from BERTSimilarity) (1.10 .0+cu111)

Requirement already satisfied: transformers in /usr/local/lib/python3.7/dist-packages (from BERTSimilarity (4.12.5)

Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from BERTSimilarity) (1.4.

Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages (from torch->BE RTSimilarity) (3.10.0.2)

Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packages (from transfor mers->BERTSimilarity) (4.8.2)

Requirement already satisfied: huggingface-hub<1.0,>=0.1.0 in /usr/local/lib/python3.7/dist-packages (from transformers->BERTSimilarity) (0.1.2)

Requirement already satisfied: tokenizers<0.11,>=0.10.1 in /usr/local/lib/python3.7/dist-packages (from tr ansformers->BERTSimilarity) (0.10.3)

Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.7/dist-packages (from transformers->B ERTSimilarity) (6.0) Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.7/dist-packages (from transformers->BE

RTSimilarity) (4.62.3)

Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from transformers->BERT Similarity) (2.23.0)

Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.7/dist-packages (from transformer s->BERTSimilarity) (21.2)

 ${\tt Requirement\ already\ satisfied:\ regex!=2019.12.17\ in\ /usr/local/lib/python3.7/dist-packages\ (from\ transform\ packages\ packages\ (from\ transform\ packages\ pa$ ers->BERTSimilarity) (2019.12.20)

Requirement already satisfied: filelock in /usr/local/lib/python3.7/dist-packages (from transformers->BERT Similarity) (3.3.2)

Requirement already satisfied: sacremoses in /usr/local/lib/python3.7/dist-packages (from transformers->BE RTSimilarity) (0.0.46)

Requirement already satisfied: pyparsing<3,>=2.0.2 in /usr/local/lib/python3.7/dist-packages (from packagi ng>=20.0->transformers->BERTSimilarity) (2.4.7)

Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-metadat a->transformers->BERTSimilarity) (3.6.0)

Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-pa ckages (from requests->transformers->BERTSimilarity) (1.24.3)

Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->tran sformers->BERTSimilarity) (2.10)

Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests ->transformers->BERTSimilarity) (2021.10.8)

 $\label{eq:condition} \textbf{Requirement already satisfied: chardet} < 4, >= 3.0.2 in /usr/local/lib/python 3.7/dist-packages (from requests-packages) (from requests-packages)$ >transformers->BERTSimilarity) (3.0.4)

Requirement already satisfied: click in /usr/local/lib/python3.7/dist-packages (from sacremoses->transform ers->BERTSimilarity) (7.1.2)

Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from sacremoses->transformer s->BERTSimilarity) (1.15.0)

Requirement already satisfied: joblib in /usr/local/lib/python3.7/dist-packages (from sacremoses->transfor mers->BERTSimilarity) (1.1.0)

Building wheels for collected packages: BERTSimilarity

Building wheel for BERTSimilarity (setup.py) ... done

Created wheel for BERTSimilarity: filename=BERTSimilarity-0.1-py3-none-any.whl size=3612 sha256=0dfd843f 7ee268ab52d5b733fc6cd868ca7593062ca9a551aae0b44ba5aadc4a

Stored in directory: /root/.cache/pip/wheels/fa/f7/22/510c1c7131e536fb02b71c619dddcce9636913654ba2f22f22 Successfully built BERTSimilarity

Installing collected packages: BERTSimilarity

Successfully installed BERTSimilarity-0.1

```
In [3]:
import torch
from transformers import BertTokenizer, BertModel
from scipy.spatial.distance import cosine
class BERTSimilarity():
    def bert_tokenize(self,data):
        self.data=data
        self.output tokens=''
        self.output tokens+='[CLS] ' +self.data+' [SEP]'
        return self.output tokens
    def sentential embeddings(self, tokenizer, tokenized text):
        self.tokenizer=tokenizer
        self.tokenized text=tokenized text
        self.idx_tokens=self.tokenizer.convert_tokens_to_ids(self.tokenized_text)
        self.segmenter_idx=[1] *len(self.tokenized_text)
        self.tokens tensor=torch.tensor([self.idx tokens])
        self.segmenter tensor=torch.tensor([self.segmenter idx])
        self.model=BertModel.from pretrained('bert-base-uncased',output hidden states=True)
        self.model.eval()
        with torch.no grad():
            self.outputs=self.model(self.tokens tensor,self.segmenter tensor)
            self.hidden state=self.outputs[2]
        self.embedding token=torch.stack(self.hidden state,dim=0)
        self.embedding_token=torch.squeeze(self.embedding_token,dim=1)
        self.embedding token=self.embedding token.permute(1,0,2)
        self.vs_sum_cat=[]
        for i in self.embedding token:
            vs li=torch.sum(i[-4:],dim=0)
            self.vs sum cat.append(vs li)
        self.token_vecs=self.hidden_state[-2][0]
        self.sentence_embeddings=torch.mean(self.token_vecs,dim=0)
        return self.sentence embeddings, self.vs sum cat
    def calculate_distance(self, sentence_1, sentence_2):
        self.sentence 1=sentence 1
        self.sentence 2=sentence 2
        self.tokenizer=BertTokenizer.from pretrained('bert-base-uncased')
        self.preprocess 1=self.bert tokenize(self.sentence 1)
        self.preprocess_2=self.bert_tokenize(self.sentence_2)
        self.tokenized_text_1=self.tokenizer.tokenize(self.preprocess_1)
        self.tokenized text 2=self.tokenizer.tokenize(self.preprocess_2)
        self.sentence 1,self.vs sum cat1=self.sentential embeddings(self.tokenizer,self.tokenized text 1)
        self.sentence_2,self.vs_sum_cat2=self.sentential_embeddings(self.tokenizer,self.tokenized_text_2)
        self.distance=1-cosine(self.sentence 1, self.sentence 2)
        return self.distance
    def corr(self, sentence 1, sentence 2):
        self.sentence 1=sentence 1
        self.sentence_2=sentence_2
        self.tokenizer=BertTokenizer.from pretrained('bert-base-uncased')
        self.preprocess 1=self.bert tokenize(self.sentence 1)
        self.preprocess 2=self.bert tokenize(self.sentence 2)
        self.tokenized_text_1=self.tokenizer.tokenize(self.preprocess_1)
        self.tokenized_text_2=self.tokenizer.tokenize(self.preprocess_2)
        \verb|self.sentence_1|, \verb|self.vs_sum_cat1| = \verb|self.sentential_embeddings(self.tokenizer, \verb|self.tokenized_text_1|)| \\
        self.sentence 2,self.vs sum cat2=self.sentential embeddings(self.tokenizer,self.tokenized text 2)
        self.spcorr=scipy.stats.spearmanr(self.sentence 1, self.sentence 2)[0]
        return self.spcorr
In [4]:
bertsimilarity=BERTSimilarity()
In [12]:
from transformers import logging
logging.set verbosity error()
In [13]:
bert df = train df[:500]
In [14]:
distances=[]
spear_corr = []
for i in range(len(bert df)):
  q1=bert df['question1'][i]
  q2=bert_df['question2'][i]
  distances.append(bertsimilarity.calculate distance(q1,q2))
  spear corr.append(bertsimilarity.corr(q1,q2))
bert df['cos sim'] = distances
```

```
bert_df['spear_corr']=spear_corr
```

```
In [29]:
```

```
bert_df['pred_res(cos_sim)'] = bert_df['cos_sim'].apply(similarity_to_predictions, threshold=0.89)
bert_df['pred_res(spear_corr)'] = bert_df['spear_corr'].apply(similarity_to_predictions, threshold=0.87)
```

In [30]:

bert_df.head()

Out[30]:

	id	qid1	qid2	question1	question2	is_duplicate	cos_sim	spear_corr	pred_res(cos_sim)	pred_res(spear_corr)
0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0	0.970151	0.937000	1	1
1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0	0.905713	0.770143	1	0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0	0.923254	0.811675	1	0
3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24}[/math] i	0	0.610703	0.356798	0	0
4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0	0.761775	0.490774	0	0

In [31]:

```
print("Accuary for BERT embeddings using cosine similarity- ", metrics.accuracy_score(bert_df['is_duplicat
e'], bert_df['pred_res(cos_sim)']))
print("Accuary for BERT embeddings using spearman's correlation- ", metrics.accuracy_score(bert_df['is_duplicate'], bert_df['pred_res(spear_corr)']))
```

Accuary for BERT embeddings using cosine similarity- 0.692 Accuary for BERT embeddings using spearman's correlation- 0.662

7. Universal Sentene Encoder embeddings

In []:

```
!pip3 install --upgrade tensorflow-gpu
!pip3 install tensorflow-hub
```

Collecting tensorflow-gpu

```
Downloading tensorflow gpu-2.7.0-cp37-cp37m-manylinux2010 x86 64.whl (489.6 MB)
```

| 489.6 MB 24 kB/s

Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.7/dist-packages (from tensorflo w-gpu) (3.3.0)

Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow -gpu) (1.1.0)

Requirement already satisfied: tensorflow-estimator<2.8, \sim =2.7.0rc0 in /usr/local/lib/python3.7/dist-packag es (from tensorflow-gpu) (2.7.0)

Requirement already satisfied: wheel<1.0,>=0.32.0 in /usr/local/lib/python3.7/dist-packages (from tensorfl ow-gpu) (0.37.0)

Requirement already satisfied: google-pasta>=0.1.1 in /usr/local/lib/python3.7/dist-packages (from tensorf low-gpu) (0.2.0)

Requirement already satisfied: keras-preprocessing>=1.1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflow-gpu) (1.1.2)

Requirement already satisfied: typing-extensions>=3.6.6 in /usr/local/lib/python3.7/dist-packages (from te nsorflow-gpu) (3.10.0.2)

Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-gpu) (1.15.0)

Requirement already satisfied: flatbuffers<3.0,>=1.12 in /usr/local/lib/python3.7/dist-packages (from tens orflow-gpu) (2.0)

Requirement already satisfied: absl-py>=0.4.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-g pu) (0.12.0)

Requirement already satisfied: gast<0.5.0,>=0.2.1 in /usr/local/lib/python3.7/dist-packages (from tensorfl ow-gpu) (0.4.0)

Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-gp u) (1.13.3)

Requirement already satisfied: astunparse>=1.6.0 in /usr/local/lib/python3.7/dist-packages (from tensorflo w-qpu) (1.6.3)

Requirement already satisfied: h5py>=2.9.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-gpu) (3.1.0)

```
Requirement already satisfied: keras<2.8,>=2.7.0rc0 in /usr/local/lib/python3.7/dist-packages (from tensor
flow-gpu) (2.7.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.7/dist-packages (from tensorf
low-gpu) (1.41.1)
Requirement already satisfied: libclang>=9.0.1 in /usr/local/lib/python3.7/dist-packages (from tensorflow-
apu) (12.0.0)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.21.0 in /usr/local/lib/python3.7/dist-packa
ges (from tensorflow-gpu) (0.22.0)
Requirement already satisfied: protobuf>=3.9.2 in /usr/local/lib/python3.7/dist-packages (from tensorflow-
gpu) (3.17.3)
Requirement already satisfied: numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages (from tensorflow-gp
u) (1.19.5)
Requirement already satisfied: tensorboard~=2.6 in /usr/local/lib/python3.7/dist-packages (from tensorflow
-gpu) (2.7.0)
Requirement already satisfied: cached-property in /usr/local/lib/python3.7/dist-packages (from h5py>=2.9.0
->tensorflow-gpu) (1.5.2)
Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in /usr/local/lib/python3.7/dist-pack
ages (from tensorboard~=2.6->tensorflow-gpu) (0.6.1)
Requirement already satisfied: google-auth<3,>=1.6.3 in /usr/local/lib/python3.7/dist-packages (from tenso
rboard~=2.6->tensorflow-gpu) (1.35.0)
Requirement already satisfied: setuptools>=41.0.0 in /usr/local/lib/python3.7/dist-packages (from tensorbo
ard~=2.6->tensorflow-gpu) (57.4.0)
{\tt Requirement\ already\ satisfied:\ tensorboard-plugin-wit>=1.6.0\ in\ /usr/local/lib/python3.7/dist-packages\ (fractional already\ satisfied:\ satisfied:\ satisfied:\ satisfied:\ satisfied:\ satisfied:\ sat
om tensorboard~=2.6->tensorflow-gpu) (1.8.0)
Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.7/dist-packages (from tensorb
oard~=2.6->tensorflow-gpu) (2.23.0)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /usr/local/lib/python3.7/dist-packages
(from tensorboard~=2.6->tensorflow-gpu) (0.4.6)
Requirement already satisfied: werkzeug>=0.11.15 in /usr/local/lib/python3.7/dist-packages (from tensorboa
rd~=2.6->tensorflow-gpu) (1.0.1)
Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.7/dist-packages (from tensorboard
\sim=2.6->tensorflow-gpu) (3.3.4)
Requirement already satisfied: rsa<5,>=3.1.4 in /usr/local/lib/python3.7/dist-packages (from google-auth<3
,>=1.6.3->tensorboard~=2.6->tensorflow-gpu) (4.7.2)
Requirement already satisfied: pyasn1-modules>=0.2.1 in /usr/local/lib/python3.7/dist-packages (from googl
e-auth<3,>=1.6.3->tensorboard\sim=2.6->tensorflow-gpu) (0.2.8)
Requirement already satisfied: cachetools<5.0,>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from goog
le-auth<3,>=1.6.3->tensorboard~=2.6->tensorflow-gpu) (4.2.4)
Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.7/dist-packages (from go
ogle-auth-oauthlib<0.5,>=0.4.1->tensorboard\sim=2.6->tensorflow-gpu) (1.3.0)
Requirement already satisfied: importlib-metadata in /usr/local/lib/python3.7/dist-packages (from markdown
>=2.6.8->tensorboard~=2.6->tensorflow-gpu) (4.8.2)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /usr/local/lib/python3.7/dist-packages (from pyasn1
-modules >= 0.2.1 - google-auth < 3, >= 1.6.3 - tensorboard ~= 2.6 - tensorflow-gpu) (0.4.8)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests
<3,>=2.21.0->tensorboard~=2.6->tensorflow-gpu) (2021.10.8)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests<3,>=2
.21.0->tensorboard~=2.6->tensorflow-gpu) (2.10)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests<
3, \geq 2.21.0 \rightarrow \text{tensorboard} = 2.6 \rightarrow \text{tensorflow-gpu} (3.0.4)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-pa
ckages (from requests<3,>=2.21.0->tensorboard~=2.6->tensorflow-gpu) (1.24.3)
Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.7/dist-packages (from requests-oa
 uthlib >= 0.7.0 - y cogle-auth-oauthlib < 0.5, >= 0.4.1 - y tensor board <= 2.6 - y tensor flow-gpu) \\ (3.1.1) 
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-packages (from importlib-metadat
a->markdown>=2.6.8->tensorboard~=2.6->tensorflow-gpu) (3.6.0)
Installing collected packages: tensorflow-gpu
Successfully installed tensorflow-gpu-2.7.0
Requirement already satisfied: tensorflow-hub in /usr/local/lib/python3.7/dist-packages (0.12.0)
Requirement already satisfied: protobuf>=3.8.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-
hub) (3.17.3)
Requirement already satisfied: numpy>=1.12.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow-hu
b) (1.19.5)
Requirement already satisfied: six >= 1.9 in /usr/local/lib/python3.7/dist-packages (from protobuf>=3.8.0->t
ensorflow-hub) (1.15.0)
In [ ]:
import tensorflow as tf
```

```
import tensorflow hub as hub
```

```
module url = "https://tfhub.dev/google/universal-sentence-encoder/4"
model = hub.load(module url)
print ("module %s loaded" % module_url)
INFO:absl:Using /tmp/tfhub_modules to cache modules.
INFO:absl:Downloading TF-Hub Module 'https://tfhub.dev/google/universal-sentence-encoder/4'.
INFO:absl:Downloaded https://tfhub.dev/google/universal-sentence-encoder/4, Total size: 987.47MB
INFO:absl:Downloaded TF-Hub Module 'https://tfhub.dev/google/universal-sentence-encoder/4'.
```

```
module https://tfhub.dev/google/universal-sentence-encoder/4 loaded
In [ ]:
use df = train df[:500]
In [ ]:
sentences question1 = list(sent for sent in use df['question1'].values)
sentences question2 = list(sent for sent in use df['question2'].values)
In [ ]:
sentence1 embeddings = model(sentences question1)
print("shape of question1 sentence embeddings:", sentence1_embeddings.shape)
sentence2 embeddings = model(sentences question2)
print("shape of question2 sentence embeddings:", sentence2 embeddings.shape)
shape of question1 sentence embeddings: (500, 512)
shape of question2 sentence embeddings: (500, 512)
Tn [ ]:
use_df['question1_sent_embeddings'] = pd.DataFrame({'question1_sent_embeddings' : list(sentence1 embedding
s)})
use_df['question2_sent_embeddings'] = pd.DataFrame({'question2_sent_embeddings' : list(sentence2_embedding
s)})
In [ ]:
\cos \sin = []
for index, row in use df.iterrows():
  cos sim.append(cosine similarity([row['question1 sent embeddings']],[row['question2 sent embeddings']]))
  spear corr.append(scipy.stats.spearmanr(row['question1 sent embeddings'],row['question2 sent embeddings
'])[0])
use df['cos_sim'] = cos_sim
use df['spear corr'] = spear corr
In [ ]:
use_df['pred_res(cos_sim)'] = use_df['cos_sim'].apply(similarity_to_predictions, threshold=0.86)
use_df['pred_res(spear_corr)'] = use_df['spear_corr'].apply(similarity_to_predictions, threshold=0.88)
In [ ]:
use df.head(3)
Out[]:
   id qid1 qid2 question1
                          question2 is_duplicate question1_sent_embeddings question2_sent_embeddings
                                                                                                  cos sim spear corr pro
                 What is
                         What is the
                the step
                           step by
                 by step
                                                 (tf.Tensor(0.0021821507,
                                                                          (tf.Tensor(0.018747559,
   0
        1
                         step guide
                                           0
                                                                                               [[0.9364382]]
                                                                                                           0.934645
                guide to
                                                  shape=(), dtype=float...
                                                                         shape=(), dtype=float3...
                         to invest in
                invest in
                              sh...
                   sh...
                 What is
                the story What would
                                                                         (tf.Tensor(-0.026330141,
                     of
                          happen if
                                                 (tf.Tensor(-0.0081168795,
        3
               Kohinoor
                          the Indian
                                           0
                                                                                                           0.675223
                                                   shape=(), dtype=floa...
                                                                          shape=(), dtype=float...
                 (Koh-i- government
                  Noor)
                             sto...
                   Dia...
                How can
                          How can
                           Internet
               I increase
              the speed
                          speed be
                                                 (tf.Tensor(-0.025076203,
                                                                         (tf.Tensor(-0.019373633,
   2
        5
                                           O
                                                                                             [[0.60938096]]
                                                                                                           0.589570
                  of my
                          increased
                                                  shape=(), dtype=float...
                                                                          shape=(), dtype=float...
                 internet
                               by
                          hacking...
                   co...
In [ ]:
print ("Accuracy for USE embeddings using cosine similarity - ", metrics.accuracy score (use df['is duplicat
e'], use_df['pred_res(cos_sim)']))
print("Accuracy for USE embeddings using spearman's correlation- ", metrics.accuracy score(use df['is dupl
icate'], use_df['pred_res(spear_corr)']))
```

```
Accuracy for USE embeddings using cosine similarity - 0.688
Accuracy for USE embeddings using spearman's correlation- 0.686
```

8. RoBERTa embeddings

sh...

```
In [ ]:
rb model = SentenceTransformer('roberta-base-nli-stsb-mean-tokens')
In [ ]:
roberta df = train df[:500]
In [ ]:
sentences question1 = list(sent for sent in roberta df['question1'].values)
sentences question2 = list(sent for sent in roberta df['question2'].values)
In [ ]:
def generate sent embeddings(data):
    return rb model.encode(data)
In [ ]:
sentence1_embeddings = generate_sent_embeddings(sentences_question1)
print("shape of question1 sentence embeddings:", sentence1 embeddings.shape)
sentence2_embeddings = generate_sent_embeddings(sentences_question2)
print("shape of question2 sentence embeddings:", sentence2 embeddings.shape)
shape of question1 sentence embeddings: (500, 768)
shape of question2 sentence embeddings: (500, 768)
In [ ]:
roberta df['question1 sent embeddings'] = pd.DataFrame({'question1 sent embeddings' : list(question1 sent
embeddings)})
roberta df['question2 sent embeddings'] = pd.DataFrame({'question2 sent embeddings' : list(question2 sent
embeddings) })
In [ ]:
\cos \sin = []
spear corr = []
for index, row in roberta df.iterrows():
  spear corr.append(scipy.stats.spearmanr(row['question1 sent embeddings'],row['question2 sent embeddings
'])[0])
roberta df['cos sim'] = cos sim
roberta df['spear corr'] = spear corr
In [ ]:
roberta df['pred res(cos sim)'] = roberta df['cos sim'].apply(similarity to predictions, threshold=0.86)
roberta_df['pred_res(spear_corr)'] = roberta_df['spear_corr'].apply(similarity_to_predictions, threshold=
0.86)
In [ ]:
roberta df.head(3)
Out[]:
  id qid1 qid2 question1
                       question2 is_duplicate question1_sent_embeddings question2_sent_embeddings
                                                                                         cos sim spear corr pro
               What is
                      What is the
               the step
                                                                  [0.15146354, -0.20154329, [[0.84010166]]
                         step by
                                          [-0.009722352, -0.32162306,
               by step
                                                                                                  0.810172
   0
                       step guide
                                               0.9211391, 0.12629...
                                                                    0.9581177, 0.0159406...
               guide to
                      to invest in
              invest in
                           sh...
```

```
id qid1 qid2 question1
What is
                                question2 is_duplicate question1_sent_embeddings question2_sent_embeddings
                                                                                                                           cos sim spear corr pro
                   the story
                              What would
                                happen if
                          of
                                                            [0.27386734, 0.47279105, -
                                                                                          [0.19313551, 0.09134984, -
                                                      O
         3
                  Kohinoor
                                the Indian
                                                                                                                       [[0.7469238]]
                                                                                                                                       0.731909
                                                               0.6623544, 0.1045286...
                                                                                             1.0451194, 0.5032031...
                     (Koh-i- government
                       Noor)
                                     sto...
                       Dia...
                   How can
                                 How can
                  I increase
                                  Internet
                                                           \hbox{[-0.20832907, -0.15172529, [0.27955115, 0.0012331137, -}\\
                  the speed
                                speed be
                                                                                                                     [[0.89106655]]
2 2
         5
                                                                                                                                       0.881655
                       of my
                                increased
                                                                1.1032256, 0.248804...
                                                                                               0.03924411, 0.3699...
                    internet
                                       by
                        co...
                                hacking...
```

In []:

print("Accuary for RoBERTa embeddings using cosine similarity- ", metrics.accuracy_score(roberta_df['is_du
plicate'], roberta_df['pred_res(cos_sim)']))
print("Accuary for RoBERTa embeddings using spearman's correlation- ", metrics.accuracy_score(roberta_df[
'is_duplicate'], roberta_df['pred_res(spear_corr)']))

Accuary for Roberta embeddings using cosine similarity- 0.728 Accuary for Roberta embeddings using spearman's correlation- 0.726

9. InferSent embeddings

```
!wget -c http://nlp.stanford.edu/data/glove.840B.300d.zip
wget -c https://dl.fbaipublicfiles.com/infersent/infersent1.pkl
wget -c https://raw.githubusercontent.com/facebookresearch/InferSent/master/models.py
--2021-11-19 05:28:05-- http://nlp.stanford.edu/data/glove.840B.300d.zip
Resolving nlp.stanford.edu (nlp.stanford.edu)... 171.64.67.140
Connecting to nlp.stanford.edu (nlp.stanford.edu) | 171.64.67.140 | : 80... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://nlp.stanford.edu/data/glove.840B.300d.zip [following]
--2021-11-19 05:28:06-- https://nlp.stanford.edu/data/glove.840B.300d.zip
Connecting to nlp.stanford.edu (nlp.stanford.edu)|171.64.67.140|:443... connected.
HTTP request sent, awaiting response... 301 Moved Permanently
Location: http://downloads.cs.stanford.edu/nlp/data/glove.840B.300d.zip [following]
--2021-11-19 05:28:06-- http://downloads.cs.stanford.edu/nlp/data/glove.840B.300d.zip
Resolving downloads.cs.stanford.edu (downloads.cs.stanford.edu)... 171.64.64.22
Connecting to downloads.cs.stanford.edu (downloads.cs.stanford.edu) | 171.64.64.22 | : 80... connected.
HTTP request sent, awaiting response... 200 \ensuremath{\text{OK}}
Length: 2176768927 (2.0G) [application/zip]
Saving to: 'glove.840B.300d.zip'
glove.840B.300d.zip 100%[============] 2.03G 5.03MB/s
                                                                   in 6m 53s
2021-11-19 05:34:59 (5.03 MB/s) - 'glove.840B.300d.zip' saved [2176768927/2176768927]
--2021-11-19 05:34:59-- https://dl.fbaipublicfiles.com/infersent/infersent1.pkl
Resolving dl.fbaipublicfiles.com (dl.fbaipublicfiles.com)... 104.22.74.142, 172.67.9.4, 104.22.75.142, ...
Connecting to dl.fbaipublicfiles.com (dl.fbaipublicfiles.com)|104.22.74.142|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 154010676 (147M) [application/octet-stream]
Saving to: 'infersent1.pkl'
infersent1.pkl
                   2021-11-19 05:35:05 (26.2 MB/s) - 'infersentl.pkl' saved [154010676/154010676]
--2021-11-19 05:35:06-- https://raw.githubusercontent.com/facebookresearch/InferSent/master/models.py
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 185.199.111.133, 185.199.109.133, 185.19
9.108.133, ...
Connecting to raw.githubusercontent.com (raw.githubusercontent.com)|185.199.111.133|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 9875 (9.6K) [text/plain]
Saving to: 'models.py'
```

9.64K --.-KB/s

in Os

models.py

2021-11-19 05:35:06 (81.2 MB/s) - 'models.py' saved [9875/9875]

```
zipread.extractall("/content/")
 zipread.close
In [ ]:
from models import InferSent
MODEL PATH = '/content/infersent1.pkl'
infersent = InferSent(params_model)
infersent.load state dict(torch.load(MODEL PATH))
infersent.set w2v path("/content/glove.840B.300d.txt")
In [ ]:
infersent df = train df[:500]
In [ ]:
import itertools
from itertools import chain
infersent.build vocab(list(chain(infersent df.question1, infersent df.question2)), tokenize=True)
Found 2665(/2739) words with w2v vectors
Vocab size : 2665
In [ ]:
embeddings1 = infersent.encode(infersent df.question1, tokenize=True)
embeddings2 = infersent.encode(infersent_df.question2, tokenize=True)
In [ ]:
```

```
cos_sim = []
spear_corr = []
for (e1, e2) in zip(embeddings1, embeddings2):
   cos_sim.append(cosine_similarity(e1.reshape(1,-1), e2.reshape(1,-1))[0][0])
   spear_corr.append(scipy.stats.spearmanr(e1, e2)[0])
infersent_df["cos_sim"] = cos_sim
infersent_df["spear_corr"] = spear_corr
```

```
infersent_df['pred_res(cos_sim)'] = infersent_df['cos_sim'].apply(similarity_to_predictions, threshold=0.
85)
infersent_df['pred_res(spear_corr)'] = infersent_df['spear_corr'].apply(similarity_to_predictions, threshold=0.85)
```

In []:

```
infersent_df.head(3)
```

Out[]:

	id	qid1	qid2	question1	question2	is_duplicate	cos_sim	spear_corr	pred_res(cos_sim)	pred_res(spear_corr)
0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0	0.953249	0.921047	1	1
1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0	0.832436	0.702478	0	0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0	0.868755	0.768777	1	0

In []:

```
print("Accuary for InferSent embeddings using cosine similarity- ", metrics.accuracy_score(infersent_df['i
s_duplicate'], infersent_df['pred_res(cos_sim)']))
print("Accuary for InferSent embeddings using spearman's correlation- ", metrics.accuracy_score(infersent_
df['is_duplicate'], infersent_df['pred_res(spear_corr)']))
```

Accuary for InferSent embeddings using cosine similarity- 0.682 Accuary for InferSent embeddings using spearman's correlation- 0.654

10. Semantic Search using SBERT

```
In [ ]:
semser df = train df[:100]
semser_df.shape
Out[]:
(100, 6)
In [ ]:
q1 = semser df.question1.tolist()
q2 = semser_df.question2.tolist()
In [ ]:
sentences = q1 + q2
sentence embeddings = st model.encode(sentences)
In [ ]:
print('Sample BERT embedding vector - length', len(sentence embeddings[0]))
Sample BERT embedding vector - length 768
In [ ]:
query = semser_df['question1'][28]
queries = [query]
query_embeddings = st_model.encode(queries)
n = 4
print("Semantic Search Results")
for query, query_embedding in zip(queries, query_embeddings):
    distances = scipy.spatial.distance.cdist([query_embedding], sentence_embeddings, "cosine")[0]
    results = zip(range(len(distances)), distances)
    results = sorted(results, key=lambda x: x[1])
    print("Query:", query)
    print("\nTop 3 most similar sentences - ")
    for idx, distance in results[0:n]:
       print(sentences[idx].strip(), "(Cosine Score: %.4f)" % (1-distance))
Semantic Search Results
Query: What is best way to make money online?
Top 3 most similar sentences -
What is best way to make money online? (Cosine Score: 1.0000)
What is best way to ask for money online? (Cosine Score: 0.9583)
How can I make money through the Internet? (Cosine Score: 0.7752)
What are some different ways to make money online, excluding selling things? (Cosine Score: 0.7561)
In [ ]:
```