Homework5

Ze Chen Chunran Yao

```
In [1]:
```

```
from keras.models import Sequential, Model
from keras.layers import Dense, Activation, Dropout, BatchNormalization, Input,
from keras import regularizers, Model
from keras.datasets import fashion mnist
import keras
from sklearn.datasets import load iris
from sklearn.model selection import GridSearchCV
from sklearn.model selection import StratifiedShuffleSplit
from keras.wrappers.scikit learn import KerasClassifier
from sklearn.model selection import train test split
import pandas as pd
import numpy as np
import glob
import os
from keras.layers import Conv2D, MaxPooling2D, Flatten
import matplotlib.pyplot as plt
```

Using TensorFlow backend.

Task 1

```
In [0]:
```

```
def make_model(optimizer = 'adam', hidden_size = 32,regularization_strength = 0.
01):
    model = Sequential()
    model.add(Dense(hidden_size, activation='relu', input_dim=4,kernel_regularizer
=regularizers.12(regularization_strength)))
    model.add(Dense(hidden_size, activation='relu', kernel_regularizer=regularizer
s.12(regularization_strength)))
    model.add(Dense(3, activation='softmax'))

    model.compile(optimizer=optimizer,loss="categorical_crossentropy",metrics=['accuracy'])
    return model
```

```
In [0]:
```

```
clf = KerasClassifier(make_model)
```

```
In [0]:
```

```
iris_X, iris_y = load_iris(return_X_y=True)
X_train, X_test, y_train,y_test = train_test_split(iris_X,iris_y)
```

```
In [0]:
```

ss = StratifiedShuffleSplit(n_splits=5, test_size=0.25)

In [0]:

```
Epoch 1/10
84/84 [============= ] - 19s 226ms/step - loss: 1.19
45 - acc: 0.3214
Epoch 2/10
84/84 [============== ] - 0s 258us/step - loss: 1.086
6 - acc: 0.3690
Epoch 3/10
84/84 [============= ] - 0s 262us/step - loss: 0.994
2 - acc: 0.6548
Epoch 4/10
84/84 [============= ] - 0s 245us/step - loss: 0.919
9 - acc: 0.6667
Epoch 5/10
84/84 [============== ] - 0s 239us/step - loss: 0.862
5 - acc: 0.6667
Epoch 6/10
84/84 [============= ] - 0s 258us/step - loss: 0.821
5 - acc: 0.6667
Epoch 7/10
84/84 [============ ] - 0s 252us/step - loss: 0.787
3 - acc: 0.6667
Epoch 8/10
84/84 [============= ] - 0s 266us/step - loss: 0.763
1 - acc: 0.6667
Epoch 9/10
84/84 [============== ] - 0s 266us/step - loss: 0.738
2 - acc: 0.6429
Epoch 10/10
84/84 [============ ] - 0s 241us/step - loss: 0.716
1 - acc: 0.7024
28/28 [========= ] - 8s 277ms/step
84/84 [======== ] - 0s 255us/step
Epoch 1/10
84/84 [============ ] - 18s 210ms/step - loss: 1.34
79 - acc: 0.2976
Epoch 2/10
84/84 [============== ] - 0s 249us/step - loss: 1.159
9 - acc: 0.3214
Epoch 3/10
84/84 [============== ] - 0s 243us/step - loss: 1.035
3 - acc: 0.4524
Epoch 4/10
84/84 [=========== ] - 0s 247us/step - loss: 0.929
1 - acc: 0.7143
Epoch 5/10
84/84 [============= ] - 0s 230us/step - loss: 0.868
9 - acc: 0.7024
Epoch 6/10
84/84 [============= ] - 0s 249us/step - loss: 0.820
4 - acc: 0.7024
Epoch 7/10
84/84 [============= ] - 0s 255us/step - loss: 0.789
0 - acc: 0.7143
Epoch 8/10
84/84 [============= ] - 0s 235us/step - loss: 0.764
5 - acc: 0.7024
Epoch 9/10
84/84 [============= ] - 0s 245us/step - loss: 0.742
8 - acc: 0.7024
Epoch 10/10
84/84 [============== ] - 0s 269us/step - loss: 0.723
```

```
2 - acc: 0.7024
28/28 [======== ] - 8s 280ms/step
84/84 [======== ] - 0s 220us/step
Epoch 1/10
84/84 [============= ] - 18s 213ms/step - loss: 1.10
36 - acc: 0.3571
Epoch 2/10
84/84 [=========== ] - 0s 232us/step - loss: 0.925
7 - acc: 0.6310
Epoch 3/10
84/84 [============= ] - 0s 235us/step - loss: 0.835
1 - acc: 0.6786
Epoch 4/10
84/84 [============ ] - 0s 235us/step - loss: 0.783
0 - acc: 0.7143
Epoch 5/10
84/84 [============ ] - 0s 237us/step - loss: 0.751
0 - acc: 0.7857
Epoch 6/10
84/84 [============= ] - 0s 234us/step - loss: 0.723
7 - acc: 0.7738
Epoch 7/10
84/84 [============= ] - 0s 258us/step - loss: 0.693
3 - acc: 0.7619
Epoch 8/10
84/84 [============= ] - 0s 244us/step - loss: 0.661
8 - acc: 0.7619
Epoch 9/10
84/84 [============== ] - 0s 239us/step - loss: 0.637
3 - acc: 0.7143
Epoch 10/10
84/84 [============= ] - 0s 258us/step - loss: 0.611
8 - acc: 0.6905
28/28 [========= ] - 8s 297ms/step
84/84 [========] - 0s 276us/step
Epoch 1/10
84/84 [============ ] - 18s 216ms/step - loss: 1.35
15 - acc: 0.3333
Epoch 2/10
84/84 [============== ] - 0s 283us/step - loss: 1.246
3 - acc: 0.3333
Epoch 3/10
84/84 [============== ] - 0s 266us/step - loss: 1.158
6 - acc: 0.3095
Epoch 4/10
84/84 [============= ] - 0s 265us/step - loss: 1.087
4 - acc: 0.4405
Epoch 5/10
84/84 [============== ] - 0s 236us/step - loss: 1.028
7 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 237us/step - loss: 0.983
4 - acc: 0.6786
Epoch 7/10
84/84 [============== ] - 0s 239us/step - loss: 0.941
0 - acc: 0.7381
Epoch 8/10
84/84 [============== ] - 0s 229us/step - loss: 0.902
0 - acc: 0.8095
Epoch 9/10
84/84 [=============== ] - 0s 247us/step - loss: 0.861
```

```
4 - acc: 0.8571
Epoch 10/10
84/84 [========== ] - 0s 263us/step - loss: 0.828
7 - acc: 0.8095
28/28 [======== ] - 8s 283ms/step
84/84 [======== ] - 0s 268us/step
Epoch 1/10
84/84 [============ ] - 18s 216ms/step - loss: 1.01
43 - acc: 0.2381
Epoch 2/10
84/84 [============== ] - 0s 252us/step - loss: 0.917
7 - acc: 0.4881
Epoch 3/10
84/84 [============ ] - 0s 241us/step - loss: 0.862
9 - acc: 0.6667
Epoch 4/10
84/84 [========= ] - 0s 231us/step - loss: 0.826
4 - acc: 0.6905
Epoch 5/10
84/84 [============= ] - 0s 253us/step - loss: 0.785
0 - acc: 0.8095
Epoch 6/10
84/84 [============= ] - 0s 236us/step - loss: 0.746
3 - acc: 0.8214
Epoch 7/10
84/84 [============== ] - 0s 232us/step - loss: 0.711
4 - acc: 0.8095
Epoch 8/10
84/84 [============== ] - 0s 234us/step - loss: 0.685
1 - acc: 0.8214
Epoch 9/10
84/84 [============= ] - 0s 260us/step - loss: 0.664
6 - acc: 0.8333
Epoch 10/10
84/84 [============= ] - 0s 247us/step - loss: 0.644
9 - acc: 0.8333
28/28 [======== ] - 8s 284ms/step
84/84 [======== ] - 0s 258us/step
Epoch 1/10
84/84 [============== ] - 18s 217ms/step - loss: 2.05
59 - acc: 0.3214
Epoch 2/10
84/84 [============== ] - 0s 232us/step - loss: 1.858
4 - acc: 0.3214
Epoch 3/10
84/84 [============= ] - 0s 267us/step - loss: 1.657
5 - acc: 0.3214
Epoch 4/10
84/84 [============ ] - 0s 245us/step - loss: 1.539
2 - acc: 0.3214
Epoch 5/10
84/84 [============ ] - 0s 228us/step - loss: 1.455
3 - acc: 0.3214
Epoch 6/10
84/84 [=============== ] - 0s 244us/step - loss: 1.402
8 - acc: 0.3571
Epoch 7/10
84/84 [============= ] - 0s 246us/step - loss: 1.349
4 - acc: 0.5357
Epoch 8/10
84/84 [=========== ] - 0s 229us/step - loss: 1.309
```

```
3 - acc: 0.6429
Epoch 9/10
84/84 [============ ] - 0s 259us/step - loss: 1.275
5 - acc: 0.6786
Epoch 10/10
84/84 [============= ] - 0s 247us/step - loss: 1.240
2 - acc: 0.6786
28/28 [======== ] - 9s 304ms/step
84/84 [=======] - 0s 263us/step
Epoch 1/10
64 - acc: 0.3452
Epoch 2/10
84/84 [============ ] - 0s 243us/step - loss: 1.572
3 - acc: 0.5238
Epoch 3/10
3 - acc: 0.6429
Epoch 4/10
84/84 [============= ] - 0s 233us/step - loss: 1.347
4 - acc: 0.6667
Epoch 5/10
84/84 [============== ] - 0s 229us/step - loss: 1.262
0 - acc: 0.6667
Epoch 6/10
84/84 [============= ] - 0s 234us/step - loss: 1.207
2 - acc: 0.7619
Epoch 7/10
9 - acc: 0.9524
Epoch 8/10
84/84 [============== ] - 0s 234us/step - loss: 1.125
2 - acc: 0.8214
Epoch 9/10
84/84 [============= ] - 0s 267us/step - loss: 1.090
2 - acc: 0.7857
Epoch 10/10
84/84 [============= ] - 0s 237us/step - loss: 1.058
6 - acc: 0.8333
28/28 [========= ] - 8s 288ms/step
84/84 [======== ] - 0s 274us/step
Epoch 1/10
84/84 [============= ] - 19s 224ms/step - loss: 1.73
37 - acc: 0.3452
Epoch 2/10
84/84 [============= ] - 0s 263us/step - loss: 1.563
8 - acc: 0.3452
Epoch 3/10
84/84 [============= ] - 0s 251us/step - loss: 1.435
4 - acc: 0.3571
Epoch 4/10
84/84 [============ ] - 0s 248us/step - loss: 1.358
2 - acc: 0.6667
Epoch 5/10
84/84 [============== ] - 0s 236us/step - loss: 1.315
4 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 236us/step - loss: 1.286
9 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 243us/step - loss: 1.260
```

```
5 - acc: 0.6786
Epoch 8/10
84/84 [========== ] - 0s 235us/step - loss: 1.234
3 - acc: 0.6786
Epoch 9/10
84/84 [============== ] - 0s 254us/step - loss: 1.208
9 - acc: 0.6786
Epoch 10/10
84/84 [============= ] - 0s 253us/step - loss: 1.184
7 - acc: 0.6786
28/28 [======== ] - 8s 289ms/step
84/84 [=======] - 0s 303us/step
Epoch 1/10
84/84 [========= ] - 19s 222ms/step - loss: 1.65
74 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 290us/step - loss: 1.568
4 - acc: 0.3810
Epoch 3/10
84/84 [============= ] - 0s 276us/step - loss: 1.499
0 - acc: 0.3333
Epoch 4/10
84/84 [============= ] - 0s 270us/step - loss: 1.449
1 - acc: 0.3333
Epoch 5/10
84/84 [============= ] - 0s 290us/step - loss: 1.405
6 - acc: 0.3333
Epoch 6/10
84/84 [============== ] - 0s 290us/step - loss: 1.359
4 - acc: 0.3333
Epoch 7/10
84/84 [============== ] - 0s 387us/step - loss: 1.325
4 - acc: 0.4643
Epoch 8/10
84/84 [============== ] - 0s 317us/step - loss: 1.291
0 - acc: 0.6667
Epoch 9/10
84/84 [============== ] - 0s 271us/step - loss: 1.257
7 - acc: 0.6786
Epoch 10/10
84/84 [=========== ] - 0s 286us/step - loss: 1.226
4 - acc: 0.6786
28/28 [=======] - 9s 313ms/step
84/84 [======== ] - 0s 285us/step
Epoch 1/10
84/84 [============= ] - 19s 223ms/step - loss: 2.17
06 - acc: 0.3333
Epoch 2/10
84/84 [============== ] - 0s 242us/step - loss: 1.979
6 - acc: 0.3333
Epoch 3/10
84/84 [============ ] - 0s 243us/step - loss: 1.803
6 - acc: 0.3333
Epoch 4/10
84/84 [============== ] - 0s 238us/step - loss: 1.669
8 - acc: 0.3333
Epoch 5/10
84/84 [============= ] - 0s 245us/step - loss: 1.567
6 - acc: 0.3333
Epoch 6/10
84/84 [============== ] - 0s 263us/step - loss: 1.491
```

```
1 - acc: 0.3333
Epoch 7/10
84/84 [============ ] - 0s 255us/step - loss: 1.443
1 - acc: 0.3333
Epoch 8/10
84/84 [============== ] - 0s 243us/step - loss: 1.414
7 - acc: 0.3571
Epoch 9/10
84/84 [============== ] - 0s 261us/step - loss: 1.386
6 - acc: 0.5833
Epoch 10/10
84/84 [============= ] - 0s 274us/step - loss: 1.366
7 - acc: 0.6548
28/28 [======== ] - 8s 294ms/step
84/84 [======== ] - 0s 257us/step
Epoch 1/10
84/84 [=========== ] - 19s 224ms/step - loss: 3.96
25 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 251us/step - loss: 3.528
7 - acc: 0.3333
Epoch 3/10
84/84 [============== ] - 0s 251us/step - loss: 3.221
9 - acc: 0.3333
Epoch 4/10
84/84 [============== ] - 0s 240us/step - loss: 3.038
5 - acc: 0.3333
Epoch 5/10
8 - acc: 0.3571
Epoch 6/10
84/84 [============== ] - 0s 240us/step - loss: 2.882
4 - acc: 0.3214
Epoch 7/10
84/84 [============= ] - 0s 277us/step - loss: 2.827
2 - acc: 0.3214
Epoch 8/10
84/84 [============ ] - 0s 253us/step - loss: 2.755
3 - acc: 0.3214
Epoch 9/10
84/84 [============== ] - 0s 246us/step - loss: 2.682
7 - acc: 0.3214
Epoch 10/10
84/84 [============== ] - 0s 270us/step - loss: 2.610
3 - acc: 0.5000
28/28 [========= ] - 8s 295ms/step
84/84 [======== ] - 0s 256us/step
Epoch 1/10
84/84 [============== ] - 19s 228ms/step - loss: 3.45
29 - acc: 0.3214
Epoch 2/10
84/84 [============ ] - 0s 294us/step - loss: 3.287
7 - acc: 0.2976
Epoch 3/10
84/84 [============== ] - 0s 301us/step - loss: 3.136
4 - acc: 0.3333
Epoch 4/10
84/84 [============== ] - 0s 295us/step - loss: 3.010
4 - acc: 0.3333
Epoch 5/10
84/84 [============== ] - 0s 269us/step - loss: 2.906
```

```
5 - acc: 0.3333
Epoch 6/10
84/84 [============= ] - 0s 271us/step - loss: 2.818
1 - acc: 0.3333
Epoch 7/10
84/84 [============ ] - 0s 274us/step - loss: 2.746
8 - acc: 0.3333
Epoch 8/10
84/84 [============= ] - 0s 319us/step - loss: 2.673
3 - acc: 0.3333
Epoch 9/10
84/84 [============= ] - 0s 316us/step - loss: 2.604
1 - acc: 0.3333
Epoch 10/10
84/84 [============= ] - 0s 305us/step - loss: 2.542
9 - acc: 0.3333
28/28 [========= ] - 9s 324ms/step
84/84 [=======] - 0s 296us/step
Epoch 1/10
84/84 [============== ] - 19s 227ms/step - loss: 3.91
67 - acc: 0.3214
Epoch 2/10
84/84 [============= ] - 0s 271us/step - loss: 3.498
0 - acc: 0.3214
Epoch 3/10
84/84 [============= ] - 0s 264us/step - loss: 3.193
5 - acc: 0.3214
Epoch 4/10
84/84 [============ ] - 0s 279us/step - loss: 2.970
5 - acc: 0.3214
Epoch 5/10
84/84 [============== ] - 0s 285us/step - loss: 2.803
7 - acc: 0.5952
Epoch 6/10
84/84 [============== ] - 0s 250us/step - loss: 2.717
9 - acc: 0.7024
Epoch 7/10
84/84 [=========== ] - 0s 242us/step - loss: 2.643
7 - acc: 0.6786
Epoch 8/10
84/84 [============ ] - 0s 236us/step - loss: 2.584
4 - acc: 0.6786
Epoch 9/10
84/84 [============= ] - 0s 247us/step - loss: 2.527
1 - acc: 0.6786
Epoch 10/10
84/84 [=========== ] - 0s 242us/step - loss: 2.470
7 - acc: 0.6786
28/28 [========= ] - 8s 300ms/step
84/84 [======== ] - 0s 258us/step
Epoch 1/10
84/84 [=========== ] - 19s 228ms/step - loss: 3.44
92 - acc: 0.3452
Epoch 2/10
84/84 [============== ] - 0s 251us/step - loss: 3.198
7 - acc: 0.3452
Epoch 3/10
84/84 [============== ] - 0s 235us/step - loss: 3.021
5 - acc: 0.3571
Epoch 4/10
84/84 [============ ] - 0s 242us/step - loss: 2.890
```

```
3 - acc: 0.5952
Epoch 5/10
84/84 [============= ] - 0s 238us/step - loss: 2.798
6 - acc: 0.6548
Epoch 6/10
84/84 [============= ] - 0s 250us/step - loss: 2.727
3 - acc: 0.6190
Epoch 7/10
84/84 [============= ] - 0s 238us/step - loss: 2.657
2 - acc: 0.6310
Epoch 8/10
84/84 [============= ] - 0s 265us/step - loss: 2.591
2 - acc: 0.6786
Epoch 9/10
84/84 [============= ] - 0s 267us/step - loss: 2.528
7 - acc: 0.6786
Epoch 10/10
84/84 [============ ] - 0s 268us/step - loss: 2.466
0 - acc: 0.6786
28/28 [======== ] - 8s 301ms/step
84/84 [======== ] - 0s 247us/step
Epoch 1/10
84/84 [============= ] - 20s 237ms/step - loss: 3.40
06 - acc: 0.2857
Epoch 2/10
84/84 [============= ] - 0s 275us/step - loss: 3.185
9 - acc: 0.2976
Epoch 3/10
7 - acc: 0.3333
Epoch 4/10
84/84 [============== ] - 0s 277us/step - loss: 2.903
2 - acc: 0.5595
Epoch 5/10
84/84 [============= ] - 0s 305us/step - loss: 2.800
7 - acc: 0.6429
Epoch 6/10
84/84 [=========== ] - 0s 292us/step - loss: 2.713
6 - acc: 0.6190
Epoch 7/10
84/84 [============ ] - 0s 315us/step - loss: 2.640
5 - acc: 0.6548
Epoch 8/10
84/84 [============ ] - 0s 290us/step - loss: 2.565
2 - acc: 0.6667
Epoch 9/10
84/84 [============= ] - 0s 298us/step - loss: 2.492
6 - acc: 0.6786
Epoch 10/10
84/84 [============== ] - 0s 293us/step - loss: 2.421
7 - acc: 0.6429
28/28 [======== ] - 8s 303ms/step
84/84 [======== ] - 0s 254us/step
Epoch 1/10
84/84 [============ ] - 20s 234ms/step - loss: 5.17
95 - acc: 0.6786
Epoch 2/10
84/84 [============= ] - 0s 274us/step - loss: 4.881
7 - acc: 0.6786
Epoch 3/10
84/84 [=============== ] - 0s 272us/step - loss: 4.643
```

```
4 - acc: 0.6786
Epoch 4/10
84/84 [========= ] - 0s 251us/step - loss: 4.454
7 - acc: 0.6786
Epoch 5/10
84/84 [============= ] - 0s 263us/step - loss: 4.301
5 - acc: 0.6786
Epoch 6/10
84/84 [============= ] - 0s 264us/step - loss: 4.197
4 - acc: 0.7381
Epoch 7/10
84/84 [============== ] - 0s 280us/step - loss: 4.121
2 - acc: 0.7500
Epoch 8/10
84/84 [============ ] - 0s 259us/step - loss: 4.041
5 - acc: 0.7262
Epoch 9/10
84/84 [=========== ] - 0s 263us/step - loss: 3.949
9 - acc: 0.7262
Epoch 10/10
84/84 [============== ] - 0s 246us/step - loss: 3.853
8 - acc: 0.8810
28/28 [======== ] - 9s 313ms/step
84/84 [========] - 0s 261us/step
Epoch 1/10
84/84 [============= ] - 20s 233ms/step - loss: 5.70
33 - acc: 0.3214
Epoch 2/10
84/84 [============== ] - 0s 267us/step - loss: 5.452
7 - acc: 0.4643
Epoch 3/10
84/84 [============= ] - 0s 258us/step - loss: 5.256
2 - acc: 0.3452
Epoch 4/10
84/84 [============= ] - 0s 243us/step - loss: 5.083
9 - acc: 0.3333
Epoch 5/10
84/84 [============ ] - 0s 258us/step - loss: 4.903
9 - acc: 0.3333
Epoch 6/10
84/84 [============= ] - 0s 245us/step - loss: 4.750
1 - acc: 0.3333
Epoch 7/10
84/84 [============== ] - 0s 245us/step - loss: 4.615
2 - acc: 0.3333
Epoch 8/10
84/84 [============= ] - 0s 252us/step - loss: 4.489
1 - acc: 0.3333
Epoch 9/10
84/84 [============ ] - 0s 267us/step - loss: 4.371
8 - acc: 0.4286
Epoch 10/10
84/84 [============= ] - 0s 248us/step - loss: 4.257
1 - acc: 0.6429
28/28 [======== ] - 9s 307ms/step
84/84 [========] - 0s 261us/step
Epoch 1/10
84/84 [============= ] - 20s 242ms/step - loss: 5.15
37 - acc: 0.3333
Epoch 2/10
84/84 [============== ] - 0s 251us/step - loss: 4.999
```

```
9 - acc: 0.3333
Epoch 3/10
84/84 [========= ] - 0s 265us/step - loss: 4.869
3 - acc: 0.3333
Epoch 4/10
84/84 [============= ] - 0s 262us/step - loss: 4.749
8 - acc: 0.1905
Epoch 5/10
84/84 [============= ] - 0s 259us/step - loss: 4.635
3 - acc: 0.1190
Epoch 6/10
84/84 [============== ] - 0s 258us/step - loss: 4.531
1 - acc: 0.0952
Epoch 7/10
84/84 [============= ] - 0s 248us/step - loss: 4.427
9 - acc: 0.3452
Epoch 8/10
84/84 [=========== ] - 0s 253us/step - loss: 4.323
2 - acc: 0.4643
Epoch 9/10
84/84 [============= ] - 0s 250us/step - loss: 4.221
7 - acc: 0.4405
Epoch 10/10
84/84 [============== ] - 0s 281us/step - loss: 4.124
4 - acc: 0.3929
28/28 [======== ] - 9s 313ms/step
84/84 [======== ] - 0s 273us/step
Epoch 1/10
84/84 [============== ] - 20s 239ms/step - loss: 5.69
42 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 266us/step - loss: 5.413
1 - acc: 0.3333
Epoch 3/10
84/84 [============= ] - 0s 250us/step - loss: 5.197
0 - acc: 0.3333
Epoch 4/10
84/84 [=========== ] - 0s 248us/step - loss: 4.996
1 - acc: 0.3333
Epoch 5/10
84/84 [============ ] - 0s 250us/step - loss: 4.835
8 - acc: 0.3333
Epoch 6/10
84/84 [=========== ] - 0s 252us/step - loss: 4.688
7 - acc: 0.3333
Epoch 7/10
84/84 [========== ] - 0s 265us/step - loss: 4.569
1 - acc: 0.3333
Epoch 8/10
84/84 [============ ] - 0s 280us/step - loss: 4.459
8 - acc: 0.3333
Epoch 9/10
84/84 [============= ] - 0s 253us/step - loss: 4.356
0 - acc: 0.3333
Epoch 10/10
84/84 [============== ] - 0s 251us/step - loss: 4.253
8 - acc: 0.3333
28/28 [======== ] - 9s 310ms/step
84/84 [========] - 0s 260us/step
Epoch 1/10
84/84 [============ ] - 20s 237ms/step - loss: 5.38
```

```
69 - acc: 0.3452
Epoch 2/10
84/84 [========= ] - 0s 272us/step - loss: 5.059
3 - acc: 0.3452
Epoch 3/10
84/84 [============= ] - 0s 242us/step - loss: 4.781
6 - acc: 0.3690
Epoch 4/10
84/84 [============= ] - 0s 269us/step - loss: 4.548
6 - acc: 0.5952
Epoch 5/10
84/84 [============= ] - 0s 276us/step - loss: 4.366
0 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 249us/step - loss: 4.216
6 - acc: 0.6786
Epoch 7/10
84/84 [=========== ] - 0s 269us/step - loss: 4.087
3 - acc: 0.6786
Epoch 8/10
84/84 [============= ] - 0s 282us/step - loss: 3.975
4 - acc: 0.6786
Epoch 9/10
84/84 [============== ] - 0s 267us/step - loss: 3.885
5 - acc: 0.7143
Epoch 10/10
84/84 [============== ] - 0s 259us/step - loss: 3.801
8 - acc: 0.8690
28/28 [======== ] - 9s 319ms/step
84/84 [======== ] - 0s 280us/step
Epoch 1/10
84/84 [============= ] - 20s 244ms/step - loss: 1.28
80 - acc: 0.3452
Epoch 2/10
84/84 [============= ] - 0s 243us/step - loss: 1.046
9 - acc: 0.4881
Epoch 3/10
84/84 [============ ] - 0s 247us/step - loss: 0.940
8 - acc: 0.7857
Epoch 4/10
84/84 [============= ] - 0s 264us/step - loss: 0.886
8 - acc: 0.7143
Epoch 5/10
84/84 [============== ] - 0s 244us/step - loss: 0.859
8 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 250us/step - loss: 0.821
3 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 244us/step - loss: 0.777
3 - acc: 0.6786
Epoch 8/10
84/84 [============= ] - 0s 249us/step - loss: 0.739
6 - acc: 0.8333
Epoch 9/10
84/84 [============= ] - 0s 246us/step - loss: 0.714
1 - acc: 0.9524
Epoch 10/10
84/84 [=========== ] - 0s 256us/step - loss: 0.685
5 - acc: 0.9762
28/28 [======== ] - 9s 314ms/step
```

```
84/84 [========= ] - 0s 261us/step
Epoch 1/10
84/84 [============= ] - 20s 239ms/step - loss: 1.52
76 - acc: 0.3214
Epoch 2/10
84/84 [============== ] - 0s 276us/step - loss: 1.243
5 - acc: 0.3214
Epoch 3/10
84/84 [============== ] - 0s 240us/step - loss: 1.084
6 - acc: 0.5000
Epoch 4/10
84/84 [============== ] - 0s 245us/step - loss: 1.006
5 - acc: 0.5833
Epoch 5/10
84/84 [============== ] - 0s 263us/step - loss: 0.976
5 - acc: 0.5833
Epoch 6/10
84/84 [=========== ] - 0s 245us/step - loss: 0.955
7 - acc: 0.6190
Epoch 7/10
84/84 [============= ] - 0s 236us/step - loss: 0.921
2 - acc: 0.6548
Epoch 8/10
84/84 [============== ] - 0s 259us/step - loss: 0.878
0 - acc: 0.6786
Epoch 9/10
84/84 [============== ] - 0s 251us/step - loss: 0.837
3 - acc: 0.6786
Epoch 10/10
84/84 [============ ] - 0s 280us/step - loss: 0.800
1 - acc: 0.6786
28/28 [======== ] - 9s 316ms/step
84/84 [======== ] - 0s 253us/step
Epoch 1/10
84/84 [============= ] - 21s 252ms/step - loss: 1.18
10 - acc: 0.3333
Epoch 2/10
84/84 [============ ] - 0s 332us/step - loss: 1.097
1 - acc: 0.4405
Epoch 3/10
84/84 [============ ] - 0s 290us/step - loss: 1.028
1 - acc: 0.5000
Epoch 4/10
84/84 [============== ] - 0s 294us/step - loss: 0.956
4 - acc: 0.5238
Epoch 5/10
84/84 [============= ] - 0s 285us/step - loss: 0.896
9 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 281us/step - loss: 0.842
4 - acc: 0.6786
Epoch 7/10
84/84 [============== ] - 0s 282us/step - loss: 0.791
2 - acc: 0.6786
Epoch 8/10
84/84 [============= ] - 0s 287us/step - loss: 0.742
4 - acc: 0.7143
Epoch 9/10
84/84 [=========== ] - 0s 300us/step - loss: 0.701
5 - acc: 0.8333
Epoch 10/10
```

```
7 - acc: 0.8929
28/28 [=======] - 9s 333ms/step
84/84 [=======] - 0s 258us/step
Epoch 1/10
84/84 [============ ] - 20s 241ms/step - loss: 1.04
14 - acc: 0.4048
Epoch 2/10
84/84 [============= ] - 0s 292us/step - loss: 0.960
6 - acc: 0.8929
Epoch 3/10
84/84 [============== ] - 0s 257us/step - loss: 0.883
2 - acc: 0.7381
Epoch 4/10
84/84 [============= ] - 0s 263us/step - loss: 0.827
8 - acc: 0.6786
Epoch 5/10
84/84 [============ ] - 0s 253us/step - loss: 0.776
3 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 251us/step - loss: 0.723
8 - acc: 0.6786
Epoch 7/10
84/84 [============== ] - 0s 243us/step - loss: 0.674
1 - acc: 0.6905
Epoch 8/10
84/84 [============= ] - 0s 257us/step - loss: 0.633
5 - acc: 0.8214
Epoch 9/10
84/84 [============ ] - 0s 248us/step - loss: 0.596
2 - acc: 0.9524
Epoch 10/10
84/84 [============= ] - 0s 253us/step - loss: 0.562
3 - acc: 0.9881
28/28 [======== ] - 9s 320ms/step
84/84 [======== ] - 0s 252us/step
Epoch 1/10
84/84 [=========== ] - 20s 241ms/step - loss: 1.61
21 - acc: 0.1071
Epoch 2/10
84/84 [============ ] - 0s 288us/step - loss: 1.360
3 - acc: 0.1548
Epoch 3/10
84/84 [=============== ] - 0s 247us/step - loss: 1.152
2 - acc: 0.0833
Epoch 4/10
84/84 [============= ] - 0s 281us/step - loss: 1.083
1 - acc: 0.3095
Epoch 5/10
84/84 [=============== ] - 0s 244us/step - loss: 1.054
9 - acc: 0.3333
Epoch 6/10
84/84 [============= ] - 0s 264us/step - loss: 1.007
8 - acc: 0.3333
Epoch 7/10
84/84 [============= ] - 0s 246us/step - loss: 0.949
5 - acc: 0.3929
Epoch 8/10
84/84 [============ ] - 0s 246us/step - loss: 0.902
7 - acc: 0.6667
Epoch 9/10
```

```
84/84 [============== ] - 0s 246us/step - loss: 0.854
5 - acc: 0.8810
Epoch 10/10
84/84 [============== ] - 0s 248us/step - loss: 0.815
5 - acc: 0.9762
28/28 [============= ] - 9s 324ms/step
84/84 [======== ] - 0s 246us/step
Epoch 1/10
84/84 [============= ] - 21s 253ms/step - loss: 1.92
11 - acc: 0.6786
Epoch 2/10
84/84 [============== ] - 0s 273us/step - loss: 1.711
1 - acc: 0.6786
Epoch 3/10
84/84 [============= ] - 0s 245us/step - loss: 1.560
8 - acc: 0.6786
Epoch 4/10
84/84 [=========== ] - 0s 248us/step - loss: 1.454
2 - acc: 0.6905
Epoch 5/10
84/84 [============== ] - 0s 258us/step - loss: 1.375
1 - acc: 0.8095
Epoch 6/10
84/84 [============== ] - 0s 242us/step - loss: 1.311
4 - acc: 0.8333
Epoch 7/10
84/84 [============= ] - 0s 248us/step - loss: 1.254
8 - acc: 0.8333
Epoch 8/10
84/84 [============ ] - 0s 250us/step - loss: 1.203
8 - acc: 0.8452
Epoch 9/10
84/84 [============== ] - 0s 255us/step - loss: 1.157
6 - acc: 0.8333
Epoch 10/10
84/84 [============= ] - 0s 266us/step - loss: 1.113
0 - acc: 0.8690
28/28 [========= ] - 9s 321ms/step
84/84 [=======] - 0s 287us/step
Epoch 1/10
84/84 [============ ] - 21s 245ms/step - loss: 2.26
71 - acc: 0.3214
Epoch 2/10
3 - acc: 0.4524
Epoch 3/10
84/84 [============= ] - 0s 251us/step - loss: 1.650
9 - acc: 0.5952
Epoch 4/10
84/84 [============ ] - 0s 260us/step - loss: 1.602
3 - acc: 0.6786
Epoch 5/10
84/84 [============= ] - 0s 271us/step - loss: 1.566
1 - acc: 0.6786
Epoch 6/10
84/84 [============= ] - 0s 249us/step - loss: 1.491
5 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 257us/step - loss: 1.390
3 - acc: 0.6786
Epoch 8/10
```

```
84/84 [============== ] - 0s 272us/step - loss: 1.317
4 - acc: 0.6786
Epoch 9/10
84/84 [============= ] - 0s 243us/step - loss: 1.275
3 - acc: 0.8095
Epoch 10/10
84/84 [=========== ] - 0s 246us/step - loss: 1.236
1 - acc: 0.7024
28/28 [======== 1 - 9s 324ms/step
84/84 [=======] - 0s 266us/step
Epoch 1/10
84/84 [============ ] - 21s 246ms/step - loss: 1.67
29 - acc: 0.6786
Epoch 2/10
84/84 [============= ] - 0s 255us/step - loss: 1.534
0 - acc: 0.7381
Epoch 3/10
84/84 [=========== ] - 0s 244us/step - loss: 1.474
0 - acc: 0.5595
Epoch 4/10
84/84 [============== ] - 0s 263us/step - loss: 1.396
4 - acc: 0.7976
Epoch 5/10
84/84 [============== ] - 0s 266us/step - loss: 1.333
2 - acc: 0.7738
Epoch 6/10
84/84 [============ ] - 0s 249us/step - loss: 1.260
8 - acc: 0.7619
Epoch 7/10
84/84 [============ ] - 0s 251us/step - loss: 1.201
4 - acc: 0.8452
Epoch 8/10
84/84 [=========== ] - 0s 265us/step - loss: 1.160
9 - acc: 0.9405
Epoch 9/10
84/84 [========== ] - 0s 259us/step - loss: 1.117
4 - acc: 0.9643
Epoch 10/10
84/84 [============ ] - 0s 252us/step - loss: 1.071
8 - acc: 0.9286
28/28 [======== ] - 9s 339ms/step
84/84 [=======] - 0s 292us/step
Epoch 1/10
20 - acc: 0.3214
Epoch 2/10
84/84 [============ ] - 0s 294us/step - loss: 1.667
2 - acc: 0.4167
Epoch 3/10
84/84 [============ ] - 0s 253us/step - loss: 1.604
4 - acc: 0.5238
Epoch 4/10
84/84 [============== ] - 0s 267us/step - loss: 1.537
2 - acc: 0.6548
Epoch 5/10
84/84 [============= ] - 0s 294us/step - loss: 1.465
4 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 279us/step - loss: 1.393
4 - acc: 0.7500
Epoch 7/10
```

```
84/84 [============== ] - 0s 254us/step - loss: 1.335
2 - acc: 0.8214
Epoch 8/10
84/84 [============= ] - 0s 274us/step - loss: 1.280
9 - acc: 0.9167
Epoch 9/10
84/84 [============== ] - 0s 262us/step - loss: 1.230
0 - acc: 0.8571
Epoch 10/10
0 - acc: 0.8095
28/28 [======== ] - 9s 330ms/step
84/84 [======== ] - 0s 343us/step
Epoch 1/10
84/84 [============== ] - 21s 250ms/step - loss: 1.83
55 - acc: 0.3452
Epoch 2/10
84/84 [=========== ] - 0s 250us/step - loss: 1.669
8 - acc: 0.5714
Epoch 3/10
84/84 [============ ] - 0s 251us/step - loss: 1.582
5 - acc: 0.6786
Epoch 4/10
84/84 [============== ] - 0s 257us/step - loss: 1.520
7 - acc: 0.6667
Epoch 5/10
84/84 [============ ] - 0s 248us/step - loss: 1.462
4 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 253us/step - loss: 1.407
7 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 264us/step - loss: 1.357
8 - acc: 0.9286
Epoch 8/10
84/84 [========== ] - 0s 262us/step - loss: 1.304
0 - acc: 0.9643
Epoch 9/10
84/84 [============ ] - 0s 285us/step - loss: 1.257
2 - acc: 0.8452
Epoch 10/10
84/84 [============== ] - 0s 286us/step - loss: 1.208
1 - acc: 0.8214
28/28 [========] - 9s 331ms/step
84/84 [========] - 0s 275us/step
Epoch 1/10
84/84 [============ ] - 22s 258ms/step - loss: 4.77
74 - acc: 0.3810
Epoch 2/10
84/84 [============ ] - 0s 293us/step - loss: 4.482
2 - acc: 0.6786
Epoch 3/10
84/84 [============= ] - 0s 303us/step - loss: 4.293
2 - acc: 0.3690
Epoch 4/10
84/84 [============= ] - 0s 290us/step - loss: 4.140
1 - acc: 0.4762
Epoch 5/10
84/84 [========== ] - 0s 291us/step - loss: 3.999
3 - acc: 0.5595
Epoch 6/10
```

```
84/84 [============== ] - 0s 288us/step - loss: 3.848
6 - acc: 0.7619
Epoch 7/10
84/84 [============= ] - 0s 308us/step - loss: 3.695
9 - acc: 0.8214
Epoch 8/10
84/84 [============== ] - 0s 302us/step - loss: 3.562
6 - acc: 0.6786
Epoch 9/10
0 - acc: 0.6786
Epoch 10/10
84/84 [============= ] - 0s 320us/step - loss: 3.336
8 - acc: 0.6786
28/28 [======== ] - 9s 334ms/step
84/84 [======== ] - 0s 260us/step
Epoch 1/10
14 - acc: 0.3452
Epoch 2/10
84/84 [============== ] - 0s 266us/step - loss: 4.608
5 - acc: 0.3452
Epoch 3/10
84/84 [============ ] - 0s 259us/step - loss: 4.334
1 - acc: 0.4762
Epoch 4/10
84/84 [============ ] - 0s 267us/step - loss: 4.146
8 - acc: 0.7024
Epoch 5/10
84/84 [============ ] - 0s 251us/step - loss: 3.995
9 - acc: 0.9405
Epoch 6/10
84/84 [============= ] - 0s 267us/step - loss: 3.850
4 - acc: 0.8929
Epoch 7/10
84/84 [========== ] - 0s 336us/step - loss: 3.726
0 - acc: 0.7262
Epoch 8/10
84/84 [=========== ] - 0s 312us/step - loss: 3.599
7 - acc: 0.6905
Epoch 9/10
84/84 [============= ] - 0s 278us/step - loss: 3.482
4 - acc: 0.6786
Epoch 10/10
84/84 [============= ] - 0s 280us/step - loss: 3.366
3 - acc: 0.6786
28/28 [======== ] - 9s 334ms/step
84/84 [========] - 0s 296us/step
Epoch 1/10
84/84 [============ ] - 22s 264ms/step - loss: 4.76
26 - acc: 0.3452
Epoch 2/10
84/84 [============= ] - 0s 256us/step - loss: 4.489
3 - acc: 0.6667
Epoch 3/10
84/84 [============== ] - 0s 279us/step - loss: 4.313
0 - acc: 0.6548
Epoch 4/10
84/84 [=========== ] - 0s 244us/step - loss: 4.139
9 - acc: 0.6667
Epoch 5/10
```

```
84/84 [============= ] - 0s 236us/step - loss: 3.984
2 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 247us/step - loss: 3.836
4 - acc: 0.6786
Epoch 7/10
84/84 [============= ] - 0s 275us/step - loss: 3.696
0 - acc: 0.6786
Epoch 8/10
5 - acc: 0.6786
Epoch 9/10
84/84 [============= ] - 0s 265us/step - loss: 3.435
1 - acc: 0.6905
Epoch 10/10
84/84 [============ ] - 0s 257us/step - loss: 3.312
0 - acc: 0.7262
28/28 [======== ] - 10s 340ms/step
84/84 [========] - 0s 297us/step
Epoch 1/10
84/84 [============ ] - 22s 263ms/step - loss: 4.83
55 - acc: 0.0833
Epoch 2/10
84/84 [============= ] - 0s 279us/step - loss: 4.563
3 - acc: 0.3333
Epoch 3/10
84/84 [============= ] - 0s 248us/step - loss: 4.377
2 - acc: 0.3333
Epoch 4/10
84/84 [============ ] - 0s 278us/step - loss: 4.204
4 - acc: 0.4286
Epoch 5/10
84/84 [=========== ] - 0s 253us/step - loss: 4.035
1 - acc: 0.6786
Epoch 6/10
84/84 [========= ] - 0s 269us/step - loss: 3.869
9 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 270us/step - loss: 3.724
3 - acc: 0.7024
Epoch 8/10
84/84 [============= ] - 0s 266us/step - loss: 3.602
6 - acc: 0.7619
Epoch 9/10
84/84 [============= ] - 0s 248us/step - loss: 3.485
0 - acc: 0.7857
Epoch 10/10
84/84 [============ ] - 0s 270us/step - loss: 3.365
0 - acc: 0.8095
28/28 [======== ] - 9s 339ms/step
84/84 [========] - 0s 319us/step
Epoch 1/10
84/84 [============= ] - 22s 258ms/step - loss: 5.66
98 - acc: 0.3452
Epoch 2/10
84/84 [============= ] - 0s 282us/step - loss: 4.992
9 - acc: 0.3452
Epoch 3/10
84/84 [=========== ] - 0s 286us/step - loss: 4.515
7 - acc: 0.3452
Epoch 4/10
```

```
84/84 [============= ] - 0s 281us/step - loss: 4.209
2 - acc: 0.5833
Epoch 5/10
84/84 [============= ] - 0s 268us/step - loss: 3.994
6 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 268us/step - loss: 3.841
7 - acc: 0.6786
Epoch 7/10
84/84 [============ ] - 0s 254us/step - loss: 3.715
8 - acc: 0.6905
Epoch 8/10
84/84 [============= ] - 0s 247us/step - loss: 3.599
4 - acc: 0.7976
Epoch 9/10
84/84 [=========== ] - 0s 279us/step - loss: 3.488
6 - acc: 0.9048
Epoch 10/10
84/84 [============ ] - 0s 265us/step - loss: 3.376
3 - acc: 0.9524
28/28 [======== ] - 10s 340ms/step
84/84 [=======] - 0s 287us/step
Epoch 1/10
84/84 [============= ] - 22s 265ms/step - loss: 8.54
20 - acc: 0.4643
Epoch 2/10
84/84 [============= ] - 0s 290us/step - loss: 8.097
5 - acc: 0.6786
Epoch 3/10
84/84 [============= ] - 0s 291us/step - loss: 7.730
9 - acc: 0.6786
Epoch 4/10
84/84 [=========== ] - 0s 314us/step - loss: 7.423
3 - acc: 0.6786
Epoch 5/10
84/84 [============ ] - 0s 307us/step - loss: 7.163
5 - acc: 0.6071
Epoch 6/10
84/84 [=========== ] - 0s 312us/step - loss: 6.912
1 - acc: 0.6667
Epoch 7/10
84/84 [============== ] - 0s 285us/step - loss: 6.667
5 - acc: 0.6667
Epoch 8/10
84/84 [============= ] - 0s 313us/step - loss: 6.428
3 - acc: 0.6667
Epoch 9/10
84/84 [=========== ] - 0s 314us/step - loss: 6.200
7 - acc: 0.6905
Epoch 10/10
84/84 [=============== ] - 0s 372us/step - loss: 5.981
7 - acc: 0.7024
28/28 [========] - 10s 348ms/step
84/84 [=======] - 0s 274us/step
Epoch 1/10
84/84 [============ ] - 22s 260ms/step - loss: 7.99
22 - acc: 0.5119
Epoch 2/10
84/84 [============== ] - 0s 264us/step - loss: 7.681
3 - acc: 0.6071
Epoch 3/10
```

```
84/84 [============= ] - 0s 264us/step - loss: 7.407
5 - acc: 0.6786
Epoch 4/10
84/84 [============= ] - 0s 257us/step - loss: 7.140
2 - acc: 0.6786
Epoch 5/10
84/84 [=========== ] - 0s 260us/step - loss: 6.887
4 - acc: 0.6905
Epoch 6/10
84/84 [========= ] - 0s 248us/step - loss: 6.638
9 - acc: 0.8571
Epoch 7/10
84/84 [============== ] - 0s 266us/step - loss: 6.404
4 - acc: 0.8214
Epoch 8/10
84/84 [=========== ] - 0s 262us/step - loss: 6.174
9 - acc: 0.6905
Epoch 9/10
84/84 [=========== ] - 0s 263us/step - loss: 5.959
8 - acc: 0.6786
Epoch 10/10
84/84 [============= ] - 0s 254us/step - loss: 5.743
9 - acc: 0.6786
28/28 [========= ] - 10s 344ms/step
84/84 [======== ] - 0s 255us/step
Epoch 1/10
84/84 [============ ] - 22s 263ms/step - loss: 8.08
72 - acc: 0.3929
Epoch 2/10
84/84 [============ ] - 0s 339us/step - loss: 7.763
9 - acc: 0.5833
Epoch 3/10
84/84 [=========== ] - 0s 291us/step - loss: 7.484
9 - acc: 0.6786
Epoch 4/10
84/84 [========== ] - 0s 328us/step - loss: 7.224
0 - acc: 0.6786
Epoch 5/10
84/84 [=========== ] - 0s 302us/step - loss: 6.968
0 - acc: 0.6786
Epoch 6/10
84/84 [============= ] - 0s 313us/step - loss: 6.710
7 - acc: 0.6786
Epoch 7/10
84/84 [=========== ] - 0s 325us/step - loss: 6.466
0 - acc: 0.6786
Epoch 8/10
84/84 [============ ] - 0s 307us/step - loss: 6.228
9 - acc: 0.6786
Epoch 9/10
7 - acc: 0.6786
Epoch 10/10
84/84 [=========== ] - 0s 340us/step - loss: 5.785
6 - acc: 0.6786
28/28 [======== ] - 10s 357ms/step
84/84 [========] - 0s 307us/step
Epoch 1/10
84/84 [============= ] - 23s 271ms/step - loss: 8.47
85 - acc: 0.3333
Epoch 2/10
```

```
84/84 [============== ] - 0s 280us/step - loss: 8.005
0 - acc: 0.3333
Epoch 3/10
84/84 [============= ] - 0s 256us/step - loss: 7.714
1 - acc: 0.3810
Epoch 4/10
84/84 [=========== ] - 0s 251us/step - loss: 7.471
0 - acc: 0.3810
Epoch 5/10
84/84 [========== ] - 0s 253us/step - loss: 7.212
2 - acc: 0.6071
Epoch 6/10
84/84 [============= ] - 0s 279us/step - loss: 6.960
4 - acc: 0.7500
Epoch 7/10
84/84 [=========== ] - 0s 268us/step - loss: 6.718
6 - acc: 0.6905
Epoch 8/10
84/84 [=========== ] - 0s 273us/step - loss: 6.489
5 - acc: 0.6667
Epoch 9/10
84/84 [============= ] - 0s 320us/step - loss: 6.269
7 - acc: 0.6667
Epoch 10/10
84/84 [=========== ] - 0s 309us/step - loss: 6.055
5 - acc: 0.6786
28/28 [======== ] - 10s 349ms/step
84/84 [======= ] - 0s 274us/step
Epoch 1/10
84/84 [============ ] - 22s 265ms/step - loss: 8.65
00 - acc: 0.3333
Epoch 2/10
84/84 [=========== ] - 0s 274us/step - loss: 8.200
7 - acc: 0.3333
Epoch 3/10
84/84 [========= ] - 0s 263us/step - loss: 7.829
9 - acc: 0.3333
Epoch 4/10
84/84 [============ ] - 0s 262us/step - loss: 7.528
4 - acc: 0.3333
Epoch 5/10
84/84 [============= ] - 0s 256us/step - loss: 7.241
5 - acc: 0.5833
Epoch 6/10
84/84 [=========== ] - 0s 253us/step - loss: 6.968
3 - acc: 0.6667
Epoch 7/10
84/84 [============ ] - 0s 256us/step - loss: 6.709
8 - acc: 0.6786
Epoch 8/10
5 - acc: 0.7738
Epoch 9/10
84/84 [============= ] - 0s 248us/step - loss: 6.227
1 - acc: 0.6905
Epoch 10/10
84/84 [============== ] - 0s 274us/step - loss: 6.003
7 - acc: 0.6786
28/28 [========= ] - 10s 351ms/step
84/84 [=======] - 0s 245us/step
Epoch 1/10
```

```
84/84 [=============== ] - 23s 274ms/step - loss: 1.00
48 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 271us/step - loss: 0.821
8 - acc: 0.6786
Epoch 3/10
84/84 [============== ] - 0s 256us/step - loss: 0.702
0 - acc: 0.6905
Epoch 4/10
84/84 [========= ] - 0s 248us/step - loss: 0.589
6 - acc: 0.9048
Epoch 5/10
84/84 [============= ] - 0s 250us/step - loss: 0.534
1 - acc: 0.7262
Epoch 6/10
84/84 [============ ] - 0s 270us/step - loss: 0.470
6 - acc: 0.8214
Epoch 7/10
84/84 [========== ] - 0s 249us/step - loss: 0.425
9 - acc: 0.9643
Epoch 8/10
84/84 [============= ] - 0s 259us/step - loss: 0.392
8 - acc: 0.9643
Epoch 9/10
84/84 [============= ] - 0s 256us/step - loss: 0.362
6 - acc: 0.9405
Epoch 10/10
84/84 [============= ] - 0s 284us/step - loss: 0.347
4 - acc: 0.9524
28/28 [========] - 10s 354ms/step
84/84 [=======] - 0s 294us/step
Epoch 1/10
61 - acc: 0.4048
Epoch 2/10
84/84 [============== ] - 0s 253us/step - loss: 0.773
7 - acc: 0.7857
Epoch 3/10
84/84 [=========== ] - 0s 249us/step - loss: 0.666
8 - acc: 0.8690
Epoch 4/10
84/84 [============= ] - 0s 259us/step - loss: 0.565
6 - acc: 0.9048
Epoch 5/10
84/84 [============= ] - 0s 267us/step - loss: 0.501
6 - acc: 0.7024
Epoch 6/10
84/84 [============= ] - 0s 275us/step - loss: 0.448
5 - acc: 0.8571
Epoch 7/10
2 - acc: 0.9762
Epoch 8/10
84/84 [============= ] - 0s 269us/step - loss: 0.354
4 - acc: 0.9762
Epoch 9/10
84/84 [============= ] - 0s 253us/step - loss: 0.323
3 - acc: 0.9643
Epoch 10/10
84/84 [============== ] - 0s 253us/step - loss: 0.290
9 - acc: 0.9762
```

```
28/28 [======== ] - 10s 369ms/step
84/84 [=======] - 0s 309us/step
Epoch 1/10
84/84 [============== ] - 23s 274ms/step - loss: 1.00
12 - acc: 0.5357
Epoch 2/10
84/84 [============== ] - 0s 268us/step - loss: 0.793
9 - acc: 0.6786
Epoch 3/10
84/84 [========= ] - 0s 245us/step - loss: 0.680
9 - acc: 0.7976
Epoch 4/10
84/84 [============= ] - 0s 263us/step - loss: 0.587
1 - acc: 0.6905
Epoch 5/10
84/84 [============ ] - 0s 264us/step - loss: 0.512
8 - acc: 0.7143
Epoch 6/10
84/84 [============ ] - 0s 271us/step - loss: 0.451
4 - acc: 0.8571
Epoch 7/10
84/84 [============= ] - 0s 249us/step - loss: 0.410
3 - acc: 0.9643
Epoch 8/10
84/84 [============== ] - 0s 257us/step - loss: 0.363
6 - acc: 0.9762
Epoch 9/10
84/84 [============ ] - 0s 276us/step - loss: 0.348
7 - acc: 0.8333
Epoch 10/10
84/84 [============== ] - 0s 252us/step - loss: 0.316
3 - acc: 0.9167
28/28 [========= ] - 11s 378ms/step
84/84 [======== ] - 0s 312us/step
Epoch 1/10
84/84 [============== ] - 23s 273ms/step - loss: 1.02
33 - acc: 0.4048
Epoch 2/10
84/84 [============ ] - 0s 287us/step - loss: 0.813
7 - acc: 0.7976
Epoch 3/10
84/84 [============= ] - 0s 260us/step - loss: 0.667
7 - acc: 0.8095
Epoch 4/10
84/84 [============= ] - 0s 254us/step - loss: 0.565
0 - acc: 0.7262
Epoch 5/10
84/84 [============ ] - 0s 268us/step - loss: 0.478
9 - acc: 0.9643
Epoch 6/10
84/84 [=============== ] - 0s 264us/step - loss: 0.418
3 - acc: 1.0000
Epoch 7/10
84/84 [============= ] - 0s 255us/step - loss: 0.389
4 - acc: 0.8929
Epoch 8/10
84/84 [============== ] - 0s 250us/step - loss: 0.328
2 - acc: 0.9643
Epoch 9/10
84/84 [============== ] - 0s 247us/step - loss: 0.316
0 - acc: 0.9762
```

```
Epoch 10/10
84/84 [============== ] - 0s 270us/step - loss: 0.292
1 - acc: 0.9524
28/28 [======== ] - 10s 359ms/step
84/84 [======== ] - 0s 373us/step
Epoch 1/10
07 - acc: 0.2738
Epoch 2/10
84/84 [============== ] - 0s 271us/step - loss: 0.854
6 - acc: 0.8810
Epoch 3/10
84/84 [============== ] - 0s 267us/step - loss: 0.751
0 - acc: 0.6786
Epoch 4/10
84/84 [=========== ] - 0s 257us/step - loss: 0.651
0 - acc: 0.7143
Epoch 5/10
84/84 [============= ] - 0s 281us/step - loss: 0.561
3 - acc: 0.9643
Epoch 6/10
84/84 [============= ] - 0s 264us/step - loss: 0.495
4 - acc: 0.9762
Epoch 7/10
84/84 [============= ] - 0s 269us/step - loss: 0.437
8 - acc: 0.8810
Epoch 8/10
84/84 [============= ] - 0s 269us/step - loss: 0.397
7 - acc: 0.9405
Epoch 9/10
84/84 [============== ] - 0s 266us/step - loss: 0.360
2 - acc: 0.9762
Epoch 10/10
84/84 [============= ] - 0s 265us/step - loss: 0.330
4 - acc: 0.9524
28/28 [======== ] - 10s 361ms/step
84/84 [======== ] - 0s 286us/step
Epoch 1/10
84/84 [============ ] - 24s 282ms/step - loss: 3.68
03 - acc: 0.2738
Epoch 2/10
84/84 [============= ] - 0s 316us/step - loss: 3.301
7 - acc: 0.6190
Epoch 3/10
84/84 [============= ] - 0s 296us/step - loss: 3.020
1 - acc: 0.7857
Epoch 4/10
84/84 [============ ] - 0s 266us/step - loss: 2.772
7 - acc: 0.7024
Epoch 5/10
84/84 [============== ] - 0s 253us/step - loss: 2.583
3 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 251us/step - loss: 2.388
8 - acc: 0.8929
Epoch 7/10
84/84 [============== ] - 0s 256us/step - loss: 2.248
2 - acc: 0.9643
Epoch 8/10
84/84 [============== ] - 0s 263us/step - loss: 2.106
2 - acc: 0.9524
```

```
Epoch 9/10
84/84 [============== ] - 0s 265us/step - loss: 1.971
1 - acc: 0.9167
Epoch 10/10
84/84 [============= ] - 0s 261us/step - loss: 1.855
5 - acc: 0.9167
28/28 [======== ] - 10s 362ms/step
84/84 [========] - 0s 273us/step
Epoch 1/10
84/84 [=========== ] - 23s 275ms/step - loss: 3.64
01 - acc: 0.4405
Epoch 2/10
84/84 [============== ] - 0s 257us/step - loss: 3.312
3 - acc: 0.6071
Epoch 3/10
84/84 [============ ] - 0s 246us/step - loss: 3.005
2 - acc: 0.6786
Epoch 4/10
84/84 [============= ] - 0s 261us/step - loss: 2.790
2 - acc: 0.6786
Epoch 5/10
8 - acc: 0.8333
Epoch 6/10
84/84 [============= ] - 0s 270us/step - loss: 2.395
6 - acc: 0.9524
Epoch 7/10
84/84 [============== ] - 0s 291us/step - loss: 2.221
9 - acc: 0.8810
Epoch 8/10
84/84 [============ ] - 0s 280us/step - loss: 2.086
7 - acc: 0.8333
Epoch 9/10
84/84 [============= ] - 0s 266us/step - loss: 1.941
6 - acc: 0.9524
Epoch 10/10
84/84 [============ ] - 0s 260us/step - loss: 1.825
6 - acc: 0.9643
28/28 [========= ] - 11s 381ms/step
84/84 [======== ] - 0s 290us/step
Epoch 1/10
84/84 [============= ] - 24s 283ms/step - loss: 3.69
03 - acc: 0.2738
Epoch 2/10
84/84 [============== ] - 0s 342us/step - loss: 3.314
1 - acc: 0.5595
Epoch 3/10
84/84 [============ ] - 0s 332us/step - loss: 3.022
5 - acc: 0.7500
Epoch 4/10
0 - acc: 0.7619
Epoch 5/10
84/84 [============= ] - 0s 295us/step - loss: 2.580
9 - acc: 0.7857
Epoch 6/10
84/84 [============= ] - 0s 317us/step - loss: 2.429
1 - acc: 0.8810
Epoch 7/10
84/84 [============== ] - 0s 291us/step - loss: 2.245
1 - acc: 0.9405
```

```
Epoch 8/10
84/84 [============== ] - 0s 322us/step - loss: 2.138
0 - acc: 0.7024
Epoch 9/10
84/84 [============= ] - 0s 343us/step - loss: 1.986
1 - acc: 0.9048
Epoch 10/10
84/84 [============ ] - 0s 342us/step - loss: 1.897
0 - acc: 0.8929
28/28 [======= ] - 11s 377ms/step
84/84 [=======] - 0s 278us/step
Epoch 1/10
84/84 [============= ] - 23s 279ms/step - loss: 3.63
53 - acc: 0.3690
Epoch 2/10
84/84 [============ ] - 0s 294us/step - loss: 3.282
4 - acc: 0.6786
Epoch 3/10
84/84 [============ ] - 0s 264us/step - loss: 3.007
4 - acc: 0.6786
Epoch 4/10
84/84 [============= ] - 0s 273us/step - loss: 2.802
8 - acc: 0.8333
Epoch 5/10
84/84 [============ ] - 0s 282us/step - loss: 2.558
7 - acc: 0.9048
Epoch 6/10
84/84 [============== ] - 0s 273us/step - loss: 2.403
7 - acc: 0.7024
Epoch 7/10
84/84 [============= ] - 0s 261us/step - loss: 2.239
9 - acc: 0.7619
Epoch 8/10
84/84 [============ ] - 0s 264us/step - loss: 2.081
6 - acc: 0.9643
Epoch 9/10
84/84 [============== ] - 0s 311us/step - loss: 1.952
9 - acc: 0.9286
Epoch 10/10
84/84 [============ ] - 0s 285us/step - loss: 1.832
3 - acc: 0.8929
28/28 [=======] - 10s 373ms/step
84/84 [======] - 0s 279us/step
Epoch 1/10
84/84 [============= ] - 24s 280ms/step - loss: 3.74
52 - acc: 0.3095
Epoch 2/10
84/84 [============ ] - 0s 281us/step - loss: 3.379
4 - acc: 0.3929
Epoch 3/10
84/84 [============== ] - 0s 258us/step - loss: 3.089
1 - acc: 0.7381
Epoch 4/10
84/84 [============= ] - 0s 268us/step - loss: 2.860
4 - acc: 0.6786
Epoch 5/10
84/84 [============= ] - 0s 255us/step - loss: 2.628
7 - acc: 0.8095
Epoch 6/10
84/84 [=============== ] - 0s 275us/step - loss: 2.459
4 - acc: 0.8095
```

```
Epoch 7/10
84/84 [============== ] - 0s 273us/step - loss: 2.277
2 - acc: 0.9643
Epoch 8/10
84/84 [============= ] - 0s 306us/step - loss: 2.144
2 - acc: 0.7738
Epoch 9/10
84/84 [============ ] - 0s 258us/step - loss: 2.004
5 - acc: 0.8571
Epoch 10/10
84/84 [============== ] - 0s 261us/step - loss: 1.888
5 - acc: 0.9881
28/28 [========= ] - 11s 391ms/step
84/84 [======== ] - 0s 313us/step
Epoch 1/10
84/84 [============= ] - 24s 284ms/step - loss: 13.9
387 - acc: 0.5952
Epoch 2/10
84/84 [============ ] - 0s 273us/step - loss: 12.85
41 - acc: 0.6786
Epoch 3/10
84/84 [============= ] - 0s 270us/step - loss: 11.89
92 - acc: 0.6786
Epoch 4/10
84/84 [=========== ] - 0s 277us/step - loss: 10.98
11 - acc: 0.8929
Epoch 5/10
84/84 [============= ] - 0s 267us/step - loss: 10.15
67 - acc: 0.9524
Epoch 6/10
84/84 [============= ] - 0s 291us/step - loss: 9.379
5 - acc: 0.8690
Epoch 7/10
84/84 [============= ] - 0s 267us/step - loss: 8.660
1 - acc: 0.7857
Epoch 8/10
84/84 [============== ] - 0s 273us/step - loss: 7.986
0 - acc: 0.9048
Epoch 9/10
84/84 [============== ] - 0s 284us/step - loss: 7.360
3 - acc: 0.9405
Epoch 10/10
84/84 [============== ] - 0s 284us/step - loss: 6.784
0 - acc: 0.9286
28/28 [======== ] - 11s 391ms/step
84/84 [========] - 0s 328us/step
Epoch 1/10
84/84 [============== ] - 24s 288ms/step - loss: 13.9
875 - acc: 0.3214
Epoch 2/10
02 - acc: 0.5595
Epoch 3/10
84/84 [============= ] - 0s 284us/step - loss: 11.90
21 - acc: 0.8452
Epoch 4/10
84/84 [============= ] - 0s 266us/step - loss: 10.99
43 - acc: 0.6786
Epoch 5/10
60 - acc: 0.6786
```

```
Epoch 6/10
84/84 [============== ] - 0s 274us/step - loss: 9.348
9 - acc: 0.6786
Epoch 7/10
84/84 [============== ] - 0s 265us/step - loss: 8.613
6 - acc: 0.7143
Epoch 8/10
84/84 [============ ] - 0s 285us/step - loss: 7.929
9 - acc: 0.7619
Epoch 9/10
84/84 [============= ] - 0s 276us/step - loss: 7.292
0 - acc: 0.8929
Epoch 10/10
84/84 [============ ] - 0s 273us/step - loss: 6.706
9 - acc: 0.9643
28/28 [======== ] - 11s 379ms/step
84/84 [=======] - 0s 297us/step
Epoch 1/10
84/84 [============ ] - 25s 293ms/step - loss: 14.0
356 - acc: 0.5476
Epoch 2/10
84/84 [============= ] - 0s 304us/step - loss: 12.94
14 - acc: 0.4881
Epoch 3/10
84/84 [=========== ] - 0s 289us/step - loss: 11.95
34 - acc: 0.9524
Epoch 4/10
84/84 [=========== ] - 0s 280us/step - loss: 11.05
03 - acc: 0.8810
Epoch 5/10
84/84 [============= ] - 0s 273us/step - loss: 10.21
90 - acc: 0.8214
Epoch 6/10
84/84 [============= ] - 0s 262us/step - loss: 9.430
1 - acc: 0.7381
Epoch 7/10
84/84 [============== ] - 0s 276us/step - loss: 8.707
8 - acc: 0.7500
Epoch 8/10
3 - acc: 0.8452
Epoch 9/10
84/84 [============== ] - 0s 285us/step - loss: 7.417
3 - acc: 0.7857
Epoch 10/10
84/84 [============= ] - 0s 282us/step - loss: 6.827
0 - acc: 0.8333
28/28 [=======] - 11s 380ms/step
84/84 [=======] - 0s 306us/step
Epoch 1/10
820 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 270us/step - loss: 12.90
74 - acc: 0.5952
Epoch 3/10
84/84 [============= ] - 0s 282us/step - loss: 11.93
62 - acc: 0.6905
Epoch 4/10
84/84 [============== ] - 0s 262us/step - loss: 11.02
53 - acc: 0.7262
```

```
Epoch 5/10
84/84 [============== ] - 0s 267us/step - loss: 10.18
97 - acc: 0.9405
Epoch 6/10
84/84 [============== ] - 0s 274us/step - loss: 9.389
9 - acc: 0.9167
Epoch 7/10
84/84 [=========== ] - 0s 270us/step - loss: 8.668
2 - acc: 0.6905
Epoch 8/10
84/84 [============= ] - 0s 271us/step - loss: 7.984
4 - acc: 0.7738
Epoch 9/10
84/84 [========== ] - 0s 271us/step - loss: 7.365
3 - acc: 0.9643
Epoch 10/10
1 - acc: 0.9762
28/28 [======== ] - 11s 382ms/step
84/84 [=======] - 0s 306us/step
Epoch 1/10
84/84 [============== ] - 25s 296ms/step - loss: 14.0
000 - acc: 0.2143
Epoch 2/10
84/84 [============== ] - 0s 276us/step - loss: 12.90
40 - acc: 0.7262
Epoch 3/10
84/84 [============= ] - 0s 279us/step - loss: 11.94
58 - acc: 0.9524
Epoch 4/10
84/84 [============= ] - 0s 280us/step - loss: 11.04
03 - acc: 0.7738
Epoch 5/10
84/84 [============= ] - 0s 276us/step - loss: 10.19
52 - acc: 0.6905
Epoch 6/10
84/84 [============== ] - 0s 276us/step - loss: 9.412
0 - acc: 0.8810
Epoch 7/10
3 - acc: 0.9762
Epoch 8/10
84/84 [============== ] - 0s 318us/step - loss: 7.991
0 - acc: 0.8571
Epoch 9/10
84/84 [============= ] - 0s 276us/step - loss: 7.365
6 - acc: 0.8333
Epoch 10/10
84/84 [============== ] - 0s 275us/step - loss: 6.794
0 - acc: 0.9048
28/28 [========] - 11s 386ms/step
84/84 [========] - 0s 279us/step
Epoch 1/10
84/84 [============ ] - 25s 294ms/step - loss: 26.7
891 - acc: 0.3333
Epoch 2/10
84/84 [============= ] - 0s 307us/step - loss: 24.67
44 - acc: 0.6667
Epoch 3/10
22 - acc: 0.7381
```

```
Epoch 4/10
84/84 [============== ] - 0s 267us/step - loss: 20.88
93 - acc: 0.6786
Epoch 5/10
84/84 [============== ] - 0s 262us/step - loss: 19.18
63 - acc: 0.6786
Epoch 6/10
84/84 [=========== ] - 0s 275us/step - loss: 17.59
99 - acc: 0.6786
Epoch 7/10
84/84 [============= ] - 0s 287us/step - loss: 16.12
68 - acc: 0.7738
Epoch 8/10
84/84 [========== ] - 0s 276us/step - loss: 14.76
88 - acc: 0.8810
Epoch 9/10
84/84 [============= ] - 0s 274us/step - loss: 13.51
15 - acc: 0.7024
Epoch 10/10
84/84 [============== ] - 0s 270us/step - loss: 12.34
57 - acc: 0.6905
28/28 [======== ] - 11s 386ms/step
84/84 [========] - 0s 288us/step
Epoch 1/10
84/84 [============= ] - 25s 294ms/step - loss: 26.9
749 - acc: 0.4286
Epoch 2/10
84/84 [=========== ] - 0s 340us/step - loss: 24.89
61 - acc: 0.7976
Epoch 3/10
84/84 [============= ] - 0s 324us/step - loss: 22.95
60 - acc: 0.8810
Epoch 4/10
84/84 [============= ] - 0s 297us/step - loss: 21.14
63 - acc: 0.6905
Epoch 5/10
84/84 [============= ] - 0s 309us/step - loss: 19.45
31 - acc: 0.6905
Epoch 6/10
84/84 [============== ] - 0s 308us/step - loss: 17.86
68 - acc: 0.8214
Epoch 7/10
84/84 [============== ] - 0s 307us/step - loss: 16.38
90 - acc: 0.8810
Epoch 8/10
84/84 [============= ] - 0s 314us/step - loss: 15.02
22 - acc: 0.8571
Epoch 9/10
84/84 [============ ] - 0s 320us/step - loss: 13.75
22 - acc: 0.8214
Epoch 10/10
84/84 [============ ] - 0s 323us/step - loss: 12.58
93 - acc: 0.7619
28/28 [============ ] - 11s 398ms/step
84/84 [========] - 0s 278us/step
Epoch 1/10
84/84 [============= ] - 25s 293ms/step - loss: 26.8
597 - acc: 0.5476
Epoch 2/10
84/84 [============== ] - 0s 292us/step - loss: 24.76
91 - acc: 0.6786
```

```
Epoch 3/10
84/84 [============== ] - 0s 260us/step - loss: 22.84
47 - acc: 0.7143
Epoch 4/10
84/84 [============= ] - 0s 264us/step - loss: 21.04
91 - acc: 0.9405
Epoch 5/10
84/84 [============ ] - 0s 265us/step - loss: 19.36
34 - acc: 0.8929
Epoch 6/10
07 - acc: 0.8452
Epoch 7/10
84/84 [============= ] - 0s 258us/step - loss: 16.32
97 - acc: 0.7976
Epoch 8/10
84/84 [============= ] - 0s 297us/step - loss: 14.96
99 - acc: 0.8214
Epoch 9/10
84/84 [============== ] - 0s 262us/step - loss: 13.70
88 - acc: 0.7976
Epoch 10/10
84/84 [============= ] - 0s 260us/step - loss: 12.53
97 - acc: 0.8214
28/28 [======== ] - 11s 393ms/step
84/84 [======== ] - 0s 284us/step
Epoch 1/10
84/84 [============ ] - 25s 295ms/step - loss: 26.7
798 - acc: 0.5714
Epoch 2/10
84/84 [============= ] - 0s 272us/step - loss: 24.70
64 - acc: 0.5476
Epoch 3/10
84/84 [============= ] - 0s 258us/step - loss: 22.80
58 - acc: 0.8571
Epoch 4/10
84/84 [============== ] - 0s 276us/step - loss: 21.00
78 - acc: 0.8095
Epoch 5/10
84/84 [============ ] - 0s 260us/step - loss: 19.34
55 - acc: 0.6786
Epoch 6/10
84/84 [============== ] - 0s 290us/step - loss: 17.78
13 - acc: 0.7619
Epoch 7/10
84/84 [============= ] - 0s 271us/step - loss: 16.33
55 - acc: 0.9643
Epoch 8/10
84/84 [============ ] - 0s 281us/step - loss: 14.97
93 - acc: 0.9524
Epoch 9/10
84/84 [============ ] - 0s 267us/step - loss: 13.73
31 - acc: 0.7500
Epoch 10/10
84/84 [============== ] - 0s 285us/step - loss: 12.59
41 - acc: 0.7857
28/28 [========= ] - 12s 417ms/step
84/84 [=======] - 0s 335us/step
Epoch 1/10
84/84 [============== ] - 26s 306ms/step - loss: 26.8
165 - acc: 0.5238
```

```
Epoch 2/10
84/84 [============= ] - 0s 346us/step - loss: 24.77
41 - acc: 0.6310
Epoch 3/10
84/84 [========== ] - 0s 310us/step - loss: 22.85
31 - acc: 0.7143
Epoch 4/10
84/84 [============ ] - 0s 324us/step - loss: 21.08
02 - acc: 0.6786
Epoch 5/10
84/84 [============= ] - 0s 320us/step - loss: 19.42
38 - acc: 0.6786
Epoch 6/10
84/84 [============ ] - 0s 320us/step - loss: 17.87
20 - acc: 0.7976
Epoch 7/10
84/84 [=========== ] - 0s 300us/step - loss: 16.41
63 - acc: 0.9167
Epoch 8/10
84/84 [============= ] - 0s 281us/step - loss: 15.07
56 - acc: 0.7143
Epoch 9/10
84/84 [============== ] - 0s 303us/step - loss: 13.82
03 - acc: 0.7738
Epoch 10/10
84/84 [============= ] - 0s 328us/step - loss: 12.68
19 - acc: 0.9167
28/28 [======== ] - 11s 396ms/step
84/84 [======== ] - 0s 295us/step
Epoch 1/10
9878 - acc: 0.4375
Epoch 2/10
112/112 [============= ] - 0s 263us/step - loss: 0.7
943 - acc: 0.6875
Epoch 3/10
042 - acc: 0.6875
Epoch 4/10
094 - acc: 0.9018
Epoch 5/10
112/112 [============] - 0s 268us/step - loss: 0.4
287 - acc: 0.8929
Epoch 6/10
840 - acc: 0.8750
Epoch 7/10
543 - acc: 0.9554
Epoch 8/10
115 - acc: 0.9375
Epoch 9/10
925 - acc: 0.9018
Epoch 10/10
581 - acc: 0.9732
```

Out[0]:

```
In [0]:
```

```
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split0 trai
n score'), which will not be available by default any more in 0.21.
If you need training scores, please set return train score=True
  warnings.warn(*warn_args, **warn_kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split1 trai
n score'), which will not be available by default any more in 0.21.
If you need training scores, please set return train score=True
  warnings.warn(*warn_args, **warn_kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split2 trai
n score'), which will not be available by default any more in 0.21.
If you need training scores, please set return train score=True
  warnings.warn(*warn_args, **warn_kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split3 trai
n score'), which will not be available by default any more in 0.21.
If you need training scores, please set return_train_score=True
  warnings.warn(*warn args, **warn kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('split4 trai
n score'), which will not be available by default any more in 0.21.
If you need training scores, please set return train score=True
  warnings.warn(*warn args, **warn kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('mean_train_
score'), which will not be available by default any more in 0.21. If
you need training scores, please set return train score=True
  warnings.warn(*warn_args, **warn_kwargs)
/usr/local/lib/python3.6/dist-packages/sklearn/utils/deprecation.py:
125: FutureWarning: You are accessing a training score ('std_train_s
core'), which will not be available by default any more in 0.21. If
you need training scores, please set return train score=True
  warnings.warn(*warn args, **warn kwargs)
```

mean test score mean train score

param_hidden_size	param_regularization_strength			
32	0.00	0.735714	0.740476	
	0.01	0.735714	0.733333	
	0.05	0.600000	0.602381	
	0.10	0.692857	0.661905	
64	0.00	0.914286	0.883333	
	0.01	0.892857	0.838095	
	0.05	0.764286	0.771429	
	0.10	0.692857	0.685714	
256	0.00	0.971429	0.961905	
	0.01	0.964286	0.966667	
	0.05	0.950000	0.923810	
	0.10	0.914286	0.902381	

Task 2

```
In [5]:
(X train, y train), (X test, y test) = fashion mnist.load data()
X train = X train.reshape(60000, 784)
X_{\text{test}} = X_{\text{test.reshape}}(10000, 784)
X train = X train.astype('float32')
X test = X test.astype('float32')
X_train /= 255
X test /= 255
y train = keras.utils.to categorical(y train, 10)
y_test = keras.utils.to_categorical(y_test, 10)
Downloading data from http://fashion-mnist.s3-website.eu-central-1.a
mazonaws.com/train-labels-idx1-ubyte.gz
Downloading data from http://fashion-mnist.s3-website.eu-central-1.a
mazonaws.com/train-images-idx3-ubyte.gz
Downloading data from http://fashion-mnist.s3-website.eu-central-1.a
mazonaws.com/t10k-labels-idx1-ubyte.gz
8192/5148 [========= ] - 0s 0u
s/step
Downloading data from http://fashion-mnist.s3-website.eu-central-1.a
mazonaws.com/t10k-images-idx3-ubyte.gz
```

```
In [0]:
```

```
def make_model(drop_out = 0, hidden_size = 32, hidden_layes = 1):
    model = Sequential()
    model.add(Dense(hidden_size, input_dim=784, activation='relu'))
    model.add(Dropout(drop_out))
    for i in range(1, hidden_layes):
        model.add(Dense(hidden_size, activation='relu'))
        model.add(Dropout(drop_out))
        model.add(Dense(10, activation='softmax'))

    model.compile(optimizer='adam',loss="categorical_crossentropy",metrics=['accuracy'])
    return model
```

Vanilla Model

```
vanilla_model = make_model()
history_callback = vanilla_model.fit(X_train, y_train, batch_size=128, epochs=50
, verbose=1,validation_split=1/6)
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tenso rflow/python/framework/op def library.py:263: colocate with (from te nsorflow.python.framework.ops) is deprecated and will be removed in a future version. Instructions for updating: Colocations handled automatically by placer. WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tenso rflow/python/ops/math ops.py:3066: to int32 (from tensorflow.python. ops.math ops) is deprecated and will be removed in a future version. Instructions for updating: Use tf.cast instead. Train on 50000 samples, validate on 10000 samples Epoch 1/50 50000/50000 [============] - 4s 77us/step - loss: 0.7045 - acc: 0.7660 - val loss: 0.5054 - val acc: 0.8231 Epoch 2/50 0.4648 - acc: 0.8390 - val loss: 0.4541 - val acc: 0.8404 Epoch 3/50 50000/50000 [============] - 1s 26us/step - loss: 0.4263 - acc: 0.8516 - val loss: 0.4214 - val acc: 0.8531 50000/50000 [============] - 1s 26us/step - loss: 0.3975 - acc: 0.8620 - val loss: 0.4068 - val acc: 0.8588 Epoch 5/50 50000/50000 [============] - 1s 26us/step - loss: 0.3823 - acc: 0.8662 - val loss: 0.4011 - val acc: 0.8563 Epoch 6/50 50000/50000 [============] - 1s 26us/step - loss: 0.3656 - acc: 0.8725 - val loss: 0.3971 - val acc: 0.8585 Epoch 7/50 50000/50000 [=============] - 1s 26us/step - loss: 0.3582 - acc: 0.8740 - val loss: 0.3795 - val acc: 0.8637 Epoch 8/50 50000/50000 [============] - 1s 26us/step - loss: 0.3471 - acc: 0.8779 - val loss: 0.3760 - val acc: 0.8669 Epoch 9/50 50000/50000 [===========] - 1s 26us/step - loss: 0.3384 - acc: 0.8794 - val loss: 0.3801 - val acc: 0.8643 Epoch 10/50 50000/50000 [===========] - 2s 30us/step - loss: 0.3327 - acc: 0.8821 - val loss: 0.3675 - val acc: 0.8707 Epoch 11/50 50000/50000 [============] - 1s 30us/step - loss: 0.3283 - acc: 0.8821 - val loss: 0.3711 - val acc: 0.8678 Epoch 12/50 50000/50000 [============] - 2s 30us/step - loss: 0.3192 - acc: 0.8874 - val_loss: 0.3836 - val_acc: 0.8667 Epoch 13/50 50000/50000 [=============] - 2s 30us/step - loss: 0.3147 - acc: 0.8871 - val loss: 0.3763 - val acc: 0.8675 Epoch 14/50 50000/50000 [===========] - 2s 30us/step - loss: 0.3125 - acc: 0.8890 - val_loss: 0.3682 - val_acc: 0.8704 Epoch 15/50 50000/50000 [===========] - 2s 30us/step - loss: 0.3068 - acc: 0.8910 - val loss: 0.3595 - val acc: 0.8709 Epoch 16/50 50000/50000 [=============] - 1s 30us/step - loss: 0.3033 - acc: 0.8908 - val_loss: 0.3626 - val_acc: 0.8718

Epoch 17/50

```
50000/50000 [============= ] - 1s 26us/step - loss:
0.2998 - acc: 0.8928 - val loss: 0.3601 - val acc: 0.8736
Epoch 18/50
50000/50000 [=========== ] - 1s 26us/step - loss:
0.2949 - acc: 0.8935 - val loss: 0.3694 - val acc: 0.8689
Epoch 19/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2918 - acc: 0.8963 - val loss: 0.3742 - val acc: 0.8653
Epoch 20/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2901 - acc: 0.8958 - val loss: 0.3751 - val acc: 0.8699
Epoch 21/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2865 - acc: 0.8974 - val loss: 0.3552 - val acc: 0.8751
Epoch 22/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2830 - acc: 0.8982 - val loss: 0.3741 - val acc: 0.8692
Epoch 23/50
50000/50000 [===========] - 1s 26us/step - loss:
0.2801 - acc: 0.8985 - val loss: 0.3705 - val acc: 0.8708
Epoch 24/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2776 - acc: 0.9010 - val loss: 0.3640 - val acc: 0.8725
Epoch 25/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2736 - acc: 0.9015 - val_loss: 0.3544 - val_acc: 0.8760
Epoch 26/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2709 - acc: 0.9012 - val_loss: 0.3520 - val acc: 0.8738
Epoch 27/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2662 - acc: 0.9048 - val loss: 0.3607 - val acc: 0.8742
Epoch 28/50
50000/50000 [===========] - 1s 26us/step - loss:
0.2667 - acc: 0.9022 - val loss: 0.3571 - val acc: 0.8777
Epoch 29/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2629 - acc: 0.9057 - val loss: 0.3761 - val acc: 0.8675
Epoch 30/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2618 - acc: 0.9064 - val loss: 0.3530 - val acc: 0.8769
Epoch 31/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2611 - acc: 0.9053 - val loss: 0.3721 - val acc: 0.8685
Epoch 32/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2588 - acc: 0.9066 - val loss: 0.3595 - val acc: 0.8768
Epoch 33/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2545 - acc: 0.9086 - val_loss: 0.3534 - val_acc: 0.8790
Epoch 34/50
50000/50000 [=========== ] - 1s 27us/step - loss:
0.2503 - acc: 0.9097 - val loss: 0.3617 - val acc: 0.8718
Epoch 35/50
50000/50000 [============= ] - 1s 27us/step - loss:
0.2521 - acc: 0.9077 - val_loss: 0.3560 - val_acc: 0.8785
Epoch 36/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2476 - acc: 0.9098 - val loss: 0.3564 - val acc: 0.8788
Epoch 37/50
50000/50000 [============= ] - 1s 26us/step - loss:
```

```
0.2445 - acc: 0.9114 - val loss: 0.3596 - val acc: 0.8776
Epoch 38/50
50000/50000 [=========== ] - 1s 26us/step - loss:
0.2445 - acc: 0.9115 - val loss: 0.3680 - val acc: 0.8752
Epoch 39/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2398 - acc: 0.9138 - val loss: 0.3535 - val acc: 0.8791
Epoch 40/50
0.2411 - acc: 0.9125 - val loss: 0.3555 - val acc: 0.8773
Epoch 41/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2383 - acc: 0.9148 - val loss: 0.3738 - val acc: 0.8758
Epoch 42/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2368 - acc: 0.9148 - val_loss: 0.3721 - val acc: 0.8740
Epoch 43/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2356 - acc: 0.9148 - val_loss: 0.3534 - val_acc: 0.8783
Epoch 44/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2347 - acc: 0.9145 - val loss: 0.3619 - val acc: 0.8783
Epoch 45/50
50000/50000 [============= ] - 1s 27us/step - loss:
0.2324 - acc: 0.9167 - val loss: 0.3672 - val acc: 0.8753
Epoch 46/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2312 - acc: 0.9163 - val loss: 0.3659 - val acc: 0.8747
Epoch 47/50
50000/50000 [===========] - 1s 26us/step - loss:
0.2299 - acc: 0.9177 - val loss: 0.3700 - val acc: 0.8770
Epoch 48/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2258 - acc: 0.9188 - val loss: 0.3774 - val acc: 0.8722
Epoch 49/50
50000/50000 [============ ] - 1s 26us/step - loss:
0.2246 - acc: 0.9185 - val loss: 0.3880 - val acc: 0.8689
Epoch 50/50
50000/50000 [============= ] - 1s 26us/step - loss:
0.2257 - acc: 0.9188 - val loss: 0.3669 - val acc: 0.8782
```

vanilla_model.summary()

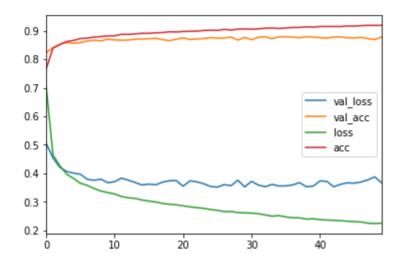
Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, 32)	25120
dropout_1 (Dropout)	(None, 32)	0
dense_2 (Dense)	(None, 10)	330

Total params: 25,450 Trainable params: 25,450 Non-trainable params: 0

pd.DataFrame(history_callback.history).plot()

Out[0]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0474626278>



A model using drop-out

In [14]:

drop_out_model = make_model(drop_out=0.5, hidden_size= 1024, hidden_layes= 2)
history_callback = drop_out_model.fit(X_train, y_train, batch_size=128, epochs=5
0, verbose=1,validation_split=1/6)

```
Train on 50000 samples, validate on 10000 samples
Epoch 1/50
50000/50000 [============ ] - 3s 54us/step - loss:
0.5923 - acc: 0.7856 - val loss: 0.4191 - val acc: 0.8435
Epoch 2/50
0.4509 - acc: 0.8359 - val loss: 0.4031 - val acc: 0.8530
Epoch 3/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.4181 - acc: 0.8492 - val loss: 0.3615 - val acc: 0.8652
Epoch 4/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3985 - acc: 0.8535 - val loss: 0.3637 - val acc: 0.8691
Epoch 5/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3898 - acc: 0.8575 - val loss: 0.3487 - val acc: 0.8739
Epoch 6/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3728 - acc: 0.8630 - val loss: 0.3483 - val acc: 0.8697
Epoch 7/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3637 - acc: 0.8666 - val loss: 0.3338 - val acc: 0.8748
Epoch 8/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3563 - acc: 0.8688 - val loss: 0.3354 - val acc: 0.8765
Epoch 9/50
50000/50000 [=========== ] - 2s 38us/step - loss:
0.3471 - acc: 0.8722 - val loss: 0.3459 - val acc: 0.8728
Epoch 10/50
50000/50000 [===========] - 2s 38us/step - loss:
0.3416 - acc: 0.8733 - val loss: 0.3214 - val acc: 0.8836
Epoch 11/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3314 - acc: 0.8772 - val loss: 0.3243 - val acc: 0.8805
Epoch 12/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3307 - acc: 0.8770 - val loss: 0.3236 - val acc: 0.8819
Epoch 13/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.3241 - acc: 0.8782 - val loss: 0.3169 - val acc: 0.8847
Epoch 14/50
50000/50000 [=========== ] - 2s 38us/step - loss:
0.3184 - acc: 0.8833 - val loss: 0.3225 - val acc: 0.8837
Epoch 15/50
50000/50000 [=========== ] - 2s 38us/step - loss:
0.3191 - acc: 0.8808 - val loss: 0.3136 - val acc: 0.8873
Epoch 16/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.3184 - acc: 0.8815 - val_loss: 0.3295 - val_acc: 0.8836
Epoch 17/50
50000/50000 [============ ] - 2s 37us/step - loss:
0.3146 - acc: 0.8832 - val_loss: 0.3206 - val_acc: 0.8854
Epoch 18/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.3049 - acc: 0.8857 - val loss: 0.3205 - val acc: 0.8864
Epoch 19/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.3013 - acc: 0.8876 - val loss: 0.3328 - val acc: 0.8828
Epoch 20/50
50000/50000 [=========== ] - 2s 38us/step - loss:
0.3000 - acc: 0.8880 - val loss: 0.3140 - val acc: 0.8866
```

```
Epoch 21/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2993 - acc: 0.8887 - val loss: 0.3085 - val acc: 0.8895
Epoch 22/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2933 - acc: 0.8904 - val loss: 0.3127 - val acc: 0.8872
Epoch 23/50
50000/50000 [=========== ] - 2s 37us/step - loss:
0.2940 - acc: 0.8902 - val loss: 0.3123 - val acc: 0.8853
Epoch 24/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2907 - acc: 0.8920 - val loss: 0.3163 - val acc: 0.8858
Epoch 25/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2853 - acc: 0.8939 - val loss: 0.3148 - val acc: 0.8896
Epoch 26/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2841 - acc: 0.8936 - val loss: 0.3053 - val acc: 0.8886
Epoch 27/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2815 - acc: 0.8937 - val loss: 0.3086 - val acc: 0.8897
Epoch 28/50
50000/50000 [===========] - 2s 40us/step - loss:
0.2823 - acc: 0.8932 - val loss: 0.3017 - val acc: 0.8950
Epoch 29/50
50000/50000 [============ ] - 2s 42us/step - loss:
0.2772 - acc: 0.8967 - val loss: 0.3199 - val acc: 0.8843
Epoch 30/50
50000/50000 [============= ] - 2s 42us/step - loss:
0.2752 - acc: 0.8952 - val_loss: 0.3064 - val acc: 0.8889
Epoch 31/50
50000/50000 [============ ] - 2s 42us/step - loss:
0.2732 - acc: 0.8987 - val loss: 0.3144 - val acc: 0.8865
Epoch 32/50
50000/50000 [============= ] - 2s 42us/step - loss:
0.2713 - acc: 0.8984 - val loss: 0.3010 - val acc: 0.8911
Epoch 33/50
50000/50000 [============ ] - 2s 39us/step - loss:
0.2702 - acc: 0.8986 - val loss: 0.2996 - val acc: 0.8928
Epoch 34/50
50000/50000 [===========] - 2s 38us/step - loss:
0.2636 - acc: 0.9004 - val loss: 0.2973 - val acc: 0.8913
Epoch 35/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2665 - acc: 0.9002 - val loss: 0.2984 - val acc: 0.8935
Epoch 36/50
50000/50000 [============ ] - 2s 37us/step - loss:
0.2631 - acc: 0.9015 - val loss: 0.3024 - val acc: 0.8896
Epoch 37/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2648 - acc: 0.9001 - val_loss: 0.2992 - val_acc: 0.8952
Epoch 38/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2631 - acc: 0.9010 - val loss: 0.2999 - val acc: 0.8940
Epoch 39/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2605 - acc: 0.9016 - val_loss: 0.2949 - val_acc: 0.8945
Epoch 40/50
50000/50000 [=========== ] - 2s 38us/step - loss:
0.2615 - acc: 0.9040 - val loss: 0.3066 - val acc: 0.8921
Epoch 41/50
```

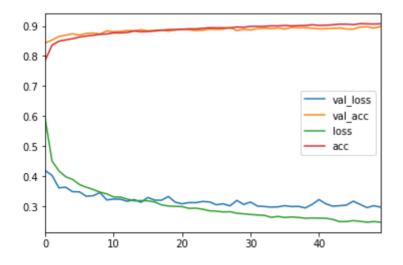
```
50000/50000 [============= ] - 2s 38us/step - loss:
0.2609 - acc: 0.9019 - val loss: 0.3226 - val acc: 0.8904
Epoch 42/50
50000/50000 [============= ] - 2s 37us/step - loss:
0.2604 - acc: 0.9024 - val loss: 0.3085 - val acc: 0.8908
Epoch 43/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2568 - acc: 0.9039 - val loss: 0.3007 - val acc: 0.8919
Epoch 44/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2492 - acc: 0.9059 - val loss: 0.3022 - val acc: 0.8933
Epoch 45/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2494 - acc: 0.9060 - val loss: 0.3043 - val acc: 0.8898
Epoch 46/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2526 - acc: 0.9043 - val loss: 0.3172 - val acc: 0.8893
Epoch 47/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2503 - acc: 0.9076 - val loss: 0.3066 - val acc: 0.8964
Epoch 48/50
50000/50000 [============= ] - 2s 38us/step - loss:
0.2474 - acc: 0.9071 - val loss: 0.2959 - val acc: 0.8973
Epoch 49/50
50000/50000 [============ ] - 2s 38us/step - loss:
0.2495 - acc: 0.9064 - val loss: 0.3022 - val acc: 0.8930
Epoch 50/50
50000/50000 [============= ] - 2s 37us/step - loss:
0.2474 - acc: 0.9072 - val loss: 0.2972 - val acc: 0.8980
```

In [15]:

```
pd.DataFrame(history callback.history).plot()
```

Out[15]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8240616080>



In [16]:

```
drop out model.summary()
```

Layer (type)	Output	Shape	Param #
dense_19 (Dense)	(None,	1024)	803840
dropout_4 (Dropout)	(None,	1024)	0
dense_20 (Dense)	(None,	1024)	1049600
dropout_5 (Dropout)	(None,	1024)	0
dense_21 (Dense)	(None,	10)	10250
Total params: 1,863,690 Trainable params: 1,863,690 Non-trainable params: 0			

a model using batch normalization and residual connections

```
inputs = Input(shape=(784,))
h1 1 = Dense(128, activation='relu')(inputs)
h1 2 = BatchNormalization()(h1 1)
h2_1 = Dense(128, activation='relu')(h1_2)
h2_2 = BatchNormalization()(h2_1)
h3_1 = Dense(128, activation='relu')(h2_2)
h3_2 = BatchNormalization()(h3_1)
skip1 = add([h1 2,h3 2])
skip1 bn = BatchNormalization()(skip1)
h4_1 = Dense(128, activation = 'relu')(skip1_bn)
h4 2 = BatchNormalization()(h4_1)
h5_1 = Dense(128, activation = 'relu')(h4_2)
h5 2 = BatchNormalization()(h5 1)
skip2 = add([skip1 bn, h5 2])
skip2 bn = BatchNormalization()(skip2)
h6_1 = Dense(128, activation = 'relu')(skip2_bn)
h6_2 = BatchNormalization()(h6_1)
predictions = Dense(10, activation='softmax')(h6 2)
```

In [7]:

```
model = Model(inputs=inputs, outputs=predictions)
model.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accurac
y'])
history_callback = model.fit(X_train, y_train, batch_size=128, epochs=50, verbos
e=1,validation_split=1/6)
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tenso
rflow/python/ops/math ops.py:3066: to int32 (from tensorflow.python.
ops.math ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 50000 samples, validate on 10000 samples
Epoch 1/50
50000/50000 [============ ] - 10s 200us/step - los
s: 0.4924 - acc: 0.8241 - val loss: 0.4546 - val acc: 0.8290
Epoch 2/50
50000/50000 [============= ] - 6s 127us/step - loss:
0.3566 - acc: 0.8695 - val loss: 0.3755 - val acc: 0.8637
Epoch 3/50
50000/50000 [============= ] - 6s 120us/step - loss:
0.3205 - acc: 0.8809 - val loss: 0.4132 - val acc: 0.8445
Epoch 4/50
50000/50000 [============= ] - 6s 113us/step - loss:
0.2944 - acc: 0.8911 - val loss: 0.3867 - val acc: 0.8544
Epoch 5/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.2785 - acc: 0.8957 - val loss: 0.3690 - val acc: 0.8661
50000/50000 [============= ] - 6s 114us/step - loss:
0.2629 - acc: 0.9028 - val_loss: 0.3555 - val_acc: 0.8674
Epoch 7/50
50000/50000 [============= ] - 6s 113us/step - loss:
0.2519 - acc: 0.9071 - val loss: 0.3475 - val acc: 0.8727
Epoch 8/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.2364 - acc: 0.9124 - val loss: 0.3445 - val acc: 0.8752
Epoch 9/50
50000/50000 [============ ] - 6s 113us/step - loss:
0.2240 - acc: 0.9161 - val loss: 0.3688 - val acc: 0.8736
Epoch 10/50
50000/50000 [============= ] - 6s 114us/step - loss:
0.2138 - acc: 0.9195 - val loss: 0.3370 - val acc: 0.8793
Epoch 11/50
50000/50000 [=========== ] - 6s 113us/step - loss:
0.2057 - acc: 0.9224 - val_loss: 0.3506 - val_acc: 0.8806
Epoch 12/50
50000/50000 [=========== ] - 6s 114us/step - loss:
0.1987 - acc: 0.9252 - val loss: 0.3312 - val acc: 0.8849
Epoch 13/50
50000/50000 [============ ] - 6s 112us/step - loss:
0.1923 - acc: 0.9268 - val loss: 0.3541 - val acc: 0.8774
Epoch 14/50
50000/50000 [============ ] - 6s 111us/step - loss:
0.1811 - acc: 0.9312 - val_loss: 0.3523 - val_acc: 0.8766
Epoch 15/50
50000/50000 [============== ] - 6s 115us/step - loss:
0.1670 - acc: 0.9371 - val loss: 0.4050 - val acc: 0.8687
Epoch 16/50
50000/50000 [=========== ] - 6s 127us/step - loss:
0.1625 - acc: 0.9377 - val_loss: 0.3836 - val_acc: 0.8747
Epoch 17/50
50000/50000 [===========] - 6s 118us/step - loss:
0.1540 - acc: 0.9413 - val loss: 0.3742 - val acc: 0.8779
Epoch 18/50
50000/50000 [============== ] - 6s 114us/step - loss:
0.1484 - acc: 0.9428 - val_loss: 0.3785 - val_acc: 0.8752
Epoch 19/50
```

```
50000/50000 [============== ] - 6s 113us/step - loss:
0.1446 - acc: 0.9449 - val loss: 0.4523 - val acc: 0.8545
Epoch 20/50
50000/50000 [============= ] - 6s 119us/step - loss:
0.1369 - acc: 0.9481 - val loss: 0.3829 - val acc: 0.8786
Epoch 21/50
50000/50000 [============== ] - 6s 118us/step - loss:
0.1286 - acc: 0.9511 - val loss: 0.3895 - val acc: 0.8816
Epoch 22/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.1263 - acc: 0.9522 - val loss: 0.3980 - val acc: 0.8885
Epoch 23/50
50000/50000 [============== ] - 6s 119us/step - loss:
0.1169 - acc: 0.9546 - val loss: 0.4135 - val acc: 0.8757
Epoch 24/50
50000/50000 [============ ] - 6s 128us/step - loss:
0.1150 - acc: 0.9556 - val loss: 0.4353 - val acc: 0.8735
Epoch 25/50
50000/50000 [============ ] - 6s 116us/step - loss:
0.1177 - acc: 0.9552 - val loss: 0.4077 - val acc: 0.8824
Epoch 26/50
50000/50000 [============== ] - 6s 112us/step - loss:
0.1050 - acc: 0.9597 - val loss: 0.4175 - val acc: 0.8828
Epoch 27/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.1003 - acc: 0.9608 - val loss: 0.4234 - val acc: 0.8856
Epoch 28/50
50000/50000 [============= ] - 6s 111us/step - loss:
0.0964 - acc: 0.9630 - val_loss: 0.4271 - val acc: 0.8836
Epoch 29/50
50000/50000 [============= ] - 6s 117us/step - loss:
0.0906 - acc: 0.9658 - val loss: 0.4557 - val acc: 0.8824
Epoch 30/50
50000/50000 [=========== ] - 6s 127us/step - loss:
0.0899 - acc: 0.9657 - val loss: 0.4676 - val acc: 0.8801
Epoch 31/50
50000/50000 [============= ] - 6s 115us/step - loss:
0.0858 - acc: 0.9675 - val loss: 0.4892 - val acc: 0.8765
Epoch 32/50
50000/50000 [============ ] - 6s 112us/step - loss:
0.0863 - acc: 0.9671 - val loss: 0.4927 - val acc: 0.8764
50000/50000 [============ ] - 6s 112us/step - loss:
0.0800 - acc: 0.9689 - val loss: 0.4682 - val acc: 0.8838
Epoch 34/50
50000/50000 [============== ] - 6s 112us/step - loss:
0.0783 - acc: 0.9704 - val_loss: 0.4880 - val_acc: 0.8803
Epoch 35/50
50000/50000 [============ ] - 6s 110us/step - loss:
0.0718 - acc: 0.9731 - val_loss: 0.4976 - val_acc: 0.8826
Epoch 36/50
50000/50000 [===========] - 6s 111us/step - loss:
0.0740 - acc: 0.9715 - val loss: 0.5178 - val acc: 0.8809
Epoch 37/50
50000/50000 [============== ] - 6s 111us/step - loss:
0.0719 - acc: 0.9723 - val_loss: 0.5041 - val_acc: 0.8840
Epoch 38/50
50000/50000 [============= ] - 6s 111us/step - loss:
0.0691 - acc: 0.9734 - val loss: 0.5224 - val acc: 0.8794
Epoch 39/50
50000/50000 [============== ] - 6s 110us/step - loss:
```

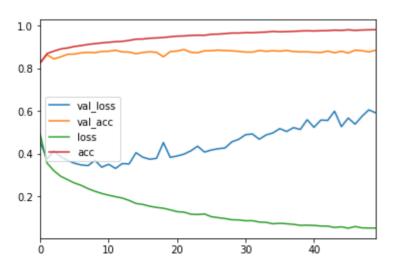
```
0.0645 - acc: 0.9754 - val loss: 0.5138 - val acc: 0.8774
Epoch 40/50
50000/50000 [============ ] - 5s 110us/step - loss:
0.0650 - acc: 0.9764 - val loss: 0.5594 - val acc: 0.8777
Epoch 41/50
50000/50000 [============= ] - 6s 111us/step - loss:
0.0640 - acc: 0.9754 - val loss: 0.5238 - val acc: 0.8756
Epoch 42/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.0614 - acc: 0.9766 - val loss: 0.5579 - val acc: 0.8744
Epoch 43/50
50000/50000 [============ ] - 6s 114us/step - loss:
0.0607 - acc: 0.9769 - val loss: 0.5562 - val acc: 0.8812
Epoch 44/50
0.0546 - acc: 0.9794 - val loss: 0.5993 - val acc: 0.8731
Epoch 45/50
50000/50000 [============== ] - 6s 118us/step - loss:
0.0574 - acc: 0.9781 - val_loss: 0.5268 - val_acc: 0.8807
Epoch 46/50
50000/50000 [============= ] - 6s 111us/step - loss:
0.0510 - acc: 0.9810 - val loss: 0.5673 - val acc: 0.8719
Epoch 47/50
50000/50000 [============== ] - 6s 111us/step - loss:
0.0589 - acc: 0.9781 - val loss: 0.5387 - val acc: 0.8854
Epoch 48/50
50000/50000 [============= ] - 6s 112us/step - loss:
0.0527 - acc: 0.9800 - val loss: 0.5749 - val acc: 0.8831
Epoch 49/50
50000/50000 [============= ] - 6s 111us/step - loss:
0.0517 - acc: 0.9809 - val loss: 0.6057 - val acc: 0.8775
Epoch 50/50
50000/50000 [============== ] - 6s 112us/step - loss:
0.0518 - acc: 0.9815 - val loss: 0.5918 - val acc: 0.8850
```

In [8]:

```
pd.DataFrame(history callback.history).plot()
```

Out[8]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8257909320>



In [9]:

model.summary()

Layer (type) nected to	Output	Shape	Param #	Con
<pre>input_2 (InputLayer)</pre>	(None,	784)	0	
dense_8 (Dense) ut_2[0][0]	(None,	128)	100480	inp
batch_normalization_9 (BatchNor se_8[0][0]	(None,	128)	512	den
dense_9 (Dense) ch_normalization_9[0][0]	(None,	128)	16512	bat
batch_normalization_10 (BatchNo se_9[0][0]	(None,	128)	512	den
dense_10 (Dense) ch_normalization_10[0][0]	(None,	128)	16512	bat
batch_normalization_11 (BatchNo se_10[0][0]	(None,	128)	512	den
add_3 (Add) ch_normalization_9[0][0]	(None,	128)	0	bat
ch_normalization_11[0][0]				bat
batch_normalization_12 (BatchNo _3[0][0]	(None,	128)	512	add
dense_11 (Dense) ch_normalization_12[0][0]	(None,	128)	16512	bat
batch_normalization_13 (BatchNo se_11[0][0]	(None,	128)	512	den
dense_12 (Dense) ch_normalization_13[0][0]	(None,	128)	16512	bat
batch_normalization_14 (BatchNo se_12[0][0]	(None,	128)	512	den
add_4 (Add) ch_normalization_12[0][0]	(None,	128)	0	bat

_	_	_
0	а	т

ch	_normalization_	14	[0]	[0]

batch_normalization_15 (BatchNo _4[0][0]	(None,	128)	512	add
dense_13 (Dense) ch_normalization_15[0][0]	(None,	128)	16512	bat
batch_normalization_16 (BatchNo se_13[0][0]	(None,	128)	512	den
dense_14 (Dense) ch_normalization_16[0][0]	(None,	10)	1290	bat
Total params: 188,426	 _			-

Total params: 188,426
Trainable params: 186,378
Non-trainable params: 2,048

HW5 Preparation

April 27, 2019

1 HW5 Data Preparation

```
In [282]: from sklearn.model_selection import train_test_split
          import pandas as pd
          import numpy as np
          import os
          from glob import glob
          import matplotlib.pylab as plt
          %matplotlib inline
  reference:
               http://www.degeneratestate.org/posts/2016/Oct/23/image-processing-with-
numpy/
In [8]: print(os.listdir("IDC_regular_ps50_idx5"))
['9036', '10268', '10257', '8913', '13613', '8914', '15510', '10259', '16165', '10292', '12951
In [20]: Data = glob('IDC_regular_ps50_idx5/**/*.png', recursive=True)
In [23]: len(Data)
Out[23]: 277524
In [27]: pwd
Out[27]: '/Users/zechen/Desktop/AML/Homework/HW5'
In [46]: import cv2
         import glob
         import numpy as np
In [71]: folder_lst = os.listdir("IDC_regular_ps50_idx5")
In [73]: train = []
         train_labels = []
In [181]: image_train = np.ndarray((277524, 50, 50,3),dtype = np.uint8)
          image_train.shape
```

```
Out[181]: (277524, 50, 50, 3)
In [62]: img = cv2.imread('IDC_regular_ps50_idx5/9036/0/9036_idx5_x2151_y1301_class0.png')
In [ ]: files = glob.glob ("/data/train/class1/*.png") # your image path
        for myFile in files:
            image = cv2.imread (myFile)
            train.append (image)
            train_labels.append([1., 0.])
In [198]: folder_lst = os.listdir("IDC_regular_ps50_idx5")
          image_train = np.ndarray((277524, 50, 50,3),dtype = np.uint8)
          train = []
          train_labels = []
          i=0
          for item in folder_lst:
              path = "IDC_regular_ps50_idx5/*"+item+"/1/*.png"
              files = glob.glob(path) # your image path
              for myFile in files:
                  image = cv2.imread(myFile)
                  #print(type(train))
                  #print(train)
                  image_train[i] = cv2.resize(image,(50,50))
                  train_labels.append([1., 0.])
                  i+=1
              path2 = "IDC_regular_ps50_idx5/*"+item+"/0/*.png"
              files2 = glob.glob(path2)
              for myFile2 in files2:
                  image2 = cv2.imread(myFile2)
                  image_train[i] = cv2.resize(image2,(50,50))
                  train_labels.append([0., 1.])
                  i+=1
In [207]: train_labels = np.array(train_labels)
          np.save('train',train)
          np.save('train_labels',train_labels)
          np.save('image_train',image_train)
In [304]: image_train = np.load('image_train.npy')
          train_labels = np.load('train_labels.npy')
In [326]: X_train_sub, X_test_sub, y_train_sub, y_test_sub = train_test_split(image_train,train)
In [327]: X_train_sub.shape
Out[327]: (61055, 50, 50, 3)
In [328]: X_test_sub.shape
```

```
Out[328]: (216469, 50, 50, 3)
In [329]: np.save('X_train_sub',X_train_sub)
          np.save('y_train_sub',y_train_sub)
  Plot numpy array to image to double check
In [261]: def plti(im, h=8, **kwargs):
             Helper function to plot an image.
             y = im.shape[0]
             x = im.shape[1]
             w = (y/x) * h
             plt.figure(figsize=(w,h))
             plt.imshow(im, interpolation="none", **kwargs)
             plt.axis('off')
In [260]: im = plt.imread("IDC_regular_ps50_idx5/9036/0/9036_idx5_x2151_y1301_class0.png")
Out[260]: array([[[0.9607843 , 0.9372549 , 0.9529412 ],
                  [0.93333334, 0.88235295, 0.9137255],
                  [0.81960785, 0.6313726, 0.74509805],
                  [0.8980392, 0.8117647, 0.88235295],
                  [0.91764706, 0.8392157, 0.88235295],
                  [0.9137255, 0.8235294, 0.8862745]],
                 [[0.95686275, 0.9411765, 0.9529412],
                  [0.9490196, 0.92941177, 0.9529412],
                  [0.87058824, 0.7607843 , 0.8235294 ],
                  [0.9647059, 0.9372549, 0.9411765],
                  [0.9372549, 0.8745098, 0.92156863],
                  [0.84313726, 0.654902 , 0.7490196 ]],
                 [[0.9647059, 0.94509804, 0.9607843],
                  [0.92941177, 0.8862745, 0.9098039],
                  [0.8980392, 0.8039216, 0.87058824],
                  [0.8627451, 0.70980394, 0.8039216],
                  [0.8235294, 0.64705884, 0.76862746],
                  [0.79607844, 0.61960787, 0.733333335]],
                 . . . ,
                 [[0.73333335, 0.50980395, 0.64705884],
                  [0.8156863, 0.5686275, 0.7058824],
```

```
[0.7921569, 0.53333336, 0.6666667],
                  . . . ,
                  [0.9529412, 0.94509804, 0.9529412],
                  [0.9490196, 0.9607843, 0.94509804],
                  [0.9607843, 0.9490196, 0.96862745]],
                 [[0.7294118 , 0.44705883 , 0.627451 ],
                  [0.8627451, 0.64705884, 0.76862746],
                  [0.72156864, 0.39607844, 0.5647059],
                  [0.95686275, 0.9607843, 0.95686275],
                  [0.9607843, 0.9490196, 0.9607843],
                  [0.95686275, 0.94509804, 0.95686275]],
                                   , 0.5803922 ],
                 [[0.7176471 , 0.4
                  [0.7411765, 0.4117647, 0.5764706],
                  [0.6156863, 0.27058825, 0.43529412],
                  [0.95686275, 0.9607843, 0.95686275],
                  [0.9647059, 0.95686275, 0.96862745],
                  [0.94509804, 0.9647059 , 0.95686275]]], dtype=float32)
In [161]: train.shape
Out[161]: (277524,)
In [262]: plti(im)
```



Task 3

```
In [0]:
from google.colab import drive
drive.mount('/content/drive')
Go to this URL in a browser: https://accounts.google.com/o/oauth2/au
th?client id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.goog
leusercontent.com&redirect uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&s
cope=email%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdocs.test%20h
ttps%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive%20https%3A%2F%2Fwww.g
oogleapis.com%2Fauth%2Fdrive.photos.readonly%20https%3A%2F%2Fwww.goo
gleapis.com%2Fauth%2Fpeopleapi.readonly&response type=code
Enter your authorization code:
Mounted at /content/drive
In [0]:
cd "/content/drive/My Drive/Colab Notebooks"
/content/drive/My Drive/Colab Notebooks
In [0]:
pwd
Out[0]:
'/content/drive/My Drive/Colab Notebooks'
In [0]:
image train =np.load('image train.npy')
In [0]:
image label = np.load('train labels.npy')
In [0]:
image train.shape
Out[0]:
(277524, 50, 50, 3)
In [0]:
image label.shape
Out[0]:
(277524, 2)
```

```
In [0]:

X_train, X_test, y_train, y_test = train_test_split(image_train,image_label)

In [0]:

X_train.shape

Out[0]:
(208143, 50, 50, 3)
```

3.1 Start with a model without residual connections (using batch normalization is likely to be helpful and you should try it, whether you use dropout is your choice)

Scenario1: use all data

Build CNN Model: use Batch Normalization and Dropout

```
from keras.layers import Conv2D, MaxPooling2D, Flatten
input\_shape = (50,50,3)
num classes = 2
cnn = Sequential()
cnn.add(Conv2D(32, kernel size=(3, 3),
                 activation='relu',
                 input shape=input shape))
cnn.add(BatchNormalization())
cnn.add(MaxPooling2D(pool size=(2, 2)))
cnn.add(Dropout(0.2))
cnn.add(Conv2D(32, (3, 3), activation='relu'))
cnn.add(BatchNormalization())
cnn.add(MaxPooling2D(pool_size=(2, 2)))
cnn.add(Dropout(0.2))
cnn.add(Flatten())
cnn.add(Dense(64, activation='relu'))
cnn.add(Dense(num classes, activation='softmax'))
```

cnn.summary()

Layer (type)	Output	Shape	Param #
conv2d_3 (Conv2D)	(None,	48, 48, 32)	896
batch_normalization_3 (Batch	(None,	48, 48, 32)	128
max_pooling2d_3 (MaxPooling2	(None,	24, 24, 32)	0
dropout_3 (Dropout)	(None,	24, 24, 32)	0
conv2d_4 (Conv2D)	(None,	22, 22, 32)	9248
batch_normalization_4 (Batch	(None,	22, 22, 32)	128
max_pooling2d_4 (MaxPooling2	(None,	11, 11, 32)	0
dropout_4 (Dropout)	(None,	11, 11, 32)	0
flatten_2 (Flatten)	(None,	3872)	0
dense_3 (Dense)	(None,	64)	247872
dense_4 (Dense)	(None,	2)	130

Total params: 258,402 Trainable params: 258,274 Non-trainable params: 128

```
cnn.compile("adam", "categorical_crossentropy", metrics=['accuracy'])
```

```
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tenso
rflow/python/ops/math ops.py:3066: to int32 (from tensorflow.python.
ops.math ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 187328 samples, validate on 20815 samples
Epoch 1/20
ss: 0.3555 - acc: 0.8501 - val loss: 0.5466 - val acc: 0.7074
Epoch 2/20
ss: 0.3185 - acc: 0.8655 - val loss: 0.4521 - val acc: 0.8310
ss: 0.3060 - acc: 0.8714 - val loss: 0.4079 - val acc: 0.8251
Epoch 4/20
ss: 0.2971 - acc: 0.8748 - val loss: 0.3517 - val acc: 0.8480
Epoch 5/20
ss: 0.2862 - acc: 0.8794 - val loss: 0.3880 - val acc: 0.8276
ss: 0.2735 - acc: 0.8851 - val_loss: 0.5506 - val acc: 0.8051
Epoch 7/20
ss: 0.2599 - acc: 0.8909 - val loss: 0.3313 - val acc: 0.8618
Epoch 8/20
ss: 0.2431 - acc: 0.8990 - val loss: 0.3370 - val acc: 0.8588
Epoch 9/20
ss: 0.2254 - acc: 0.9071 - val loss: 0.6006 - val acc: 0.7821
Epoch 10/20
ss: 0.2078 - acc: 0.9148 - val loss: 0.4527 - val acc: 0.8449
Epoch 11/20
ss: 0.1908 - acc: 0.9223 - val loss: 0.6321 - val acc: 0.8492
Epoch 12/20
ss: 0.1746 - acc: 0.9289 - val loss: 0.4963 - val acc: 0.8359
Epoch 13/20
ss: 0.1601 - acc: 0.9352 - val loss: 0.5364 - val acc: 0.8464
Epoch 14/20
ss: 0.1467 - acc: 0.9413 - val_loss: 0.5272 - val_acc: 0.8429
Epoch 15/20
ss: 0.1345 - acc: 0.9464 - val loss: 0.5776 - val acc: 0.8341
Epoch 16/20
ss: 0.1232 - acc: 0.9505 - val_loss: 0.6100 - val_acc: 0.7970
Epoch 17/20
ss: 0.1142 - acc: 0.9541 - val loss: 0.6027 - val acc: 0.8312
Epoch 18/20
ss: 0.1076 - acc: 0.9570 - val_loss: 0.7718 - val_acc: 0.8174
Epoch 19/20
```

```
ss: 0.0994 - acc: 0.9605 - val loss: 0.7617 - val acc: 0.8479
Epoch 20/20
ss: 0.0925 - acc: 0.9635 - val loss: 0.7913 - val acc: 0.8189
In [0]:
score = cnn.evaluate(X test, y test, verbose=0)
print("Test loss: {:.3f}".format(score[0]))
print("Test Accuracy: {:.3f}".format(score[1]))
Test loss: 0.780
Test Accuracy: 0.824
Scenario2: Use Subsample for Task3.1
In [0]:
X_train_sub =np.load('X_train_sub.npy')
In [0]:
y_train_sub =np.load('y_train_sub.npy')
In [0]:
X_train_sub, X_test_sub, y_train_sub, y_test_sub = train_test_split(X_train_sub,
y_train_sub)
In [0]:
X_train_sub.shape
Out[0]:
(45791, 50, 50, 3)
```

```
Train on 45791 samples, validate on 15264 samples
Epoch 1/50
45791/45791 [============== ] - 15s 317us/step - los
s: 0.3916 - acc: 0.8350 - val_loss: 0.9729 - val acc: 0.7536
Epoch 2/50
0.3479 - acc: 0.8515 - val loss: 0.3942 - val acc: 0.8346
Epoch 3/50
0.3359 - acc: 0.8572 - val loss: 0.5593 - val acc: 0.7846
Epoch 4/50
0.3291 - acc: 0.8598 - val loss: 0.4644 - val acc: 0.8318
Epoch 5/50
0.3241 - acc: 0.8634 - val_loss: 0.6799 - val_acc: 0.7937
Epoch 6/50
0.3136 - acc: 0.8687 - val loss: 0.6957 - val acc: 0.7936
Epoch 7/50
0.3054 - acc: 0.8733 - val loss: 0.5825 - val acc: 0.7936
Epoch 8/50
0.2937 - acc: 0.8765 - val loss: 0.4609 - val acc: 0.7934
Epoch 9/50
0.2875 - acc: 0.8798 - val loss: 0.4353 - val acc: 0.8219
Epoch 10/50
0.2784 - acc: 0.8838 - val loss: 0.3600 - val acc: 0.8497
Epoch 11/50
0.2649 - acc: 0.8905 - val loss: 0.7892 - val acc: 0.6985
Epoch 12/50
0.2537 - acc: 0.8953 - val_loss: 0.5498 - val_acc: 0.7740
Epoch 13/50
0.2414 - acc: 0.9012 - val loss: 0.4692 - val acc: 0.8339
Epoch 14/50
0.2309 - acc: 0.9057 - val_loss: 1.0478 - val_acc: 0.6468
Epoch 15/50
0.2196 - acc: 0.9117 - val loss: 0.4777 - val acc: 0.8388
Epoch 16/50
0.2045 - acc: 0.9164 - val_loss: 0.5328 - val_acc: 0.8244
Epoch 17/50
0.1940 - acc: 0.9210 - val_loss: 0.4414 - val_acc: 0.8407
Epoch 18/50
0.1864 - acc: 0.9254 - val loss: 0.4724 - val acc: 0.8318
Epoch 19/50
0.1712 - acc: 0.9310 - val loss: 0.4837 - val acc: 0.8349
Epoch 20/50
0.1688 - acc: 0.9320 - val loss: 1.1887 - val acc: 0.6885
```

```
Epoch 21/50
0.1597 - acc: 0.9352 - val loss: 0.6234 - val acc: 0.8413
Epoch 22/50
0.1514 - acc: 0.9397 - val loss: 1.4490 - val acc: 0.6929
Epoch 23/50
0.1415 - acc: 0.9434 - val loss: 0.6517 - val acc: 0.8443
Epoch 24/50
0.1375 - acc: 0.9456 - val loss: 0.4853 - val acc: 0.8277
Epoch 25/50
0.1302 - acc: 0.9481 - val loss: 1.1942 - val acc: 0.7744
Epoch 26/50
0.1234 - acc: 0.9505 - val loss: 0.6204 - val acc: 0.8064
Epoch 27/50
0.1198 - acc: 0.9518 - val loss: 1.1603 - val acc: 0.7250
Epoch 28/50
0.1193 - acc: 0.9522 - val loss: 0.8342 - val acc: 0.7938
Epoch 29/50
0.1113 - acc: 0.9556 - val loss: 1.2644 - val acc: 0.8013
Epoch 30/50
0.1086 - acc: 0.9565 - val loss: 0.7301 - val acc: 0.8362
Epoch 31/50
0.1017 - acc: 0.9598 - val_loss: 0.8843 - val_acc: 0.8333
Epoch 32/50
0.0982 - acc: 0.9610 - val loss: 1.1191 - val acc: 0.7497
Epoch 33/50
0.0975 - acc: 0.9618 - val loss: 1.8316 - val acc: 0.7798
0.0941 - acc: 0.9630 - val_loss: 1.1207 - val_acc: 0.7819
Epoch 35/50
0.0912 - acc: 0.9643 - val loss: 0.9404 - val acc: 0.7990
Epoch 36/50
0.0932 - acc: 0.9636 - val loss: 0.9926 - val acc: 0.8153
Epoch 37/50
0.0892 - acc: 0.9648 - val_loss: 2.0138 - val_acc: 0.7678
Epoch 38/50
0.0866 - acc: 0.9656 - val loss: 0.6101 - val acc: 0.8394
Epoch 39/50
0.0817 - acc: 0.9680 - val_loss: 0.6695 - val_acc: 0.8276
Epoch 40/50
0.0792 - acc: 0.9696 - val loss: 0.7427 - val acc: 0.8411
Epoch 41/50
```

```
0.0792 - acc: 0.9686 - val loss: 0.8608 - val acc: 0.8241
Epoch 42/50
0.0786 - acc: 0.9693 - val loss: 0.6481 - val acc: 0.8126
Epoch 43/50
0.0737 - acc: 0.9714 - val loss: 0.7935 - val acc: 0.8001
Epoch 44/50
0.0768 - acc: 0.9696 - val loss: 0.9980 - val acc: 0.8176
Epoch 45/50
0.0735 - acc: 0.9702 - val loss: 1.1493 - val acc: 0.7911
Epoch 46/50
0.0725 - acc: 0.9721 - val loss: 0.6474 - val acc: 0.8382
Epoch 47/50
0.0671 - acc: 0.9731 - val loss: 0.6795 - val acc: 0.8429
Epoch 48/50
0.0699 - acc: 0.9728 - val loss: 0.7817 - val acc: 0.8214
Epoch 49/50
0.0676 - acc: 0.9733 - val loss: 0.7673 - val acc: 0.8484
Epoch 50/50
0.0674 - acc: 0.9738 - val_loss: 1.2057 - val_acc: 0.8204
```

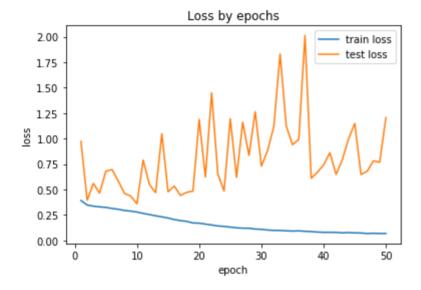
```
score = cnn.evaluate(X_test_sub, y_test_sub, verbose=0)
print("Test loss: {:.3f}".format(score[0]))
print("Test Accuracy: {:.3f}".format(score[1]))
```

Test loss: 1.206
Test Accuracy: 0.820

```
x=np.arange(1,51)
plt.plot(x,history_cnn_subsample.history['loss'],label='train loss')
plt.plot(x,history_cnn_subsample.history['val_loss'], label = 'test loss')
plt.legend()
plt.xlabel('epoch')
plt.ylabel('loss')
plt.title('Loss by epochs')
```

Out[0]:

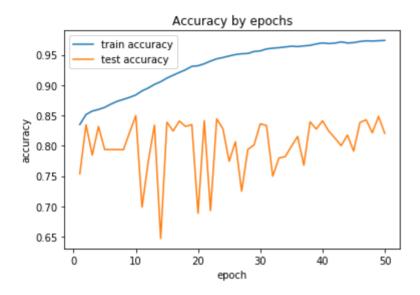
Text(0.5, 1.0, 'Loss by epochs')



```
x=np.arange(1,51)
plt.plot(x,history_cnn_subsample.history['acc'],label='train accuracy')
plt.plot(x,history_cnn_subsample.history['val_acc'], label = 'test accuracy')
plt.legend()
plt.xlabel('epoch')
plt.ylabel('accuracy')
plt.title('Accuracy by epochs')
```

Out[0]:

Text(0.5, 1.0, 'Accuracy by epochs')



3.2 Augment the data using rotations, mirroring and possibly other transformations. How much can you improve your original model by data augmentation?

```
input shape = (50, 50, 3)
num_classes = 2
cnn = Sequential()
cnn.add(Conv2D(32, kernel size=(3, 3),
                 activation='relu',
                 input shape=input shape))
cnn.add(BatchNormalization())
cnn.add(MaxPooling2D(pool_size=(2, 2)))
cnn.add(Dropout(0.2))
cnn.add(Conv2D(32, (3, 3), activation='relu'))
cnn.add(BatchNormalization())
cnn.add(MaxPooling2D(pool size=(2, 2)))
cnn.add(Dropout(0.2))
cnn.add(Flatten())
cnn.add(Dense(64, activation='relu'))
cnn.add(Dense(num classes, activation='softmax'))
cnn.compile("adam", "categorical crossentropy", metrics=['accuracy'])
```

```
from keras.preprocessing.image import ImageDataGenerator, array_to_img, img_to_a
rray, load_img

datagen2 = ImageDataGenerator(
    featurewise_center=True,
    featurewise_std_normalization=True,
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True)
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tenso rflow/python/ops/math_ops.py:3066: to_int32 (from tensorflow.python.ops.math_ops) is deprecated and will be removed in a future version. Instructions for updating:

Use tf.cast instead.

Epoch 1/20

/usr/local/lib/python3.6/dist-packages/keras_preprocessing/image/image_data_generator.py:699: UserWarning: This ImageDataGenerator specifies `featurewise_center`, but it hasn't been fit on any training data. Fit it first by calling `.fit(numpy data)`.

warnings.warn('This ImageDataGenerator specifies '/usr/local/lib/python3.6/dist-packages/keras_preprocessing/image/image_data_generator.py:707: UserWarning: This ImageDataGenerator specifies `featurewise std normalization`, but it hasn't been fit on any

training data. Fit it first by calling `.fit(numpy_data)`.
 warnings.warn('This ImageDataGenerator specifies '

```
0.3690 - acc: 0.8442 - val loss: 0.3614 - val acc: 0.8470
Epoch 2/20
0.3417 - acc: 0.8543 - val loss: 0.4654 - val acc: 0.8274
Epoch 3/20
0.3308 - acc: 0.8589 - val loss: 0.3374 - val acc: 0.8609
Epoch 4/20
0.3258 - acc: 0.8613 - val loss: 0.4197 - val acc: 0.8475
Epoch 5/20
0.3219 - acc: 0.8627 - val loss: 0.8340 - val acc: 0.7742
0.3189 - acc: 0.8644 - val loss: 0.3524 - val acc: 0.8578
Epoch 7/20
0.3160 - acc: 0.8654 - val loss: 0.3353 - val acc: 0.8566
Epoch 8/20
0.3131 - acc: 0.8664 - val loss: 0.4391 - val acc: 0.8380
Epoch 9/20
0.3101 - acc: 0.8676 - val_loss: 0.4995 - val_acc: 0.8466
Epoch 10/20
0.3086 - acc: 0.8682 - val loss: 0.4923 - val acc: 0.8445
Epoch 11/20
0.3059 - acc: 0.8689 - val_loss: 0.6132 - val_acc: 0.6944
Epoch 12/20
0.3043 - acc: 0.8696 - val loss: 0.3482 - val acc: 0.8549
Epoch 13/20
0.3009 - acc: 0.8714 - val_loss: 0.3905 - val_acc: 0.8362
Epoch 14/20
0.3004 - acc: 0.8714 - val_loss: 0.3578 - val_acc: 0.8620
Epoch 15/20
0.2980 - acc: 0.8724 - val loss: 0.6110 - val acc: 0.8394
Epoch 16/20
0.2955 - acc: 0.8738 - val_loss: 0.4092 - val_acc: 0.8272
Epoch 17/20
0.2951 - acc: 0.8726 - val loss: 0.4071 - val acc: 0.8286
Epoch 18/20
0.2931 - acc: 0.8745 - val loss: 0.3563 - val acc: 0.8500
Epoch 19/20
0.2917 - acc: 0.8753 - val loss: 0.2986 - val acc: 0.8761
Epoch 20/20
0.2911 - acc: 0.8756 - val_loss: 0.3873 - val_acc: 0.8469
```

```
score = cnn.evaluate(X_test_sub, y_test_sub, verbose=0)
print("Test loss: {:.3f}".format(score[0]))
print("Test Accuracy: {:.3f}".format(score[1]))
```

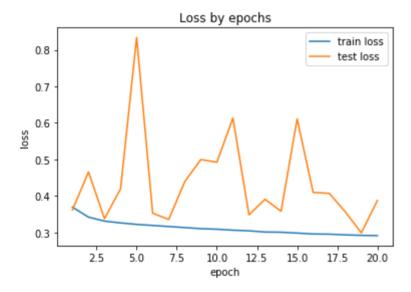
Test loss: 0.387
Test Accuracy: 0.847

In [0]:

```
x=np.arange(1,21)
plt.plot(x,history_cnn_datagen2.history['loss'],label='train loss')
plt.plot(x,history_cnn_datagen2.history['val_loss'], label = 'test loss')
plt.legend()
plt.xlabel('epoch')
plt.ylabel('loss')
plt.title('Loss by epochs')
```

Out[0]:

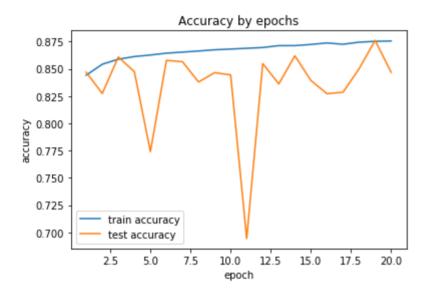
Text(0.5, 1.0, 'Loss by epochs')



```
x=np.arange(1,21)
plt.plot(x,history_cnn_datagen2.history['acc'],label='train accuracy')
plt.plot(x,history_cnn_datagen2.history['val_acc'], label = 'test accuracy')
plt.legend()
plt.xlabel('epoch')
plt.ylabel('accuracy')
plt.title('Accuracy by epochs')
```

Out[0]:

Text(0.5, 1.0, 'Accuracy by epochs')



Comments:

Yes, data augmentation improves accuracy a bit, from 0.82 to 0.84.

3.3 Build a deeper model using residual connections. Show that you can build a deep model that would not be able to learn if you remove the residual connections (i.e. compare a deep model with and without residual connections while the rest of the architecture is constant).

With residual connection

```
from keras.layers import Input, Conv2D, MaxPooling2D, Flatten, add
from keras.models import Model
num classes = 2
inputs = Input(shape=(50, 50, 3))
#next
conv1 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(inputs)
conv1 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv1_1)
conv1 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv1 2)
maxpool1 = MaxPooling2D(pool_size=(2, 2))(conv1_3)
conv2 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool1)
conv2 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv2 1)
skip1 = add([maxpool1, conv2 2])
conv2_3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(skip1)
maxpool2 = MaxPooling2D(pool size=(2, 2))(conv2 3)
conv3 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool2)
conv3 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv3 1)
conv3_3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv3 2)
skip2 = add([conv3 1, conv3 3])
maxpool3 = MaxPooling2D(pool size=(2, 2))(skip2)
conv4 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool3)
conv4 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv4_1)
conv4 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv4 2)
skip3 = add([conv4 1, conv4 3])
maxpool4 = MaxPooling2D(pool size=(2, 2))(skip3)
conv5_1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool4)
conv5 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv5 1)
conv5 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv5_2)
skip4 = add([conv5 1, conv5 3])
maxpool5 = MaxPooling2D(pool size=(2, 2))(skip4)
flat = Flatten()(maxpool5)
dense = Dense(64, activation='relu')(flat)
predictions = Dense(num_classes, activation='softmax')(dense)
model resnet = Model(inputs=inputs, outputs=predictions)
```

```
In [0]:
```

```
Train on 45791 samples, validate on 15264 samples
Epoch 1/20
45791/45791 [============= ] - 20s 445us/step - los
s: 0.5364 - acc: 0.7674 - val_loss: 0.3943 - val acc: 0.8253
Epoch 2/20
s: 0.3963 - acc: 0.8278 - val loss: 0.3739 - val acc: 0.8364
Epoch 3/20
s: 0.3751 - acc: 0.8367 - val loss: 0.3689 - val acc: 0.8434
Epoch 4/20
s: 0.3609 - acc: 0.8449 - val loss: 0.3617 - val acc: 0.8446
Epoch 5/20
45791/45791 [============= ] - 14s 295us/step - los
s: 0.3504 - acc: 0.8494 - val_loss: 0.3481 - val_acc: 0.8501
Epoch 6/20
s: 0.3464 - acc: 0.8502 - val_loss: 0.3358 - val_acc: 0.8568
Epoch 7/20
s: 0.3354 - acc: 0.8549 - val_loss: 0.3387 - val acc: 0.8518
Epoch 8/20
s: 0.3307 - acc: 0.8581 - val loss: 0.3684 - val acc: 0.8391
Epoch 9/20
s: 0.3244 - acc: 0.8609 - val loss: 0.3393 - val acc: 0.8558
Epoch 10/20
s: 0.3249 - acc: 0.8625 - val loss: 0.3304 - val acc: 0.8540
Epoch 11/20
45791/45791 [============= ] - 13s 291us/step - los
s: 0.3121 - acc: 0.8663 - val loss: 0.3261 - val acc: 0.8575
Epoch 12/20
45791/45791 [============= ] - 13s 291us/step - los
s: 0.3088 - acc: 0.8679 - val_loss: 0.3318 - val acc: 0.8565
Epoch 13/20
s: 0.3003 - acc: 0.8731 - val loss: 0.3253 - val acc: 0.8593
Epoch 14/20
s: 0.2990 - acc: 0.8727 - val loss: 0.3358 - val acc: 0.8549
Epoch 15/20
45791/45791 [============= ] - 13s 293us/step - los
s: 0.2915 - acc: 0.8772 - val loss: 0.3439 - val acc: 0.8494
Epoch 16/20
s: 0.2892 - acc: 0.8783 - val_loss: 0.3465 - val_acc: 0.8531
Epoch 17/20
s: 0.2800 - acc: 0.8819 - val_loss: 0.3437 - val_acc: 0.8531
Epoch 18/20
s: 0.2749 - acc: 0.8855 - val loss: 0.3570 - val acc: 0.8481
Epoch 19/20
45791/45791 [============== ] - 13s 290us/step - los
s: 0.2697 - acc: 0.8875 - val loss: 0.3527 - val acc: 0.8494
Epoch 20/20
s: 0.2631 - acc: 0.8903 - val loss: 0.3736 - val acc: 0.8477
```

```
In [0]:
```

```
score = model_resnet.evaluate(X_test_sub, y_test_sub, verbose=0)
print("Test loss: {:.3f}".format(score[0]))
print("Test Accuracy: {:.3f}".format(score[1]))
```

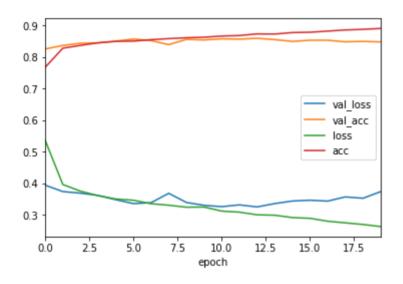
Test loss: 0.374
Test Accuracy: 0.848

In [0]:

```
pd.DataFrame(history_callback_resnet.history).plot()
plt.xlabel('epoch')
```

Out[0]:

Text(0.5, 0, 'epoch')



Without Residual Connection

```
from keras.layers import Input, Conv2D, MaxPooling2D, Flatten, add
from keras.models import Model
num classes = 2
inputs = Input(shape=(50, 50, 3))
#next
conv1 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(inputs)
conv1 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv1_1)
conv1 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv1 2)
maxpool1 = MaxPooling2D(pool_size=(2, 2))(conv1_3)
conv2 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool1)
conv2 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv2 1)
conv2 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv2 2)
maxpool2 = MaxPooling2D(pool size=(2, 2))(conv2 3)
conv3 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool2)
conv3_2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv3 1)
conv3 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv3 2)
maxpool3 = MaxPooling2D(pool size=(2, 2))(conv3 3)
conv4 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool3)
conv4 2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv4 1)
conv4 3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv4_2)
maxpool4 = MaxPooling2D(pool size=(2, 2))(conv4 3)
conv5 1 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(maxpool4)
conv5_2 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv5_1)
conv5_3 = Conv2D(32, (3, 3), activation='relu',
                 padding='same')(conv5 2)
maxpool5 = MaxPooling2D(pool_size=(2, 2))(conv5_3)
flat = Flatten()(maxpool5)
dense = Dense(64, activation='relu')(flat)
predictions = Dense(num classes, activation='softmax')(dense)
model noresnet = Model(inputs=inputs, outputs=predictions)
```

```
Train on 45791 samples, validate on 15264 samples
Epoch 1/20
45791/45791 [============== ] - 21s 469us/step - los
s: 0.5039 - acc: 0.7578 - val loss: 0.3937 - val acc: 0.8252
Epoch 2/20
s: 0.3948 - acc: 0.8270 - val loss: 0.3657 - val acc: 0.8391
Epoch 3/20
s: 0.3705 - acc: 0.8387 - val loss: 0.3603 - val acc: 0.8411
Epoch 4/20
45791/45791 [============== ] - 14s 307us/step - los
s: 0.3612 - acc: 0.8437 - val loss: 0.3851 - val acc: 0.8362
Epoch 5/20
45791/45791 [============= ] - 14s 310us/step - los
s: 0.3592 - acc: 0.8430 - val_loss: 0.3338 - val_acc: 0.8551
Epoch 6/20
s: 0.3496 - acc: 0.8491 - val loss: 0.3410 - val acc: 0.8529
Epoch 7/20
s: 0.3449 - acc: 0.8522 - val_loss: 0.3331 - val acc: 0.8523
Epoch 8/20
s: 0.3399 - acc: 0.8544 - val loss: 0.3658 - val acc: 0.8348
Epoch 9/20
s: 0.3327 - acc: 0.8568 - val loss: 0.3436 - val acc: 0.8517
Epoch 10/20
s: 0.3302 - acc: 0.8587 - val loss: 0.3309 - val acc: 0.8579
Epoch 11/20
45791/45791 [============= ] - 14s 308us/step - los
s: 0.3231 - acc: 0.8619 - val loss: 0.3395 - val acc: 0.8555
Epoch 12/20
45791/45791 [============= ] - 14s 308us/step - los
s: 0.3171 - acc: 0.8646 - val loss: 0.3304 - val acc: 0.8539
Epoch 13/20
s: 0.3093 - acc: 0.8674 - val loss: 0.3261 - val acc: 0.8595
Epoch 14/20
s: 0.3118 - acc: 0.8667 - val loss: 0.3098 - val acc: 0.8652
Epoch 15/20
45791/45791 [============= ] - 14s 309us/step - los
s: 0.3032 - acc: 0.8704 - val loss: 0.3105 - val acc: 0.8665
Epoch 16/20
45791/45791 [============== ] - 14s 307us/step - los
s: 0.2977 - acc: 0.8750 - val_loss: 0.3295 - val_acc: 0.8564
Epoch 17/20
s: 0.2955 - acc: 0.8750 - val_loss: 0.3245 - val_acc: 0.8572
Epoch 18/20
s: 0.2914 - acc: 0.8772 - val loss: 0.3152 - val acc: 0.8674
Epoch 19/20
45791/45791 [============== ] - 14s 309us/step - los
s: 0.2863 - acc: 0.8784 - val loss: 0.3061 - val acc: 0.8706
Epoch 20/20
s: 0.2844 - acc: 0.8807 - val loss: 0.3133 - val acc: 0.8645
```

```
score = model_noresnet.evaluate(X_test_sub, y_test_sub, verbose=0)
print("Test loss: {:.3f}".format(score[0]))
print("Test Accuracy: {:.3f}".format(score[1]))
```

Test loss: 0.313
Test Accuracy: 0.864

In [0]:

```
pd.DataFrame(history_callback_noresnet.history).plot()
plt.xlabel('epoch')
```

Out[0]:

Text(0.5, 0, 'epoch')

