

```
In [1]: from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense
        from tensorflow.keras.datasets import reuters
        from tensorflow.keras.layers import Dense, LSTM, Embedding, Activation
        import numpy as np
        from tensorflow.keras.datasets import imdb
```

```
In [2]: (X_train, y_train), (X_test, y_test) = imdb.load_data(num_words = None, index_
        from=3)
```

```
In [3]: print(X_train[0])
```

```
[1, 14, 22, 16, 43, 530, 973, 1622, 1385, 65, 458, 4468, 66, 3941, 4, 173, 3
6, 256, 5, 25, 100, 43, 838, 112, 50, 670, 22665, 9, 35, 480, 284, 5, 150, 4,
172, 112, 167, 21631, 336, 385, 39, 4, 172, 4536, 1111, 17, 546, 38, 13, 447,
4, 192, 50, 16, 6, 147, 2025, 19, 14, 22, 4, 1920, 4613, 469, 4, 22, 71, 87,
12, 16, 43, 530, 38, 76, 15, 13, 1247, 4, 22, 17, 515, 17, 12, 16, 626, 18, 1
9193, 5, 62, 386, 12, 8, 316, 8, 106, 5, 4, 2223, 5244, 16, 480, 66, 3785, 3
3, 4, 130, 12, 16, 38, 619, 5, 25, 124, 51, 36, 135, 48, 25, 1415, 33, 6, 22,
12, 215, 28, 77, 52, 5, 14, 407, 16, 82, 10311, 8, 4, 107, 117, 5952, 15, 25
6, 4, 31050, 7, 3766, 5, 723, 36, 71, 43, 530, 476, 26, 400, 317, 46, 7, 4, 1
2118, 1029, 13, 104, 88, 4, 381, 15, 297, 98, 32, 2071, 56, 26, 141, 6, 194,
7486, 18, 4, 226, 22, 21, 134, 476, 26, 480, 5, 144, 30, 5535, 18, 51, 36, 2
8, 224, 92, 25, 104, 4, 226, 65, 16, 38, 1334, 88, 12, 16, 283, 5, 16, 4472,
113, 103, 32, 15, 16, 5345, 19, 178, 32]
```

```
In [4]: word_index = imdb.get_word_index() # key_word,value_index dic
        reverse_index = dict([(value, key) for (key, value) in word_index.items()]) #
        key_index,value_word dic
        decoded = " ".join( [reverse_index.get(i-3, "#") for i in X_train[0]] )
        print(decoded)
```

```
# this film was just brilliant casting location scenery story direction every
one's really suited the part they played and you could just imagine being the
re robert redford's is an amazing actor and now the same being director norma
n's father came from the same scottish island as myself so i loved the fact t
here was a real connection with this film the witty remarks throughout the fi
lm were great it was just brilliant so much that i bought the film as soon as
it was released for retail and would recommend it to everyone to watch and th
e fly fishing was amazing really cried at the end it was so sad and you know
what they say if you cry at a film it must have been good and this definitely
was also congratulations to the two little boy's that played the part's of no
rman and paul they were just brilliant children are often left out of the pra
ising list i think because the stars that play them all grown up are such a b
ig profile for the whole film but these children are amazing and should be pr
aised for what they have done don't you think the whole story was so lovely b
ecause it was true and was someone's life after all that was shared with us a
ll
```

```
In [5]: print(len(X_train[0]))
```

```
218
```

```
In [6]: np.unique(y_train)
```

```
Out[6]: array([0, 1], dtype=int64)
```

```
In [7]: y_train[0]
```

```
Out[7]: 1
```

```
In [8]: num_classes = max(y_train) + 1  
print(num_classes)
```

```
2
```

```
In [9]: from tensorflow.keras.preprocessing.text import Tokenizer  
from tensorflow.keras.utils import to_categorical
```

```
In [10]: max_num = 10000  
tokenizer = Tokenizer(num_words=max_num)  
tokenizer.fit_on_sequences(X_train)  
X_train = tokenizer.sequences_to_matrix(X_train, mode = 'tfidf')  
X_test = tokenizer.sequences_to_matrix(X_test, mode = 'tfidf')
```

```
In [11]: print(X_train[0][4])
```

```
2.5855967311039936
```

```
In [12]: y_train = to_categorical(y_train, num_classes)  
y_test = to_categorical(y_test, num_classes)  
y_train.shape
```

```
Out[12]: (25000, 2)
```

```
In [13]: from sklearn.model_selection import KFold  
kf = KFold(n_splits = 3, shuffle = True)  
kf
```

```
Out[13]: KFold(n_splits=3, random_state=None, shuffle=True)
```

find out the overfitting model

```
In [15]: units = range(100,300,10)
loss_train = []
acc_train = []
loss_val = []
acc_val = []
for num in units:
    model = Sequential()
    model.add(Dense(num,input_shape = (max_num,)))
    model.add(Activation('relu'))
    model.add(Dense(num_classes))
    model.add(Activation('softmax'))
    model.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metri
cs = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3)
        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model.evaluate(X_val_cv,y_val_cv, batch_size=200, verbo
se = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val.append(loss_val_cv/3)
    acc_val.append(acc_val_cv/3)
    loss_train.append(loss_train_cv/3)
    acc_train.append(acc_train_cv/3)
```

```
Epoch 1/3
16666/16666 [=====] - 6s 348us/step - loss: 0.3507 -
acc: 0.8497
Epoch 2/3
16666/16666 [=====] - 5s 313us/step - loss: 0.0815 -
acc: 0.9764
Epoch 3/3
16666/16666 [=====] - 5s 311us/step - loss: 0.0231 -
acc: 0.9962
8334/8334 [=====] - 2s 197us/step
Epoch 1/3
16667/16667 [=====] - 5s 312us/step - loss: 0.1807 -
acc: 0.9381
Epoch 2/3
16667/16667 [=====] - 5s 306us/step - loss: 0.0318 -
acc: 0.9924
Epoch 3/3
16667/16667 [=====] - 5s 303us/step - loss: 0.0068 -
acc: 0.9993
8333/8333 [=====] - 1s 167us/step
Epoch 1/3
16667/16667 [=====] - 5s 305us/step - loss: 0.0291 -
acc: 0.9914
Epoch 2/3
16667/16667 [=====] - 5s 310us/step - loss: 0.0068 -
acc: 0.9994
Epoch 3/3
16667/16667 [=====] - 5s 311us/step - loss: 0.0018 -
acc: 1.0000
8333/8333 [=====] - 2s 191us/step
Epoch 1/3
16666/16666 [=====] - 6s 375us/step - loss: 0.3582 -
acc: 0.8471
Epoch 2/3
16666/16666 [=====] - 5s 327us/step - loss: 0.0808 -
acc: 0.9758
Epoch 3/3
16666/16666 [=====] - 5s 320us/step - loss: 0.0230 -
acc: 0.9966 1s - loss: 0.0
8334/8334 [=====] - 2s 185us/step
Epoch 1/3
16667/16667 [=====] - 5s 322us/step - loss: 0.1744 -
acc: 0.9376
Epoch 2/3
16667/16667 [=====] - 5s 326us/step - loss: 0.0292 -
acc: 0.9941
Epoch 3/3
16667/16667 [=====] - 5s 326us/step - loss: 0.0081 -
acc: 0.9993
8333/8333 [=====] - 2s 195us/step
Epoch 1/3
16667/16667 [=====] - 5s 328us/step - loss: 0.0243 -
acc: 0.9942
Epoch 2/3
16667/16667 [=====] - 5s 324us/step - loss: 0.0066 -
acc: 0.9995
Epoch 3/3
```

```
16667/16667 [=====] - 5s 323us/step - loss: 0.0035 -  
acc: 0.9999  
8333/8333 [=====] - 1s 171us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 376us/step - loss: 0.3464 -  
acc: 0.8511  
Epoch 2/3  
16666/16666 [=====] - 6s 335us/step - loss: 0.0717 -  
acc: 0.9793  
Epoch 3/3  
16666/16666 [=====] - 6s 337us/step - loss: 0.0189 -  
acc: 0.9968  
8334/8334 [=====] - 2s 231us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 331us/step - loss: 0.1824 -  
acc: 0.9396  
Epoch 2/3  
16667/16667 [=====] - 6s 337us/step - loss: 0.0323 -  
acc: 0.9924  
Epoch 3/3  
16667/16667 [=====] - 6s 334us/step - loss: 0.0071 -  
acc: 0.9994  
8333/8333 [=====] - 2s 215us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 337us/step - loss: 0.0240 -  
acc: 0.9935  
Epoch 2/3  
16667/16667 [=====] - 6s 337us/step - loss: 0.0043 -  
acc: 0.9999  
Epoch 3/3  
16667/16667 [=====] - 5s 330us/step - loss: 0.0015 -  
acc: 1.0000  
8333/8333 [=====] - 1s 171us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 401us/step - loss: 0.3568 -  
acc: 0.8478  
Epoch 2/3  
16666/16666 [=====] - 6s 346us/step - loss: 0.0784 -  
acc: 0.9767  
Epoch 3/3  
16666/16666 [=====] - 6s 346us/step - loss: 0.0211 -  
acc: 0.9965  
8334/8334 [=====] - 2s 207us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 350us/step - loss: 0.1641 -  
acc: 0.9431  
Epoch 2/3  
16667/16667 [=====] - 6s 346us/step - loss: 0.0249 -  
acc: 0.9950  
Epoch 3/3  
16667/16667 [=====] - 6s 346us/step - loss: 0.0066 -  
acc: 0.9997  
8333/8333 [=====] - 2s 194us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 351us/step - loss: 0.0221 -  
acc: 0.9935  
Epoch 2/3
```

```
16667/16667 [=====] - 6s 348us/step - loss: 0.0041 -  
acc: 0.9999  
Epoch 3/3  
16667/16667 [=====] - 6s 335us/step - loss: 0.0013 -  
acc: 1.0000  
8333/8333 [=====] - 2s 180us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 410us/step - loss: 0.3449 -  
acc: 0.8558  
Epoch 2/3  
16666/16666 [=====] - 6s 367us/step - loss: 0.0700 -  
acc: 0.9797  
Epoch 3/3  
16666/16666 [=====] - 6s 361us/step - loss: 0.0191 -  
acc: 0.9971  
8334/8334 [=====] - 2s 189us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 363us/step - loss: 0.1831 -  
acc: 0.9359  
Epoch 2/3  
16667/16667 [=====] - 6s 362us/step - loss: 0.0282 -  
acc: 0.9932  
Epoch 3/3  
16667/16667 [=====] - 6s 360us/step - loss: 0.0061 -  
acc: 0.9993  
8333/8333 [=====] - 1s 172us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 365us/step - loss: 0.0232 -  
acc: 0.9926  
Epoch 2/3  
16667/16667 [=====] - 6s 365us/step - loss: 0.0044 -  
acc: 0.9999  
Epoch 3/3  
16667/16667 [=====] - 6s 365us/step - loss: 0.0013 -  
acc: 1.0000  
8333/8333 [=====] - 1s 178us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 431us/step - loss: 0.3515 -  
acc: 0.8508  
Epoch 2/3  
16666/16666 [=====] - 6s 374us/step - loss: 0.0702 -  
acc: 0.9798  
Epoch 3/3  
16666/16666 [=====] - 6s 376us/step - loss: 0.0171 -  
acc: 0.9976  
8334/8334 [=====] - 2s 249us/step  
Epoch 1/3  
16667/16667 [=====] - 6s 378us/step - loss: 0.1710 -  
acc: 0.9398  
Epoch 2/3  
16667/16667 [=====] - 6s 372us/step - loss: 0.0269 -  
acc: 0.9941  
Epoch 3/3  
16667/16667 [=====] - 6s 376us/step - loss: 0.0053 -  
acc: 0.9996  
8333/8333 [=====] - 2s 187us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 6s 384us/step - loss: 0.0214 -  
acc: 0.9936  
Epoch 2/3  
16667/16667 [=====] - 7s 391us/step - loss: 0.0044 -  
acc: 0.9998  
Epoch 3/3  
16667/16667 [=====] - 6s 379us/step - loss: 0.0013 -  
acc: 1.0000  
8333/8333 [=====] - 1s 179us/step  
Epoch 1/3  
16666/16666 [=====] - 8s 450us/step - loss: 0.3481 -  
acc: 0.8507  
Epoch 2/3  
16666/16666 [=====] - 7s 393us/step - loss: 0.0658 -  
acc: 0.9822  
Epoch 3/3  
16666/16666 [=====] - 6s 389us/step - loss: 0.0162 -  
acc: 0.9979  
8334/8334 [=====] - 2s 237us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 394us/step - loss: 0.1803 -  
acc: 0.9387  
Epoch 2/3  
16667/16667 [=====] - 7s 391us/step - loss: 0.0298 -  
acc: 0.9933  
Epoch 3/3  
16667/16667 [=====] - 6s 388us/step - loss: 0.0070 -  
acc: 0.9995  
8333/8333 [=====] - 2s 180us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0217 -  
acc: 0.9947  
Epoch 2/3  
16667/16667 [=====] - 6s 386us/step - loss: 0.0049 -  
acc: 0.9999  
Epoch 3/3  
16667/16667 [=====] - 6s 380us/step - loss: 0.0023 -  
acc: 0.9999  
8333/8333 [=====] - 2s 230us/step  
Epoch 1/3  
16666/16666 [=====] - 8s 461us/step - loss: 0.3559 -  
acc: 0.8495  
Epoch 2/3  
16666/16666 [=====] - 7s 405us/step - loss: 0.0674 -  
acc: 0.9803  
Epoch 3/3  
16666/16666 [=====] - 7s 399us/step - loss: 0.0162 -  
acc: 0.9977  
8334/8334 [=====] - 2s 239us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 400us/step - loss: 0.1739 -  
acc: 0.9401  
Epoch 2/3  
16667/16667 [=====] - 7s 405us/step - loss: 0.0252 -  
acc: 0.9944  
Epoch 3/3  
16667/16667 [=====] - 7s 401us/step - loss: 0.0054 -
```

```
acc: 0.9998
8333/8333 [=====] - 1s 180us/step
Epoch 1/3
16667/16667 [=====] - 7s 407us/step - loss: 0.0233 -
acc: 0.9939
Epoch 2/3
16667/16667 [=====] - 7s 405us/step - loss: 0.0055 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 7s 401us/step - loss: 0.0013 -
acc: 1.0000
8333/8333 [=====] - 1s 177us/step
Epoch 1/3
16666/16666 [=====] - 8s 492us/step - loss: 0.3523 -
acc: 0.8539 0s - loss: 0.3579 - acc:
Epoch 2/3
16666/16666 [=====] - 7s 415us/step - loss: 0.0618 -
acc: 0.9835
Epoch 3/3
16666/16666 [=====] - 7s 413us/step - loss: 0.0140 -
acc: 0.9983
8334/8334 [=====] - 2s 244us/step
Epoch 1/3
16667/16667 [=====] - 7s 420us/step - loss: 0.1693 -
acc: 0.9417
Epoch 2/3
16667/16667 [=====] - 7s 421us/step - loss: 0.0241 -
acc: 0.9951
Epoch 3/3
16667/16667 [=====] - 7s 419us/step - loss: 0.0050 -
acc: 0.9995
8333/8333 [=====] - 2s 199us/step
Epoch 1/3
16667/16667 [=====] - 7s 420us/step - loss: 0.0183 -
acc: 0.9954
Epoch 2/3
16667/16667 [=====] - 7s 420us/step - loss: 0.0040 -
acc: 0.9999
Epoch 3/3
16667/16667 [=====] - 7s 424us/step - loss: 0.0012 -
acc: 0.9999
8333/8333 [=====] - 2s 185us/step
Epoch 1/3
16666/16666 [=====] - 8s 500us/step - loss: 0.3536 -
acc: 0.8477
Epoch 2/3
16666/16666 [=====] - 7s 436us/step - loss: 0.0664 -
acc: 0.9811
Epoch 3/3
16666/16666 [=====] - 7s 434us/step - loss: 0.0167 -
acc: 0.9975
8334/8334 [=====] - 2s 249us/step
Epoch 1/3
16667/16667 [=====] - 7s 434us/step - loss: 0.1722 -
acc: 0.9413
Epoch 2/3
16667/16667 [=====] - 7s 435us/step - loss: 0.0246 -
```



```
acc: 0.9943
Epoch 3/3
16667/16667 [=====] - 7s 436us/step - loss: 0.0060 -
acc: 0.9994
8333/8333 [=====] - 2s 188us/step
Epoch 1/3
16667/16667 [=====] - 7s 435us/step - loss: 0.0163 -
acc: 0.9962 2s - 1
Epoch 2/3
16667/16667 [=====] - 7s 431us/step - loss: 0.0030 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 7s 432us/step - loss: 0.0012 -
acc: 0.9999
8333/8333 [=====] - 2s 193us/step
Epoch 1/3
16666/16666 [=====] - 9s 515us/step - loss: 0.3518 -
acc: 0.8531
Epoch 2/3
16666/16666 [=====] - 7s 443us/step - loss: 0.0624 -
acc: 0.9822
Epoch 3/3
16666/16666 [=====] - 7s 444us/step - loss: 0.0149 -
acc: 0.9977
8334/8334 [=====] - 2s 266us/step
Epoch 1/3
16667/16667 [=====] - 7s 438us/step - loss: 0.1750 -
acc: 0.9394
Epoch 2/3
16667/16667 [=====] - 7s 443us/step - loss: 0.0226 -
acc: 0.9957
Epoch 3/3
16667/16667 [=====] - 7s 443us/step - loss: 0.0046 -
acc: 0.9996
8333/8333 [=====] - 2s 196us/step
Epoch 1/3
16667/16667 [=====] - 8s 450us/step - loss: 0.0198 -
acc: 0.9945
Epoch 2/3
16667/16667 [=====] - 8s 452us/step - loss: 0.0033 -
acc: 0.9999 0s - loss: 0.0034 -
Epoch 3/3
16667/16667 [=====] - 7s 443us/step - loss: 9.6005e-
04 - acc: 1.0000
8333/8333 [=====] - 2s 202us/step
Epoch 1/3
16666/16666 [=====] - 9s 546us/step - loss: 0.3499 -
acc: 0.8516
Epoch 2/3
16666/16666 [=====] - 8s 455us/step - loss: 0.0636 -
acc: 0.9821
Epoch 3/3
16666/16666 [=====] - 8s 457us/step - loss: 0.0140 -
acc: 0.9984
8334/8334 [=====] - 2s 231us/step
Epoch 1/3
16667/16667 [=====] - 8s 455us/step - loss: 0.1757 -
```

```
acc: 0.9401 4s - - ETA: 2s
Epoch 2/3
16667/16667 [=====] - 8s 459us/step - loss: 0.0237 -
acc: 0.9953 1s - loss: 0.0247 - a - ETA: 0s - loss: 0.0247 - acc
Epoch 3/3
16667/16667 [=====] - 8s 458us/step - loss: 0.0051 -
acc: 0.9996
8333/8333 [=====] - 2s 193us/step
Epoch 1/3
16667/16667 [=====] - 8s 461us/step - loss: 0.0194 -
acc: 0.9944 0s - loss: 0.0197 - acc:
Epoch 2/3
16667/16667 [=====] - 8s 458us/step - loss: 0.0047 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 8s 452us/step - loss: 0.0020 -
acc: 0.9999
8333/8333 [=====] - 2s 200us/step
Epoch 1/3
16666/16666 [=====] - 9s 557us/step - loss: 0.3528 -
acc: 0.8481
Epoch 2/3
16666/16666 [=====] - 8s 472us/step - loss: 0.0625 -
acc: 0.9827
Epoch 3/3
16666/16666 [=====] - 8s 478us/step - loss: 0.0147 -
acc: 0.9981
8334/8334 [=====] - 2s 239us/step
Epoch 1/3
16667/16667 [=====] - 8s 471us/step - loss: 0.1722 -
acc: 0.9393
Epoch 2/3
16667/16667 [=====] - 8s 471us/step - loss: 0.0225 -
acc: 0.9951
Epoch 3/3
16667/16667 [=====] - 8s 472us/step - loss: 0.0043 -
acc: 0.9997
8333/8333 [=====] - 2s 224us/step
Epoch 1/3
16667/16667 [=====] - 8s 469us/step - loss: 0.0200 -
acc: 0.9943
Epoch 2/3
16667/16667 [=====] - 8s 476us/step - loss: 0.0029 -
acc: 1.0000
Epoch 3/3
16667/16667 [=====] - 8s 479us/step - loss: 9.7745e-
04 - acc: 1.0000
8333/8333 [=====] - 2s 192us/step
Epoch 1/3
16666/16666 [=====] - 9s 567us/step - loss: 0.3629 -
acc: 0.8415
Epoch 2/3
16666/16666 [=====] - 8s 490us/step - loss: 0.0611 -
acc: 0.9840
Epoch 3/3
16666/16666 [=====] - 8s 489us/step - loss: 0.0144 -
acc: 0.9983
```

```
8334/8334 [=====] - 2s 239us/step
Epoch 1/3
16667/16667 [=====] - 8s 489us/step - loss: 0.1723 -
acc: 0.9405
Epoch 2/3
16667/16667 [=====] - 8s 497us/step - loss: 0.0212 -
acc: 0.9954
Epoch 3/3
16667/16667 [=====] - 8s 486us/step - loss: 0.0044 -
acc: 0.9999
8333/8333 [=====] - 2s 202us/step
Epoch 1/3
16667/16667 [=====] - 8s 487us/step - loss: 0.0198 -
acc: 0.9952
Epoch 2/3
16667/16667 [=====] - 8s 492us/step - loss: 0.0038 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 8s 485us/step - loss: 0.0012 -
acc: 0.9999
8333/8333 [=====] - 2s 200us/step
Epoch 1/3
16666/16666 [=====] - 10s 581us/step - loss: 0.3549
- acc: 0.8484
Epoch 2/3
16666/16666 [=====] - 8s 502us/step - loss: 0.0604 -
acc: 0.9838
Epoch 3/3
16666/16666 [=====] - 8s 494us/step - loss: 0.0137 -
acc: 0.9983
8334/8334 [=====] - 2s 260us/step
Epoch 1/3
16667/16667 [=====] - 8s 502us/step - loss: 0.1755 -
acc: 0.9399
Epoch 2/3
16667/16667 [=====] - 8s 499us/step - loss: 0.0243 -
acc: 0.9953
Epoch 3/3
16667/16667 [=====] - 8s 501us/step - loss: 0.0053 -
acc: 0.9995
8333/8333 [=====] - 2s 201us/step
Epoch 1/3
16667/16667 [=====] - 8s 498us/step - loss: 0.0192 -
acc: 0.9953
Epoch 2/3
16667/16667 [=====] - 8s 505us/step - loss: 0.0044 -
acc: 0.9999
Epoch 3/3
16667/16667 [=====] - 8s 505us/step - loss: 0.0020 -
acc: 0.9999
8333/8333 [=====] - 2s 210us/step
Epoch 1/3
16666/16666 [=====] - 10s 600us/step - loss: 0.3489
- acc: 0.8510
Epoch 2/3
16666/16666 [=====] - 9s 513us/step - loss: 0.0567 -
acc: 0.9848
```

```
Epoch 3/3
16666/16666 [=====] - 9s 511us/step - loss: 0.0125 -
acc: 0.9984
8334/8334 [=====] - 2s 257us/step
Epoch 1/3
16667/16667 [=====] - 9s 512us/step - loss: 0.1816 -
acc: 0.9365
Epoch 2/3
16667/16667 [=====] - 9s 511us/step - loss: 0.0247 -
acc: 0.9956
Epoch 3/3
16667/16667 [=====] - 9s 530us/step - loss: 0.0054 -
acc: 0.9998
8333/8333 [=====] - 2s 213us/step
Epoch 1/3
16667/16667 [=====] - 9s 521us/step - loss: 0.0194 -
acc: 0.9952
Epoch 2/3
16667/16667 [=====] - 9s 523us/step - loss: 0.0038 -
acc: 0.9999
Epoch 3/3
16667/16667 [=====] - 9s 526us/step - loss: 0.0019 -
acc: 0.9999
8333/8333 [=====] - 2s 198us/step
Epoch 1/3
16666/16666 [=====] - ETA: 0s - loss: 0.3548 - acc:
0.848 - 11s 636us/step - loss: 0.3543 - acc: 0.8486
Epoch 2/3
16666/16666 [=====] - 9s 545us/step - loss: 0.0561 -
acc: 0.9840
Epoch 3/3
16666/16666 [=====] - 9s 561us/step - loss: 0.0121 -
acc: 0.9987
8334/8334 [=====] - 2s 260us/step
Epoch 1/3
16667/16667 [=====] - 9s 546us/step - loss: 0.1709 -
acc: 0.9416
Epoch 2/3
16667/16667 [=====] - 9s 548us/step - loss: 0.0223 -
acc: 0.9954
Epoch 3/3
16667/16667 [=====] - 9s 544us/step - loss: 0.0049 -
acc: 0.9997
8333/8333 [=====] - 2s 246us/step
Epoch 1/3
16667/16667 [=====] - 9s 564us/step - loss: 0.0199 -
acc: 0.9946
Epoch 2/3
16667/16667 [=====] - 9s 546us/step - loss: 0.0046 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 9s 561us/step - loss: 0.0019 -
acc: 0.9999
8333/8333 [=====] - 2s 212us/step
Epoch 1/3
16666/16666 [=====] - 11s 655us/step - loss: 0.3657
- acc: 0.8369
```

```
Epoch 2/3
16666/16666 [=====] - 9s 558us/step - loss: 0.0616 -
acc: 0.9836
Epoch 3/3
16666/16666 [=====] - 9s 560us/step - loss: 0.0145 -
acc: 0.9985
8334/8334 [=====] - 2s 286us/step
Epoch 1/3
16667/16667 [=====] - 9s 555us/step - loss: 0.1766 -
acc: 0.9380
Epoch 2/3
16667/16667 [=====] - 9s 556us/step - loss: 0.0222 -
acc: 0.9956
Epoch 3/3
16667/16667 [=====] - 9s 551us/step - loss: 0.0044 -
acc: 0.9998
8333/8333 [=====] - 2s 209us/step
Epoch 1/3
16667/16667 [=====] - 9s 552us/step - loss: 0.0182 -
acc: 0.9949
Epoch 2/3
16667/16667 [=====] - 9s 552us/step - loss: 0.0040 -
acc: 0.9999
Epoch 3/3
16667/16667 [=====] - 9s 552us/step - loss: 0.0010 -
acc: 1.0000
8333/8333 [=====] - 2s 224us/step
Epoch 1/3
16666/16666 [=====] - 11s 665us/step - loss: 0.3492
- acc: 0.8510
Epoch 2/3
16666/16666 [=====] - 10s 570us/step - loss: 0.0551
- acc: 0.9851
Epoch 3/3
16666/16666 [=====] - 10s 571us/step - loss: 0.0112
- acc: 0.9987
8334/8334 [=====] - 2s 266us/step
Epoch 1/3
16667/16667 [=====] - 9s 561us/step - loss: 0.1771 -
acc: 0.9399
Epoch 2/3
16667/16667 [=====] - 10s 571us/step - loss: 0.0243
- acc: 0.9953
Epoch 3/3
16667/16667 [=====] - 10s 575us/step - loss: 0.0049
- acc: 0.9996
8333/8333 [=====] - 2s 215us/step
Epoch 1/3
16667/16667 [=====] - 9s 564us/step - loss: 0.0173 -
acc: 0.9956
Epoch 2/3
16667/16667 [=====] - 9s 567us/step - loss: 0.0024 -
acc: 1.0000
Epoch 3/3
16667/16667 [=====] - 9s 562us/step - loss: 8.4070e-
04 - acc: 1.0000
8333/8333 [=====] - 2s 212us/step
```

```

Epoch 1/3
16666/16666 [=====] - 11s 686us/step - loss: 0.3447
- acc: 0.8502
Epoch 2/3
16666/16666 [=====] - 10s 582us/step - loss: 0.0500
- acc: 0.9859
Epoch 3/3
16666/16666 [=====] - 10s 583us/step - loss: 0.0109
- acc: 0.9989
8334/8334 [=====] - 2s 270us/step
Epoch 1/3
16667/16667 [=====] - 10s 577us/step - loss: 0.1847
- acc: 0.9339
Epoch 2/3
16667/16667 [=====] - 10s 582us/step - loss: 0.0222
- acc: 0.9953
Epoch 3/3
16667/16667 [=====] - 10s 578us/step - loss: 0.0040
- acc: 0.9998
8333/8333 [=====] - 2s 219us/step
Epoch 1/3
16667/16667 [=====] - 10s 574us/step - loss: 0.0183
- acc: 0.9957
Epoch 2/3
16667/16667 [=====] - 9s 569us/step - loss: 0.0023 -
acc: 1.0000
Epoch 3/3
16667/16667 [=====] - 10s 577us/step - loss: 8.6907e
-04 - acc: 1.0000
8333/8333 [=====] - 2s 217us/step

```

```

In [14]: %matplotlib inline
import matplotlib.pyplot as plt
plt.style.use('seaborn-whitegrid')
import pandas as pd

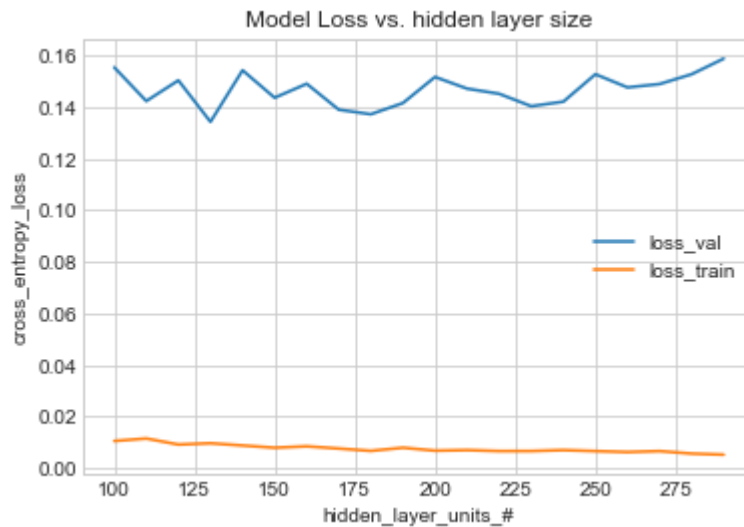
```

```

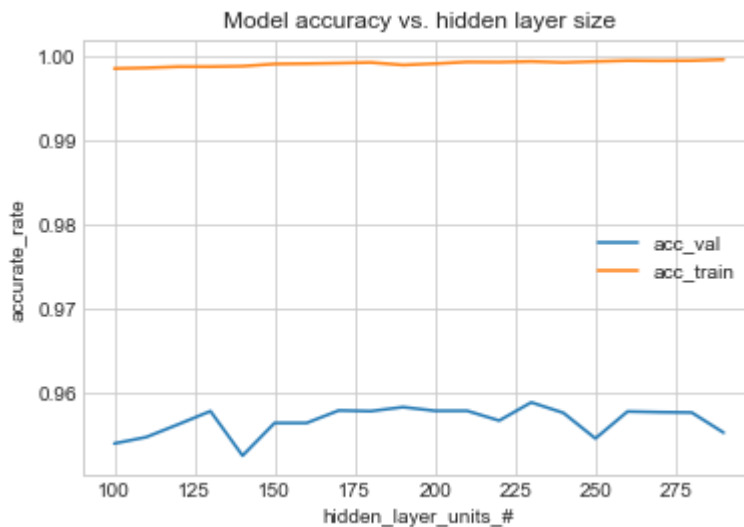
In [17]: ix = units
loss_train= pd.Series(loss_train, index = ix)
loss_val = pd.Series(loss_val, index = ix)
acc_train = pd.Series(acc_train, index = ix)
acc_val = pd.Series(acc_val, index = ix)

```

```
In [27]: fig = plt.figure()
ax = plt.axes()
ax.plot(loss_val, label = 'loss_val')
ax.plot(loss_train, label = 'loss_train')
ax.legend()
plt.xlabel('hidden_layer_units_#')
plt.ylabel('cross_entropy_loss')
plt.title('Model Loss vs. hidden layer size')
fig.savefig("loss_cv.png",dpi = 400)
```



```
In [28]: fig = plt.figure()
ax = plt.axes()
ax.plot(acc_val, label = 'acc_val')
ax.plot(acc_train, label = 'acc_train')
ax.legend()
plt.xlabel('hidden_layer_units_#')
plt.ylabel('accurate_rate')
plt.title('Model accuracy vs. hidden layer size')
fig.savefig("acc_cv.png",dpi=400)
```



```
In [ ]: # overfitting, while hidden layer contains 250 nodes in this case
# first kind of regularization: dropout
# second kind of regularization: l1_norm_kernel_regularizer
# third kind of regularization: l2_norm_kernel_regularizer
```

```
In [48]: # the baseline to compare with
model_overfitting = Sequential()
model_overfitting.add(Dense(250,input_shape = (max_num,)))
model_overfitting.add(Activation('relu'))
model_overfitting.add(Dense(num_classes))
model_overfitting.add(Activation('softmax'))
model_overfitting.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_overfitting = model_overfitting.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_overfitting = model_overfitting.evaluate(X_test,y_test, batch_size=200, verbose = 1)

base_train_acc = hist_overfitting.history.get('acc')[-1]
base_test_acc = score_overfitting[1]
```

```
Epoch 1/3
25000/25000 [=====] - 13s 536us/step - loss: 0.3210
- acc: 0.8646
Epoch 2/3
25000/25000 [=====] - 11s 435us/step - loss: 0.0665
- acc: 0.9803
Epoch 3/3
25000/25000 [=====] - 10s 417us/step - loss: 0.0141
- acc: 0.9978
25000/25000 [=====] - 6s 229us/step
```

```
In [49]: print('The accurate rate of the overfitting model on training dataset is')
print(base_train_acc)
print('The accurate rate of the overfitting model on test dataset is')
print(base_test_acc)
```

```
The accurate rate of the overfitting model on training dataset is
0.9978400020599365
The accurate rate of the overfitting model on test dataset is
0.8704400014877319
```

```
In [29]: # Save the weights
model_overfitting.save_weights('model_base_weights.h5')

# Save the model architecture
with open('model_base_architecture.json', 'w') as f:
    f.write(model_overfitting.to_json())
```



```
In [ ]: from keras.models import model_from_json

# Model reconstruction from JSON file
with open('model_base_architecture.json', 'r') as f:
    model_overfitting = model_from_json(f.read())

# Load weights into the new model
model_overfitting.load_weights('model_base_weights.h5')
```

```
In [ ]:
```

```
In [44]: from tensorflow.keras.layers import Dropout
rate = [0.1,0.2,0.3,0.4,0.5,0.6,0.7,0.8,0.9]
loss_train_drop = []
acc_train_drop = []
loss_val_drop = []
acc_val_drop = []
for r in rate:
    model_drop = Sequential()
    model_drop.add(Dense(250,input_shape = (max_num,)))
    model_drop.add(Activation('relu'))
    model_drop.add(Dropout(r))
    model_drop.add(Dense(num_classes))
    model_drop.add(Activation('softmax'))
    model_drop.compile(loss = 'categorical_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_drop.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3
    )
        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_drop.evaluate(X_val_cv,y_val_cv, batch_size=200,
verbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_drop.append(loss_val_cv/3)
    acc_val_drop.append(acc_val_cv/3)
    loss_train_drop.append(loss_train_cv/3)
    acc_train_drop.append(acc_train_cv/3)
```

```
Epoch 1/3
16666/16666 [=====] - 10s 575us/step - loss: 0.3509
- acc: 0.8487
Epoch 2/3
16666/16666 [=====] - 7s 439us/step - loss: 0.0665 -
acc: 0.9796
Epoch 3/3
16666/16666 [=====] - 7s 417us/step - loss: 0.0151 -
acc: 0.9976
8334/8334 [=====] - 2s 235us/step
Epoch 1/3
16667/16667 [=====] - 7s 420us/step - loss: 0.1679 -
acc: 0.9443
Epoch 2/3
16667/16667 [=====] - 7s 417us/step - loss: 0.0240 -
acc: 0.9945
Epoch 3/3
16667/16667 [=====] - 7s 413us/step - loss: 0.0043 -
acc: 0.9999
8333/8333 [=====] - 1s 124us/step
Epoch 1/3
16667/16667 [=====] - 7s 412us/step - loss: 0.0190 -
acc: 0.9951
Epoch 2/3
16667/16667 [=====] - 7s 422us/step - loss: 0.0036 -
acc: 0.9998
Epoch 3/3
16667/16667 [=====] - 7s 418us/step - loss: 0.0016 -
acc: 0.9999
8333/8333 [=====] - 1s 123us/step
Epoch 1/3
16666/16666 [=====] - 9s 554us/step - loss: 0.3457 -
acc: 0.8537
Epoch 2/3
16666/16666 [=====] - 7s 412us/step - loss: 0.0728 -
acc: 0.9778
Epoch 3/3
16666/16666 [=====] - 7s 432us/step - loss: 0.0186 -
acc: 0.9969 2
8334/8334 [=====] - 2s 256us/step
Epoch 1/3
16667/16667 [=====] - 7s 434us/step - loss: 0.1829 -
acc: 0.9414
Epoch 2/3
16667/16667 [=====] - 7s 444us/step - loss: 0.0335 -
acc: 0.9917
Epoch 3/3
16667/16667 [=====] - 7s 439us/step - loss: 0.0084 -
acc: 0.9990
8333/8333 [=====] - 1s 138us/step
Epoch 1/3
16667/16667 [=====] - 7s 421us/step - loss: 0.0229 -
acc: 0.9932
Epoch 2/3
16667/16667 [=====] - 7s 414us/step - loss: 0.0064 -
acc: 0.9998
Epoch 3/3
```

```
16667/16667 [=====] - 7s 411us/step - loss: 0.0032 -  
acc: 0.9998  
8333/8333 [=====] - 1s 132us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 554us/step - loss: 0.3603 -  
acc: 0.8462  
Epoch 2/3  
16666/16666 [=====] - 7s 411us/step - loss: 0.0846 -  
acc: 0.9735  
Epoch 3/3  
16666/16666 [=====] - 7s 413us/step - loss: 0.0265 -  
acc: 0.9945  
8334/8334 [=====] - 2s 226us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.1752 -  
acc: 0.9429  
Epoch 2/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.0341 -  
acc: 0.9919  
Epoch 3/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.0082 -  
acc: 0.9991  
8333/8333 [=====] - 1s 127us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 417us/step - loss: 0.0210 -  
acc: 0.9939  
Epoch 2/3  
16667/16667 [=====] - 7s 418us/step - loss: 0.0067 -  
acc: 0.9995  
Epoch 3/3  
16667/16667 [=====] - 7s 420us/step - loss: 0.0031 -  
acc: 0.9996  
8333/8333 [=====] - 1s 141us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 555us/step - loss: 0.3671 -  
acc: 0.8420  
Epoch 2/3  
16666/16666 [=====] - 7s 412us/step - loss: 0.0941 -  
acc: 0.9708  
Epoch 3/3  
16666/16666 [=====] - 7s 409us/step - loss: 0.0308 -  
acc: 0.9935  
8334/8334 [=====] - 2s 226us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.1784 -  
acc: 0.9383  
Epoch 2/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.0409 -  
acc: 0.9898  
Epoch 3/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0141 -  
acc: 0.9978  
8333/8333 [=====] - 1s 123us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.0287 -  
acc: 0.9920  
Epoch 2/3
```

```
16667/16667 [=====] - 7s 413us/step - loss: 0.0091 -  
acc: 0.9989  
Epoch 3/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0055 -  
acc: 0.9996  
8333/8333 [=====] - 1s 133us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 558us/step - loss: 0.3780 -  
acc: 0.8363  
Epoch 2/3  
16666/16666 [=====] - 7s 432us/step - loss: 0.1128 -  
acc: 0.9608  
Epoch 3/3  
16666/16666 [=====] - 7s 432us/step - loss: 0.0445 -  
acc: 0.9884  
8334/8334 [=====] - 2s 246us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.1829 -  
acc: 0.9386  
Epoch 2/3  
16667/16667 [=====] - 7s 415us/step - loss: 0.0575 -  
acc: 0.9849  
Epoch 3/3  
16667/16667 [=====] - 7s 414us/step - loss: 0.0210 -  
acc: 0.9962  
8333/8333 [=====] - 1s 129us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 421us/step - loss: 0.0391 -  
acc: 0.9882  
Epoch 2/3  
16667/16667 [=====] - 7s 431us/step - loss: 0.0153 -  
acc: 0.9977  
Epoch 3/3  
16667/16667 [=====] - 7s 427us/step - loss: 0.0076 -  
acc: 0.9986  
8333/8333 [=====] - 1s 150us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 562us/step - loss: 0.3803 -  
acc: 0.8366  
Epoch 2/3  
16666/16666 [=====] - 7s 410us/step - loss: 0.1395 -  
acc: 0.9504  
Epoch 3/3  
16666/16666 [=====] - 7s 413us/step - loss: 0.0698 -  
acc: 0.9794  
8334/8334 [=====] - 2s 233us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 409us/step - loss: 0.1915 -  
acc: 0.9342  
Epoch 2/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.0792 -  
acc: 0.9764  
Epoch 3/3  
16667/16667 [=====] - 7s 417us/step - loss: 0.0386 -  
acc: 0.9897  
8333/8333 [=====] - 1s 131us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 7s 411us/step - loss: 0.0574 -  
acc: 0.9815  
Epoch 2/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0248 -  
acc: 0.9935  
Epoch 3/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0161 -  
acc: 0.9960  
8333/8333 [=====] - 1s 140us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 562us/step - loss: 0.4089 -  
acc: 0.8185  
Epoch 2/3  
16666/16666 [=====] - 7s 409us/step - loss: 0.1725 -  
acc: 0.9378  
Epoch 3/3  
16666/16666 [=====] - 7s 408us/step - loss: 0.0985 -  
acc: 0.9662  
8334/8334 [=====] - 2s 229us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.2037 -  
acc: 0.9276 0s - loss: 0.2014 - ac  
Epoch 2/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.1088 -  
acc: 0.9640  
Epoch 3/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.0641 -  
acc: 0.9798  
8333/8333 [=====] - 1s 122us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 412us/step - loss: 0.0792 -  
acc: 0.9730  
Epoch 2/3  
16667/16667 [=====] - 7s 411us/step - loss: 0.0532 -  
acc: 0.9847  
Epoch 3/3  
16667/16667 [=====] - 7s 413us/step - loss: 0.0350 -  
acc: 0.9908  
8333/8333 [=====] - 1s 135us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 568us/step - loss: 0.4576 -  
acc: 0.7972  
Epoch 2/3  
16666/16666 [=====] - 7s 410us/step - loss: 0.2261 -  
acc: 0.9171  
Epoch 3/3  
16666/16666 [=====] - 7s 421us/step - loss: 0.1576 -  
acc: 0.9432  
8334/8334 [=====] - 2s 251us/step  
Epoch 1/3  
16667/16667 [=====] - 7s 423us/step - loss: 0.2300 -  
acc: 0.9158  
Epoch 2/3  
16667/16667 [=====] - 7s 423us/step - loss: 0.1624 -  
acc: 0.9420  
Epoch 3/3  
16667/16667 [=====] - 7s 420us/step - loss: 0.1271 -
```

```

acc: 0.9572
8333/8333 [=====] - 1s 138us/step
Epoch 1/3
16667/16667 [=====] - 7s 409us/step - loss: 0.1395 -
acc: 0.9500
Epoch 2/3
16667/16667 [=====] - 7s 410us/step - loss: 0.1138 -
acc: 0.9620
Epoch 3/3
16667/16667 [=====] - 7s 410us/step - loss: 0.0836 -
acc: 0.9724
8333/8333 [=====] - 1s 123us/step
Epoch 1/3
16666/16666 [=====] - 10s 577us/step - loss: 0.6025
- acc: 0.7201
Epoch 2/3
16666/16666 [=====] - 7s 420us/step - loss: 0.3487 -
acc: 0.8590
Epoch 3/3
16666/16666 [=====] - 7s 418us/step - loss: 0.2781 -
acc: 0.8943
8334/8334 [=====] - 2s 255us/step
Epoch 1/3
16667/16667 [=====] - 7s 417us/step - loss: 0.3046 -
acc: 0.8816
Epoch 2/3
16667/16667 [=====] - 8s 464us/step - loss: 0.2533 -
acc: 0.9060
Epoch 3/3
16667/16667 [=====] - 7s 430us/step - loss: 0.2226 -
acc: 0.9167
8333/8333 [=====] - 1s 139us/step
Epoch 1/3
16667/16667 [=====] - 7s 407us/step - loss: 0.2397 -
acc: 0.9089
Epoch 2/3
16667/16667 [=====] - 7s 413us/step - loss: 0.2120 -
acc: 0.9220
Epoch 3/3
16667/16667 [=====] - 7s 441us/step - loss: 0.1995 -
acc: 0.9277
8333/8333 [=====] - 1s 147us/step

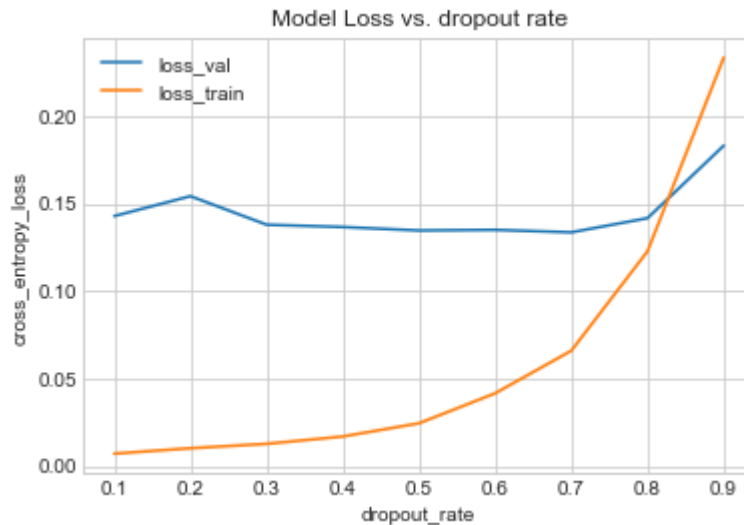
```

```

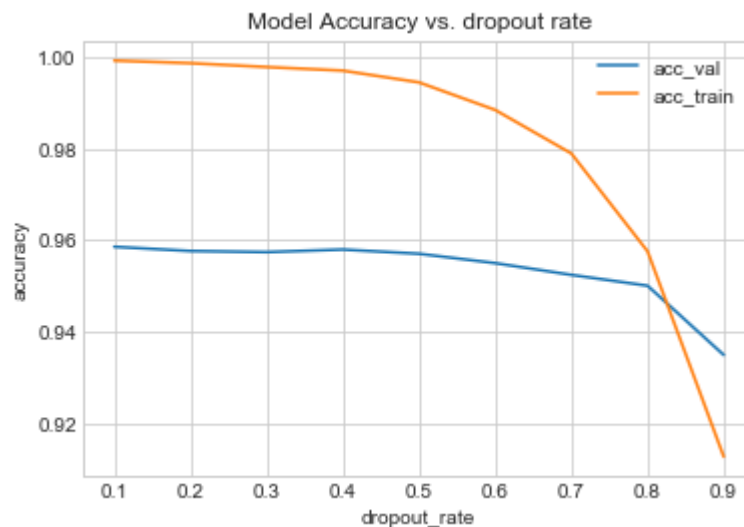
In [45]: ix = rate
loss_train_drop= pd.Series(loss_train_drop, index = ix)
loss_val_drop = pd.Series(loss_val_drop, index = ix)
acc_train_drop = pd.Series(acc_train_drop, index = ix)
acc_val_drop = pd.Series(acc_val_drop, index = ix)

```

```
In [46]: fig = plt.figure()
ax = plt.axes()
ax.plot(loss_val_drop, label = 'loss_val')
ax.plot(loss_train_drop, label = 'loss_train')
ax.legend()
plt.xlabel('dropout_rate')
plt.ylabel('cross_entropy_loss')
plt.title('Model Loss vs. dropout rate')
fig.savefig("drop_loss_cv.png",dpi = 400)
```



```
In [47]: fig = plt.figure()
ax = plt.axes()
ax.plot(acc_val_drop, label = 'acc_val')
ax.plot(acc_train_drop, label = 'acc_train')
ax.legend()
plt.xlabel('dropout_rate')
plt.ylabel('accuracy')
plt.title('Model Accuracy vs. dropout rate')
fig.savefig("drop_acc_cv.png",dpi = 400)
```



In []:


```
In [26]: from tensorflow.keras.layers import Dropout
model_drop = Sequential()
model_drop.add(Dense(250,input_shape = (max_num,)))
model_drop.add(Dropout(0.5))
model_drop.add(Activation('relu'))
model_drop.add(Dense(num_classes))
model_drop.add(Activation('softmax'))
model_drop.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_drop = model_drop.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_drop = model_drop.evaluate(X_test,y_test, batch_size=200, verbose = 1)

drop_train_acc = hist_drop.history.get('acc')[-1]
drop_test_acc = score_drop[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 568us/step - loss: 0.3484
- acc: 0.8506
Epoch 2/3
25000/25000 [=====] - 13s 505us/step - loss: 0.1211
- acc: 0.9577
Epoch 3/3
25000/25000 [=====] - 13s 507us/step - loss: 0.0536
- acc: 0.9844
25000/25000 [=====] - 6s 250us/step
```

```
In [27]: print('The accurate rate of the model with dropout(0.5) on training dataset is')
print(drop_train_acc)
print('The accurate rate of the model with dropout(0.5) on test dataset is')
print(drop_test_acc)
```

```
The accurate rate of the model with dropout(0.5) on training dataset is
0.9844000105857849
The accurate rate of the model with dropout(0.5) on test dataset is
0.8729600014686585
```

```
In [92]: from tensorflow.keras.layers import Dropout
model_125 = Sequential()
model_125.add(Dense(125,input_shape = (max_num,)))
model_125.add(Activation('relu'))
model_125.add(Dense(num_classes))
model_125.add(Activation('softmax'))
model_125.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_125 = model_125.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_125 = model_125.evaluate(X_test,y_test, batch_size=200, verbose = 1)
nn125_train_acc = hist_125.history.get('acc')[-1]
nn125_test_acc = score_125[1]
```

```
Epoch 1/3
25000/25000 [=====] - 7s 283us/step - loss: 0.3292 - acc: 0.8593
Epoch 2/3
25000/25000 [=====] - 6s 246us/step - loss: 0.0769 - acc: 0.9771
Epoch 3/3
25000/25000 [=====] - 6s 244us/step - loss: 0.0193 - acc: 0.9965
25000/25000 [=====] - 3s 105us/step
```

```
In [93]: print('The accurate rate of the model with 125 hidden units on training dataset is')
print(nn125_train_acc)
print('The accurate rate of the model with 125 hidden units on test dataset is')
print(nn125_test_acc)
```

```
The accurate rate of the model with 125 hidden units on training dataset is
0.9965200033187867
The accurate rate of the model with 125 hidden units on test dataset is
0.8692400031089783
```

```
In [28]: from tensorflow.keras.layers import Dropout
model_drop = Sequential()
model_drop.add(Dense(250,input_shape = (max_num,)))
model_drop.add(Dropout(0.4))
model_drop.add(Activation('relu'))
model_drop.add(Dense(num_classes))
model_drop.add(Activation('softmax'))
model_drop.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_drop = model_drop.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_drop = model_drop.evaluate(X_test,y_test, batch_size=200, verbose = 1)

drop_train_acc = hist_drop.history.get('acc')[-1]
drop_test_acc = score_drop[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 567us/step - loss: 0.3492
- acc: 0.8533
Epoch 2/3
25000/25000 [=====] - 13s 520us/step - loss: 0.1024
- acc: 0.9656
Epoch 3/3
25000/25000 [=====] - 13s 516us/step - loss: 0.0349
- acc: 0.9922
25000/25000 [=====] - 5s 213us/step
```

```
In [29]: print('The accurate rate of the model with dropout(0.4) on training dataset is')
print(drop_train_acc)
print('The accurate rate of the model with dropout(0.4) on test dataset is')
print(drop_test_acc)
```

```
The accurate rate of the model with dropout(0.4) on training dataset is
0.9922400074005127
The accurate rate of the model with dropout(0.4) on test dataset is
0.8701600012779236
```

```
In [30]: from tensorflow.keras.layers import Dropout
model_drop = Sequential()
model_drop.add(Dense(250,input_shape = (max_num,)))
model_drop.add(Dropout(0.6))
model_drop.add(Activation('relu'))
model_drop.add(Dense(num_classes))
model_drop.add(Activation('softmax'))
model_drop.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_drop = model_drop.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_drop = model_drop.evaluate(X_test,y_test, batch_size=200, verbose = 1)

drop_train_acc = hist_drop.history.get('acc')[-1]
drop_test_acc = score_drop[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 568us/step - loss: 0.3573
- acc: 0.8468
Epoch 2/3
25000/25000 [=====] - 13s 503us/step - loss: 0.1421
- acc: 0.9507
Epoch 3/3
25000/25000 [=====] - 13s 507us/step - loss: 0.0742
- acc: 0.9769
25000/25000 [=====] - 6s 223us/step
```

```
In [31]: print('The accurate rate of the model with dropout(0.6) on training dataset is')
print(drop_train_acc)
print('The accurate rate of the model with dropout(0.6) on test dataset is')
print(drop_test_acc)
```

```
The accurate rate of the model with dropout(0.6) on training dataset is
0.9768800086975098
The accurate rate of the model with dropout(0.6) on test dataset is
0.8719600014686585
```

In []:

In []:

In []:

In []:

```
In [66]: from tensorflow.keras.regularizers import l1
from tensorflow.keras.regularizers import l2
```

```

In [72]: lambda_list = [1e-6,1e-5,1e-4,1e-3,1e-2,1e-1]
loss_train_l1 = []
acc_train_l1 = []
loss_val_l1 = []
acc_val_l1 = []
for par in lambda_list:
    model_l1 = Sequential()
    model_l1.add(Dense(250,input_shape = (max_num,), kernel_regularizer = l1(p
ar), bias_regularizer=l1(par)))
    model_l1.add(Activation('relu'))
    model_l1.add(Dense(num_classes, kernel_regularizer = l1(par), bias_regular
izer=l1(par)))
    model_l1.add(Activation('softmax'))
    model_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_l1.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3)
        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_l1.evaluate(X_val_cv,y_val_cv, batch_size=200, ve
rbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_l1.append(loss_val_cv/3)
    acc_val_l1.append(acc_val_cv/3)
    loss_train_l1.append(loss_train_cv/3)
    acc_train_l1.append(acc_train_cv/3)

```

```
Epoch 1/3
16666/16666 [=====] - 9s 526us/step - loss: 0.3740 -
acc: 0.8564
Epoch 2/3
16666/16666 [=====] - 8s 476us/step - loss: 0.0917 -
acc: 0.9828
Epoch 3/3
16666/16666 [=====] - 8s 478us/step - loss: 0.0435 -
acc: 0.9976
8334/8334 [=====] - 1s 168us/step
Epoch 1/3
16667/16667 [=====] - 8s 480us/step - loss: 0.2051 -
acc: 0.9390 2s - 1
Epoch 2/3
16667/16667 [=====] - 8s 477us/step - loss: 0.0574 -
acc: 0.9950
Epoch 3/3
16667/16667 [=====] - 8s 481us/step - loss: 0.0378 -
acc: 0.9996
8333/8333 [=====] - 1s 161us/step
Epoch 1/3
16667/16667 [=====] - 8s 476us/step - loss: 0.0578 -
acc: 0.9931
Epoch 2/3
16667/16667 [=====] - 8s 477us/step - loss: 0.0388 -
acc: 0.9997
Epoch 3/3
16667/16667 [=====] - 8s 480us/step - loss: 0.0341 -
acc: 1.0000
8333/8333 [=====] - 1s 156us/step
Epoch 1/3
16666/16666 [=====] - 9s 519us/step - loss: 0.5924 -
acc: 0.8473
Epoch 2/3
16666/16666 [=====] - 8s 475us/step - loss: 0.2856 -
acc: 0.9798
Epoch 3/3
16666/16666 [=====] - 8s 475us/step - loss: 0.2049 -
acc: 0.9967
8334/8334 [=====] - 1s 168us/step - ETA
Epoch 1/3
16667/16667 [=====] - 8s 474us/step - loss: 0.3510 -
acc: 0.9398
Epoch 2/3
16667/16667 [=====] - 8s 478us/step - loss: 0.2218 -
acc: 0.9926
Epoch 3/3
16667/16667 [=====] - 8s 481us/step - loss: 0.1700 -
acc: 0.9995
8333/8333 [=====] - 1s 156us/step
Epoch 1/3
16667/16667 [=====] - 8s 475us/step - loss: 0.2129 -
acc: 0.9810
Epoch 2/3
16667/16667 [=====] - 8s 478us/step - loss: 0.1758 -
acc: 0.9969
Epoch 3/3
```

```
16667/16667 [=====] - 8s 475us/step - loss: 0.1414 -  
acc: 1.0000  
8333/8333 [=====] - 1s 157us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 528us/step - loss: 1.8369 -  
acc: 0.8526  
Epoch 2/3  
16666/16666 [=====] - 8s 487us/step - loss: 0.7665 -  
acc: 0.9507  
Epoch 3/3  
16666/16666 [=====] - 8s 486us/step - loss: 0.5812 -  
acc: 0.9566  
8334/8334 [=====] - 1s 169us/step  
Epoch 1/3  
16667/16667 [=====] - 8s 497us/step - loss: 0.7104 -  
acc: 0.9189  
Epoch 2/3  
16667/16667 [=====] - 8s 476us/step - loss: 0.5210 -  
acc: 0.9674  
Epoch 3/3  
16667/16667 [=====] - 8s 480us/step - loss: 0.3922 -  
acc: 0.9801  
8333/8333 [=====] - 1s 154us/step  
Epoch 1/3  
16667/16667 [=====] - 8s 475us/step - loss: 0.5506 -  
acc: 0.9392  
Epoch 2/3  
16667/16667 [=====] - 8s 479us/step - loss: 0.4241 -  
acc: 0.9803 4s - 1 - ETA: 1s - loss:  
Epoch 3/3  
16667/16667 [=====] - 8s 472us/step - loss: 0.3144 -  
acc: 0.9908  
8333/8333 [=====] - 1s 152us/step  
Epoch 1/3  
16666/16666 [=====] - 9s 534us/step - loss: 6.3297 -  
acc: 0.8357  
Epoch 2/3  
16666/16666 [=====] - 8s 503us/step - loss: 1.3332 -  
acc: 0.8669  
Epoch 3/3  
16666/16666 [=====] - 8s 494us/step - loss: 1.2293 -  
acc: 0.8711  
8334/8334 [=====] - 1s 177us/step  
Epoch 1/3  
16667/16667 [=====] - 8s 495us/step - loss: 1.1430 -  
acc: 0.8764  
Epoch 2/3  
16667/16667 [=====] - 8s 502us/step - loss: 1.0341 -  
acc: 0.8869  
Epoch 3/3  
16667/16667 [=====] - 8s 493us/step - loss: 0.9667 -  
acc: 0.8915  
8333/8333 [=====] - 1s 163us/step  
Epoch 1/3  
16667/16667 [=====] - 8s 482us/step - loss: 0.9496 -  
acc: 0.8910  
Epoch 2/3
```

```
16667/16667 [=====] - 9s 538us/step - loss: 0.8667 -  
acc: 0.9035  
Epoch 3/3  
16667/16667 [=====] - 10s 577us/step - loss: 0.8105  
- acc: 0.9111  
8333/8333 [=====] - 2s 217us/step  
Epoch 1/3  
16666/16666 [=====] - 10s 600us/step - loss: 42.1275  
- acc: 0.7672  
Epoch 2/3  
16666/16666 [=====] - 9s 557us/step - loss: 3.8045 -  
acc: 0.8068  
Epoch 3/3  
16666/16666 [=====] - 9s 552us/step - loss: 3.5522 -  
acc: 0.8283  
8334/8334 [=====] - 2s 233us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 532us/step - loss: 3.5191 -  
acc: 0.8345  
Epoch 2/3  
16667/16667 [=====] - 9s 547us/step - loss: 3.5023 -  
acc: 0.8403  
Epoch 3/3  
16667/16667 [=====] - 9s 551us/step - loss: 3.4889 -  
acc: 0.8462  
8333/8333 [=====] - 2s 203us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 533us/step - loss: 3.4878 -  
acc: 0.8511  
Epoch 2/3  
16667/16667 [=====] - 9s 561us/step - loss: 3.4870 -  
acc: 0.8526  
Epoch 3/3  
16667/16667 [=====] - 9s 567us/step - loss: 3.4820 -  
acc: 0.8585  
8333/8333 [=====] - 2s 220us/step  
Epoch 1/3  
16666/16666 [=====] - 10s 625us/step - loss: 412.166  
0 - acc: 0.5414  
Epoch 2/3  
16666/16666 [=====] - 10s 583us/step - loss: 31.7424  
- acc: 0.4999  
Epoch 3/3  
16666/16666 [=====] - 9s 560us/step - loss: 30.8289  
- acc: 0.4989  
8334/8334 [=====] - 2s 239us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 543us/step - loss: 30.8347  
- acc: 0.4955  
Epoch 2/3  
16667/16667 [=====] - 10s 579us/step - loss: 30.8026  
- acc: 0.5010  
Epoch 3/3  
16667/16667 [=====] - 11s 673us/step - loss: 30.8211  
- acc: 0.5049  
8333/8333 [=====] - 2s 218us/step  
Epoch 1/3
```



```

16667/16667 [=====] - 9s 545us/step - loss: 30.7826
- acc: 0.5021
Epoch 2/3
16667/16667 [=====] - 9s 559us/step - loss: 30.8032
- acc: 0.5048
Epoch 3/3
16667/16667 [=====] - 9s 560us/step - loss: 30.7899
- acc: 0.5039
8333/8333 [=====] - 2s 219us/step

```

```

In [73]: print(loss_train_l1)
         print(loss_val_l1)

```

```

[0.03845995333584134, 0.17209522215436823, 0.42926743624753066, 1.00215903385
95625, 3.507723528102316, 30.813288159613112]
[0.17665963721892683, 0.297276239302909, 0.5640312226375032, 1.02416921316438
85, 3.496250938156813, 30.906951873829343]

```

```

In [74]: print(acc_train_l1)
         print(acc_val_l1)

```

```

[0.9990799712760805, 0.9987399603998445, 0.9758396216811569, 0.89123959722478
43, 0.8443396800387178, 0.5025999266247423]
[0.956642814160677, 0.9464422868833272, 0.9096412133018205, 0.878440065659191
9, 0.8461607332347469, 0.504719904790355]

```

```

In [95]: l1_reg_cv = pd.DataFrame({'loss_train': loss_train_l1,
                                   'loss_val': loss_val_l1,
                                   'acc_train': acc_train_l1,
                                   'acc_val': acc_val_l1,
                                   'lambda': lambda_list}).set_index('lambda')

l1_reg_cv

```

Out[95]:

	loss_train	loss_val	acc_train	acc_val
lambda				
0.000001	0.038460	0.176660	0.99908	0.956643
0.000010	0.172095	0.297276	0.99874	0.946442
0.000100	0.429267	0.564031	0.97584	0.909641
0.001000	1.002159	1.024169	0.89124	0.878440
0.010000	3.507724	3.496251	0.84434	0.846161
0.100000	30.813288	30.906952	0.50260	0.504720

```
In [87]: # indicating by the result of l1_reg_cv data frame, will set the lambda as 1e-6 for l1 norm regularization
model_l1 = Sequential()
model_l1.add(Dense(250, input_shape = (max_num,), kernel_regularizer = l1(1e-6), bias_regularizer=l1(1e-6)))
model_l1.add(Activation('relu'))
model_l1.add(Dense(num_classes, kernel_regularizer = l1(1e-6), bias_regularizer=l1(1e-6)))
model_l1.add(Activation('softmax'))
model_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_l1 = model_l1.fit(X_train, y_train, batch_size=200, epochs = 3, verbose = 1)
score_l1 = model_l1.evaluate(X_test, y_test, batch_size=200, verbose = 1)

l1_train_acc = hist_l1.history.get('acc')[-1]
l1_test_acc = score_l1[1]
```

```
Epoch 1/3
25000/25000 [=====] - 13s 522us/step - loss: 0.3630
- acc: 0.85690s - loss: 0.3652 - acc: 0
Epoch 2/3
25000/25000 [=====] - 12s 471us/step - loss: 0.0964
- acc: 0.9801
Epoch 3/3
25000/25000 [=====] - 12s 469us/step - loss: 0.0456
- acc: 0.9980
25000/25000 [=====] - 4s 166us/step
```

```
In [88]: print('The accurate rate of the model with l1 regularization on training datasets is')
print(l1_train_acc)
print('The accurate rate of the model with l1 regularization on test datasets is')
print(l1_test_acc)
```

```
The accurate rate of the model with l1 regularization on training dataset is
0.9979600019454956
The accurate rate of the model with l1 regularization on test dataset is
0.8711199998855591
```

```
In [108]: # indicating by the result of l1_reg_cv data frame, will set the lambda as 1e-5 for l1 norm regularization
model_l1 = Sequential()
model_l1.add(Dense(250, input_shape = (max_num,), kernel_regularizer = l1(1e-5), bias_regularizer=l1(1e-5)))
model_l1.add(Activation('relu'))
model_l1.add(Dense(num_classes, kernel_regularizer = l1(1e-5), bias_regularizer=l1(1e-5)))
model_l1.add(Activation('softmax'))
model_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_l1 = model_l1.fit(X_train, y_train, batch_size=200, epochs = 3, verbose = 1)
score_l1 = model_l1.evaluate(X_test, y_test, batch_size=200, verbose = 1)

l1_train_acc = hist_l1.history.get('acc')[-1]
l1_test_acc = score_l1[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 563us/step - loss: 0.5563
- acc: 0.8622
Epoch 2/3
25000/25000 [=====] - 12s 476us/step - loss: 0.2818
- acc: 0.9758
Epoch 3/3
25000/25000 [=====] - 12s 478us/step - loss: 0.1938
- acc: 0.9958
25000/25000 [=====] - 4s 175us/step
```

```
In [109]: print('The accurate rate of the model with l1 regularization on training dataset is')
print(l1_train_acc)
print('The accurate rate of the model with l1 regularization on test dataset is')
print(l1_test_acc)
```

```
The accurate rate of the model with l1 regularization on training dataset is
0.9958400039672851
The accurate rate of the model with l1 regularization on test dataset is
0.8743999996185303
```

```
In [ ]:
```

```
In [100]: lambda_list = [1e-6,1e-5,1e-4,1e-3,1e-2,1e-1]
loss_train_l2 = []
acc_train_l2 = []
loss_val_l2 = []
acc_val_l2 = []
for par in lambda_list:
    model_l2 = Sequential()
    model_l2.add(Dense(250,input_shape = (max_num,), kernel_regularizer = l2(p
ar), bias_regularizer=l2(par)))
    model_l2.add(Activation('relu'))
    model_l2.add(Dense(num_classes, kernel_regularizer = l2(par), bias_regular
izer=l2(par)))
    model_l2.add(Activation('softmax'))
    model_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_l2.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3)
        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_l2.evaluate(X_val_cv,y_val_cv, batch_size=200, ve
rbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_l2.append(loss_val_cv/3)
    acc_val_l2.append(acc_val_cv/3)
    loss_train_l2.append(loss_train_cv/3)
    acc_train_l2.append(acc_train_cv/3)
```

```
Epoch 1/3
16666/16666 [=====] - 10s 583us/step - loss: 0.3717
- acc: 0.8422
Epoch 2/3
16666/16666 [=====] - 8s 473us/step - loss: 0.0603 -
acc: 0.9833
Epoch 3/3
16666/16666 [=====] - 8s 473us/step - loss: 0.0151 -
acc: 0.9984
8334/8334 [=====] - 2s 208us/step
Epoch 1/3
16667/16667 [=====] - 8s 487us/step - loss: 0.1704 -
acc: 0.9417
Epoch 2/3
16667/16667 [=====] - 8s 496us/step - loss: 0.0239 -
acc: 0.9950
Epoch 3/3
16667/16667 [=====] - 8s 486us/step - loss: 0.0062 -
acc: 0.9997
8333/8333 [=====] - 1s 158us/step
Epoch 1/3
16667/16667 [=====] - 8s 474us/step - loss: 0.0190 -
acc: 0.9953
Epoch 2/3
16667/16667 [=====] - 9s 529us/step - loss: 0.0043 -
acc: 0.9999
Epoch 3/3
16667/16667 [=====] - 9s 570us/step - loss: 0.0022 -
acc: 1.0000
8333/8333 [=====] - 2s 187us/step
Epoch 1/3
16666/16666 [=====] - 12s 712us/step - loss: 0.3552
- acc: 0.8497
Epoch 2/3
16666/16666 [=====] - 13s 802us/step - loss: 0.0652
- acc: 0.98342s - loss: 0.
Epoch 3/3
16666/16666 [=====] - 14s 848us/step - loss: 0.0200
- acc: 0.9981
8334/8334 [=====] - 3s 330us/step
Epoch 1/3
16667/16667 [=====] - 9s 562us/step - loss: 0.1804 -
acc: 0.9425
Epoch 2/3
16667/16667 [=====] - 9s 557us/step - loss: 0.0324 -
acc: 0.9948
Epoch 3/3
16667/16667 [=====] - 10s 616us/step - loss: 0.0139
- acc: 0.9998
8333/8333 [=====] - 2s 275us/step
Epoch 1/3
16667/16667 [=====] - 9s 565us/step - loss: 0.0300 -
acc: 0.9944
Epoch 2/3
16667/16667 [=====] - 10s 614us/step - loss: 0.0152
- acc: 0.9996
Epoch 3/3
```

```
16667/16667 [=====] - 10s 583us/step - loss: 0.0114
- acc: 0.9999
8333/8333 [=====] - 2s 221us/step
Epoch 1/3
16666/16666 [=====] - 11s 664us/step - loss: 0.3909
- acc: 0.8548
Epoch 2/3
16666/16666 [=====] - 9s 556us/step - loss: 0.1128 -
acc: 0.9804
Epoch 3/3
16666/16666 [=====] - 9s 561us/step - loss: 0.0632 -
acc: 0.9982
8334/8334 [=====] - 2s 269us/step
Epoch 1/3
16667/16667 [=====] - 9s 542us/step - loss: 0.2255 -
acc: 0.9409
Epoch 2/3
16667/16667 [=====] - 9s 560us/step - loss: 0.0857 -
acc: 0.9935
Epoch 3/3
16667/16667 [=====] - 9s 555us/step - loss: 0.0590 -
acc: 0.9996
8333/8333 [=====] - 2s 214us/step
Epoch 1/3
16667/16667 [=====] - 9s 538us/step - loss: 0.0807 -
acc: 0.9914
Epoch 2/3
16667/16667 [=====] - 9s 569us/step - loss: 0.0585 -
acc: 0.9994
Epoch 3/3
16667/16667 [=====] - 10s 578us/step - loss: 0.0497
- acc: 0.9999
8333/8333 [=====] - 2s 228us/step
Epoch 1/3
16666/16666 [=====] - 11s 679us/step - loss: 0.6667
- acc: 0.8456
Epoch 2/3
16666/16666 [=====] - 9s 563us/step - loss: 0.2985 -
acc: 0.9766
Epoch 3/3
16666/16666 [=====] - 9s 555us/step - loss: 0.1959 -
acc: 0.9917
8334/8334 [=====] - 2s 266us/step
Epoch 1/3
16667/16667 [=====] - 9s 545us/step - loss: 0.3584 -
acc: 0.9272
Epoch 2/3
16667/16667 [=====] - 9s 561us/step - loss: 0.2332 -
acc: 0.9839
Epoch 3/3
16667/16667 [=====] - 9s 557us/step - loss: 0.1580 -
acc: 0.9972
8333/8333 [=====] - 2s 215us/step
Epoch 1/3
16667/16667 [=====] - 9s 537us/step - loss: 0.2632 -
acc: 0.9570
Epoch 2/3
```

```
16667/16667 [=====] - 9s 570us/step - loss: 0.2040 -  
acc: 0.9899  
Epoch 3/3  
16667/16667 [=====] - 10s 572us/step - loss: 0.1464  
- acc: 0.9978  
8333/8333 [=====] - 2s 209us/step  
Epoch 1/3  
16666/16666 [=====] - 11s 661us/step - loss: 1.5827  
- acc: 0.8511  
Epoch 2/3  
16666/16666 [=====] - 9s 563us/step - loss: 0.5100 -  
acc: 0.9179  
Epoch 3/3  
16666/16666 [=====] - 9s 563us/step - loss: 0.4485 -  
acc: 0.9183  
8334/8334 [=====] - 2s 270us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 551us/step - loss: 0.4926 -  
acc: 0.9015  
Epoch 2/3  
16667/16667 [=====] - 9s 561us/step - loss: 0.3963 -  
acc: 0.9225  
Epoch 3/3  
16667/16667 [=====] - 9s 552us/step - loss: 0.3761 -  
acc: 0.9270  
8333/8333 [=====] - 2s 212us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 541us/step - loss: 0.4299 -  
acc: 0.9061  
Epoch 2/3  
16667/16667 [=====] - 9s 563us/step - loss: 0.3382 -  
acc: 0.9375  
Epoch 3/3  
16667/16667 [=====] - 10s 573us/step - loss: 0.3179  
- acc: 0.9410  
8333/8333 [=====] - 2s 228us/step  
Epoch 1/3  
16666/16666 [=====] - 11s 673us/step - loss: 5.3259  
- acc: 0.8403  
Epoch 2/3  
16666/16666 [=====] - 10s 573us/step - loss: 0.7179  
- acc: 0.8597  
Epoch 3/3  
16666/16666 [=====] - 9s 551us/step - loss: 0.6690 -  
acc: 0.8648  
8334/8334 [=====] - 2s 269us/step  
Epoch 1/3  
16667/16667 [=====] - 9s 539us/step - loss: 0.6765 -  
acc: 0.8616  
Epoch 2/3  
16667/16667 [=====] - 9s 567us/step - loss: 0.6584 -  
acc: 0.8635  
Epoch 3/3  
16667/16667 [=====] - 10s 584us/step - loss: 0.6530  
- acc: 0.8642  
8333/8333 [=====] - 2s 235us/step  
Epoch 1/3
```

```

16667/16667 [=====] - 10s 576us/step - loss: 0.6417
- acc: 0.8679
Epoch 2/3
16667/16667 [=====] - 10s 573us/step - loss: 0.6301
- acc: 0.8741
Epoch 3/3
16667/16667 [=====] - 9s 558us/step - loss: 0.6249 -
acc: 0.8689
8333/8333 [=====] - 2s 211us/step

```

```

In [101]: print(loss_train_l2)
          print(loss_val_l2)

```

```

[0.007824130377315422, 0.015088598974607422, 0.057281762232496654, 0.16676480
56307641, 0.3808560894824569, 0.6489676033762257]
[0.1390206992715748, 0.15104453086864839, 0.19869745330250457, 0.330773005932
91253, 0.48299898957397297, 0.6537691378268679]

```

```

In [102]: print(acc_train_l2)
          print(acc_val_l2)

```

```

[0.9993599810094662, 0.9992599771022933, 0.9992399799237882, 0.99559992659116
89, 0.9287797917134272, 0.8659399778324643]
[0.959482759669288, 0.9571628738318507, 0.9539227791629714, 0.921921695183919
2, 0.8826001151991272, 0.8591601228896826]

```

```

In [103]: l2_reg_cv = pd.DataFrame({'loss_train': loss_train_l2,
                                   'loss_val': loss_val_l2,
                                   'acc_train': acc_train_l2,
                                   'acc_val': acc_val_l2,
                                   'lambda': lambda_list}).set_index('lambda')

l2_reg_cv

```

Out[103]:

	loss_train	loss_val	acc_train	acc_val
lambda				
0.000001	0.007824	0.139021	0.99936	0.959483
0.000010	0.015089	0.151045	0.99926	0.957163
0.000100	0.057282	0.198697	0.99924	0.953923
0.001000	0.166765	0.330773	0.99560	0.921922
0.010000	0.380856	0.482999	0.92878	0.882600
0.100000	0.648968	0.653769	0.86594	0.859160


```
In [104]: model_l2 = Sequential()
model_l2.add(Dense(250,input_shape = (max_num,), kernel_regularizer = l2(1e-6), bias_regularizer=l2(1e-6)))
model_l2.add(Activation('relu'))
model_l2.add(Dense(num_classes, kernel_regularizer = l2(1e-6), bias_regularizer=l2(1e-6)))
model_l2.add(Activation('softmax'))
model_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_l2 = model_l2.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
score_l2 = model_l2.evaluate(X_test,y_test, batch_size=200, verbose = 1)

l2_train_acc = hist_l2.history.get('acc')[-1]
l2_test_acc = score_l2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 569us/step - loss: 0.3231
- acc: 0.8651
Epoch 2/3
25000/25000 [=====] - 12s 485us/step - loss: 0.0702
- acc: 0.9788
Epoch 3/3
25000/25000 [=====] - 13s 511us/step - loss: 0.0148
- acc: 0.9978
25000/25000 [=====] - 8s 323us/step
```

```
In [105]: print('The accurate rate of the model with l2 regularization on training dataset is')
print(l2_train_acc)
print('The accurate rate of the model with l2 regularization on test dataset is')
print(l2_test_acc)
```

```
The accurate rate of the model with l2 regularization on training dataset is
0.9977600021362305
The accurate rate of the model with l2 regularization on test dataset is
0.8718800010681153
```

```
In [106]: model_l2 = Sequential()
model_l2.add(Dense(250, input_shape = (max_num,), kernel_regularizer = l2(1e-5), bias_regularizer=l2(1e-5)))
model_l2.add(Activation('relu'))
model_l2.add(Dense(num_classes, kernel_regularizer = l2(1e-5), bias_regularizer=l2(1e-5)))
model_l2.add(Activation('softmax'))
model_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_l2 = model_l2.fit(X_train, y_train, batch_size=200, epochs = 3, verbose = 1)
score_l2 = model_l2.evaluate(X_test, y_test, batch_size=200, verbose = 1)

l2_train_acc = hist_l2.history.get('acc')[-1]
l2_test_acc = score_l2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 14s 555us/step - loss: 0.3362
- acc: 0.8628
Epoch 2/3
25000/25000 [=====] - 12s 473us/step - loss: 0.0740
- acc: 0.9798
Epoch 3/3
25000/25000 [=====] - 13s 508us/step - loss: 0.0220
- acc: 0.9979
25000/25000 [=====] - 6s 249us/step
```

```
In [107]: print('The accurate rate of the model with l2 regularization on training dataset is')
print(l2_train_acc)
print('The accurate rate of the model with l2 regularization on test dataset is')
print(l2_test_acc)
```

```
The accurate rate of the model with l2 regularization on training dataset is
0.9978800020217895
The accurate rate of the model with l2 regularization on test dataset is
0.8700399985313415
```

```
In [ ]:
```

two hidden layers

```

In [36]: units = range(10,200,10)
loss_train_2 = []
acc_train_2 = []
loss_val_2 = []
acc_val_2 = []
for num in units:
    model_2 = Sequential()
    model_2.add(Dense(num,input_shape = (max_num,)))
    model_2.add(Activation('relu'))
    model_2.add(Dense(num))
    model_2.add(Activation('relu'))
    model_2.add(Dense(num_classes))
    model_2.add(Activation('softmax'))
    model_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_2.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3)
        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_2.evaluate(X_val_cv,y_val_cv, batch_size=200, verbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_2.append(loss_val_cv/3)
    acc_val_2.append(acc_val_cv/3)
    loss_train_2.append(loss_train_cv/3)
    acc_train_2.append(acc_train_cv/3)

```

```
Epoch 1/3
16666/16666 [=====] - 6s 383us/step - loss: 0.4289 -
acc: 0.8081
Epoch 2/3
16666/16666 [=====] - 4s 255us/step - loss: 0.1823 -
acc: 0.9371
Epoch 3/3
16666/16666 [=====] - 4s 259us/step - loss: 0.1024 -
acc: 0.9699
8334/8334 [=====] - 2s 296us/step
Epoch 1/3
16667/16667 [=====] - 4s 237us/step - loss: 0.2068 -
acc: 0.9312
Epoch 2/3
16667/16667 [=====] - 4s 258us/step - loss: 0.0942 -
acc: 0.9723
Epoch 3/3
16667/16667 [=====] - 4s 236us/step - loss: 0.0511 -
acc: 0.9888
8333/8333 [=====] - 2s 200us/step
Epoch 1/3
16667/16667 [=====] - 4s 240us/step - loss: 0.1141 -
acc: 0.9618
Epoch 2/3
16667/16667 [=====] - 4s 245us/step - loss: 0.0478 -
acc: 0.9867
Epoch 3/3
16667/16667 [=====] - 4s 254us/step - loss: 0.0232 -
acc: 0.9957
8333/8333 [=====] - 2s 199us/step
Epoch 1/3
16666/16666 [=====] - 7s 397us/step - loss: 0.4025 -
acc: 0.8220
Epoch 2/3
16666/16666 [=====] - 5s 302us/step - loss: 0.1520 -
acc: 0.9473
Epoch 3/3
16666/16666 [=====] - 4s 265us/step - loss: 0.0669 -
acc: 0.9798
8334/8334 [=====] - 2s 287us/step
Epoch 1/3
16667/16667 [=====] - 5s 291us/step - loss: 0.1979 -
acc: 0.9341
Epoch 2/3
16667/16667 [=====] - 5s 281us/step - loss: 0.0547 -
acc: 0.9849
Epoch 3/3
16667/16667 [=====] - 5s 301us/step - loss: 0.0186 -
acc: 0.9960
8333/8333 [=====] - ETA: - 2s 239us/step
Epoch 1/3
16667/16667 [=====] - 4s 268us/step - loss: 0.0529 -
acc: 0.9829
Epoch 2/3
16667/16667 [=====] - 5s 275us/step - loss: 0.0157 -
acc: 0.9959
Epoch 3/3
```

```
16667/16667 [=====] - 5s 303us/step - loss: 0.0050 -  
acc: 0.9994  
8333/8333 [=====] - 2s 206us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 405us/step - loss: 0.3917 -  
acc: 0.8298  
Epoch 2/3  
16666/16666 [=====] - 5s 300us/step - loss: 0.1285 -  
acc: 0.9566  
Epoch 3/3  
16666/16666 [=====] - 5s 277us/step - loss: 0.0437 -  
acc: 0.9885  
8334/8334 [=====] - 2s 253us/step  
Epoch 1/3  
16667/16667 [=====] - 4s 267us/step - loss: 0.1928 -  
acc: 0.9354  
Epoch 2/3  
16667/16667 [=====] - 4s 268us/step - loss: 0.0467 -  
acc: 0.9864  
Epoch 3/3  
16667/16667 [=====] - 5s 301us/step - loss: 0.0111 -  
acc: 0.9975  
8333/8333 [=====] - 1s 167us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 286us/step - loss: 0.0426 -  
acc: 0.9852  
Epoch 2/3  
16667/16667 [=====] - 5s 314us/step - loss: 0.0126 -  
acc: 0.9977 0s - loss: 0.0125 -  
Epoch 3/3  
16667/16667 [=====] - 5s 293us/step - loss: 0.0025 -  
acc: 0.9997  
8333/8333 [=====] - 1s 176us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 446us/step - loss: 0.3854 -  
acc: 0.8284  
Epoch 2/3  
16666/16666 [=====] - 4s 269us/step - loss: 0.1142 -  
acc: 0.9625  
Epoch 3/3  
16666/16666 [=====] - 4s 270us/step - loss: 0.0369 -  
acc: 0.9906  
8334/8334 [=====] - 2s 247us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 273us/step - loss: 0.1895 -  
acc: 0.9350  
Epoch 2/3  
16667/16667 [=====] - 5s 287us/step - loss: 0.0427 -  
acc: 0.9876  
Epoch 3/3  
16667/16667 [=====] - 5s 301us/step - loss: 0.0096 -  
acc: 0.9982  
8333/8333 [=====] - 1s 158us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 327us/step - loss: 0.0393 -  
acc: 0.9873  
Epoch 2/3
```

```
16667/16667 [=====] - 5s 327us/step - loss: 0.0097 -  
acc: 0.9980  
Epoch 3/3  
16667/16667 [=====] - 5s 317us/step - loss: 0.0020 -  
acc: 0.9999  
8333/8333 [=====] - 2s 218us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 415us/step - loss: 0.3762 -  
acc: 0.8432  
Epoch 2/3  
16666/16666 [=====] - 5s 294us/step - loss: 0.1147 -  
acc: 0.9599  
Epoch 3/3  
16666/16666 [=====] - 5s 305us/step - loss: 0.0350 -  
acc: 0.9902  
8334/8334 [=====] - 2s 268us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 300us/step - loss: 0.2004 -  
acc: 0.9330  
Epoch 2/3  
16667/16667 [=====] - 5s 295us/step - loss: 0.0420 -  
acc: 0.9876  
Epoch 3/3  
16667/16667 [=====] - 5s 324us/step - loss: 0.0089 -  
acc: 0.9983  
8333/8333 [=====] - 2s 228us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 326us/step - loss: 0.0373 -  
acc: 0.9863  
Epoch 2/3  
16667/16667 [=====] - 5s 325us/step - loss: 0.0067 -  
acc: 0.9986  
Epoch 3/3  
16667/16667 [=====] - 5s 312us/step - loss: 0.0013 -  
acc: 0.9999  
8333/8333 [=====] - 1s 159us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 417us/step - loss: 0.3677 -  
acc: 0.8450  
Epoch 2/3  
16666/16666 [=====] - 5s 295us/step - loss: 0.1025 -  
acc: 0.9635  
Epoch 3/3  
16666/16666 [=====] - 6s 335us/step - loss: 0.0273 -  
acc: 0.9925  
8334/8334 [=====] - 2s 260us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 284us/step - loss: 0.1889 -  
acc: 0.9350  
Epoch 2/3  
16667/16667 [=====] - 6s 335us/step - loss: 0.0360 -  
acc: 0.9901  
Epoch 3/3  
16667/16667 [=====] - 5s 324us/step - loss: 0.0068 -  
acc: 0.9983  
8333/8333 [=====] - 2s 203us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 5s 305us/step - loss: 0.0383 -  
acc: 0.9869  
Epoch 2/3  
16667/16667 [=====] - 5s 330us/step - loss: 0.0085 -  
acc: 0.9983  
Epoch 3/3  
16667/16667 [=====] - 6s 333us/step - loss: 0.0017 -  
acc: 0.9998  
8333/8333 [=====] - 2s 182us/step  
Epoch 1/3  
16666/16666 [=====] - 8s 498us/step - loss: 0.3656 -  
acc: 0.8422  
Epoch 2/3  
16666/16666 [=====] - 5s 285us/step - loss: 0.0924 -  
acc: 0.9689  
Epoch 3/3  
16666/16666 [=====] - 5s 272us/step - loss: 0.0243 -  
acc: 0.9939  
8334/8334 [=====] - 2s 275us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 275us/step - loss: 0.1868 -  
acc: 0.9369  
Epoch 2/3  
16667/16667 [=====] - 4s 269us/step - loss: 0.0363 -  
acc: 0.9905  
Epoch 3/3  
16667/16667 [=====] - 4s 266us/step - loss: 0.0061 -  
acc: 0.9988  
8333/8333 [=====] - 1s 169us/step  
Epoch 1/3  
16667/16667 [=====] - 4s 263us/step - loss: 0.0346 -  
acc: 0.9888  
Epoch 2/3  
16667/16667 [=====] - 4s 266us/step - loss: 0.0077 -  
acc: 0.9987  
Epoch 3/3  
16667/16667 [=====] - 4s 265us/step - loss: 0.0013 -  
acc: 0.9999  
8333/8333 [=====] - 1s 175us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 441us/step - loss: 0.3555 -  
acc: 0.8471  
Epoch 2/3  
16666/16666 [=====] - 5s 279us/step - loss: 0.0886 -  
acc: 0.9713  
Epoch 3/3  
16666/16666 [=====] - 5s 278us/step - loss: 0.0203 -  
acc: 0.9950  
8334/8334 [=====] - 2s 297us/step  
Epoch 1/3  
16667/16667 [=====] - 5s 277us/step - loss: 0.2033 -  
acc: 0.9318 1s - loss:  
Epoch 2/3  
16667/16667 [=====] - 5s 279us/step - loss: 0.0363 -  
acc: 0.9901  
Epoch 3/3  
16667/16667 [=====] - 5s 278us/step - loss: 0.0055 -
```

```
acc: 0.9992
8333/8333 [=====] - 1s 156us/step
Epoch 1/3
16667/16667 [=====] - 5s 272us/step - loss: 0.0288 -
acc: 0.9906
Epoch 2/3
16667/16667 [=====] - 4s 261us/step - loss: 0.0050 -
acc: 0.9987 0s - loss: 0.0049 - acc:
Epoch 3/3
16667/16667 [=====] - 4s 222us/step - loss: 0.0011 -
acc: 0.9999
8333/8333 [=====] - 1s 118us/step
Epoch 1/3
16666/16666 [=====] - 5s 309us/step - loss: 0.3562 -
acc: 0.8495 0s - loss: 0.3712 - acc
Epoch 2/3
16666/16666 [=====] - 4s 230us/step - loss: 0.0855 -
acc: 0.9736
Epoch 3/3
16666/16666 [=====] - 4s 265us/step - loss: 0.0193 -
acc: 0.9953
8334/8334 [=====] - 2s 222us/step
Epoch 1/3
16667/16667 [=====] - 4s 256us/step - loss: 0.1993 -
acc: 0.9316
Epoch 2/3
16667/16667 [=====] - 4s 237us/step - loss: 0.0347 -
acc: 0.9903
Epoch 3/3
16667/16667 [=====] - 4s 235us/step - loss: 0.0039 -
acc: 0.9993
8333/8333 [=====] - 1s 125us/step
Epoch 1/3
16667/16667 [=====] - 4s 260us/step - loss: 0.0294 -
acc: 0.9904
Epoch 2/3
16667/16667 [=====] - 4s 237us/step - loss: 0.0052 -
acc: 0.9993
Epoch 3/3
16667/16667 [=====] - 4s 237us/step - loss: 6.2547e-
04 - acc: 1.0000 3s - loss: 8.0275e-04 - ac - ETA: 2s -
8333/8333 [=====] - 1s 127us/step
Epoch 1/3
16666/16666 [=====] - 6s 359us/step - loss: 0.3608 -
acc: 0.8440
Epoch 2/3
16666/16666 [=====] - 4s 258us/step - loss: 0.0880 -
acc: 0.9710
Epoch 3/3
16666/16666 [=====] - 4s 252us/step - loss: 0.0174 -
acc: 0.9959
8334/8334 [=====] - 2s 209us/step
Epoch 1/3
16667/16667 [=====] - 4s 262us/step - loss: 0.1841 -
acc: 0.9374
Epoch 2/3
16667/16667 [=====] - 4s 249us/step - loss: 0.0273 -
```



```
acc: 0.9924
Epoch 3/3
16667/16667 [=====] - 4s 249us/step - loss: 0.0036 -
acc: 0.9996
8333/8333 [=====] - 1s 121us/step
Epoch 1/3
16667/16667 [=====] - 4s 243us/step - loss: 0.0284 -
acc: 0.9906
Epoch 2/3
16667/16667 [=====] - 4s 251us/step - loss: 0.0050 -
acc: 0.9992
Epoch 3/3
16667/16667 [=====] - 4s 227us/step - loss: 0.0010 -
acc: 0.9999
8333/8333 [=====] - 1s 105us/step
Epoch 1/3
16666/16666 [=====] - 6s 353us/step - loss: 0.3571 -
acc: 0.8460
Epoch 2/3
16666/16666 [=====] - 4s 267us/step - loss: 0.0891 -
acc: 0.9703
Epoch 3/3
16666/16666 [=====] - 5s 328us/step - loss: 0.0174 -
acc: 0.9963
8334/8334 [=====] - 2s 277us/step
Epoch 1/3
16667/16667 [=====] - 5s 294us/step - loss: 0.1916 -
acc: 0.9356
Epoch 2/3
16667/16667 [=====] - 5s 280us/step - loss: 0.0285 -
acc: 0.9930
Epoch 3/3
16667/16667 [=====] - 4s 258us/step - loss: 0.0041 -
acc: 0.9993
8333/8333 [=====] - 1s 125us/step
Epoch 1/3
16667/16667 [=====] - 4s 251us/step - loss: 0.0306 -
acc: 0.9900
Epoch 2/3
16667/16667 [=====] - 4s 259us/step - loss: 0.0061 -
acc: 0.9991
Epoch 3/3
16667/16667 [=====] - 5s 272us/step - loss: 0.0016 -
acc: 0.9999
8333/8333 [=====] - 1s 124us/step
Epoch 1/3
16666/16666 [=====] - 6s 389us/step - loss: 0.3650 -
acc: 0.8420
Epoch 2/3
16666/16666 [=====] - 4s 255us/step - loss: 0.0840 -
acc: 0.9713
Epoch 3/3
16666/16666 [=====] - 4s 252us/step - loss: 0.0162 -
acc: 0.9963
8334/8334 [=====] - 1s 176us/step
Epoch 1/3
16667/16667 [=====] - 4s 246us/step - loss: 0.1858 -
```

```
acc: 0.9376
Epoch 2/3
16667/16667 [=====] - 4s 250us/step - loss: 0.0279 -
acc: 0.9928
Epoch 3/3
16667/16667 [=====] - 4s 248us/step - loss: 0.0037 -
acc: 0.9996
8333/8333 [=====] - 1s 110us/step
Epoch 1/3
16667/16667 [=====] - 4s 243us/step - loss: 0.0234 -
acc: 0.9922
Epoch 2/3
16667/16667 [=====] - 4s 250us/step - loss: 0.0035 -
acc: 0.9993
Epoch 3/3
16667/16667 [=====] - 4s 247us/step - loss: 4.6936e-
04 - acc: 1.0000
8333/8333 [=====] - 1s 112us/step
Epoch 1/3
16666/16666 [=====] - 6s 351us/step - loss: 0.3584 -
acc: 0.8507
Epoch 2/3
16666/16666 [=====] - 4s 260us/step - loss: 0.0752 -
acc: 0.9761
Epoch 3/3
16666/16666 [=====] - 4s 261us/step - loss: 0.0126 -
acc: 0.9972
8334/8334 [=====] - 1s 176us/step
Epoch 1/3
16667/16667 [=====] - 4s 263us/step - loss: 0.1928 -
acc: 0.9369
Epoch 2/3
16667/16667 [=====] - 5s 273us/step - loss: 0.0301 -
acc: 0.9914
Epoch 3/3
16667/16667 [=====] - 5s 270us/step - loss: 0.0037 -
acc: 0.9996
8333/8333 [=====] - 1s 116us/step
Epoch 1/3
16667/16667 [=====] - 4s 262us/step - loss: 0.0269 -
acc: 0.9916
Epoch 2/3
16667/16667 [=====] - 5s 284us/step - loss: 0.0064 -
acc: 0.9986
Epoch 3/3
16667/16667 [=====] - 5s 280us/step - loss: 0.0023 -
acc: 0.9998
8333/8333 [=====] - 1s 119us/step
Epoch 1/3
16666/16666 [=====] - 6s 374us/step - loss: 0.3570 -
acc: 0.8446
Epoch 2/3
16666/16666 [=====] - 5s 277us/step - loss: 0.0753 -
acc: 0.9758
Epoch 3/3
16666/16666 [=====] - 5s 277us/step - loss: 0.0102 -
acc: 0.9980
```

```
8334/8334 [=====] - 2s 190us/step
Epoch 1/3
16667/16667 [=====] - 5s 281us/step - loss: 0.1900 -
acc: 0.9363
Epoch 2/3
16667/16667 [=====] - 5s 282us/step - loss: 0.0272 -
acc: 0.9929
Epoch 3/3
16667/16667 [=====] - 5s 287us/step - loss: 0.0041 -
acc: 0.9993
8333/8333 [=====] - 1s 118us/step
Epoch 1/3
16667/16667 [=====] - 5s 285us/step - loss: 0.0258 -
acc: 0.9923
Epoch 2/3
16667/16667 [=====] - 5s 286us/step - loss: 0.0034 -
acc: 0.9995
Epoch 3/3
16667/16667 [=====] - 5s 282us/step - loss: 7.0690e-
04 - acc: 0.9999
8333/8333 [=====] - 1s 117us/step
Epoch 1/3
16666/16666 [=====] - 6s 383us/step - loss: 0.3725 -
acc: 0.8379
Epoch 2/3
16666/16666 [=====] - 5s 285us/step - loss: 0.0819 -
acc: 0.9738
Epoch 3/3
16666/16666 [=====] - 5s 282us/step - loss: 0.0150 -
acc: 0.9964
8334/8334 [=====] - 1s 170us/step
Epoch 1/3
16667/16667 [=====] - 5s 287us/step - loss: 0.1949 -
acc: 0.9350
Epoch 2/3
16667/16667 [=====] - 5s 313us/step - loss: 0.0264 -
acc: 0.9932
Epoch 3/3
16667/16667 [=====] - 5s 294us/step - loss: 0.0036 -
acc: 0.9996
8333/8333 [=====] - 1s 100us/step
Epoch 1/3
16667/16667 [=====] - 5s 294us/step - loss: 0.0237 -
acc: 0.9929
Epoch 2/3
16667/16667 [=====] - 5s 298us/step - loss: 0.0047 -
acc: 0.9992
Epoch 3/3
16667/16667 [=====] - 5s 291us/step - loss: 0.0014 -
acc: 0.9999 0s - loss: 0.0015 -
8333/8333 [=====] - 1s 101us/step
Epoch 1/3
16666/16666 [=====] - 7s 398us/step - loss: 0.3564 -
acc: 0.8509
Epoch 2/3
16666/16666 [=====] - 5s 298us/step - loss: 0.0728 -
acc: 0.9780
```

```
Epoch 3/3
16666/16666 [=====] - 5s 298us/step - loss: 0.0105 -
acc: 0.9977 3s - loss: 0.0130 - acc: - ETA: 2s - los - ETA: 1s -
8334/8334 [=====] - 2s 182us/step
Epoch 1/3
16667/16667 [=====] - 5s 318us/step - loss: 0.1929 -
acc: 0.9319
Epoch 2/3
16667/16667 [=====] - 5s 330us/step - loss: 0.0256 -
acc: 0.9936
Epoch 3/3
16667/16667 [=====] - 5s 306us/step - loss: 0.0025 -
acc: 0.9997
8333/8333 [=====] - 1s 100us/step
Epoch 1/3
16667/16667 [=====] - 5s 311us/step - loss: 0.0263 -
acc: 0.9913
Epoch 2/3
16667/16667 [=====] - 5s 304us/step - loss: 0.0040 -
acc: 0.9990
Epoch 3/3
16667/16667 [=====] - 5s 301us/step - loss: 5.2137e-
04 - acc: 1.0000 1s - loss: 6 - ETA: 0s - loss: 5.4008e-04 - acc: - ETA: 0s
- loss: 5.2645e-04 - acc: 1.
8333/8333 [=====] - 1s 103us/step
Epoch 1/3
16666/16666 [=====] - 7s 419us/step - loss: 0.3570 -
acc: 0.8432
Epoch 2/3
16666/16666 [=====] - 5s 323us/step - loss: 0.0659 -
acc: 0.9791
Epoch 3/3
16666/16666 [=====] - 5s 312us/step - loss: 0.0094 -
acc: 0.9984 0s - loss: 0.0089 - acc: 0. - ETA: 0s - loss: 0.0096 - acc: 0.9
8334/8334 [=====] - 1s 177us/step
Epoch 1/3
16667/16667 [=====] - 5s 309us/step - loss: 0.1972 -
acc: 0.9358
Epoch 2/3
16667/16667 [=====] - 5s 308us/step - loss: 0.0258 -
acc: 0.9926
Epoch 3/3
16667/16667 [=====] - 5s 311us/step - loss: 0.0021 -
acc: 0.9996
8333/8333 [=====] - 1s 110us/step
Epoch 1/3
16667/16667 [=====] - 6s 332us/step - loss: 0.0234 -
acc: 0.9925
Epoch 2/3
16667/16667 [=====] - 5s 317us/step - loss: 0.0073 -
acc: 0.9982
Epoch 3/3
16667/16667 [=====] - 5s 320us/step - loss: 0.0013 -
acc: 0.9999
8333/8333 [=====] - 1s 107us/step
Epoch 1/3
16666/16666 [=====] - 7s 447us/step - loss: 0.3605 -
```

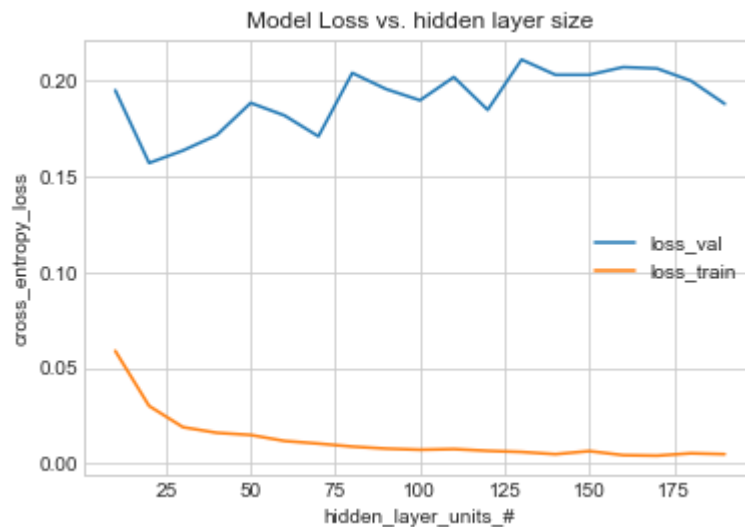
```
acc: 0.8429
Epoch 2/3
16666/16666 [=====] - 5s 329us/step - loss: 0.0709 -
acc: 0.9782
Epoch 3/3
16666/16666 [=====] - 6s 336us/step - loss: 0.0098 -
acc: 0.9979
8334/8334 [=====] - 2s 194us/step
Epoch 1/3
16667/16667 [=====] - 6s 335us/step - loss: 0.1951 -
acc: 0.9347
Epoch 2/3
16667/16667 [=====] - 6s 335us/step - loss: 0.0267 -
acc: 0.9934
Epoch 3/3
16667/16667 [=====] - 6s 335us/step - loss: 0.0050 -
acc: 0.9994
8333/8333 [=====] - 1s 129us/step
Epoch 1/3
16667/16667 [=====] - 6s 332us/step - loss: 0.0237 -
acc: 0.9935
Epoch 2/3
16667/16667 [=====] - 6s 332us/step - loss: 0.0037 -
acc: 0.9995
Epoch 3/3
16667/16667 [=====] - 6s 331us/step - loss: 0.0017 -
acc: 0.9998
8333/8333 [=====] - 1s 122us/step
Epoch 1/3
16666/16666 [=====] - 8s 456us/step - loss: 0.3742 -
acc: 0.8391
Epoch 2/3
16666/16666 [=====] - 6s 345us/step - loss: 0.0818 -
acc: 0.9737
Epoch 3/3
16666/16666 [=====] - 6s 343us/step - loss: 0.0123 -
acc: 0.9973
8334/8334 [=====] - 2s 195us/step
Epoch 1/3
16667/16667 [=====] - 6s 345us/step - loss: 0.1820 -
acc: 0.9398
Epoch 2/3
16667/16667 [=====] - 6s 344us/step - loss: 0.0247 -
acc: 0.9935
Epoch 3/3
16667/16667 [=====] - 6s 354us/step - loss: 0.0025 -
acc: 0.9998
8333/8333 [=====] - 1s 162us/step
Epoch 1/3
16667/16667 [=====] - 6s 358us/step - loss: 0.0228 -
acc: 0.9927
Epoch 2/3
16667/16667 [=====] - 6s 347us/step - loss: 0.0035 -
acc: 0.9993
Epoch 3/3
16667/16667 [=====] - 6s 350us/step - loss: 3.2763e-
```

04 - acc: 1.0000

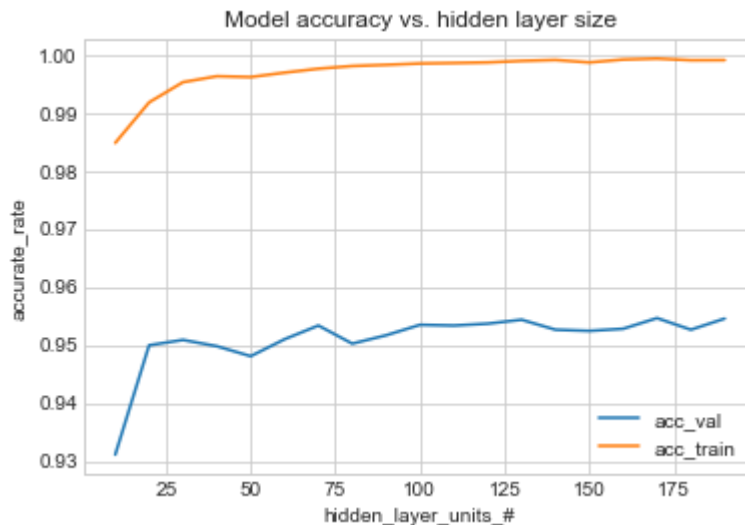
8333/8333 [=====] - 1s 128us/step

```
In [37]: ix = units
loss_train_2 = pd.Series(loss_train_2, index = ix)
loss_val_2 = pd.Series(loss_val_2, index = ix)
acc_train_2 = pd.Series(acc_train_2, index = ix)
acc_val_2 = pd.Series(acc_val_2, index = ix)
```

```
In [38]: fig = plt.figure()
ax = plt.axes()
ax.plot(loss_val_2, label = 'loss_val')
ax.plot(loss_train_2, label = 'loss_train')
ax.legend()
plt.xlabel('hidden_layer_units_#')
plt.ylabel('cross_entropy_loss')
plt.title('Model Loss vs. hidden layer size')
fig.savefig("loss_cv_2.png",dpi = 400)
```



```
In [39]: fig = plt.figure()
ax = plt.axes()
ax.plot(acc_val_2,label = 'acc_val')
ax.plot(acc_train_2, label = 'acc_train')
ax.legend()
plt.xlabel('hidden_layer_units_#')
plt.ylabel('accurate_rate')
plt.title('Model accuracy vs. hidden layer size')
fig.savefig("acc_cv_2.png",dpi=400)
```



```
In [52]: # baseline model with two hidden layers
model_overfitting_2 = Sequential()
model_overfitting_2.add(Dense(50,input_shape = (max_num,)))
model_overfitting_2.add(Activation('relu'))
model_overfitting_2.add(Dense(50))
model_overfitting_2.add(Activation('relu'))
model_overfitting_2.add(Dense(num_classes))
model_overfitting_2.add(Activation('softmax'))
model_overfitting_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_overfitting_2 = model_overfitting_2.fit(X_train,y_train, batch_size=200,
epochs = 3, verbose = 1)
score_overfitting_2 = model_overfitting_2.evaluate(X_test,y_test, batch_size=200, verbose = 1)

base_train_acc_2 = hist_overfitting_2.history.get('acc')[-1]
base_test_acc_2 = score_overfitting_2[1]
```

Epoch 1/3

25000/25000 [=====] - 7s 270us/step - loss: 0.3421 - acc: 0.8544

Epoch 2/3

25000/25000 [=====] - 4s 158us/step - loss: 0.1223 - acc: 0.9559

Epoch 3/3

25000/25000 [=====] - 4s 159us/step - loss: 0.0409 - acc: 0.9875

25000/25000 [=====] - 3s 131us/step

```
In [53]: print('The accurate rate of the overfitting model with two hidden layers on training dataset is')
print(base_train_acc_2)
print('The accurate rate of the overfitting model with two hidden layers on test dataset is')
print(base_test_acc_2)
```

The accurate rate of the overfitting model with two hidden layers on training dataset is

0.987520010471344

The accurate rate of the overfitting model with two hidden layers on test dataset is

0.8638800020217896


```

In [54]: from tensorflow.keras.layers import Dropout
rate = [0.1,0.2,0.3,0.4,0.5,0.6,0.7,0.8,0.9]
loss_train_drop = []
acc_train_drop = []
loss_val_drop = []
acc_val_drop = []
for r in rate:
    model_drop = Sequential()
    model_drop.add(Dense(50,input_shape = (max_num,)))
    model_drop.add(Activation('relu'))
    model_drop.add(Dropout(r))
    model_drop.add(Dense(50))
    model_drop.add(Activation('relu'))
    model_drop.add(Dropout(r))
    model_drop.add(Dense(num_classes))
    model_drop.add(Activation('softmax'))
    model_drop.compile(loss = 'categorical_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_drop.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3
    )

    loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
    acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
    score_val_cv = model_drop.evaluate(X_val_cv,y_val_cv, batch_size=200,
verbose = 1)
    loss_val_cv = loss_val_cv + score_val_cv[0]
    acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_drop.append(loss_val_cv/3)
    acc_val_drop.append(acc_val_cv/3)
    loss_train_drop.append(loss_train_cv/3)
    acc_train_drop.append(acc_train_cv/3)

```

```
Epoch 1/3
16666/16666 [=====] - 6s 342us/step - loss: 0.3881 -
acc: 0.8276
Epoch 2/3
16666/16666 [=====] - 3s 163us/step - loss: 0.1257 -
acc: 0.9543
Epoch 3/3
16666/16666 [=====] - 3s 164us/step - loss: 0.0485 -
acc: 0.9853
8334/8334 [=====] - 2s 222us/step
Epoch 1/3
16667/16667 [=====] - 3s 165us/step - loss: 0.1950 -
acc: 0.9328
Epoch 2/3
16667/16667 [=====] - 3s 163us/step - loss: 0.0455 -
acc: 0.9866
Epoch 3/3
16667/16667 [=====] - 3s 163us/step - loss: 0.0120 -
acc: 0.9972
8333/8333 [=====] - 1s 90us/step
Epoch 1/3
16667/16667 [=====] - 3s 165us/step - loss: 0.0411 -
acc: 0.9861
Epoch 2/3
16667/16667 [=====] - 3s 165us/step - loss: 0.0135 -
acc: 0.9966
Epoch 3/3
16667/16667 [=====] - 3s 164us/step - loss: 0.0051 -
acc: 0.9990
8333/8333 [=====] - 1s 93us/step
Epoch 1/3
16666/16666 [=====] - 6s 338us/step - loss: 0.4107 -
acc: 0.8174
Epoch 2/3
16666/16666 [=====] - 3s 167us/step - loss: 0.1479 -
acc: 0.9449
Epoch 3/3
16666/16666 [=====] - 3s 166us/step - loss: 0.0667 -
acc: 0.9781
8334/8334 [=====] - 2s 236us/step
Epoch 1/3
16667/16667 [=====] - 3s 180us/step - loss: 0.2052 -
acc: 0.9295
Epoch 2/3
16667/16667 [=====] - 3s 173us/step - loss: 0.0701 -
acc: 0.9787
Epoch 3/3
16667/16667 [=====] - 3s 163us/step - loss: 0.0299 -
acc: 0.9915
8333/8333 [=====] - 1s 95us/step
Epoch 1/3
16667/16667 [=====] - 3s 163us/step - loss: 0.0590 -
acc: 0.9803
Epoch 2/3
16667/16667 [=====] - 3s 163us/step - loss: 0.0192 -
acc: 0.9940
Epoch 3/3
```

```
16667/16667 [=====] - 3s 170us/step - loss: 0.0122 -  
acc: 0.9962  
8333/8333 [=====] - 1s 100us/step  
Epoch 1/3  
16666/16666 [=====] - 5s 330us/step - loss: 0.4484 -  
acc: 0.7904  
Epoch 2/3  
16666/16666 [=====] - 3s 163us/step - loss: 0.1867 -  
acc: 0.9331  
Epoch 3/3  
16666/16666 [=====] - 3s 163us/step - loss: 0.0990 -  
acc: 0.9657  
8334/8334 [=====] - 2s 232us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 165us/step - loss: 0.2042 -  
acc: 0.9294  
Epoch 2/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.0825 -  
acc: 0.9728  
Epoch 3/3  
16667/16667 [=====] - 3s 167us/step - loss: 0.0396 -  
acc: 0.9877  
8333/8333 [=====] - 1s 95us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 166us/step - loss: 0.0699 -  
acc: 0.9758  
Epoch 2/3  
16667/16667 [=====] - 3s 176us/step - loss: 0.0340 -  
acc: 0.9887  
Epoch 3/3  
16667/16667 [=====] - 3s 177us/step - loss: 0.0258 -  
acc: 0.9914  
8333/8333 [=====] - 1s 98us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 334us/step - loss: 0.5186 -  
acc: 0.7434  
Epoch 2/3  
16666/16666 [=====] - 3s 164us/step - loss: 0.2548 -  
acc: 0.9022  
Epoch 3/3  
16666/16666 [=====] - 3s 165us/step - loss: 0.1512 -  
acc: 0.9467  
8334/8334 [=====] - 2s 231us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 165us/step - loss: 0.2253 -  
acc: 0.9206  
Epoch 2/3  
16667/16667 [=====] - 3s 167us/step - loss: 0.1334 -  
acc: 0.9532  
Epoch 3/3  
16667/16667 [=====] - 3s 163us/step - loss: 0.0824 -  
acc: 0.9713 2s - loss: 0.0777 - ETA: 1s - loss  
8333/8333 [=====] - 1s 93us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 166us/step - loss: 0.1143 -  
acc: 0.9585  
Epoch 2/3
```

```
16667/16667 [=====] - 3s 164us/step - loss: 0.0768 -  
acc: 0.9717  
Epoch 3/3  
16667/16667 [=====] - 3s 169us/step - loss: 0.0584 -  
acc: 0.9770  
8333/8333 [=====] - 1s 101us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 338us/step - loss: 0.5539 -  
acc: 0.7104  
Epoch 2/3  
16666/16666 [=====] - 3s 165us/step - loss: 0.2923 -  
acc: 0.8876  
Epoch 3/3  
16666/16666 [=====] - 3s 163us/step - loss: 0.2031 -  
acc: 0.9292  
8334/8334 [=====] - 2s 234us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 169us/step - loss: 0.2449 -  
acc: 0.9124  
Epoch 2/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.1635 -  
acc: 0.9419  
Epoch 3/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.1282 -  
acc: 0.9549  
8333/8333 [=====] - 1s 95us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 166us/step - loss: 0.1519 -  
acc: 0.9444  
Epoch 2/3  
16667/16667 [=====] - 3s 169us/step - loss: 0.1098 -  
acc: 0.9603  
Epoch 3/3  
16667/16667 [=====] - 3s 170us/step - loss: 0.0814 -  
acc: 0.9691  
8333/8333 [=====] - 1s 99us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 340us/step - loss: 0.6292 -  
acc: 0.6495  
Epoch 2/3  
16666/16666 [=====] - 3s 162us/step - loss: 0.3972 -  
acc: 0.8274  
Epoch 3/3  
16666/16666 [=====] - 3s 163us/step - loss: 0.2877 -  
acc: 0.8838  
8334/8334 [=====] - 2s 237us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 166us/step - loss: 0.2923 -  
acc: 0.8877  
Epoch 2/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.2353 -  
acc: 0.9125  
Epoch 3/3  
16667/16667 [=====] - 3s 165us/step - loss: 0.1884 -  
acc: 0.9283  
8333/8333 [=====] - 1s 92us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 3s 174us/step - loss: 0.2161 -  
acc: 0.9180  
Epoch 2/3  
16667/16667 [=====] - 3s 169us/step - loss: 0.1821 -  
acc: 0.9317  
Epoch 3/3  
16667/16667 [=====] - 3s 173us/step - loss: 0.1515 -  
acc: 0.9402  
8333/8333 [=====] - 1s 100us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 345us/step - loss: 0.7058 -  
acc: 0.5620  
Epoch 2/3  
16666/16666 [=====] - 3s 184us/step - loss: 0.5380 -  
acc: 0.7335  
Epoch 3/3  
16666/16666 [=====] - 3s 176us/step - loss: 0.4289 -  
acc: 0.8109  
8334/8334 [=====] - 2s 253us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 165us/step - loss: 0.3966 -  
acc: 0.8354  
Epoch 2/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.3399 -  
acc: 0.8655  
Epoch 3/3  
16667/16667 [=====] - 3s 168us/step - loss: 0.2849 -  
acc: 0.8848  
8333/8333 [=====] - 1s 98us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 166us/step - loss: 0.2971 -  
acc: 0.8806  
Epoch 2/3  
16667/16667 [=====] - 3s 164us/step - loss: 0.2690 -  
acc: 0.8962  
Epoch 3/3  
16667/16667 [=====] - 3s 183us/step - loss: 0.2382 -  
acc: 0.9065  
8333/8333 [=====] - 1s 101us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 349us/step - loss: 0.7501 -  
acc: 0.4933  
Epoch 2/3  
16666/16666 [=====] - 3s 163us/step - loss: 0.6910 -  
acc: 0.5371  
Epoch 3/3  
16666/16666 [=====] - 3s 164us/step - loss: 0.6498 -  
acc: 0.6076  
8334/8334 [=====] - 2s 250us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 165us/step - loss: 0.6137 -  
acc: 0.6412  
Epoch 2/3  
16667/16667 [=====] - 3s 168us/step - loss: 0.5689 -  
acc: 0.6894  
Epoch 3/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.5337 -
```

```

acc: 0.7151
8333/8333 [=====] - 1s 99us/step
Epoch 1/3
16667/16667 [=====] - 3s 167us/step - loss: 0.5102 -
acc: 0.7403
Epoch 2/3
16667/16667 [=====] - 3s 169us/step - loss: 0.4880 -
acc: 0.7601
Epoch 3/3
16667/16667 [=====] - 3s 168us/step - loss: 0.4587 -
acc: 0.7805
8333/8333 [=====] - 1s 98us/step
Epoch 1/3
16666/16666 [=====] - 6s 350us/step - loss: 0.8428 -
acc: 0.4963
Epoch 2/3
16666/16666 [=====] - 3s 163us/step - loss: 0.7036 -
acc: 0.5007
Epoch 3/3
16666/16666 [=====] - 3s 166us/step - loss: 0.7065 -
acc: 0.5006
8334/8334 [=====] - 2s 241us/step
Epoch 1/3
16667/16667 [=====] - 3s 164us/step - loss: 0.6975 -
acc: 0.4968
Epoch 2/3
16667/16667 [=====] - 3s 163us/step - loss: 0.6984 -
acc: 0.4940
Epoch 3/3
16667/16667 [=====] - 3s 162us/step - loss: 0.7001 -
acc: 0.4988
8333/8333 [=====] - 1s 89us/step
Epoch 1/3
16667/16667 [=====] - 3s 164us/step - loss: 0.6972 -
acc: 0.4994
Epoch 2/3
16667/16667 [=====] - 3s 165us/step - loss: 0.6988 -
acc: 0.5005
Epoch 3/3
16667/16667 [=====] - 3s 164us/step - loss: 0.6974 -
acc: 0.5032
8333/8333 [=====] - 1s 100us/step

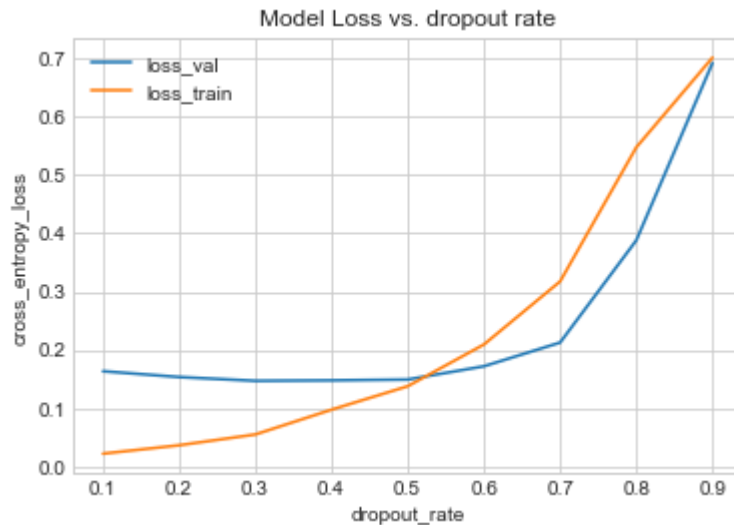
```

```

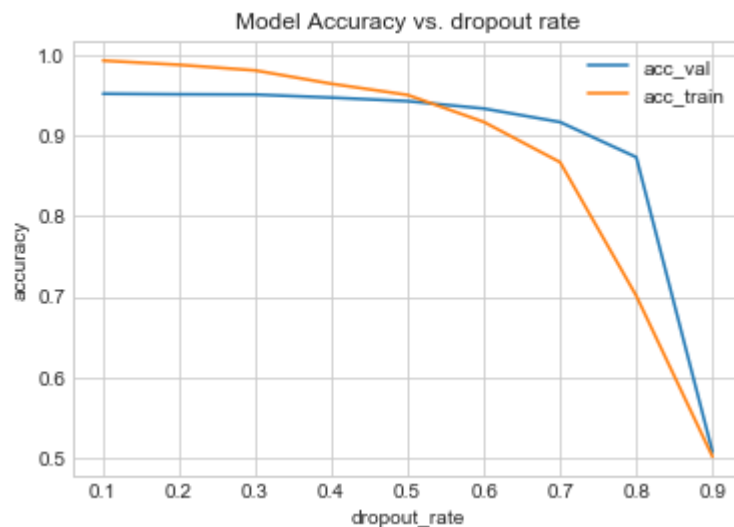
In [55]: ix = rate
loss_train_drop= pd.Series(loss_train_drop, index = ix)
loss_val_drop = pd.Series(loss_val_drop, index = ix)
acc_train_drop = pd.Series(acc_train_drop, index = ix)
acc_val_drop = pd.Series(acc_val_drop, index = ix)

```

```
In [56]: fig = plt.figure()
ax = plt.axes()
ax.plot(loss_val_drop, label = 'loss_val')
ax.plot(loss_train_drop, label = 'loss_train')
ax.legend()
plt.xlabel('dropout_rate')
plt.ylabel('cross_entropy_loss')
plt.title('Model Loss vs. dropout rate')
fig.savefig("drop_loss_cv_2.png",dpi = 400)
```



```
In [57]: fig = plt.figure()
ax = plt.axes()
ax.plot(acc_val_drop, label = 'acc_val')
ax.plot(acc_train_drop, label = 'acc_train')
ax.legend()
plt.xlabel('dropout_rate')
plt.ylabel('accuracy')
plt.title('Model Accuracy vs. dropout rate')
fig.savefig("drop_acc_cv_2.png",dpi = 400)
```



```
In [60]: model_drop_2 = Sequential()
model_drop_2.add(Dense(50,input_shape = (max_num,)))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.5))
model_drop_2.add(Dense(50))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.5))
model_drop_2.add(Dense(num_classes))
model_drop_2.add(Activation('softmax'))
model_drop_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])

hist_drop_2 = model_drop_2.fit(X_train,y_train, batch_size=200, epochs = 3, ve
rbose = 1)
score_drop_2 = model_drop_2.evaluate(X_test,y_test, batch_size=200, verbose =
1)

drop_train_acc_2 = hist_drop_2.history.get('acc')[-1]
drop_test_acc_2 = score_drop_2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 8s 304us/step - loss: 0.4674 -
acc: 0.7780
Epoch 2/3
25000/25000 [=====] - 4s 172us/step - loss: 0.2513 -
acc: 0.9074
Epoch 3/3
25000/25000 [=====] - 4s 173us/step - loss: 0.1687 -
acc: 0.9395
25000/25000 [=====] - 4s 155us/step
```

```
In [61]: print('The accurate rate of the model with two hidden layers and dropout(0.5)
on training dataset is')
print(drop_train_acc_2)
print('The accurate rate of the model with two hidden layers and dropout(0.5)
on test dataset is')
print(drop_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and dropout(0.5) on tra
ining dataset is
0.939479998588562
The accurate rate of the model with two hidden layers and dropout(0.5) on tes
t dataset is
0.877079999923706
```



```
In [84]: model_nn25_2 = Sequential()
model_nn25_2.add(Dense(25,input_shape = (max_num,)))
model_nn25_2.add(Activation('relu'))
model_nn25_2.add(Dense(25))
model_nn25_2.add(Activation('relu'))
model_nn25_2.add(Dense(num_classes))
model_nn25_2.add(Activation('softmax'))
model_nn25_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])

hist_nn25_2 = model_nn25_2.fit(X_train,y_train, batch_size=200, epochs = 3, ve
rbose = 1)
score_nn25_2 = model_nn25_2.evaluate(X_test,y_test, batch_size=200, verbose =
1)

nn25_train_acc_2 = hist_nn25_2.history.get('acc')[-1]
nn25_test_acc_2 = score_nn25_2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 22s 869us/step - loss: 0.3517
- acc: 0.8536
Epoch 2/3
25000/25000 [=====] - 6s 249us/step - loss: 0.1441 -
acc: 0.9484
Epoch 3/3
25000/25000 [=====] - 6s 232us/step - loss: 0.0743 -
acc: 0.9756
25000/25000 [=====] - 12s 487us/step
```

```
In [85]: print('The accurate rate of the model with two hidden layers(25units) on train
ing dataset is')
print(drop_train_acc_2)
print('The accurate rate of the model with two hidden layers(25units) on test
dataset is')
print(drop_test_acc_2)
```

```
The accurate rate of the model with two hidden layers(25units) on training da
taset is
0.9673600010871887
The accurate rate of the model with two hidden layers(25units) on test datase
t is
0.8725199995040893
```

```
In [62]: model_drop_2 = Sequential()
model_drop_2.add(Dense(50,input_shape = (max_num,)))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.4))
model_drop_2.add(Dense(50))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.4))
model_drop_2.add(Dense(num_classes))
model_drop_2.add(Activation('softmax'))
model_drop_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])

hist_drop_2 = model_drop_2.fit(X_train,y_train, batch_size=200, epochs = 3, ve
rbose = 1)
score_drop_2 = model_drop_2.evaluate(X_test,y_test, batch_size=200, verbose =
1)

drop_train_acc_2 = hist_drop_2.history.get('acc')[-1]
drop_test_acc_2 = score_drop_2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 8s 309us/step - loss: 0.4405 -
acc: 0.7981
Epoch 2/3
25000/25000 [=====] - 5s 182us/step - loss: 0.2181 -
acc: 0.9203
Epoch 3/3
25000/25000 [=====] - 4s 170us/step - loss: 0.1437 -
acc: 0.9492
25000/25000 [=====] - 4s 150us/step
```

```
In [63]: print('The accurate rate of the model with two hidden layers and dropout(0.4)
on training dataset is')
print(drop_train_acc_2)
print('The accurate rate of the model with two hidden layers and dropout(0.4)
on test dataset is')
print(drop_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and dropout(0.4) on tra
ining dataset is
0.9491999969482422
The accurate rate of the model with two hidden layers and dropout(0.4) on tes
t dataset is
0.8730800008773804
```

```
In [64]: model_drop_2 = Sequential()
model_drop_2.add(Dense(50,input_shape = (max_num,)))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.3))
model_drop_2.add(Dense(50))
model_drop_2.add(Activation('relu'))
model_drop_2.add(Dropout(0.3))
model_drop_2.add(Dense(num_classes))
model_drop_2.add(Activation('softmax'))
model_drop_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', me
trics = ['accuracy'])

hist_drop_2 = model_drop_2.fit(X_train,y_train, batch_size=200, epochs = 3, ve
rbose = 1)
score_drop_2 = model_drop_2.evaluate(X_test,y_test, batch_size=200, verbose =
1)

drop_train_acc_2 = hist_drop_2.history.get('acc')[-1]
drop_test_acc_2 = score_drop_2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 8s 308us/step - loss: 0.4090 -
acc: 0.8146
Epoch 2/3
25000/25000 [=====] - 4s 166us/step - loss: 0.1834 -
acc: 0.9328
Epoch 3/3
25000/25000 [=====] - 4s 167us/step - loss: 0.0976 -
acc: 0.9674
25000/25000 [=====] - 4s 148us/step
```

```
In [65]: print('The accurate rate of the model with two hidden layers and dropout(0.3)
on training dataset is')
print(drop_train_acc_2)
print('The accurate rate of the model with two hidden layers and dropout(0.3)
on test dataset is')
print(drop_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and dropout(0.3) on tra
ining dataset is
0.9673600010871887
The accurate rate of the model with two hidden layers and dropout(0.3) on tes
t dataset is
0.8725199995040893
```

```

In [67]: lambda_list = [1e-6,1e-5,1e-4,1e-3,1e-2,1e-1]
loss_train_2_l1 = []
acc_train_2_l1 = []
loss_val_2_l1 = []
acc_val_2_l1 = []
for par in lambda_list:
    model_2_l1 = Sequential()
    model_2_l1.add(Dense(50,input_shape = (max_num,), kernel_regularizer = l1(
par), bias_regularizer=l1(par)))
    model_2_l1.add(Activation('relu'))
    model_2_l1.add(Dense(50, kernel_regularizer = l1(par), bias_regularizer=l1
(par)))
    model_2_l1.add(Activation('relu'))
    model_2_l1.add(Dense(num_classes, kernel_regularizer = l1(par), bias_regul
arizer=l1(par)))
    model_2_l1.add(Activation('softmax'))
    model_2_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_2_l1.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3
)

        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_2_l1.evaluate(X_val_cv,y_val_cv, batch_size=200,
verbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_2_l1.append(loss_val_cv/3)
    acc_val_2_l1.append(acc_val_cv/3)
    loss_train_2_l1.append(loss_train_cv/3)
    acc_train_2_l1.append(acc_train_cv/3)

```

```
Epoch 1/3
16666/16666 [=====] - 7s 402us/step - loss: 0.3948 -
acc: 0.8295
Epoch 2/3
16666/16666 [=====] - 3s 184us/step - loss: 0.1339 -
acc: 0.9548
Epoch 3/3
16666/16666 [=====] - 3s 191us/step - loss: 0.0534 -
acc: 0.9863
8334/8334 [=====] - 3s 311us/step
Epoch 1/3
16667/16667 [=====] - 3s 189us/step - loss: 0.2059 -
acc: 0.9323
Epoch 2/3
16667/16667 [=====] - 3s 184us/step - loss: 0.0505 -
acc: 0.9873
Epoch 3/3
16667/16667 [=====] - 3s 182us/step - loss: 0.0162 -
acc: 0.9983
8333/8333 [=====] - 1s 102us/step
Epoch 1/3
16667/16667 [=====] - 3s 176us/step - loss: 0.0477 -
acc: 0.9878
Epoch 2/3
16667/16667 [=====] - 3s 173us/step - loss: 0.0177 -
acc: 0.9980
Epoch 3/3
16667/16667 [=====] - 3s 174us/step - loss: 0.0117 -
acc: 0.9996
8333/8333 [=====] - 1s 99us/step
Epoch 1/3
16666/16666 [=====] - 6s 379us/step - loss: 0.4162 -
acc: 0.8459
Epoch 2/3
16666/16666 [=====] - 3s 173us/step - loss: 0.1735 -
acc: 0.9605
Epoch 3/3
16666/16666 [=====] - 3s 173us/step - loss: 0.1021 -
acc: 0.9875
8334/8334 [=====] - 2s 268us/step
Epoch 1/3
16667/16667 [=====] - 3s 174us/step - loss: 0.2741 -
acc: 0.9276
Epoch 2/3
16667/16667 [=====] - 3s 176us/step - loss: 0.1165 -
acc: 0.9852
Epoch 3/3
16667/16667 [=====] - 3s 176us/step - loss: 0.0776 -
acc: 0.9974
8333/8333 [=====] - 1s 104us/step
Epoch 1/3
16667/16667 [=====] - 3s 179us/step - loss: 0.1257 -
acc: 0.9779
Epoch 2/3
16667/16667 [=====] - 3s 196us/step - loss: 0.0900 -
acc: 0.9947
Epoch 3/3
```

```
16667/16667 [=====] - 3s 193us/step - loss: 0.0729 -  
acc: 0.9994  
8333/8333 [=====] - 1s 104us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 393us/step - loss: 0.8180 -  
acc: 0.8366  
Epoch 2/3  
16666/16666 [=====] - 3s 178us/step - loss: 0.4571 -  
acc: 0.9492  
Epoch 3/3  
16666/16666 [=====] - 3s 176us/step - loss: 0.3510 -  
acc: 0.9708  
8334/8334 [=====] - 2s 287us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 176us/step - loss: 0.4823 -  
acc: 0.9266  
Epoch 2/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.3380 -  
acc: 0.9761  
Epoch 3/3  
16667/16667 [=====] - 3s 177us/step - loss: 0.2568 -  
acc: 0.9903  
8333/8333 [=====] - 1s 100us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 176us/step - loss: 0.4007 -  
acc: 0.9435  
Epoch 2/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.3340 -  
acc: 0.9817  
Epoch 3/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.2488 -  
acc: 0.9956  
8333/8333 [=====] - 1s 101us/step  
Epoch 1/3  
16666/16666 [=====] - 6s 390us/step - loss: 2.3336 -  
acc: 0.8399  
Epoch 2/3  
16666/16666 [=====] - 3s 175us/step - loss: 0.9719 -  
acc: 0.8942  
Epoch 3/3  
16666/16666 [=====] - 3s 180us/step - loss: 0.8172 -  
acc: 0.9024  
8334/8334 [=====] - 2s 284us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 175us/step - loss: 0.7263 -  
acc: 0.9030  
Epoch 2/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.5863 -  
acc: 0.9260  
Epoch 3/3  
16667/16667 [=====] - 3s 193us/step - loss: 0.5042 -  
acc: 0.9384  
8333/8333 [=====] - 1s 124us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 189us/step - loss: 0.5779 -  
acc: 0.9121  
Epoch 2/3
```

```
16667/16667 [=====] - 3s 182us/step - loss: 0.4615 -  
acc: 0.9431  
Epoch 3/3  
16667/16667 [=====] - 3s 178us/step - loss: 0.4252 -  
acc: 0.9490  
8333/8333 [=====] - 1s 103us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 390us/step - loss: 11.5090  
- acc: 0.7734  
Epoch 2/3  
16666/16666 [=====] - 3s 176us/step - loss: 2.2685 -  
acc: 0.8452  
Epoch 3/3  
16666/16666 [=====] - 3s 174us/step - loss: 1.4654 -  
acc: 0.8604  
8334/8334 [=====] - 2s 282us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 180us/step - loss: 1.2861 -  
acc: 0.8702  
Epoch 2/3  
16667/16667 [=====] - 3s 175us/step - loss: 1.2563 -  
acc: 0.8748  
Epoch 3/3  
16667/16667 [=====] - 3s 173us/step - loss: 1.2545 -  
acc: 0.8780  
8333/8333 [=====] - 1s 98us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 176us/step - loss: 1.2579 -  
acc: 0.8733  
Epoch 2/3  
16667/16667 [=====] - 3s 175us/step - loss: 1.2486 -  
acc: 0.8759  
Epoch 3/3  
16667/16667 [=====] - 3s 179us/step - loss: 1.2477 -  
acc: 0.8806  
8333/8333 [=====] - 1s 103us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 398us/step - loss: 106.2643  
- acc: 0.5230  
Epoch 2/3  
16666/16666 [=====] - 3s 173us/step - loss: 14.8354  
- acc: 0.5011  
Epoch 3/3  
16666/16666 [=====] - 3s 176us/step - loss: 7.6500 -  
acc: 0.5087  
8334/8334 [=====] - 2s 283us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 175us/step - loss: 6.4992 -  
acc: 0.4978  
Epoch 2/3  
16667/16667 [=====] - 3s 175us/step - loss: 6.4632 -  
acc: 0.4998  
Epoch 3/3  
16667/16667 [=====] - 3s 174us/step - loss: 6.4630 -  
acc: 0.5037  
8333/8333 [=====] - 1s 95us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 3s 181us/step - loss: 6.4624 -  
acc: 0.4986  
Epoch 2/3  
16667/16667 [=====] - 3s 175us/step - loss: 6.4635 -  
acc: 0.4963  
Epoch 3/3  
16667/16667 [=====] - 3s 179us/step - loss: 6.4633 -  
acc: 0.4979  
8333/8333 [=====] - 1s 103us/step
```

```
In [68]: l1_2_reg_cv = pd.DataFrame({'loss_train': loss_train_2_l1,  
                                     'loss_val': loss_val_2_l1,  
                                     'acc_train': acc_train_2_l1,  
                                     'acc_val': acc_val_2_l1,  
                                     'lambda': lambda_list}).set_index('lambda')  
  
l1_2_reg_cv
```

Out[68]:

	loss_train	loss_val	acc_train	acc_val
lambda				
0.000001	0.027111	0.197342	0.994740	0.946803
0.000010	0.084224	0.267125	0.994780	0.942803
0.000100	0.285536	0.462369	0.985540	0.918402
0.001000	0.582214	0.661997	0.929919	0.886720
0.010000	1.322553	1.279183	0.873020	0.869440
0.100000	6.858770	6.491953	0.503460	0.500000


```
In [69]: # indicating by the result of l1_reg_cv data frame, will set the lambda as 1e-5 for l1 norm regularization
model_2_l1 = Sequential()
model_2_l1.add(Dense(50, input_shape = (max_num,), kernel_regularizer = l1(1e-5), bias_regularizer=l1(1e-5)))
model_2_l1.add(Activation('relu'))
model_2_l1.add(Dense(50, kernel_regularizer = l1(1e-5), bias_regularizer=l1(1e-5)))
model_2_l1.add(Activation('relu'))
model_2_l1.add(Dense(num_classes, kernel_regularizer = l1(1e-5), bias_regularizer=l1(1e-5)))
model_2_l1.add(Activation('softmax'))
model_2_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_2_l1 = model_2_l1.fit(X_train, y_train, batch_size=200, epochs = 3, verbose = 1)
score_2_l1 = model_2_l1.evaluate(X_test, y_test, batch_size=200, verbose = 1)

l1_train_acc_2 = hist_2_l1.history.get('acc')[-1]
l1_test_acc_2 = score_2_l1[1]
```

```
Epoch 1/3
25000/25000 [=====] - 9s 351us/step - loss: 0.3950 - acc: 0.8576
Epoch 2/3
25000/25000 [=====] - 4s 177us/step - loss: 0.1798 - acc: 0.9588
Epoch 3/3
25000/25000 [=====] - 4s 172us/step - loss: 0.1058 - acc: 0.9873
25000/25000 [=====] - 4s 163us/step
```

```
In [70]: print('The accurate rate of the model with two hidden layers and l1 regularization on training dataset is')
print(l1_train_acc_2)
print('The accurate rate of the model with two hidden layers and l1 regularization on test dataset is')
print(l1_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and l1 regularization on training dataset is
0.987320011138916
The accurate rate of the model with two hidden layers and l1 regularization on test dataset is
0.865840000629425
```

```
In [71]: # indicating by the result of l1_reg_cv data frame, will set the lambda as 1e-6 for l1 norm regularization
model_2_l1 = Sequential()
model_2_l1.add(Dense(50, input_shape = (max_num,), kernel_regularizer = l1(1e-6), bias_regularizer=l1(1e-6)))
model_2_l1.add(Activation('relu'))
model_2_l1.add(Dense(50, kernel_regularizer = l1(1e-6), bias_regularizer=l1(1e-6)))
model_2_l1.add(Activation('relu'))
model_2_l1.add(Dense(num_classes, kernel_regularizer = l1(1e-6), bias_regularizer=l1(1e-6)))
model_2_l1.add(Activation('softmax'))
model_2_l1.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_2_l1 = model_2_l1.fit(X_train, y_train, batch_size=200, epochs = 3, verbose = 1)
score_2_l1 = model_2_l1.evaluate(X_test, y_test, batch_size=200, verbose = 1)

l1_train_acc_2 = hist_2_l1.history.get('acc')[-1]
l1_test_acc_2 = score_2_l1[1]
```

```
Epoch 1/3
25000/25000 [=====] - 8s 338us/step - loss: 0.3470 - acc: 0.8552
Epoch 2/3
25000/25000 [=====] - 4s 173us/step - loss: 0.1427 - acc: 0.9525
Epoch 3/3
25000/25000 [=====] - 4s 175us/step - loss: 0.0596 - acc: 0.9826
25000/25000 [=====] - 4s 161us/step
```

```
In [72]: print('The accurate rate of the model with two hidden layers and l1 regularization on training dataset is')
print(l1_train_acc_2)
print('The accurate rate of the model with two hidden layers and l1 regularization on test dataset is')
print(l1_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and l1 regularization on training dataset is
0.9826000151634217
The accurate rate of the model with two hidden layers and l1 regularization on test dataset is
0.8631199975013732
```

```

In [75]: lambda_list = [1e-6,1e-5,1e-4,1e-3,1e-2,1e-1]
loss_train_2_l2 = []
acc_train_2_l2 = []
loss_val_2_l2 = []
acc_val_2_l2 = []
for par in lambda_list:
    model_2_l2 = Sequential()
    model_2_l2.add(Dense(50,input_shape = (max_num,), kernel_regularizer = l2(
par), bias_regularizer=l2(par)))
    model_2_l2.add(Activation('relu'))
    model_2_l2.add(Dense(50, kernel_regularizer = l2(par), bias_regularizer=l2
(par)))
    model_2_l2.add(Activation('relu'))
    model_2_l2.add(Dense(num_classes, kernel_regularizer = l2(par), bias_regul
arizer=l2(par)))
    model_2_l2.add(Activation('softmax'))
    model_2_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam',
metrics = ['accuracy'])
    loss_val_cv = 0
    acc_val_cv = 0
    loss_train_cv = 0
    acc_train_cv = 0
    for train_cv_index, val_index in kf.split(X_train):
        X_train_cv = X_train[train_cv_index]
        y_train_cv = y_train[train_cv_index]
        X_val_cv = X_train[val_index]
        y_val_cv = y_train[val_index]
        hist = model_2_l2.fit(X_train_cv,y_train_cv,batch_size=200, epochs = 3
)

        loss_train_cv = loss_train_cv + hist.history.get('loss')[-1]
        acc_train_cv = acc_train_cv + hist.history.get('acc')[-1]
        score_val_cv = model_2_l2.evaluate(X_val_cv,y_val_cv, batch_size=200,
verbose = 1)
        loss_val_cv = loss_val_cv + score_val_cv[0]
        acc_val_cv = acc_val_cv + score_val_cv[1]
    loss_val_2_l2.append(loss_val_cv/3)
    acc_val_2_l2.append(acc_val_cv/3)
    loss_train_2_l2.append(loss_train_cv/3)
    acc_train_2_l2.append(acc_train_cv/3)

```

```
Epoch 1/3
16666/16666 [=====] - 7s 414us/step - loss: 0.3684 -
acc: 0.8426
Epoch 2/3
16666/16666 [=====] - 3s 172us/step - loss: 0.1094 -
acc: 0.9639
Epoch 3/3
16666/16666 [=====] - 3s 172us/step - loss: 0.0346 -
acc: 0.9913
8334/8334 [=====] - 2s 298us/step
Epoch 1/3
16667/16667 [=====] - 3s 176us/step - loss: 0.1944 -
acc: 0.9338
Epoch 2/3
16667/16667 [=====] - 3s 175us/step - loss: 0.0430 -
acc: 0.9873
Epoch 3/3
16667/16667 [=====] - 3s 173us/step - loss: 0.0093 -
acc: 0.9977
8333/8333 [=====] - 1s 100us/step
Epoch 1/3
16667/16667 [=====] - 3s 175us/step - loss: 0.0411 -
acc: 0.9859
Epoch 2/3
16667/16667 [=====] - 3s 174us/step - loss: 0.0106 -
acc: 0.9978
Epoch 3/3
16667/16667 [=====] - 3s 175us/step - loss: 0.0028 -
acc: 0.9994
8333/8333 [=====] - 1s 99us/step
Epoch 1/3
16666/16666 [=====] - 7s 419us/step - loss: 0.3726 -
acc: 0.8359
Epoch 2/3
16666/16666 [=====] - 3s 177us/step - loss: 0.1180 -
acc: 0.9606
Epoch 3/3
16666/16666 [=====] - 3s 174us/step - loss: 0.0384 -
acc: 0.9899
8334/8334 [=====] - 3s 306us/step
Epoch 1/3
16667/16667 [=====] - 3s 175us/step - loss: 0.1952 -
acc: 0.9341
Epoch 2/3
16667/16667 [=====] - 3s 174us/step - loss: 0.0457 -
acc: 0.9872
Epoch 3/3
16667/16667 [=====] - 3s 175us/step - loss: 0.0114 -
acc: 0.9985
8333/8333 [=====] - 1s 100us/step
Epoch 1/3
16667/16667 [=====] - 3s 176us/step - loss: 0.0434 -
acc: 0.9857
Epoch 2/3
16667/16667 [=====] - 3s 175us/step - loss: 0.0125 -
acc: 0.9984
Epoch 3/3
```

```
16667/16667 [=====] - 3s 175us/step - loss: 0.0059 -  
acc: 0.9996  
8333/8333 [=====] - 1s 98us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 426us/step - loss: 0.3803 -  
acc: 0.8379  
Epoch 2/3  
16666/16666 [=====] - 3s 174us/step - loss: 0.1244 -  
acc: 0.9643  
Epoch 3/3  
16666/16666 [=====] - 3s 178us/step - loss: 0.0503 -  
acc: 0.9917  
8334/8334 [=====] - 3s 309us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 175us/step - loss: 0.2144 -  
acc: 0.9367  
Epoch 2/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.0651 -  
acc: 0.9880  
Epoch 3/3  
16667/16667 [=====] - 3s 175us/step - loss: 0.0322 -  
acc: 0.9981  
8333/8333 [=====] - 1s 100us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 176us/step - loss: 0.0654 -  
acc: 0.9855  
Epoch 2/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.0382 -  
acc: 0.9968  
Epoch 3/3  
16667/16667 [=====] - 3s 174us/step - loss: 0.0293 -  
acc: 0.9995  
8333/8333 [=====] - 1s 101us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 432us/step - loss: 0.4978 -  
acc: 0.8374  
Epoch 2/3  
16666/16666 [=====] - 3s 176us/step - loss: 0.2396 -  
acc: 0.9585  
Epoch 3/3  
16666/16666 [=====] - 3s 173us/step - loss: 0.1606 -  
acc: 0.9849  
8334/8334 [=====] - 3s 320us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 189us/step - loss: 0.3119 -  
acc: 0.9294  
Epoch 2/3  
16667/16667 [=====] - 3s 184us/step - loss: 0.1693 -  
acc: 0.9826  
Epoch 3/3  
16667/16667 [=====] - 3s 183us/step - loss: 0.1187 -  
acc: 0.9964  
8333/8333 [=====] - 1s 105us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 187us/step - loss: 0.1946 -  
acc: 0.9657  
Epoch 2/3
```

```
16667/16667 [=====] - 3s 178us/step - loss: 0.1472 -  
acc: 0.9900  
Epoch 3/3  
16667/16667 [=====] - 3s 177us/step - loss: 0.1209 -  
acc: 0.9975  
8333/8333 [=====] - 1s 103us/step  
Epoch 1/3  
16666/16666 [=====] - 7s 427us/step - loss: 1.1380 -  
acc: 0.8364  
Epoch 2/3  
16666/16666 [=====] - 3s 175us/step - loss: 0.5568 -  
acc: 0.9267  
Epoch 3/3  
16666/16666 [=====] - 3s 173us/step - loss: 0.4381 -  
acc: 0.9302  
8334/8334 [=====] - 3s 316us/step  
Epoch 1/3  
16667/16667 [=====] - 3s 175us/step - loss: 0.4591 -  
acc: 0.9032  
Epoch 2/3  
16667/16667 [=====] - 3s 176us/step - loss: 0.3410 -  
acc: 0.9408  
Epoch 3/3  
16667/16667 [=====] - 3s 175us/step - loss: 0.3277 -  
acc: 0.9428  
8333/8333 [=====] - 1s 99us/step  
Epoch 1/3  
16667/16667 [=====] - 4s 233us/step - loss: 0.3976 -  
acc: 0.9152  
Epoch 2/3  
16667/16667 [=====] - 4s 215us/step - loss: 0.3011 -  
acc: 0.9511  
Epoch 3/3  
16667/16667 [=====] - 3s 187us/step - loss: 0.2914 -  
acc: 0.9526  
8333/8333 [=====] - 1s 120us/step  
Epoch 1/3  
16666/16666 [=====] - 11s 644us/step - loss: 4.8265  
- acc: 0.8372  
Epoch 2/3  
16666/16666 [=====] - 3s 200us/step - loss: 1.6492 -  
acc: 0.8780  
Epoch 3/3  
16666/16666 [=====] - 4s 250us/step - loss: 0.9525 -  
acc: 0.8799  
8334/8334 [=====] - 3s 402us/step  
Epoch 1/3  
16667/16667 [=====] - 4s 221us/step - loss: 0.7578 -  
acc: 0.8780  
Epoch 2/3  
16667/16667 [=====] - 4s 220us/step - loss: 0.7033 -  
acc: 0.8783  
Epoch 3/3  
16667/16667 [=====] - 4s 235us/step - loss: 0.6863 -  
acc: 0.8861  
8333/8333 [=====] - 1s 154us/step  
Epoch 1/3
```

```
16667/16667 [=====] - 4s 217us/step - loss: 0.6862 -  
acc: 0.8844  
Epoch 2/3  
16667/16667 [=====] - 3s 194us/step - loss: 0.6809 -  
acc: 0.8838  
Epoch 3/3  
16667/16667 [=====] - 3s 191us/step - loss: 0.6703 -  
acc: 0.8899  
8333/8333 [=====] - 1s 109us/step
```

```
In [76]: l2_2_reg_cv = pd.DataFrame({'loss_train': loss_train_2_l2,  
                                     'loss_val': loss_val_2_l2,  
                                     'acc_train': acc_train_2_l2,  
                                     'acc_val': acc_val_2_l2,  
                                     'lambda': lambda_list}).set_index('lambda')  
  
l2_2_reg_cv
```

Out[76]:

	loss_train	loss_val	acc_train	acc_val
lambda				
0.000001	0.015571	0.182387	0.99614	0.950123
0.000010	0.018547	0.176901	0.99598	0.950683
0.000100	0.037256	0.205451	0.99642	0.951043
0.001000	0.133404	0.314742	0.99294	0.930482
0.010000	0.352425	0.496134	0.94188	0.877400
0.100000	0.769707	0.733524	0.88528	0.873720

```
In [78]: # indicating by the result of l2_2_reg_cv data frame, will set the lambda as 1
e-4 for l2 norm regularization
model_2_l2 = Sequential()
model_2_l2.add(Dense(50,input_shape = (max_num,), kernel_regularizer = l2(1e-4
), bias_regularizer=l2(1e-4)))
model_2_l2.add(Activation('relu'))
model_2_l2.add(Dense(50, kernel_regularizer = l2(1e-4), bias_regularizer=l2(1e
-4)))
model_2_l2.add(Activation('relu'))
model_2_l2.add(Dense(num_classes, kernel_regularizer = l2(1e-4), bias_regulari
zer=l2(1e-4)))
model_2_l2.add(Activation('softmax'))
model_2_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metr
ics = ['accuracy'])

hist_2_l2 = model_2_l2.fit(X_train,y_train, batch_size=200, epochs = 3, verbos
e = 1)
score_2_l2 = model_2_l2.evaluate(X_test,y_test, batch_size=200, verbose = 1)

l2_train_acc_2 = hist_2_l2.history.get('acc')[-1]
l2_test_acc_2 = score_2_l2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 9s 342us/step - loss: 0.3527 -
acc: 0.8592
Epoch 2/3
25000/25000 [=====] - 4s 164us/step - loss: 0.1344 -
acc: 0.9594
Epoch 3/3
25000/25000 [=====] - 4s 163us/step - loss: 0.0587 -
acc: 0.9888
25000/25000 [=====] - 4s 167us/step
```

```
In [79]: print('The accurate rate of the model with two hidden layers and l2 regulariza
tion on training dataset is')
print(l2_train_acc_2)
print('The accurate rate of the model with two hidden layers and l2 regulariza
tion on test dataset is')
print(l2_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and l2 regularization o
n training dataset is
0.988760009765625
The accurate rate of the model with two hidden layers and l2 regularization o
n test dataset is
0.862720000743866
```



```
In [80]: # indicating by the result of l2_2_reg_cv data frame, will set the lambda as 1
e-5 for l2 norm regularization
model_2_l2 = Sequential()
model_2_l2.add(Dense(50,input_shape = (max_num,), kernel_regularizer = l2(1e-5
), bias_regularizer=l2(1e-5)))
model_2_l2.add(Activation('relu'))
model_2_l2.add(Dense(50, kernel_regularizer = l2(1e-5), bias_regularizer=l2(1e
-5)))
model_2_l2.add(Activation('relu'))
model_2_l2.add(Dense(num_classes, kernel_regularizer = l2(1e-5), bias_regulari
zer=l2(1e-5)))
model_2_l2.add(Activation('softmax'))
model_2_l2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metr
ics = ['accuracy'])

hist_2_l2 = model_2_l2.fit(X_train,y_train, batch_size=200, epochs = 3, verbos
e = 1)
score_2_l2 = model_2_l2.evaluate(X_test,y_test, batch_size=200, verbose = 1)

l2_train_acc_2 = hist_2_l2.history.get('acc')[-1]
l2_test_acc_2 = score_2_l2[1]
```

```
Epoch 1/3
25000/25000 [=====] - 8s 340us/step - loss: 0.3385 -
acc: 0.8563
Epoch 2/3
25000/25000 [=====] - 4s 167us/step - loss: 0.1322 -
acc: 0.9526
Epoch 3/3
25000/25000 [=====] - 4s 166us/step - loss: 0.0576 -
acc: 0.9823
25000/25000 [=====] - 4s 165us/step
```

```
In [81]: print('The accurate rate of the model with two hidden layers and l2 regulariza
tion on training dataset is')
print(l2_train_acc_2)
print('The accurate rate of the model with two hidden layers and l2 regulariza
tion on test dataset is')
print(l2_test_acc_2)
```

```
The accurate rate of the model with two hidden layers and l2 regularization o
n training dataset is
0.9823200106620789
The accurate rate of the model with two hidden layers and l2 regularization o
n test dataset is
0.8600399994850159
```

In [83]: `print(model_2_12.summary())`

Layer (type)	Output Shape	Param #
dense_258 (Dense)	(None, 50)	500050
activation_210 (Activation)	(None, 50)	0
dense_259 (Dense)	(None, 50)	2550
activation_211 (Activation)	(None, 50)	0
dense_260 (Dense)	(None, 2)	102
activation_212 (Activation)	(None, 2)	0

=====
 Total params: 502,702
 Trainable params: 502,702
 Non-trainable params: 0
 =====
 None

In []:

augment the training dataset

In [3]: `data = np.concatenate((X_train, X_test), axis=0)`
`targets = np.concatenate((y_train, y_test), axis=0)`

In [4]: `from sklearn.model_selection import train_test_split`
`X_train, X_test, y_train, y_test = train_test_split(data, targets, test_size=0.2)`

```
In [5]: num_classes = max(y_train) + 1
        print(num_classes)

        from tensorflow.keras.preprocessing.text import Tokenizer
        from tensorflow.keras.utils import to_categorical

        max_num = 10000
        tokenizer = Tokenizer(num_words=max_num)
        tokenizer.fit_on_sequences(X_train)
        X_train = tokenizer.sequences_to_matrix(X_train, mode = 'tfidf')
        X_test = tokenizer.sequences_to_matrix(X_test, mode = 'tfidf')

        y_train = to_categorical(y_train, num_classes)
        y_test = to_categorical(y_test, num_classes)
        y_train.shape
```

2

Out[5]: KFold(n_splits=3, random_state=None, shuffle=True)

```
In [6]: # the baseline to compare with
        model_overfitting = Sequential()
        model_overfitting.add(Dense(250,input_shape = (max_num,)))
        model_overfitting.add(Activation('relu'))
        model_overfitting.add(Dense(num_classes))
        model_overfitting.add(Activation('softmax'))
        model_overfitting.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

        hist_overfitting = model_overfitting.fit(X_train,y_train, batch_size=200, epochs = 3, verbose = 1)
        score_overfitting = model_overfitting.evaluate(X_test,y_test, batch_size=200, verbose = 1)

        base_train_acc = hist_overfitting.history.get('acc')[-1]
        base_test_acc = score_overfitting[1]

        print('The accurate rate of the overfitting model on training dataset is')
        print(base_train_acc)
        print('The accurate rate of the overfitting model on test dataset is')
        print(base_test_acc)
```

```
Epoch 1/3
40000/40000 [=====] - 22s 553us/step - loss: 0.3048
- acc: 0.8711
Epoch 2/3
40000/40000 [=====] - 21s 518us/step - loss: 0.0810
- acc: 0.9741
Epoch 3/3
40000/40000 [=====] - 20s 510us/step - loss: 0.0160
- acc: 0.9971
10000/10000 [=====] - 2s 232us/step
The accurate rate of the overfitting model on training dataset is
0.9971000027656555
The accurate rate of the overfitting model on test dataset is
0.8960999977588654
```

```
In [7]: # baseline model with two hidden layers
model_overfitting_2 = Sequential()
model_overfitting_2.add(Dense(50,input_shape = (max_num,)))
model_overfitting_2.add(Activation('relu'))
model_overfitting_2.add(Dense(50))
model_overfitting_2.add(Activation('relu'))
model_overfitting_2.add(Dense(num_classes))
model_overfitting_2.add(Activation('softmax'))
model_overfitting_2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

hist_overfitting_2 = model_overfitting_2.fit(X_train,y_train, batch_size=200,
epochs = 3, verbose = 1)
score_overfitting_2 = model_overfitting_2.evaluate(X_test,y_test, batch_size=200, verbose = 1)

base_train_acc_2 = hist_overfitting_2.history.get('acc')[-1]
base_test_acc_2 = score_overfitting_2[1]

print('The accurate rate of the overfitting model with two hidden layers on training dataset is')
print(base_train_acc_2)
print('The accurate rate of the overfitting model with two hidden layers on test dataset is')
print(base_test_acc_2)
```

```
Epoch 1/3
40000/40000 [=====] - 10s 258us/step - loss: 0.3205
- acc: 0.8643
Epoch 2/3
40000/40000 [=====] - 9s 227us/step - loss: 0.1456 -
acc: 0.9445
Epoch 3/3
40000/40000 [=====] - 9s 233us/step - loss: 0.0584 -
acc: 0.9799
10000/10000 [=====] - 2s 169us/step
The accurate rate of the overfitting model with two hidden layers on training
dataset is
0.9799250122904778
The accurate rate of the overfitting model with two hidden layers on test dat
aset is
0.8881999969482421
```