2.PyTorch Text-CNN with GloVe(F1-Score)

November 4, 2020

Acknowledgments: https://www.kaggle.com/ziliwang/pytorch-text-cnn

Acknowledgments: I would also like to thank Youwen Wang for helping me out through out for technicalities and places where I was stuck.

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

Mounted at /content/gdrive

```
[]: import pandas as pd
  import numpy as np
  import torch
  from torch import nn
# from sklearn.metrics import fbeta_score, make_scorer
# f2score=make_scorer(fbeta_score, beta=2)
  from sklearn.metrics import f1_score
  import torchtext
  from tqdm import tqdm, tqdm_notebook
  from nltk import word_tokenize
  import random
  from torch import optim
  import nltk
  nltk.download('punkt')
```

[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Unzipping tokenizers/punkt.zip.

[]: True

```
[]: !pip install -q kaggle
[]: from google.colab import files
[]: files.upload()
```

<IPython.core.display.HTML object>

```
Saving kaggle.json to kaggle.json
[]: {'kaggle.json':
     b'{"username": "nabhsanjaymehtautd", "key": "ee0e2e2e8b50d345f23e44404b090088"}'}
[]: |mkdir ~/.kaggle/
[]: !cp kaggle.json ~/.kaggle/
    !chmod 60 ~/.kaggle/kaggle.json
[]: | !kaggle competitions download -c transferlearning-dl-spring2020
    Warning: Your Kaggle API key is readable by other users on this system! To fix
    this, you can run 'chmod 600 /root/.kaggle/kaggle.json'
    Warning: Looks like you're using an outdated API Version, please consider
    updating (server 1.5.9 / client 1.5.4)
    Downloading sample_submission.csv to /content
      0% 0.00/34.2k [00:00<?, ?B/s]
    100% 34.2k/34.2k [00:00<00:00, 30.1MB/s]
    Downloading train.csv.zip to /content
      0% 0.00/527k [00:00<?, ?B/s]
    100% 527k/527k [00:00<00:00, 74.4MB/s]
    Downloading test.csv to /content
      0% 0.00/506k [00:00<?, ?B/s]
    100% 506k/506k [00:00<00:00, 70.1MB/s]
[]: !unzip train.csv.zip -d train
    Archive: train.csv.zip
      inflating: train/train.csv
[]: data = pd.read_csv('/content/train/train.csv', encoding = "ISO-8859-1")
     testdata = pd.read_csv('/content/test.csv', encoding="ISO-8859-1")
[]: data.head()
[]:
          id
                                                            text target
     0 86426 @USER She should ask a few native Americans wh...
     1 16820 Amazon is investigating Chinese employees who ...
     2 62688 @USER Someone should'veTaken" this piece of sh...
                                                                     1
     3 43605
              QUSER QUSER Obama wanted liberals & amp; illega...
     4 97670
                               QUSER Liberals are all Kookoo!!!
                                                                       1
[]: text = torchtext.data.Field(lower=True, batch_first=True,_
     →tokenize=word_tokenize, fix_length=70)
     id = torchtext.data.Field()
     target = torchtext.data.Field(sequential=False, use_vocab=False, is_target=True)
```

```
[]: train = torchtext.data.TabularDataset(path='/content/train/train.csv',
                                          format='csv',
                                          fields={'text': ('text',text),
                                                  'target': ('target',target)})
    test = torchtext.data.TabularDataset(path='/content/test.csv',
                                         format='csv',
                                         fields={'id': ('id', id),
                                                 'text': ('text', text)})
[]: #Build Vocabulary
    text.build_vocab(train, test, min_freq=3)
    id.build_vocab(test)
[]: !wget http://nlp.stanford.edu/data/glove.6B.zip
    --2020-11-04 05:04:51-- http://nlp.stanford.edu/data/glove.6B.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.6B.zip [following]
    --2020-11-04 05:04:51-- https://nlp.stanford.edu/data/glove.6B.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.6B.zip [following]
    --2020-11-04 05:04:51-- http://downloads.cs.stanford.edu/nlp/data/glove.6B.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 OK
    Length: 862182613 (822M) [application/zip]
    Saving to: 'glove.6B.zip'
    glove.6B.zip
                        in 6m 26s
    2020-11-04 05:11:17 (2.13 MB/s) - 'glove.6B.zip' saved [862182613/862182613]
[]: !unzip glove*.zip
    Archive: glove.6B.zip
      inflating: glove.6B.50d.txt
      inflating: glove.6B.100d.txt
      inflating: glove.6B.200d.txt
      inflating: glove.6B.300d.txt
```

```
[]:!ls
     ! pwd
                       glove.6B.300d.txt kaggle.json
    gdrive
                                                                  test.csv
    glove.6B.100d.txt glove.6B.50d.txt
                                          sample_data
                                                                  train
    glove.6B.200d.txt glove.6B.zip
                                          sample_submission.csv train.csv.zip
    /content
[]:|glove = glove = torchtext.vocab.GloVe(name='840B', dim=300)#torchtext.vocab.
     → Vectors('/content/glove.6B.200d.txt')
     text.vocab.set_vectors(glove.stoi, glove.vectors, dim=300)
    .vector_cache/glove.840B.300d.zip: 2.18GB [16:53, 2.15MB/s]
              2195480/2196017 [04:11<00:00, 9036.46it/s]
[]: class TextCNN(nn.Module):
         def __init__(self, lm, padding_idx, static=True, kernel_num=128,__
      →fixed_length=50, kernel_size=[2, 5, 10], dropout=0.2):
             super(TextCNN, self).__init__()
             self.dropout = nn.Dropout(p=dropout)
             self.embedding = nn.Embedding.from_pretrained(lm)
             if static:
                 self.embedding.weight.requires_grad = False
             self.embedding.padding idx = padding idx
             self.conv = nn.ModuleList([nn.Conv2d(1, kernel_num, (i, self.embedding.
      →embedding_dim)) for i in kernel_size])
             self.maxpools = [nn.MaxPool2d((fixed_length+1-i,1)) for i in_
      →kernel_size]
             self.fc = nn.Linear(len(kernel_size)*kernel_num, 1)
         def forward(self, input):
             x = self.embedding(input).unsqueeze(1) # B X Ci X H X W
            x = [self.maxpools[i](torch.tanh(cov(x))).squeeze(3).squeeze(2) for i,_{\sqcup}
      →cov in enumerate(self.conv)] # B X Kn
             x = torch.cat(x, dim=1) # B X Kn * len(Kz)
             y = self.fc(self.dropout(x))
            return y
[ ]: def search_best_f1(true, pred):
         tmp = [0,0,0] # idx, cur, max
         delta = 0
         for tmp[0] in np.arange(0.1, 0.501, 0.01):
             tmp[1] = f1_score(true, np.array(pred)>tmp[0])
             if tmp[1] > tmp[2]:
                 delta = tmp[0]
                 tmp[2] = tmp[1]
```

```
[]: def training(epoch, model, loss_func, optimizer, train_iter):
         while e < epoch:
             train_iter.init_epoch()
             losses, preds, true = [], [], []
             for train_batch in tqdm(list(iter(train_iter)), 'epcoh {} training'.
      →format(e)):
                 model.train()
                 x = train_batch.text.cuda()
                 y = train_batch.target.type(torch.Tensor).cuda()
                 true.append(train_batch.target.numpy())
                 model.zero grad()
                 pred = model.forward(x).view(-1)
                 loss = loss_function(pred, y)
                 preds.append(torch.sigmoid(pred).cpu().data.numpy())
                 losses.append(loss.cpu().data.numpy())
                 loss.backward()
                   clip_grad_norm_(model.parameters(), 2)
     #
                 optimizer.step()
             train_f1, alpha_train = search_best_f1([j for i in true for j in i], [ju
      →for i in preds for j in i])
             print('epcoh {:02} - train loss {:.4f} - train f1 {:.4f} - delta {:.
      \rightarrow4f}'.format(
                                 e, np.mean(losses), train_f1, alpha_train))
             e += 1
         return alpha_train
[]: random.seed(1234)
     batch size = 512
     train_iter = torchtext.data.BucketIterator(dataset=train,
                                                     batch_size=batch_size,
                                                     shuffle=True,
                                                     sort=False)
[]: def init_network(model, method='xavier', exclude='embedding', seed=123):
         torch.manual_seed(seed)
         if torch.cuda.is_available():
             torch.cuda.manual_seed_all(seed)
         for name, w in model.named_parameters():
             if not exclude in name:
                 if 'weight' in name:
                     if method is 'xavier':
                         nn.init.xavier_normal_(w)
```

return tmp[2], delta

```
elif method is 'kaiming':
                         nn.init.kaiming_normal_(w)
                        nn.init.normal_(w)
                 elif 'bias' in name:
                    nn.init.constant_(w, 0.0)
                 else:
                    pass
[]: def print_model(model, ignore='embedding'):
        total = 0
        for name, w in model.named_parameters():
            if not ignore or ignore not in name:
                total += w.nelement()
                print('{}: {} parameters'.format(name, w.shape, w.nelement()))
        print('-----*4)
         print('Total {} parameters'.format(total))
[]: text.fix_length = 70
    model = TextCNN(text.vocab.vectors,
                    padding_idx=text.vocab.stoi[text.pad_token],
                    kernel_size=[1, 2, 3, 5], kernel_num=128,
                     static=False, fixed_length=text.fix_length,
                     dropout=0.1).cuda()
    init_network(model)
    optimizer = optim.Adam(params=model.parameters(), lr=1e-3)
    loss_function = nn.MSELoss()
    print_model(model, ignore=None)
    embedding.weight: torch.Size([6484, 300]) 1945200 parameters
    conv.0.weight : torch.Size([128, 1, 1, 300]) 38400 parameters
    conv.0.bias : torch.Size([128]) 128 parameters
    conv.1.weight : torch.Size([128, 1, 2, 300]) 76800 parameters
    conv.1.bias : torch.Size([128]) 128 parameters
    conv.2.weight : torch.Size([128, 1, 3, 300]) 115200 parameters
    conv.2.bias : torch.Size([128]) 128 parameters
    conv.3.weight: torch.Size([128, 1, 5, 300]) 192000 parameters
    conv.3.bias : torch.Size([128]) 128 parameters
    fc.weight : torch.Size([1, 512]) 512 parameters
    fc.bias : torch.Size([1]) 1 parameters
    Total 2368625 parameters
```

```
[]: alpha = training(3, model, loss_function, optimizer, train_iter)
                                     | 0/19 [00:00<?, ?it/s]
    epcoh 0 training:
                        0%1
    epcoh 0 training:
                        5% l
                                     | 1/19 [00:00<00:07, 2.47it/s]
                                   | 6/19 [00:00<00:03, 3.44it/s]
    epcoh 0 training:
                       32%|
                                  | 11/19 [00:00<00:01, 4.77it/s]
    epcoh 0 training: 58%
    epcoh 0 training: 100%
                                 | 19/19 [00:00<00:00, 23.37it/s]
    epcoh 1 training:
                        0%1
                                     | 0/19 [00:00<?, ?it/s]
    epcoh 00 - train_loss 0.4193 - train f1 0.5021 - delta 0.4700
    epcoh 1 training: 26%
                                    | 5/19 [00:00<00:00, 48.56it/s]
    epcoh 1 training: 58%
                                  | 11/19 [00:00<00:00, 48.85it/s]
                                | 19/19 [00:00<00:00, 46.88it/s]
    epcoh 1 training: 100%
    epcoh 2 training:
                        0%1
                                     | 0/19 [00:00<?, ?it/s]
    epcoh 01 - train_loss 0.1632 - train f1 0.5250 - delta 0.5000
                                    | 5/19 [00:00<00:00, 45.40it/s]
    epcoh 2 training:
                       26%|
    epcoh 2 training: 53%
                                  | 10/19 [00:00<00:00, 45.00it/s]
                                | 19/19 [00:00<00:00, 45.87it/s]
    epcoh 2 training: 100%
    epcoh 02 - train_loss 0.1384 - train f1 0.5279 - delta 0.5000
[ ]: def predict(model, test_list):
        pred = []
        with torch.no_grad():
             for test_batch in test_list:
                 model.eval()
                 x = test_batch.text.cuda()
                 pred += torch.sigmoid(model.forward(x).view(-1)).cpu().data.numpy().
      →tolist()
        return pred
[]: |test_list = list(torchtext.data.BucketIterator(dataset=test,
                                         batch_size=batch_size,
                                         sort=False,
                                         train=False))
[ ]: preds = predict(model, test_list)
     sub = pd.DataFrame()
     sub['id'] = [id.vocab.itos[j] for i in test_list for j in i.id.view(-1).numpy()]
     sub['prediction'] = (preds > alpha).astype(int)
     sub.head()
```