Predictor

July 14, 2020

1 Initialization

```
[]: # Simple fork of the original git to fix tqdm progress bar in google collab
!pip install --no-cache-dir git+https://github.com/lowener/sentence-transformers
!pip install ngt
```

```
[]: import numpy as np
  import torch
  import math
  from torch.utils.data import Dataset, DataLoader
  from sklearn import datasets, model_selection
  from matplotlib import pyplot as plt
  import itertools
  from sentence_transformers import SentenceTransformer, losses, evaluation
  import ngt
  from sklearn import metrics
  import seaborn as sn
  import ngtpy
```

```
[3]: SEED = 2020
np.random.seed(SEED)
torch.manual_seed(SEED)
torch.backends.cudnn.enabled = True
torch.backends.cudnn.benchmark = True
```

2 Data Loader for Triplet Learning on 20 Newsgroup dataset

```
focus_negative = {
    0: [15, 19],
    1: [2,3,4,5, 12],
    2: [1,3,4,5, 12],
    3: [1,2,4,5, 12],
    4: [1,2,3,5, 12],
    5: [1,2,3,4, 12],
   7: [8,9,10],
    8: [7,9,10],
    9: [7,8,10],
    10: [7,8,9],
    11: [12,13,14],
    12: [1, 2, 3, 4, 5,11,13,14],
    13: [11,12,14],
   14: [11,12,13],
    15: [0, 19],
    16: [17,18,19],
    17: [16,18,19],
    18: [16,17,19],
    19: [0, 15, 16,17,18],
}
combinations = itertools.combinations(positive_idx, 2)
combinations = list(combinations)
triplets = []
np.random.shuffle(negative_idx)
choose_focus = np.random.random((len(combinations)))
neg = negative_idx[:len(combinations)]
for c, n, cf in zip(combinations, neg, choose_focus):
  # Choose random example 60% of the time
  # Choose a focus example 40% of the time
  if classid in focus_negative and cf < 0.4:</pre>
    neg_array = focus_negative[classid]
    neg_category = np.random.choice(neg_array)
    n = np.random.choice(np.where(Y == neg_category)[0])
    triplets.append(list(c) + [n])
  else:
    triplets.append(list(c) + [n])
return triplets
```

```
if subset == 'test': # Reduce the validation data by 2/3
    self.X = self.X[:len(self.X)//3]
    self.y = self.y[:len(self.y)//3]
  for i in range(len(self.X)):
    self.X[i] = model.tokenize(self.X[i])
  self.triplets = []
  for classid in np.unique(self.y):
    self.triplets += build_triplets(classid, self.X, self.y)
  self.y = torch.tensor(self.y, dtype=torch.long)
def __len__(self,):
  return len(self.triplets)
def __getitem__(self, index):
  triplet_idx = self.triplets[index]
  anchor = self.X[triplet_idx[0]]
  pos = self.X[triplet_idx[1]]
  neg = self.X[triplet_idx[2]]
  label = self.y[triplet_idx[0]]
  return [anchor, pos, neg], label
```

```
[]: from google.colab import drive drive.mount('/content/gdrive')
```

3 Training

```
evaluation_steps = math.ceil(len(train_data) * 0.34 / batch_size)
     print(len(train_data) // batch_size)
     print(evaluation_steps)
     print(len(test_data) // batch_size)
    /usr/local/lib/python3.6/dist-packages/transformers/tokenization_utils.py:831:
    FutureWarning: Parameter max_len is deprecated and will be removed in a future
    release. Use model_max_length instead.
      category=FutureWarning,
    6717
    2285
    1490
[]: torch.cuda.is available()
[]: True
[]: model.fit(train_objectives=[(train_dataloader, train_loss)],
               evaluator=evaluator,
               epochs=num epochs,
               evaluation steps=evaluation steps,
               warmup steps=warmup steps,
               output_path=model_save_path)
```

4 Approximate Nearest Neighbor Classification

4.1 Encode the sentences

```
[9]: model_save_path = '/content/gdrive/My Drive/Ubisoft/save_model_opti_triplets2' model = SentenceTransformer(model_save_path)
```

/usr/local/lib/python3.6/dist-packages/transformers/tokenization_utils.py:831: FutureWarning: Parameter max_len is deprecated and will be removed in a future release. Use model_max_length instead. category=FutureWarning,

```
[13]: np.save('/content/gdrive/My Drive/Ubisoft/X_train_encoded', X_train_encoded)
np.save('/content/gdrive/My Drive/Ubisoft/X_test_encoded', X_test_encoded)
np.save('/content/gdrive/My Drive/Ubisoft/y_train', y_train)
```

4.2 Build kNN index

```
[14]: dim = len(X_train_encoded[0])
   ngtpy.create(b"IndexEncoded", dim, distance_type='L1')
   index = ngtpy.Index(b"IndexEncoded")
   index.batch_insert(X_train_encoded)
   index.save()
```

4.3 Query Index

```
[15]: def classify(query: np.array, index: ngtpy.Index, y_index: np.array, size_k:___
       →int = 3, weight: bool = True) -> np.array:
        result = np.zeros((len(query)))
        for i in range(len(query)):
          vote_array = np.zeros(20)
          query_res = index.search(query[i], size=size_k)
          for neighbor in query_res:
            if neighbor[1] == 0:
              vote_array[y_index[int(neighbor[0])]] = 1
              break
            if weight:
              vote_array[y_index[int(neighbor[0])]] += 1 / neighbor[1]
            else:
              vote_array[y_index[int(neighbor[0])]] += 1
            #print('Neighbor: ' + str(i) + ', Index: ' + str(o[0]) + ', Class: ' +
       \hookrightarrow str(y\_index[o[0]]) + ', Distance: ' + str(o[1]))
          result[i] = np.argmax(vote_array)
        return result
```

```
[16]: res = classify(X_test_encoded, index, y_train, size_k=9, weight=True)
```

4.4 Classification report

```
[17]: print(metrics.classification_report(y_test, res))
```

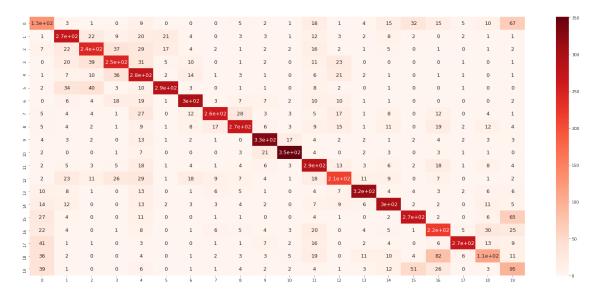
```
precision recall f1-score support

0 0.37 0.41 0.39 319
1 0.63 0.70 0.66 389
```

```
2
                    0.63
                               0.62
                                          0.63
                                                      394
           3
                    0.64
                               0.63
                                          0.64
                                                      392
           4
                    0.50
                               0.72
                                          0.59
                                                      385
           5
                    0.85
                               0.73
                                          0.79
                                                      395
           6
                               0.77
                    0.78
                                          0.77
                                                      390
           7
                    0.83
                               0.66
                                          0.73
                                                      396
           8
                    0.76
                               0.68
                                          0.71
                                                      398
           9
                    0.81
                               0.83
                                          0.82
                                                      397
          10
                    0.89
                               0.88
                                          0.88
                                                      399
          11
                    0.60
                               0.73
                                          0.66
                                                      396
          12
                    0.63
                               0.54
                                          0.58
                                                      393
          13
                    0.85
                               0.80
                                          0.82
                                                      396
          14
                    0.74
                               0.76
                                          0.75
                                                      394
          15
                    0.73
                               0.69
                                          0.71
                                                      398
          16
                    0.52
                               0.60
                                          0.56
                                                      364
          17
                    0.90
                               0.72
                                          0.80
                                                      376
          18
                    0.50
                               0.36
                                          0.42
                                                      310
          19
                    0.31
                               0.38
                                          0.34
                                                      251
                                                     7532
    accuracy
                                          0.67
                                          0.66
                                                     7532
   macro avg
                    0.67
                               0.66
weighted avg
                    0.69
                               0.67
                                          0.67
                                                     7532
```

```
[18]: cfm = metrics.confusion_matrix(y_test, res)
plt.figure(None,(28,12))
sn.heatmap(cfm,annot=True, annot_kws={"size": 13}, cmap=plt.cm.Reds)
```

[18]: <matplotlib.axes._subplots.AxesSubplot at 0x7f4f852ae9e8>



4.5 Compare with base model

```
[]: model_base = SentenceTransformer('distilbert-base-nli-mean-tokens')
#res=model.evaluate(evaluator)

/usr/local/lib/python3.6/dist-packages/transformers/tokenization_utils.py:831:
```

FutureWarning: Parameter max_len is deprecated and will be removed in a future release. Use model_max_length instead.

category=FutureWarning,

```
[]: ngtpy.create(b"IndexEncoded_base", dim, distance_type='L1')
  index_base = ngtpy.Index(b"IndexEncoded_base")
  index_base.batch_insert(X_train_encoded_base)
  index_base.save()
```

```
[]: res_base = classify(X_test_encoded_base, index_base, y_train, size_k=9, __ →weight=True)
```

```
[]: print(metrics.classification_report(y_test, res_base))
```

	precision	recall	f1-score	support
0	0.33	0.28	0.30	319
1	0.42	0.41	0.42	389
2	0.42	0.47	0.44	394
3	0.37	0.39	0.38	392
4	0.34	0.49	0.40	385
5	0.42	0.49	0.45	395
6	0.63	0.70	0.66	390
7	0.68	0.56	0.61	396
8	0.67	0.50	0.57	398
9	0.82	0.70	0.75	397
10	0.82	0.76	0.79	399
11	0.41	0.52	0.46	396
12	0.45	0.27	0.34	393
13	0.70	0.66	0.68	396
14	0.71	0.66	0.68	394
15	0.52	0.71	0.60	398
16	0.40	0.54	0.46	364
17	0.67	0.61	0.64	376
18	0.43	0.32	0.37	310
19	0.15	0.09	0.11	251

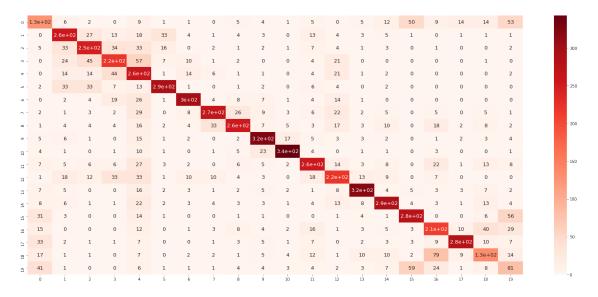
```
      accuracy
      0.52
      7532

      macro avg
      0.52
      0.51
      0.51
      7532

      weighted avg
      0.53
      0.52
      0.52
      7532
```

```
[]: cfm_base = metrics.confusion_matrix(y_test, res_base)
plt.figure(None,(28,12))
sn.heatmap(cfm_base,annot=True, annot_kws={"size": 13}, cmap=plt.cm.Reds)
```

[]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8d937b9978>



[]: labels

```
'soc.religion.christian',
'talk.politics.guns',
'talk.politics.mideast',
'talk.politics.misc',
'talk.religion.misc']
```

[20]: X_test[29]

[20]: "\n\nI don't know about Canada, but I have heard from people\ndoing translation work in Papua New Quinea, that they\nlike them and have had good response on service.\n\nAnother is seriously considering buying one.\n"