```
In [3]:
            label to ix=np.load('label to ix.npy', allow pickle=True).item()
            ix_to_label=np.load('ix_to_label.npy', allow_pickle=True)
          3 training_data=np.load('training_data.npy', allow_pickle=True)
          4 test_data=np.load('test_data.npy', allow_pickle=True)
          5 | val_data=np.load('val_data.npy', allow_pickle=True)
          6 | word_to_ix=np.load('word_to_ix.npy', allow_pickle=True).item()
          7
             ix_to_word=np.load('ix_to_word.npy', allow_pickle=True)
            newwikivec=np.load('newwikivec.npy', allow_pickle=True)
             wikivoc=np.load('wikivoc.npy', allow_pickle=True).item()
          9
         10
         11
         12
         13
             wikisize=newwikivec.shape[0]
         14
             rvocsize=newwikivec.shape[1]
         15
             wikivec=autograd.Variable(torch.FloatTensor(newwikivec))
         16
         17
             batchsize=32
         18
         19
         20
         21
             def preprocessing(data):
         22
         23
                 new data=[]
         24
                 for i, note, j in data:
         25
                     templabel=[0.0]*len(label_to_ix)
         26
                     for jj in j:
         27
                         if jj in wikivoc:
         28
                             templabel[label_to_ix[jj]]=1.0
         29
                     templabel=np.array(templabel,dtype=float)
         30
                     new_data.append((i, note, templabel))
         31
                 new_data=np.array(new_data)
         32
         33
                 lenlist=[]
         34
                 for i in new_data:
         35
                     lenlist.append(len(i[0]))
         36
                 sortlen=sorted(range(len(lenlist)), key=lambda k: lenlist[k])
         37
                 new data=new data[sortlen]
         38
         39
                 batch data=[]
         40
         41
                 for start_ix in range(0, len(new_data)-batchsize+1, batchsize):
         42
                     thisblock=new data[start ix:start ix+batchsize]
         43
                     mybsize= len(thisblock)
         44
                     numword=np.max([len(ii[0]) for ii in thisblock])
         45
                     main matrix = np.zeros((mybsize, numword), dtype= np.int)
         46
                     for i in range(main matrix.shape[0]):
         47
                         for j in range(main_matrix.shape[1]):
         48
                             try:
         49
                                  if thisblock[i][0][j] in word to ix:
         50
                                      main_matrix[i,j] = word_to_ix[thisblock[i][0][j]]
         51
         52
                             except IndexError:
         53
                                  pass
                                             # because initialze with 0, so you pad with 0
         54
         55
                     xxx2=[]
         56
                     yyy=[]
```

```
for ii in thisblock:
57
58
                xxx2.append(ii[1])
59
                yyy.append(ii[2])
60
            xxx2=np.array(xxx2)
61
62
            yyy=np.array(yyy)
63
            batch_data.append((autograd.Variable(torch.from_numpy(main_matrix)),
64
        return batch_data
   batchtraining_data=preprocessing(training_data)
65
   batchtest data=preprocessing(test data)
   batchval_data=preprocessing(val_data)
67
68
69
70
```

/opt/conda/lib/pytnon3.//site-packages/ipykernei\_launcher.py:31: visibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray

Create the model:

```
In [4]:
             Embeddingsize=100
             hidden dim=200
          2
          3
             class CNN(nn.Module):
          4
          5
                 def __init__(self, batch_size, vocab_size, tagset_size):
          6
                     super(CNN, self).__init__()
          7
                     self.hidden dim = hidden dim
          8
                     self.word embeddings = nn.Embedding(vocab size+1, Embeddingsize, pad
          9
                     self.embed drop = nn.Dropout(p=0.2)
         10
                     self.hidden2tag = nn.Linear(300, tagset size)
         11
         12
         13
                     self.convs1 = nn.Conv1d(Embeddingsize,100,3)
         14
                     self.convs2 = nn.Conv1d(Embeddingsize,100,4)
         15
         16
                     self.convs3 = nn.Conv1d(Embeddingsize,100,5)
         17
         18
         19
                     self.layer2 = nn.Linear(Embeddingsize, 1,bias=False)
                     self.embedding=nn.Linear(rvocsize,Embeddingsize)
         20
                     self.vattention=nn.Linear(Embeddingsize,Embeddingsize)
         21
         22
         23
                     self.sigmoid = nn.Sigmoid()
                     self.tanh = nn.Tanh()
         24
         25
                     self.dropout = nn.Dropout(0.2)
         26
         27
                 def forward(self, vec1, nvec, wiki, simlearning):
         28
         29
                     thisembeddings=self.word embeddings(vec1)
                     thisembeddings = self.embed drop(thisembeddings)
         30
         31
                     thisembeddings=thisembeddings.transpose(1,2)
         32
         33
                     output1=self.tanh(self.convs1(thisembeddings))
                     output1=nn.MaxPool1d(output1.size()[2])(output1)
         34
         35
                     output2=self.tanh(self.convs2(thisembeddings))
         36
         37
                     output2=nn.MaxPool1d(output2.size()[2])(output2)
         38
         39
                     output3=self.tanh(self.convs3(thisembeddings))
         40
                     output3=nn.MaxPool1d(output3.size()[2])(output3)
         41
         42
                     output4 = torch.cat([output1,output2,output3], 1).squeeze(2)
         43
                     if simlearning==1:
         44
         45
                         nvec=nvec.view(batchsize,1,-1)
         46
                         nvec=nvec.expand(batchsize,wiki.size()[0],-1)
         47
                         wiki=wiki.view(1,wiki.size()[0],-1)
         48
                         wiki=wiki.expand(nvec.size()[0],wiki.size()[1],-1)
         49
                         new=wiki*nvec
         50
                         new=self.embedding(new)
         51
                         vattention=self.sigmoid(self.vattention(new))
         52
                         new=new*vattention
         53
                         vec3=self.layer2(new)
         54
                         vec3=vec3.view(batchsize,-1)
         55
         56
```

```
vec2 = self.hidden2tag(output4)
if simlearning==1:
    tag_scores = self.sigmoid(vec2.detach()+vec3)
else:
    tag_scores = self.sigmoid(vec2)

return tag_scores
return tag_scores
```

Train the model:

```
In [6]:
             topk=10
          1
          2
          3
             def trainmodel(model, sim):
          4
                 print ('start training')
          5
                 modelsaved=[]
          6
                 modelperform=[]
          7
                 topk=10
          8
          9
         10
                 bestresults=-1
         11
                 bestiter=-1
                 for epoch in range(1000):
         12
         13
                     model.train()
         14
         15
                     lossestrain = []
         16
                     recall=[]
         17
                     for mysentence in batchtraining data:
         18
                         model.zero_grad()
         19
         20
                         targets = mysentence[2].cuda()
         21
                         tag scores = model(mysentence[0].cuda(),mysentence[1].cuda(),wik
         22
                         loss = loss_function(tag_scores, targets)
         23
                         loss.backward()
         24
                         optimizer.step()
         25
                         lossestrain.append(loss.data.mean())
         26
                     print (epoch)
         27
                     modelsaved.append(copy.deepcopy(model.state dict()))
         28
                     29
                     model.eval()
         30
         31
                     recall=[]
         32
                     for inputs in batchval data:
         33
         34
                         targets = inputs[2].cuda()
         35
                         tag_scores = model(inputs[0].cuda(),inputs[1].cuda() ,wikivec.cu
         36
         37
                         loss = loss_function(tag_scores, targets)
         38
         39
                         targets=targets.data.cpu().numpy()
         40
                         tag_scores tag_scores.data.cpu().numpy()
         41
         42
         43
                         for iii in range(0,len(tag_scores)):
         44
                             temp={}
         45
                             for iiii in range(0,len(tag scores[iii])):
         46
                                 temp[iiii]=tag_scores[iii][iiii]
         47
                             temp1=[(k, temp[k]) for k in sorted(temp, key=temp.get, reve
         48
                             thistop=int(np.sum(targets[iii]))
         49
                             hit=0.0
         50
                             for ii in temp1[0:max(thistop,topk)]:
         51
                                 if targets[iii][ii[0]]==1.0:
         52
                                     hit=hit+1
                             if thistop!=0:
         53
         54
                                 recall.append(hit/thistop)
         55
         56
                     print ('validation top-',topk, np.mean(recall))
```

```
57
 58
 59
             modelperform.append(np.mean(recall))
 60
             if modelperform[-1]>bestresults:
 61
 62
                 bestresults=modelperform[-1]
 63
                 bestiter=len(modelperform)-1
 64
 65
             if (len(modelperform)-bestiter)>5:
                 print (modelperform, bestiter)
 66
 67
                 return modelsaved[bestiter]
 68
 69
    model = CNN(batchsize, len(word to ix), len(label to ix))
 70
    model.cuda()
 71
 72
    loss function = nn.BCELoss()
 73
    optimizer = optim.Adam(model.parameters())
 74
 75
    basemodel= trainmodel(model, 0)
 76
    torch.save(basemodel, 'CNN model')
 77
 78 | model = CNN(batchsize, len(word to ix), len(label to ix))
 79
    model.cuda()
 80
    model.load_state_dict(basemodel)
 81 loss_function = nn.BCELoss()
    optimizer = optim.Adam(model.parameters())
 82
 83 | KSImodel= trainmodel(model, 1)
 84
    torch.save(KSImodel, 'KSI_CNN_model')
 85
 86
    def testmodel(modelstate, sim):
 87
         model = CNN(batchsize, len(word_to_ix), len(label_to_ix))
 88
         model.cuda()
         model.load state dict(modelstate)
 89
         loss function = nn.BCELoss()
 90
 91
         model.eval()
 92
         recall=[]
 93
         lossestest = []
 94
 95
         y true=[]
 96
         y_scores=[]
 97
 98
 99
         for inputs in batchtest_data:
100
             targets = inputs[2].cuda()
101
102
103
             tag_scores = model(inputs[0].cuda(),inputs[1].cuda() ,wikivec.cuda()
104
105
             loss = loss_function(tag_scores, targets)
106
107
             targets=targets.data.cpu().numpy()
108
             tag_scores= tag_scores.data.cpu().numpy()
109
110
             lossestest.append(loss.data.mean())
111
112
             y_true.append(targets)
113
             y scores.append(tag scores)
```

```
114
115
            for iii in range(0,len(tag scores)):
116
                temp={}
                for iiii in range(0,len(tag scores[iii])):
117
                    temp[iiii]=tag scores[iii][iiii]
118
                temp1=[(k, temp[k]) for k in sorted(temp, key=temp.get, reverse=
119
120
                thistop=int(np.sum(targets[iii]))
                hit=0.0
121
122
123
                for ii in temp1[0:max(thistop,topk)]:
124
                    if targets[iii][ii[0]]==1.0:
125
                        hit=hit+1
126
                if thistop!=0:
127
                    recall.append(hit/thistop)
128
        y_true=np.concatenate(y_true,axis=0)
129
        y scores=np.concatenate(y scores,axis=0)
130
        y true=y true.T
131
        y_scores=y_scores.T
132
        temptrue=[]
133
        tempscores=[]
134
        for col in range(0,len(y_true)):
135
            if np.sum(y true[col])!=0:
136
                temptrue.append(y true[col])
137
                tempscores.append(y_scores[col])
138
        temptrue=np.array(temptrue)
139
        tempscores=np.array(tempscores)
140
        y true=temptrue.T
141
        y scores=tempscores.T
142
        y pred=(y scores>0.5).astype(np.int)
143
        print ('test loss', torch.stack(lossestest).mean().item())
144
        print ('top-',topk, np.mean(recall))
145
        print ('macro AUC', roc_auc_score(y_true, y_scores,average='macro'))
        print ('micro AUC', roc_auc_score(y_true, y_scores,average='micro'))
146
147
        print ('macro F1', f1_score(y_true, y_pred, average='macro')
148
        print ('micro F1', f1 score(y true, y pred, average='micro') )
149
150 print ('CNN alone:
                                 ')
151 testmodel(basemodel, 0)
153 print ('KSI+CNN:
                               ')
154 testmodel(KSImodel, 1)
```

validation top- 10 0.6467904998800238 validation top- 10 0.6673111386677142 validation top- 10 0.6812441074304616 validation top- 10 0.6933596956867344 validation top- 10 0.7052186240553053 validation top- 10 0.7117489006787834 validation top- 10 0.722211327818717 validation top- 10 0.7265657968833666 validation top- 10 0.7334972301355923 13 validation top- 10 0.7345698864388871 14 validation top- 10 0.7347385604593721 validation top- 10 0.7408095989262027 16 validation top- 10 0.7420631058486372 validation top- 10 0.7462089367848072 validation top- 10 0.7501318022145766 19 validation top- 10 0.7528999172460481 20 validation top- 10 0.7534285951564181 21 validation top- 10 0.7498198864222271 22 validation top- 10 0.7532788638686233 23

```
validation top- 10 0.7540027411409296
validation top- 10 0.7573006911292467
validation top- 10 0.7541092517584836
validation top- 10 0.7552991371761425
27
validation top- 10 0.7589011476791745
28
validation top- 10 0.7577853193291497
validation top- 10 0.7583405388902464
validation top- 10 0.7560631161889517
validation top- 10 0.7577583138492285
32
validation top- 10 0.7598690505225836
33
validation top- 10 0.7560855487936969
validation top- 10 0.756507376786185
35
validation top- 10 0.7564485341974435
validation top- 10 0.7546294812010665
validation top- 10 0.7564334654936617
[0.4530655205021126, 0.529073778538287, 0.5752237865553222, 0.6170548555043338,
0.6467904998800238, 0.6673111386677142, 0.6812441074304616, 0.6933596956867344,
0.7052186240553053, 0.7117489006787834, 0.722211327818717, 0.7265657968833666,
0.7334972301355923, 0.7345698864388871, 0.7347385604593721, 0.7408095989262027,
0.7420631058486372, 0.7462089367848072, 0.7501318022145766, 0.7528999172460481,
0.7534285951564181, 0.7498198864222271, 0.7532788638686233, 0.7540027411409296,
0.7573006911292467, 0.7541092517584836, 0.7552991371761425, 0.7589011476791745,
0.7577853193291497, 0.7583405388902464, 0.7560631161889517, 0.7577583138492285,
0.7598690505225836, 0.7560855487936969, 0.756507376786185, 0.7564485341974435,
0.7546294812010665, 0.7564334654936617] 32
start_training
```

```
validation top- 10 0.7669781847454861
1
validation top- 10 0.7734568574207574
validation top- 10 0.7768123484218621
validation top- 10 0.7797745185621022
validation top- 10 0.780834126290448
validation top- 10 0.7800837381359998
validation top- 10 0.7795087269621294
7
validation top- 10 0.7794720793975843
validation top- 10 0.7781662562211387
validation top- 10 0.7783857329042567
[0.7669781847454861, 0.7734568574207574, 0.7768123484218621, 0.779774518562102
2, 0.780834126290448, 0.7800837381359998, 0.7795087269621294, 0.779472079397584
3, 0.7781662562211387, 0.7783857329042567] 4
CNN alone:
test loss 0.03949494659900665
top- 10 0.753674492427492
macro AUC 0.8306164538048904
micro AUC 0.9666265882181901
macro F1 0.21344930909581109
micro F1 0.6257646033743762
KSI+CNN:
test loss 0.03750767931342125
top- 10 0.7747399300742249
macro AUC 0.8609588009174831
micro AUC 0.9723801813466743
macro F1 0.24036851193298106
micro F1 0.6399044546248596
```

## In [ ]:

1