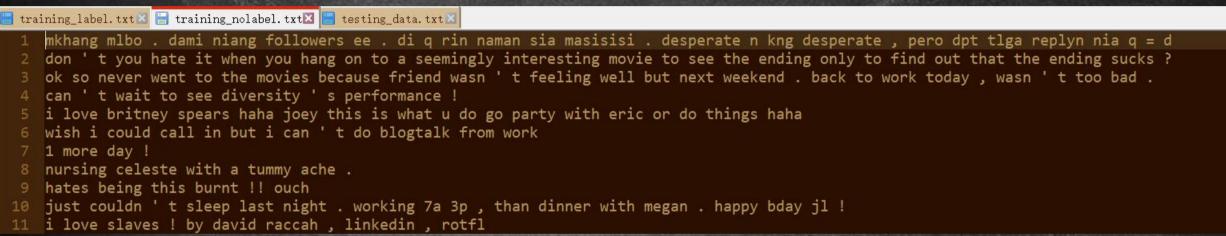
学习汇报

文本分类

通过循环神经网络对句子进行情感分类。

给定一句子, 判断这句子是正面还是负面的

```
training_label.txt training_nolabel.txt testing_data.txt testing_data.txt
```



training_label.txt training_nolabel.txt testing_data.txt testing_data.txt to did, text 2 0,my dog ate our dinner . no , seriously ... he ate it . 3 1,omg last day sooon n of primary nooooooo x im gona be swimming out of school wif the amount of tears am gona cry 4 2,stupid boys .. they 're so .. stupid ! 5 3,hi ! do u know if the nurburgring is open for tourists today ? we want to go , but there is an event today 4 4,having lunch in the office , and thinking of how to resolve this discount form issue 5 5,shopping was fun 8 6,wondering where all the nice weather has gone . 9 7,morning ! yeeesssssss new mimi in aug 10 8,umm ... maybe that 's how the british spell it ? 11 9,yes it 's 3 : 50 am . yes i 'm still awake . yes i can 't sleep . yes i 'll regret it tomorrow . haha i love you mr saturday 10 10,cute heart shaped portal cube . my baby is playing games , im reading fan fictions !

```
| def | load_training_data(path='training_label.txt'):
| # 读版 training 需要的数据
| # 如果是 'training_label.txt', 需要读版 label, 如果是 'training_nolabel.txt', 不需要读版 label
| if 'training_label' in path:
| with open(path, 'r') as f:
| lines = f.readlines()
| # lines是二维数组,第一维是行line(按回年分割),第二维是每行的单词(按空格分割)
| lines = [line.strip('\n').split(' ') for line in lines]
| # 每行按空格分割后,第2个符号之后都是何子的单词
| x = [line[2:] for line in lines]
| # 每行按空格分割后,第0个符号是label
| y = [line[0] for line in lines]
| return x, y
| else:
| with open(path, 'r') as f:
| lines 是二维数组,第一维是行line(按回车分割),第二维是每行的单词(按空格分割)
| x = [line.strip('\n').split(' ') for line in lines]
| return x
```

```
def load_testing_data(path='testing_data'):
# 读取 testing 需要的数据
with open(path, 'r') as f:
lines = f.readlines()
# 第0行是表头,从第1行开始是数据
# 第0列是id,第1列是文本,按逗号分割,需要逗号之后的文本
X = ["".join(line.strip('\n').split(",")[1:]).strip() for line in lines[1:]]
X = [sen.split(' ') for sen in X]
return X

Eddef evaluation(outputs, labels):
# outputs => 预测值,概率 (float)
# labels => 真实值,标签 (0或1)
outputs[outputs>=0.5] = 1 # 大子等于 0.5 为正面
outputs[outputs<0.5] = 0 # 小于 0.5 为负面
accuracy = torch.sum(torch.eq(outputs, labels)).item()
return accuracy
```

```
from gensim.models import Word2Vec
def train word2vec(x):
   # 训练 word to vector 的 word embedding
   model = Word2Vec(x, size=250, window=5, min count=5, workers=12, iter=10, sg=1)
   return model
# 读取 training 数据
print("loading training data ...")
train x, y = load training data('training label.txt')
train x no label = load training data('training nolabel.txt')
# 读取 testing 数据
print("loading testing data ...")
test x = load testing data('testing data.txt')
# 把 trainina 中的 word 变成 vector
# model = train word2vec(train x + train x no label + test x) # w2v all
model = train word2vec(train x + test x) # w2v
# 保存 vector
print("saving model ...")
model.save('w2v.model')
```

```
loading training data ...
loading testing data ...
saving model ...
```

```
class Preprocess():
    def __init__(self, sentences, sen_len, w2v_path):
        self.w2v_path = w2v_path # word2vec 的存储路径
        self.sentences = sentences # 句子
        self.sen_len = sen_len # 句子的固定长度
        self.idx2word = []
        self.word2idx = {}
        self.embedding_matrix = []
```

```
def add_embedding(self, word):
    # 把一个随机生成的表征向量 vector
    vector = torch.empty(1, self.embedding_dim)
    torch.nn.init.uniform_(vector)
    # 它的 index 是 word2idx 这个词典的长度,即最后一个
    self.word2idx[word] = len(self.word2idx)
    self.idx2word.append(word)
    self.embedding_matrix = torch.cat([self.embedding_matrix, vector], 0)
```

```
from torch import nn
class LSTM Net(nn.Module):
   def init (self, embedding, embedding dim, hidden dim, num layers, dropout=0.5, fix embedding=True):
       super(LSTM Net, self). init ()
       # embedding layer
       self.embedding = torch.nn.Embedding(embedding.size(0),embedding.size(1))
       self.embedding.weight = torch.nn.Parameter(embedding)
       # 是否将 embedding 固定住,如果 fix embedding 为 False,在训练过程中,embedding 也会跟着被训练
       self.embedding.weight.requires grad = False if fix embedding else True
       self.embedding dim = embedding.size(1)
       self.hidden dim = hidden dim
       self.num layers = num layers
       self.dropout = dropout
       self.lstm = nn.LSTM(embedding dim, hidden dim, num layers=num_layers, batch_first=True)
       self.classifier = nn.Sequential(
               nn.Dropout(dropout),
               nn.Linear(hidden dim, 1),
               nn.Sigmoid()
   def forward(self, inputs):
       inputs = self.embedding(inputs)
       x, = self.lstm(inputs, None)
       # 取用 LSTM 最后一层的 hidden state 丢到分类器中
       x = x[:, -1, :]
       x = self.classifier(x)
       return x
```

```
def training(batch size, n epoch, lr, train, valid, model, device):
    total = sum(p.numel() for p in model.parameters())
   trainable = sum(p.numel() for p in model.parameters() if p.requires grad)
    print('\nstart training, parameter total:{}, trainable:{}\n'.format(total, trainable))
   loss = nn.BCELoss() # 定义损失函数为二元交叉熵损失 binary cross entropy loss
   t batch = len(train) # training 数据的batch size大小
   v batch = len(valid) # validation 数据的batch size大小
   optimizer = optim.Adam(model.parameters(), lr=lr) # Adam, 学习率Lr
   total loss, total acc, best acc = 0, 0, 0
    for epoch in range(n epoch):
       total loss, total acc = 0, 0
       # training
       model.train()
       for i, (inputs, labels) in enumerate(train):
           inputs = inputs.to(device, dtype=torch.long)
           labels = labels.to(device, dtype=torch.float)
           optimizer.zero grad()
           outputs = model(inputs)
           outputs = outputs.squeeze()
           batch loss = loss(outputs, labels)
           batch loss.backward()
           optimizer.step()
           accuracy = evaluation(outputs, labels)
           total acc += (accuracy / batch size)
           total loss += batch loss.item()
       print('Epoch | {}/{}'.format(epoch+1, n epoch))
       print('Train | Loss:{:.5f} Acc: {:.3f}'.format(total loss/t batch, total acc/t batch*100))
```

```
# 定义模型
model = LSTM Net(embedding, embedding dim=250, hidden dim=150, num layers=1, dropout=0.5, fix embedding=fix embedding)
model = model.to(device) # device为 "cuda", model 使用 GPU 来训练 (inputs 也需要是 cuda tensor)
# 把 data 分为 training data 和 validation data (将一部分 training data 作为 validation data)
X train, X val, y train, y val = train test split(train x, y, test size = 0.1, random state = 1, stratify = y)
print('Train | Len:{} \nValid | Len:{}'.format(len(y train), len(y val)))
# 把 data 做成 dataset 供 dataloader 取用
train dataset = TwitterDataset(X=X train, y=y train)
val dataset = TwitterDataset(X=X val, y=y val)
# 把 data 转成 batch of tensors
train loader = DataLoader(train dataset, batch size = batch size, shuffle = True, num workers = 0)
val loader = DataLoader(val dataset, batch size = batch size, shuffle = False, num workers = 0)
# 开始训练
                                                                             loading data ...
training(batch size, epoch, lr, train loader, val loader, model, device)
```

Get embedding ... loading word to vec model ... get words #24694 total words: 24696 start training, parameter total:6415351, trainable:241351 Train | Loss: 0. 50217 Acc: 74. 707 Valid | Loss: 0. 46004 Acc: 77. 876 saving model with acc 77.876 Train | Loss: 0. 44169 Acc: 79. 298 Valid | Loss: 0. 43643 Acc: 79. 225 saving model with acc 79,225 Train | Loss: 0. 42586 Acc: 80. 206 Valid | Loss: 0. 44654 Acc: 78. 608 Train | Loss: 0. 41382 Acc: 80. 915 Valid Loss: 0. 42438 Acc: 80. 210 saving model with acc 80.210 Train | Loss: 0. 40071 Acc: 81. 617 Valid | Loss: 0. 43327 Acc: 80. 120

```
model.eval()
                              # 将 model 的模式设为 eval, 这样 model 的参数就会被固定住
               ret output = [] #返回的output
               with torch.no grad():
                  for i, inputs in enumerate(test loader):
                      inputs = inputs.to(device, dtype=torch.long)
                      outputs = model(inputs)
                      outputs = outputs.squeeze()
                      outputs[outputs>=0.5] = 1 # 大于等于0.5为正面
                      outputs[outputs<0.5] = 0 # 小于0.5为负面
                      ret output += outputs.int().tolist()
              return ret output
# 读取测试数据test x
print("loading testing data ...")
test x = load testing data('testing data.txt')
preprocess = Preprocess(test x, sen len, w2v path=w2v path)
embedding = preprocess.make embedding(load=True)
test x = preprocess.sentence word2idx()
test dataset = TwitterDataset(X=test x, y=None)
test loader = DataLoader(test dataset, batch size = batch size, shuffle = False, num workers = 0)
# 读取模型
print('\nload model ...')
model = torch.load('ckpt.model')
#测试模型
outputs = testing(batch size, test loader, model, device)
# 保存为 csv
tmp = pd.DataFrame({"id":[str(i) for i in range(len(test x))],"label":outputs})
print("save csv ...")
tmp.to csv('predict.csv', index=False)
print("Finish Predicting")
```

def testing(batch size, test loader, model, device):

loading testing data ...

Get embedding ...
loading word to vec model ...

get words #24694
total words: 24696
sentence count #200000
load model ...
save csv ...
Finish Predicting

4	Α	В	С
1	id	label	
	0	0	
3	1	0	
4	2	0	
5	3	1	
5	4	1	
2 4 5 7 8	5	1	
8	6	0	
9	7	1	
0	8	1	
1	9	1	
2	10	1	
3	11	1	
4	12	1	
5	13	0	
6	14	1	
7	15	1	
8	16	0	
9	17	0	
0	18	1	
1	19	0	
2	20	0	
3	21	1	
4	22	0	
5	23	0	
6	24	1	
7	25	0	
8	26	0	
0	0.7		