

## 6. PyTorch Text-CNN with GloVe(F2-Score\_BCEWithLogitsLoss)

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PyTorch Text-CNN with GloVe(F2-Score:BCEWithLogitsLoss)

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.

Trying different Loss functions worked great for me - <https://pytorch.org/docs/stable/nn.html>

Kaggle Public Score: 0.75187

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

Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive", force\_remount=True).

```
[34]: 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] Package punkt is already up-to-date!

```
[34]: True
```

```
[35]: !pip install -q kaggle
```

```
[36]: from google.colab import files
```

```
[37]: files.upload()
```

```
<IPython.core.display.HTML object>
```

```
Saving kaggle.json to kaggle (1).json
```

```
[37]: {'kaggle.json':  
      b'{"username":"nabhsanjaymehtautd","key":"ee0e2e2e8b50d345f23e44404b090088"}'}
```

```
[38]: !mkdir ~/.kaggle/
```

```
mkdir: cannot create directory '/root/.kaggle/': File exists
```

```
[39]: !cp kaggle.json ~/.kaggle/
```

```
[40]: !chmod 60 ~/.kaggle/kaggle.json
```

```
[41]: !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Warning: Looks like you're using an outdated API Version, please consider  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)
```

```
test.csv: Skipping, found more recently modified local copy (use --force to  
force download)
```

```
sample_submission.csv: Skipping, found more recently modified local copy (use  
--force to force download)
```

```
train.csv.zip: Skipping, found more recently modified local copy (use --force to  
force download)
```

```
[42]: !unzip train.csv.zip -d train
```

```
Archive: train.csv.zip
```

```
replace train/train.csv? [y]es, [n]o, [A]ll, [N]one, [r]ename: y
```

```
inflating: train/train.csv
```

```
[43]: data = pd.read_csv('/content/train/train.csv', encoding = "ISO-8859-1")  
      testdata = pd.read_csv('/content/test.csv', encoding="ISO-8859-1")
```

```
[44]: data.head()
```

```
[44]:
```

	id	text	target
0	86426	@USER She should ask a few native Americans wh...	1
1	16820	Amazon is investigating Chinese employees who ...	0
2	62688	@USER Someone should'veTaken" this piece of sh...	1
3	43605	@USER @USER Obama wanted liberals & illeg...	0
4	97670	@USER Liberals are all Kookoo !!!	1

```
[45]: 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)
```

```
[46]: 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)})
```

```
[47]: #Build Vocabulary
text.build_vocab(train, test, min_freq=3)
id.build_vocab(test)
```

```
[48]: !wget http://nlp.stanford.edu/data/glove.6B.zip
```

```
--2020-11-05 00:46:05-- 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]
--2020-11-05 00:46:05-- 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]
--2020-11-05 00:46:05-- 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.1'
```

```
glove.6B.zip.1      100%[=====>] 822.24M  2.00MB/s    in 6m 28s
```

```
2020-11-05 00:52:34 (2.12 MB/s) - 'glove.6B.zip.1' saved [862182613/862182613]
```

```
[49]: !unzip glove*.zip
```

```

Archive:  glove.6B.zip
replace glove.6B.50d.txt? [y]es, [n]o, [A]ll, [N]one, [r]ename: y
  inflating: glove.6B.50d.txt
replace glove.6B.100d.txt? [y]es, [n]o, [A]ll, [N]one, [r]ename: y
  inflating: glove.6B.100d.txt      y

replace glove.6B.200d.txt? [y]es, [n]o, [A]ll, [N]one, [r]ename:  inflating:
glove.6B.200d.txt      y

replace glove.6B.300d.txt? [y]es, [n]o, [A]ll, [N]one, [r]ename:  inflating:
glove.6B.300d.txt      y

```

```
[50]: !ls
      !pwd
```

```

gdrive          glove.6B.50d.txt  kaggle.json      train
glove.6B.100d.txt glove.6B.zip    sample_data      train.csv.zip
glove.6B.200d.txt glove.6B.zip.1  sample_submission.csv
glove.6B.300d.txt 'kaggle (1).json' test.csv
/content

```

```
[51]: 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)
```

```
[52]: 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,
      ↪ 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

```

```

[53]: 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] = fbeta_score(true, np.array(pred)>tmp[0], beta=2)
        if tmp[1] > tmp[2]:
            delta = tmp[0]
            tmp[2] = tmp[1]
    return tmp[2], delta

```

```

[54]: def training(epoch, model, loss_func, optimizer, train_iter):
    e = 0

    while e < epoch:
        train_iter.init_epoch()
        losses, preds, true = [], [], []
        for train_batch in tqdm(list(iter(train_iter)), 'epoch {} 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], [j
→for i in preds for j in i])
            print('epoch {:02} - train_loss {:.4f} - train f1 {:.4f} - delta {:.
→4f}'.format(
                e, np.mean(losses), train_f1, alpha_train))

            e += 1
    return alpha_train

```

```

[55]: random.seed(1234)
batch_size = 512
train_iter = torchtext.data.BucketIterator(dataset=train,
                                           batch_size=batch_size,
                                           shuffle=True,

```

```
sort=False)
```

```
[56]: 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)
                elif method is 'kaiming':
                    nn.init.kaiming_normal_(w)
                else:
                    nn.init.normal_(w)
            elif 'bias' in name:
                nn.init.constant_(w, 0.0)
            else:
                pass
```

```
[57]: 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))
```

```
[104]: 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=0.003)

loss_function = nn.BCEWithLogitsLoss()

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

```

```
[116]: alpha = training(3, model, loss_function, optimizer, train_iter)
```

```

epoch 0 training: 0%|          | 0/19 [00:00<?, ?it/s]
epoch 0 training: 16%|         | 3/19 [00:00<00:00, 24.71it/s]
epoch 0 training: 26%|         | 5/19 [00:00<00:00, 22.23it/s]
epoch 0 training: 42%|         | 8/19 [00:00<00:00, 22.65it/s]
epoch 0 training: 58%|         | 11/19 [00:00<00:00, 22.85it/s]
epoch 0 training: 74%|         | 14/19 [00:00<00:00, 23.10it/s]
epoch 0 training: 100%|        | 19/19 [00:00<00:00, 23.03it/s]

```

```
epoch 1 training: 0%|          | 0/19 [00:00<?, ?it/s]
```

```
epoch 00 - train_loss 0.0397 - train f1 0.9896 - delta 0.2500
```

```

epoch 1 training: 21%|         | 4/19 [00:00<00:00, 31.34it/s]
epoch 1 training: 37%|         | 7/19 [00:00<00:00, 30.09it/s]
epoch 1 training: 53%|         | 10/19 [00:00<00:00, 27.82it/s]
epoch 1 training: 68%|         | 13/19 [00:00<00:00, 26.32it/s]
epoch 1 training: 84%|         | 16/19 [00:00<00:00, 25.37it/s]
epoch 1 training: 100%|        | 19/19 [00:00<00:00, 25.23it/s]

```

```
epoch 2 training: 0%|          | 0/19 [00:00<?, ?it/s]
```

```
epoch 01 - train_loss 0.0310 - train f1 0.9911 - delta 0.2500
```

```

epoch 2 training: 21%|         | 4/19 [00:00<00:00, 28.45it/s]
epoch 2 training: 37%|         | 7/19 [00:00<00:00, 26.67it/s]
epoch 2 training: 58%|         | 11/19 [00:00<00:00, 27.27it/s]
epoch 2 training: 74%|         | 14/19 [00:00<00:00, 26.06it/s]
epoch 2 training: 100%|        | 19/19 [00:00<00:00, 25.33it/s]

```

```
epoch 02 - train_loss 0.0313 - train f1 0.9907 - delta 0.2900
```

```
[117]: 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

```

```

[118]: test_list = list(torchtext.data.BucketIterator(dataset=test,
                                                    batch_size=batch_size,
                                                    sort=False,
                                                    train=False))

```

```

[119]: 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()

```

```

[119]:      id  prediction
0  90194           0
1  77444           1
2  13384           0
3  54920           0
4  56117           1

```

```

[120]: pd.DataFrame({'Id': sub.id, 'Target': sub.prediction}).to_csv('submission1.
→csv', index=False)

```

```

[64]: pwd

```

```

[64]: '/content'

```