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Steps:

- 1) Read the data form file.
- 2) Split the train and validation data.
- 3) Preprocessing and Augmenting data.
- 4) Built two models.
- 5) Train the models.
- 6) Test on Validation data.
- 7) Test on test data.
- 8) Analyze the results.

I have tried two different Architecture first one is given in assignment and other is modified version where I increased the number of kernels, decreased kernel size and increased number of nodes in fully connected layers, increased the batch size.

Overall, I trained the models for 35 epochs each, but I have taken only first 10 epochs into consideration.

Results:

Model-1:

Training:

Loss=2.4429 and Validation Accuracy=34.9%

Testing:

Accuracy for 100 labels= 35.13%

Accuracy for 20 labels= 48.41%

Considering Top-3 predictions:

Accuracy for 100 classes= 35.13%

Accuracy for 20 classes= 48.41%

Considering Top-5 predictions:

Accuracy for 100 classes= 65.22%

Accuracy for 20 classes= 77.59%

Model-2:

Training:

Loss=1.2183 and Validation Accuracy=38.41%

Testing:

Accuracy for 100 labels= 39.08%

Accuracy for 20 labels= 51.69%

Considering Top-3 predictions:

Accuracy for 100 classes= 39.08%

Accuracy for 20 classes= 51.69%

Considering Top-5 predictions:

Accuracy for 100 classes= 67.66%

Accuracy for 20 classes= 80.71%

From the above results we can observe that model-2 has better results compared to model-1, we can add some more layers make the model complex and get the accuracy increased some more.

```
In [ ]: import pickle
import numpy as np
import tensorflow as tf
import matplotlib
matplotlib.use('agg')
import pylab as plt
```

```
In [ ]: count=1
epc=1
loss_graph_list=[]

#Reading the data
def unpickle(file):

    with open(file, 'rb') as fo:
        dict = pickle.load(fo, encoding='bytes')
    return dict

train=unpickle("cifar-100-python/train")
test=unpickle("cifar-100-python/test")

print(train.keys())
print(test.keys())

#Train data
Xtrain=train[b'data'].reshape((len(train[b'data']), 3, 32, 32)).transpose(0, 2, 3, 1)
ytrain=np.array(train[b'fine_labels'])

#Test data
Xtest=test[b'data'].reshape((len(test[b'data']), 3, 32, 32)).transpose(0, 2, 3, 1)
ytest=np.array(test[b'fine_labels'])

Xtrain=Xtrain.astype(np.float32)
Xtest=Xtest.astype(np.float32)

Xval=Xtrain[40000:,:,:,:]
Xtrain=Xtrain[0:40000]

yval=ytrain[40000:]
ytrain=ytrain[0:40000]

# Reset graph parameters
tf.reset_default_graph()

x = tf.placeholder(tf.float32, shape=(None, 32, 32, 3), name='input_x')
y = tf.placeholder(tf.int32, shape=(None,), name='output_y')
```

```

In [ ]: # model Architecture
def leNet(features):
    # layer 1 input
    # input_layer=tf.reshape(features['x'],[-1,32,32,3])

    # conv_layer #1
    conv1 = tf.layers.conv2d(inputs=features, filters=6,
                              kernel_size=[5, 5], strides=1,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initializer())

    # pooling_layer #1
    pool1 = tf.layers.max_pooling2d(inputs=conv1,
                                      pool_size=[2, 2],
                                      strides=2)

    batch1=tf.layers.batch_normalization(inputs=pool1)

    # conv_layer #2
    conv2 = tf.layers.conv2d(inputs=batch1, filters=16,
                              kernel_size=[5, 5], strides=1,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initializer())

    # pooling_layer #2
    pool2 = tf.layers.max_pooling2d(inputs=conv2,
                                      pool_size=[2, 2],
                                      strides=2)

    batch2=tf.layers.batch_normalization(inputs=pool2)

    # Dense_layer #1
    pool2_flat = tf.reshape(batch2, [-1, 400])
    dense1 = tf.layers.dense(inputs=pool2_flat, units=120,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initializer())

    batch3=tf.layers.batch_normalization(inputs=dense1)

    # Dense_layer #2
    dense2 = tf.layers.dense(inputs=batch3, units=84,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initializer())

    batch4=tf.layers.batch_normalization(inputs=dense2)

```

```

# logits_final_layer
logits = tf.layers.dense(inputs=batch4, units=100)

return logits

```

```

In [ ]: #epochs, batchSize, Learning rate
epochs = 2
batch_size = 64
learning_rate = 0.001

# Output of model
logits = leNet(x)
model = tf.identity(logits, name='logits')

#Loss function & Optimization Algorithm
cost = tf.reduce_mean(tf.nn.sparse_softmax_cross_entropy_with_logits(logits=logits, labels=y))
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate).minimize(cost)

#Prediction and Accuracy
correct_pred = tf.equal(tf.argmax(logits, 1),tf.cast(y,tf.int64))
accuracy = tf.reduce_mean(tf.cast(correct_pred, tf.float32), name='accuracy')

```

```

In [ ]: # Shuffling Data
def batch_features_labels(features, labels, batch_size):

    rand_index=np.random.choice(len(features),size=len(features))
    for start in range(0,len(features),batch_size):
        end=min(start+batch_size,len(features))
        tmp=np.array(rand_index[start:end])
        yield features[tmp],labels[tmp]

```

```

In [ ]: #saving model
        model_path = './image_classification'

#Creating Session
print('Training...')
with tf.Session() as sess:
    #inistalizing Global_variables
    sess.run(tf.global_variables_initializer())
    #####

    # Data Augmentation
    data_tf=tf.convert_to_tensor(Xtrain,np.float32)
    big=tf.image.resize_images(Xtrain,(36,36))
    top_r=tf.image.crop_to_bounding_box(big,0,0,32,32)
    top_l=tf.image.crop_to_bounding_box(big,0,4,32,32)
    bot_r=tf.image.crop_to_bounding_box(big,4,0,32,32)
    bot_l=tf.image.crop_to_bounding_box(big, 4, 4, 32, 32)
    cen=tf.image.crop_to_bounding_box(big,2,2,32,32)
    flip=tf.image.flip_left_right(data_tf)

    fbig = tf.image.resize_images(flip, (36, 36))
    ftop_r = tf.image.crop_to_bounding_box(fbig, 0, 0, 32, 32)
    ftop_l = tf.image.crop_to_bounding_box(fbig, 0, 4, 32, 32)
    fbot_r = tf.image.crop_to_bounding_box(fbig, 4, 0, 32, 32)
    fbot_l = tf.image.crop_to_bounding_box(fbig, 4, 4, 32, 32)
    fcen = tf.image.crop_to_bounding_box(fbig, 2, 2, 32, 32)

    Xtrain1 = tf.concat([data_tf, top_r, top_l, bot_l, bot_r, cen, flip], axis
=0)
    Xtrain2=tf.concat([ftop_l,ftop_r,fbot_l,fbot_r,fcen],axis=0)

    # Xtrain=tf.concat([Xtrain1,Xtrain2],axis=0)

    sess.run(Xtrain1)
    sess.run(Xtrain2)
    Xtrain1=Xtrain1.eval()
    Xtrain2=Xtrain2.eval()
    Xtrain=np.concatenate((Xtrain1,Xtrain2))
    ytrain=np.concatenate((ytrain,ytrain,ytrain,ytrain,ytrain,ytrain,ytrain,yt
rain,ytrain,ytrain,ytrain,ytrain))
    # Xtrain=Xtrain.eval()
    print(type(Xtrain))
    print(Xtrain.shape)
    print(ytrain.shape)

    #Calculating mean of the dataSet
    sub = np.mean(Xtrain, axis=0)
    Xtrain = Xtrain - sub
    Xval=Xval-sub
    np.save("mean_vec",sub)

    # Traning and Validation
    for epoch in range(0,epochs):
        for batch_features, batch_labels in batch_features_labels(Xtrain, ytra

```

```

in, batch_size):
    sess.run(optimizer,
               feed_dict={
                   x: batch_features,
                   y: batch_labels
               })

    loss = sess.run(cost,
                    feed_dict={
                        x: batch_features,
                        y: batch_labels
                    })
    loss_graph_list.append(loss)

    if(epoch==epc):
        epc+=1
        print('Epoch {:>2}\n'.format(epoch), end='')
        print(sum(loss_graph_list)/len(loss_graph_list))

    if(epoch==count):
        count+=1
        valid_acc = sess.run(accuracy,
                             feed_dict={
                                 x: Xval,
                                 y: yval})

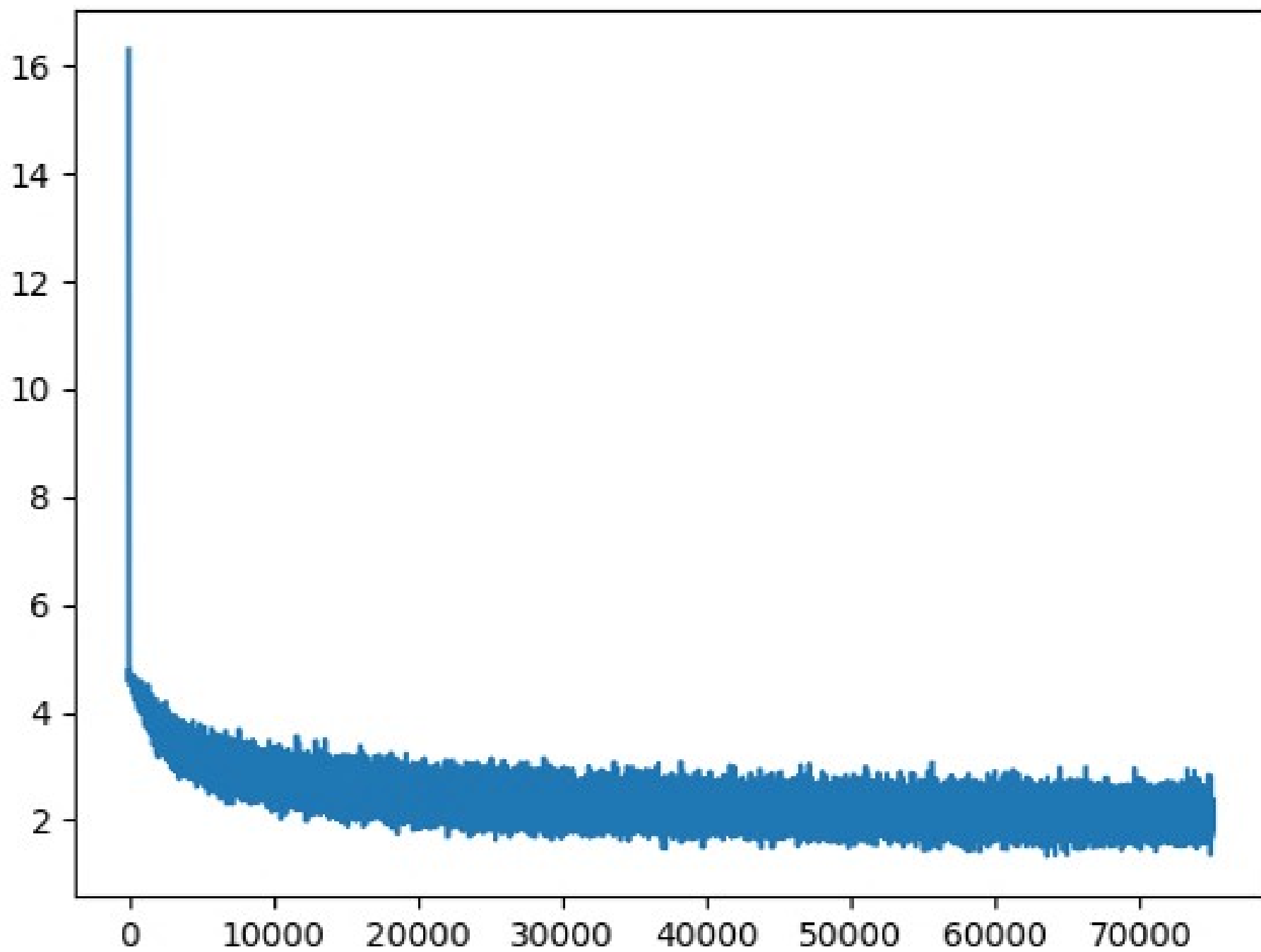
        print('Loss: {:>10.4f} Validation Accuracy: {:.6f}'.format(loss, valid_acc))

# Saving model and plotting Loss Graph
plt.plot(loss_graph_list)
plt.show()
plt.savefig("myfig")
model_saver = tf.train.Saver()
save_path_model = model_saver.save(sess, model_path)

```

```
Training...
2018-11-13 10:05:43.519192: I tensorflow/core/common_runtime
name: Tesla K80 major: 3 minor: 7 memoryClockRate(GHz): 0
pciBusID: 0adf:00:00.0
totalMemory: 11.92GiB freeMemory: 11.85GiB
2018-11-13 10:05:43.519242: I tensorflow/core/common_runtime
2018-11-13 10:05:43.795209: I tensorflow/core/common_runtime
2018-11-13 10:05:43.795276: I tensorflow/core/common_runtime
2018-11-13 10:05:43.795297: I tensorflow/core/common_runtime
2018-11-13 10:05:43.795558: I tensorflow/core/common_runtime
MB memory) -> physical GPU (device: 0, name: Tesla K80, p
<class 'numpy.ndarray'>
(480000, 32, 32, 3)
(480000,)
Epoch 1
Loss= 3.485201018696673
Loss: 3.4852 Validation Accuracy: 0.242100
Epoch 2
Loss= 3.1050059004995267
Loss: 3.1050 Validation Accuracy: 0.284400
Epoch 3
Loss= 2.8987279997354696
Loss: 2.8987 Validation Accuracy: 0.299100
Epoch 4
Loss= 2.766079532333638
Loss: 2.7661 Validation Accuracy: 0.315500
Epoch 5
Loss= 2.668010677939361
Loss: 2.6680 Validation Accuracy: 0.330500
Epoch 6
Loss= 2.5926890315719398
Loss: 2.5927 Validation Accuracy: 0.337100
Epoch 7
Loss= 2.5337450712154608
Loss: 2.5337 Validation Accuracy: 0.339600
Epoch 8
Loss= 2.4846214733668797
Loss: 2.4846 Validation Accuracy: 0.343300
Epoch 9
Loss= 2.442906861813785
Loss: 2.4429 Validation Accuracy: 0.349000
[41] Killed python mod.py
```





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```
In [1]: import tensorflow as tf
import pickle
import random
import pickle
import numpy as np
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix, accuracy_score
np.set_printoptions(threshold=np.nan)
```

```
/anaconda/envs/py35/lib/python3.5/site-packages/h5py/__init__.py:36: FutureWarning: Conversion of the second argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will be treated as `np.float64 == np.dtype(float).type`.
```

```
from ._conv import register_converters as _register_converters
```

```

In [2]: model_path = './image_classification'
        batch_size = 64
        n_samples = 10
        top_n_predictions = 5

        def unpickle(file):

            with open(file, 'rb') as fo:
                dict = pickle.load(fo, encoding='bytes')
            return dict

        train=unpickle("cifar-100-python/train")
        test=unpickle("cifar-100-python/test")
        meta=unpickle("cifar-100-python/meta")

        Xtest=test[b'data'].reshape((len(test[b'data']), 3, 32, 32)).transpose(0, 2, 3
        , 1)
        ytest=np.array(test[b'fine_labels'])

        Xtest=Xtest.astype(np.float32)

        sub=np.load("mean_vec.npy")
        Xtest=Xtest-sub

        ytestC=np.array(test[b'coarse_labels'])

        lab={}
        for i in range(0,20):
            lab[i]=set()
        for i in range(0,len(ytest)):
            lab[ytestC[i]].add(ytest[i])

        name_100_labels=[str(i)[2:-1] for i in meta[b'fine_label_names']]
        name_20_labels=[str(i)[2:-1] for i in meta[b'coarse_label_names']]

```

```

In [3]: def batch_features_labels(features, labels, batch_size):
        for start in range(0, len(features), batch_size):
            end = min(start + batch_size, len(features))
            yield features[start:end], labels[start:end]

```

```

In [4]: loaded_graph = tf.Graph()
        with tf.Session(graph=loaded_graph) as sess:

            loader = tf.train.import_meta_graph(model_path + '.meta')
            loader.restore(sess, model_path)
            loaded_x = loaded_graph.get_tensor_by_name('input_x:0')
            loaded_y = loaded_graph.get_tensor_by_name('output_y:0')
            loaded_logits = loaded_graph.get_tensor_by_name('logits:0')
            loaded_acc = loaded_graph.get_tensor_by_name('accuracy:0')

            label_predictor = sess.run(
                tf.nn.top_k(tf.nn.softmax(loaded_logits),1),
                feed_dict={loaded_x: Xtest, loaded_y: ytest})
            # print(label_predictor[1])

            matFine=confusion_matrix(ytest,label_predictor[1])
            accFine=accuracy_score(ytest,label_predictor[1])

            print("-----100 classes-----")

            print("\nAccuracy and confusion matrix for 100 labels")
            print("-----Accuracy=",100*accFine,"-----")
            print("\nConfusion-Matrix:\n",matFine)

            # print("\nPredicted_Labels:\n",label_predictor[1].reshape(len(label_predictor[1])).tolist())
            # print("\n Correct_Labels:\n",ytest)
            print("\n\n\n\n\n")

            coarse_pred=[]
            pred=np.reshape(label_predictor[1],(len(label_predictor[1]))).tolist()
            for i in range(0,len(pred)):
                for pos in lab.keys():
                    if(pred[i] in lab[pos]):
                        coarse_pred.append(pos)
            matCoa = confusion_matrix(ytestC,coarse_pred)
            accCoa = accuracy_score(ytestC, coarse_pred)
            print("-----20 classes-----")
            print("Accuracy and confusion matrix for 20 labels")
            print("-----Accuracy=",100*accCoa,"-----")
            print("\nConfusion-Matrix:\n",matCoa)

            # random_test_features, random_test_labels = tuple(zip(*random.sample(list
            (zip(Xtest, ytest)), n_samples)))

```

```

Top_3_pred= sess.run(
    tf.nn.top_k(tf.nn.softmax(loaded_logits), 1),
    feed_dict={loaded_x: Xtest, loaded_y: ytest})
print("-----Considering Top-3-----\n")
# print("Top 5 Classes:\n"Top_5_pred[1])
acc_top_3=0
for i in range(0,len(Top_3_pred[1])):
    if(ytest[i] in Top_3_pred[1][i]):
        acc_top_3+=1
print("-----Accuracy for 100 classes= ",100*acc_top_3/len(ytest))

cacc_top_3=0
for i in range(0,len(ytest)):
    temp=[]
    for j in lab.keys():
        if(ytest[i] in lab[j]):
            temp.append(lab[j])
            break
# print(temp)
# print(Top_5_pred[1][i])
if(len(temp)!=0):
    temp2=set(temp[0]).intersection(Top_3_pred[1][i])
# print(temp2)
if(len(temp2)!=0):
    cacc_top_3+=1
print("-----Accuracy for 20 classes= ",100*cacc_top_3/len(ytest))

Top_5_pred= sess.run(
    tf.nn.top_k(tf.nn.softmax(loaded_logits), top_n_predictions),
    feed_dict={loaded_x: Xtest, loaded_y: ytest})
print("-----Considering Top-5-----\n")
# print("Top 5 Classes:\n"Top_5_pred[1])
acc_top_5=0
for i in range(0,len(Top_5_pred[1])):
    if(ytest[i] in Top_5_pred[1][i]):
        acc_top_5+=1
print("-----Accuracy for 100 classes= ",100*acc_top_5/len(ytest))

cacc_top_5=0
for i in range(0,len(ytest)):
    temp=[]
    for j in lab.keys():
        if(ytest[i] in lab[j]):
            temp.append(list(lab[j]))
            break
    if(len(temp)!=0):
        temp2=set(temp[0]).intersection(Top_5_pred[1][i])

```

```
        if(len(temp2)!=0):
            cacc_top_5+=1
    print("-----Accuracy for 20 classes= ",100*cacc_top_5/len(ytest))
```

INFO:tensorflow:Restoring parameters from ./image\_classification  
-----100 classes-----

Accuracy and confusion matrix for 100 labels

-----Accuracy= 35.13 -----

Confusion-Matrix:

```
[[62 3 1 0 0 0 0 0 0 0 0 0 0 0 3 0 0 0 0 0 1 0 0 0
 1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 0
 0 0 0 0 0 3 0 0 0 0 6 0 0 0 0 1 0 0 0 0 0 0 1 0
 0 0 0 0 0 0 0 0 0 0 1 6 0 0 0 0 0 0 0 0 8 0 0 0
 0 0 1 0]
[ 0 51 0 0 0 0 1 0 0 0 1 1 0 1 3 0 0 0 1 0 0 0 1 0
 0 0 2 1 0 0 0 0 0 1 1 0 0 1 1 1 0 1 1 0 1 1 0 0
 4 0 0 1 0 0 1 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 3 0
 0 1 1 1 0 1 1 0 0 0 2 2 0 0 0 1 0 0 1 2 5 0 0 0
 0 0 0 0]
[ 0 0 23 0 0 1 1 0 0 0 0 2 0 0 2 0 0 0 0 1 1 0 0 0
 0 0 4 0 0 0 0 0 1 0 0 10 2 0 0 0 0 1 0 0 0 1 3 0
 2 0 0 1 0 1 2 0 1 1 0 0 0 2 0 0 0 2 1 0 0 0 0 0
 0 0 0 0 0 2 0 0 0 0 0 3 2 0 2 2 0 0 0 1 2 0 0 0
 0 1 18 1]
[ 0 0 0 15 6 0 4 1 1 0 0 0 0 1 1 0 1 0 0 6 0 7 0 0
 0 0 1 0 1 3 2 3 0 0 0 0 0 0 0 0 0 0 4 0 0 2 0 0
 3 0 0 3 1 0 0 1 1 0 1 1 0 2 0 1 0 3 3 0 0 0 0 0
 4 0 2 4 0 1 2 1 0 0 1 0 0 0 1 0 1 0 0 0 0 0 0 1
 0 2 1 0]
[ 0 0 0 1 13 0 1 1 0 0 1 0 0 1 0 1 1 0 1 2 0 3 0 0
 1 0 2 1 0 1 0 0 1 4 0 1 0 0 3 0 0 0 3 4 1 1 0 1
 0 0 3 4 0 0 0 0 3 0 1 0 0 0 0 3 3 0 7 0 0 0 0 0
 2 0 7 1 0 3 1 0 0 2 0 0 0 0 0 1 1 3 1 1 0 2 0 1
 0 0 0 0]
[ 0 0 0 0 0 29 0 0 2 0 1 0 3 4 1 0 1 0 0 1 8 0 1 2
 1 3 1 0 1 0 0 0 1 0 0 0 6 2 0 0 1 1 0 0 0 3 0 0
 0 0 0 0 0 0 1 0 0 1 1 0 0 1 0 0 0 1 0 0 0 0 0 0
 0 0 0 0 0 1 0 1 0 4 0 0 3 0 0 6 0 1 1 2 0 0 2 0
 0 0 1 0]
[ 0 1 1 1 0 0 46 4 0 0 0 0 0 0 6 0 0 0 3 1 1 0 0 0
 2 0 0 0 0 2 0 0 0 0 2 0 0 0 0 0 0 1 0 2 0 3 1 0
 2 0 1 0 0 0 2 0 0 1 0 0 0 0 1 0 0 0 0 0 0 0 0 0
 1 0 0 1 0 0 0 1 0 0 7 0 1 0 1 0 1 0 0 0 1 0 0 0
 0 1 1 0]
[ 0 0 0 0 0 0 3 43 1 0 1 0 0 0 6 0 1 0 0 1 2 2 0 0
 8 0 1 0 1 1 0 0 0 0 0 0 1 0 0 1 0 0 0 0 1 1 1
 6 0 0 0 0 0 0 1 1 0 0 0 0 0 2 0 1 0 0 0 1 0 0 0
 0 0 0 0 0 0 4 2 0 0 0 0 0 0 2 0 1 0 0 0 0 0 2
 0 0 0 1]
[ 0 1 1 1 0 0 2 2 30 0 0 0 0 0 0 0 0 0 1 2 0 0 0 0
 1 0 0 0 0 0 1 0 0 1 1 1 0 0 0 0 0 0 0 1 0 0 2 0
 24 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 0 0 0 0
 1 0 0 0 0 0 1 1 0 1 0 0 3 0 2 1 2 5 1 0 0 1 0 2
 0 1 1 2]
[ 0 0 0 1 0 1 0 2 0 48 0 0 1 0 0 0 2 1 1 1 0 1 0 0
 0 0 0 0 2 1 0 0 0 0 0 1 0 0 0 0 4 2 0 0 0 0 2 0
 1 0 0 1 0 1 0 1 0 0 0 0 0 2 0 0 1 1 1 0 0 1 0 0
 0 0 0 0 2 0 1 1 0 1 0 3 2 0 2 4 0 1 0 0 1 0 0 0
 0 0 1 0]
```

```

[ 3 1 1 0 0 0 0 1 0 0 20 1 0 1 2 0 3 0 0 1 3 0 5 0
  0 0 3 0 6 0 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0
  0 0 0 1 0 5 0 0 1 2 0 0 0 0 11 2 0 0 0 1 3 0 0 1 0
  0 0 0 1 0 0 0 0 0 1 0 2 2 0 1 2 1 0 0 2 2 0 0 1
  0 1 1 1]
[ 0 2 8 0 1 0 1 0 1 0 0 18 0 2 3 0 0 0 0 0 2 5 1 1
  0 1 0 0 1 1 0 0 0 0 0 15 2 1 2 0 1 0 0 1 0 1 2 0
  0 0 0 1 0 1 0 0 0 0 4 0 0 0 0 0 0 0 1 0 0 0 0 0
  0 0 0 0 0 1 0 0 0 0 0 1 5 1 0 1 0 1 1 0 0 0 0 0
  0 0 9 0]
[ 0 0 0 0 0 1 0 0 2 0 0 0 43 3 0 0 0 3 2 2 2 0 1 0
  0 1 1 0 0 0 0 0 0 0 0 0 0 4 0 0 0 0 0 0 1 0 0 1
  4 0 0 0 0 0 0 0 1 0 4 1 0 0 0 0 0 0 0 0 1 1 0 0
  0 0 0 0 0 0 0 0 1 2 0 0 3 0 1 2 1 2 5 0 0 0 0 3
  0 0 0 1]
[ 0 0 0 0 0 1 0 0 0 0 1 1 2 31 1 0 2 0 0 0 0 1 0 0
  0 1 0 0 0 0 0 0 0 1 0 0 0 4 0 0 0 2 0 0 0 0 0 0
  1 0 0 0 0 0 0 0 1 0 14 0 1 0 0 0 0 0 1 0 0 0 0 0
  0 0 0 0 0 0 1 0 0 14 0 0 4 0 1 5 1 2 3 0 2 0 0 1
  0 0 0 0]
[ 0 2 0 1 0 0 4 3 1 1 1 0 0 0 40 0 0 0 4 2 0 1 0 0
  1 0 5 0 0 0 0 0 0 2 0 0 0 0 1 0 1 0 1 0 0 2 0 0
  4 0 0 1 0 0 2 0 1 0 0 0 0 0 4 0 0 0 0 0 0 0 0 0
  0 0 1 2 0 1 0 1 0 0 0 1 1 0 3 0 0 1 0 1 2 0 0 0
  0 0 0 1]
[ 0 1 0 2 1 0 2 1 1 2 1 1 2 0 2 26 1 3 0 5 1 0 1 0
  2 0 0 0 0 8 0 0 0 1 0 0 2 0 2 0 0 0 1 5 1 0 0 0
  1 0 1 1 0 0 0 0 1 0 0 0 0 0 0 0 2 0 1 0 0 0 0 0
  1 0 0 4 0 0 0 2 0 0 0 0 2 0 1 1 1 1 0 0 0 2 0 0
  0 0 3 0]
[ 0 0 1 0 0 2 0 0 1 5 0 0 0 3 0 0 31 0 1 0 1 0 5 0
  0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 2 1 0 0 0 0 1 0 0
  4 0 1 0 0 0 0 0 0 0 2 0 0 4 1 0 0 0 2 0 0 0 0 0
  1 0 0 0 1 0 0 0 0 3 0 1 4 0 7 5 0 0 0 0 1 0 2 3
  0 1 1 0]
[ 0 0 0 0 0 0 0 0 0 1 0 0 2 2 0 2 0 46 0 4 0 0 0 0
  0 0 0 0 0 0 0 0 0 1 0 0 0 0 22 0 0 0 0 1 1 0 0 0
  0 2 0 0 1 0 0 0 1 0 1 1 0 0 0 0 0 0 2 0 1 1 0 0
  0 0 0 0 1 0 0 0 0 1 0 0 0 1 0 0 0 0 2 1 0 0 0 0
  1 1 0 0]
[ 0 1 0 0 0 1 4 1 0 1 2 0 0 1 8 2 3 0 32 1 0 1 0 0
  0 1 0 0 0 0 0 0 0 1 0 0 0 1 1 0 0 2 1 0 3 1 0 0
  0 0 0 0 0 0 1 0 0 1 1 0 0 0 1 0 2 0 0 0 0 2 0 0
  0 0 2 0 0 3 1 1 0 0 1 2 2 0 1 0 0 0 0 4 1 0 0 1
  3 0 0 1]
[ 0 1 0 1 1 0 0 2 1 0 0 0 0 0 1 11 0 1 0 33 0 0 1 0
  0 1 1 0 0 4 0 4 0 1 3 0 0 1 1 0 0 1 1 4 0 2 1 0
  5 0 0 0 1 0 0 0 0 0 2 0 0 0 0 0 0 1 4 0 0 0 0 1
  0 0 1 0 0 0 2 1 0 0 0 0 1 0 0 0 0 1 2 0 0 0 0 0
  0 0 0 0]
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	0	0	1	0	0	0	0	0	1	3	0	0	0	3	0	1	0	0	1	0	0	0	0	0
	1	0	0	0	0	0	0	0	1	0	1	0	0	0	0	0	0	0	1	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	43	0	0	2	0	0	4	0	3	7	0	0	0	0	0
	0	0	0	0]																				
[	0	1	0	0	0	0	3	0	0	0	0	0	0	1	2	0	0	0	0	0	0	0	0	0
	1	0	0	0	0	2	0	0	0	1	1	0	0	0	0	0	0	0	0	1	0	1	0	0
	2	1	0	0	0	3	1	0	0	1	0	0	0	0	6	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0	67	3	0	0	0	0	0	1	0	0	1	0	0	0
	0	0	0	0]																				
[	7	2	0	0	0	0	2	2	0	1	1	0	0	0	2	0	3	0	1	0	0	0	0	0
	0	0	1	0	0	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	3	0	0
	1	0	0	0	0	13	0	0	0	1	0	0	0	0	4	0	0	0	0	0	0	0	5	0
	0	0	0	0	0	0	0	0	0	0	1	35	0	0	0	0	2	0	1	0	8	0	1	0
	0	0	0	1]																				
[	0	0	0	0	0	7	2	1	1	1	1	2	0	0	0	0	4	1	0	3	6	0	1	0
	0	1	2	0	1	0	0	0	0	0	1	1	0	1	0	0	2	0	0	0	1	0	0	0
	1	0	1	1	0	2	0	0	1	0	1	0	1	1	0	0	0	0	3	0	0	0	0	0
	0	0	0	1	0	1	0	0	0	2	0	0	28	0	4	4	2	0	3	0	0	0	1	0
	0	1	1	0]																				
[	0	0	0	0	1	0	0	0	3	0	0	0	3	5	1	0	0	1	0	1	0	0	0	0
	0	0	1	1	0	0	0	0	1	1	0	0	0	2	0	1	0	3	0	0	0	0	0	0
	5	0	0	1	0	0	0	0	0	1	7	0	1	0	0	0	0	0	3	0	0	1	0	0
	0	0	0	3	0	0	0	0	0	4	0	0	1	34	1	0	0	6	6	0	0	0	0	0
	0	0	1	0]																				
[	0	0	0	0	0	1	0	4	0	1	2	2	2	0	0	0	1	0	0	0	1	1	2	0
	3	2	0	0	2	0	0	0	0	0	0	0	1	1	0	0	1	1	0	0	1	0	0	0
	0	0	0	0	0	0	0	0	0	0	1	0	0	1	0	0	0	1	2	0	0	1	0	0
	0	0	0	2	0	0	1	1	0	1	2	0	1	0	42	7	0	0	1	2	0	0	3	0
	0	0	2	0]																				
[	0	0	0	0	0	1	0	0	0	2	1	0	2	4	0	0	1	0	0	1	2	1	2	0
	0	1	0	0	0	0	0	0	0	1	0	0	1	1	0	0	1	0	0	0	0	0	1	0
	1	0	0	0	0	0	0	0	1	0	0	0	0	0	2	0	1	0	0	0	0	0	0	0
	0	0	0	2	0	0	0	0	0	2	0	0	5	0	2	51	0	0	0	0	0	0	7	0
	0	0	3	0]																				
[	0	1	0	0	1	1	3	2	2	0	2	0	0	0	6	1	0	0	0	2	0	2	0	0
	0	0	3	0	1	3	0	0	0	3	2	0	1	0	0	0	0	0	4	3	0	0	1	0
	3	0	0	1	0	0	0	1	1	0	0	0	0	0	0	0	2	0	3	0	0	0	0	0
	0	0	1	0	0	1	5	1	0	1	1	0	3	0	1	1	25	1	1	0	1	0	0	0
	0	2	0	0]																				
[	0	0	0	0	0	0	2	1	4	0	0	0	0	3	0	1	0	0	1	0	1	0	0	0
	0	0	1	0	0	0	0	0	0	2	0	0	0	0	0	0	0	1	0	0	0	0	0	0
16	0	0	0	0	0	0	0	0	2	0	7	0	0	0	0	0	0	0	2	0	0	1	0	0
	0	0	1	0	0	1	0	1	0	6	2	0	0	1	0	0	0	38	3	0	1	0	0	0

```

1 0 0 0]
[ 0 0 0 0 0 0 0 0 0 0 0 0 2 5 1 0 0 1 0 0 0 2 0 0
0 0 0 0 0 1 0 1 1 2 0 0 0 3 0 0 0 1 0 0 0 1 0 0
7 0 0 0 0 0 0 0 1 0 4 1 0 0 0 0 1 0 1 0 0 0 0 0
0 0 0 3 0 0 1 1 0 14 0 0 0 2 0 1 0 4 37 0 1 0 0 0
0 0 0 0]
[ 0 3 0 0 1 0 1 0 1 0 1 1 1 1 0 0 1 0 4 0 2 0 0 0
0 0 2 3 0 2 2 0 0 0 2 0 0 0 0 0 1 0 2 0 1 0 0 0
0 0 0 0 0 0 0 0 2 0 0 0 0 0 0 0 0 0 0 1 0 0 1
0 2 1 6 0 0 1 0 0 0 0 0 4 0 0 1 0 1 1 44 1 0 0 0
0 0 0 2]
[ 1 2 0 0 0 0 1 1 0 0 0 0 0 0 3 0 0 0 1 2 0 0 2 0
2 0 0 0 1 0 0 0 1 0 1 0 1 0 0 2 0 0 1 0 0 1 0 0
3 0 0 1 0 4 7 0 0 0 0 0 0 0 10 0 0 0 0 0 0 0 4 0
0 0 0 0 0 0 0 0 0 0 3 4 0 0 0 0 0 1 0 0 40 0 0 0
0 0 0 0]
[ 0 2 0 1 0 0 0 2 2 0 0 0 1 1 1 0 0 1 0 0 0 0 0 0
1 0 7 2 0 2 3 0 1 0 0 0 2 0 0 0 0 1 0 0 3 1 0 0
1 0 0 1 3 0 0 1 4 1 0 1 0 0 0 1 1 2 2 1 0 0 0 0
3 4 3 1 0 0 1 1 0 2 0 0 0 0 2 0 0 0 2 0 1 19 0 6
0 1 0 1]
[ 0 0 0 0 0 1 0 0 0 2 0 0 0 0 0 0 1 0 0 0 0 0 0
0 1 0 0 0 0 0 1 0 1 0 0 0 0 0 1 0 0 0 1 0 0 1 0
0 0 0 0 0 0 0 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 4 0 4 9 0 0 0 0 0 0 68 0
1 0 2 0]
[ 0 0 0 1 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 1
2 1 0 0 1 0 12 0 1 0 0 0 0 0 3 1 0 0 0 0 0 0 0
5 1 0 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 1 0 1 0 0
2 5 0 2 0 0 0 0 0 0 0 0 1 0 2 1 0 0 0 1 0 1 0 48
0 0 0 2]
[ 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 2 1 0 0 0 0 0
0 0 0 0 1 0 0 0 0 11 0 0 0 0 0 0 1 1 0 1 0 11
0 2 0 2 17 0 0 0 4 0 0 6 1 0 0 1 1 0 0 0 0 0 0
0 0 0 0 3 0 0 0 0 0 0 0 1 0 0 0 2 0 0 1 1 0 0
28 0 0 0]
[ 0 0 0 1 0 0 0 1 1 0 1 0 0 0 1 2 0 0 0 1 0 2 1 0
0 0 1 2 0 2 1 0 0 0 0 0 0 3 2 0 0 0 2 3 1 0 0
0 0 0 1 0 0 0 0 1 0 1 0 0 0 2 2 1 11 0 0 0 0 0
0 1 5 2 0 0 4 0 1 2 0 0 2 0 1 1 1 0 0 2 0 1 0 2
0 27 1 0]
[ 0 1 5 1 0 3 5 0 0 2 1 9 0 1 0 0 1 0 0 2 1 1 1 0
1 1 6 0 0 1 0 0 0 0 0 10 0 0 0 1 0 1 0 0 0 2 3 0
2 0 0 1 0 2 0 0 0 1 0 0 0 0 0 0 1 2 0 0 0 1 0
0 0 1 2 0 0 0 0 0 1 0 0 0 0 2 1 0 0 1 0 0 0 0 0
0 0 21 1]
[ 1 0 1 0 0 1 1 1 0 2 0 0 0 0 3 0 0 0 0 0 0 0 3 2
1 0 2 0 1 0 1 0 0 1 0 0 1 0 1 7 2 1 0 1 2 2 0 0
4 0 0 0 0 0 0 0 0 1 0 1 0 8 0 0 1 1 1 0 0 0 0 0
1 1 0 2 0 0 5 0 1 0 0 0 2 0 2 1 1 0 0 3 1 0 0 3
0 1 1 20]]

```

-----20 classes-----

Accuracy and confusion matrix for 20 labels

-----Accuracy= 48.41 -----

Confusion-Matrix:

```
[[183 21 1 6 8 18 5 19 23 8 13 29 34 24 12 28 19 11
 20 18]
[ 61 200 15 7 9 19 9 21 5 7 4 6 22 24 10 33 11 3
 27 7]
[ 0 10 350 8 34 7 2 27 5 0 2 2 5 17 6 4 4 0
 13 4]
[ 7 11 13 251 21 70 22 9 5 7 1 4 8 12 20 5 5 2
 15 12]
[ 2 9 51 12 287 4 7 40 9 2 5 7 9 20 15 8 2 1
 7 3]
[ 5 8 11 42 9 231 39 13 8 14 9 8 13 16 19 11 7 1
 23 13]
[ 2 9 5 25 7 45 274 8 4 16 8 9 7 15 14 1 10 2
 26 13]
[ 9 10 25 13 7 10 9 295 9 4 3 13 8 29 7 13 5 7
 18 6]
[ 23 5 6 9 10 10 8 34 180 3 8 40 46 18 6 35 24 8
 20 7]
[ 14 3 2 12 0 13 13 3 9 283 17 16 5 5 4 5 3 17
 50 26]
[ 18 17 1 12 4 6 13 8 5 39 301 6 3 7 1 5 4 24
 12 14]
[ 8 4 1 7 3 11 8 22 46 14 6 197 40 18 16 39 19 7
 25 9]
[ 9 8 4 3 8 13 7 34 52 1 5 26 214 14 15 24 30 3
 24 6]
[ 13 16 13 19 18 22 13 71 14 4 11 11 18 151 13 39 16 8
 19 11]
[ 2 11 9 15 20 25 24 24 14 1 1 19 14 23 247 5 11 4
 24 7]
[ 23 13 5 8 8 19 11 50 19 5 5 23 28 49 13 145 19 15
 31 11]
[ 11 12 14 15 13 16 10 46 48 2 5 25 50 37 29 24 106 8
 18 11]
[ 6 0 4 0 6 4 1 7 5 17 22 6 4 6 2 14 1 375
 8 12]
[ 6 2 5 5 0 14 13 12 7 16 5 6 11 8 6 6 0 4
 319 55]
[ 8 6 6 4 4 17 12 16 4 18 10 5 10 13 4 5 2 6
 98 252]]
```

-----Considering Top-3-----  
-----

-----Accuracy for 100 classes= 35.13

-----Accuracy for 20 classes= 48.41

-----Considering Top-5-----  
-----

-----Accuracy for 100 classes= 65.22

-----Accuracy for 20 classes= 77.59



## Model-2

```
In [ ]: import pickle
import numpy as np
import tensorflow as tf
import matplotlib
matplotlib.use('agg')
import pylab as plt
```

```
In [ ]: count=1
epc=1
loss_graph_list=[]

#Reading the data
def unpickle(file):

    with open(file, 'rb') as fo:
        dict = pickle.load(fo, encoding='bytes')
    return dict

train=unpickle("cifar-100-python/train")
test=unpickle("cifar-100-python/test")

print(train.keys())
print(test.keys())

#Train data
Xtrain=train[b'data'].reshape((len(train[b'data']), 3, 32, 32)).transpose(0, 2, 3, 1)
ytrain=np.array(train[b'fine_labels'])

#Test data
Xtest=test[b'data'].reshape((len(test[b'data']), 3, 32, 32)).transpose(0, 2, 3, 1)
ytest=np.array(test[b'fine_labels'])

Xtrain=Xtrain.astype(np.float32)
Xtest=Xtest.astype(np.float32)

Xval=Xtrain[40000:,:,:,:]
Xtrain=Xtrain[0:40000]

yval=ytrain[40000:]
ytrain=ytrain[0:40000]

# Reset graph parameters
tf.reset_default_graph()

x = tf.placeholder(tf.float32, shape=(None, 32, 32, 3), name='input_x')
y = tf.placeholder(tf.int32, shape=(None,), name='output_y')
```

```

In [ ]: # model Architecture
def leNet(features):
    # layer 1 input
    # input_layer=tf.reshape(features['x'],[-1,32,32,3])

    # conv_layer #1
    conv1 = tf.layers.conv2d(inputs=features, filters=64,
                              kernel_size=[3,3], strides=1,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initi
alizer())

    # pooling_layer #1
    pool1 = tf.layers.max_pooling2d(inputs=conv1,
                                     pool_size=[2, 2],
                                     strides=2)

    batch1=tf.layers.batch_normalization(inputs=pool1)

    # conv_layer #2
    conv2 = tf.layers.conv2d(inputs=batch1, filters=128,
                              kernel_size=[3,3], strides=1,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initi
alizer())

    # pooling_layer #2
    pool2 = tf.layers.max_pooling2d(inputs=conv2,
                                     pool_size=[2, 2],
                                     strides=2)

    batch2=tf.layers.batch_normalization(inputs=pool2)

    # Dense_layer #1
    pool2_flat = tf.reshape(batch2, [-1, 4608])
    dense1 = tf.layers.dense(inputs=pool2_flat, units=256,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initi
alizer())

    batch3=tf.layers.batch_normalization(inputs=dense1)

    # Dense_layer #2
    dense2 = tf.layers.dense(inputs=batch3, units=512,
                              activation=tf.nn.relu,
                              kernel_initializer=tf.contrib.layers.xavier_initi
alizer())

    batch4=tf.layers.batch_normalization(inputs=dense2)

```

```
# logits_final_layer
logits = tf.layers.dense(inputs=batch4, units=100)

return logits
```

```
In [ ]: #epochs, batchSize, Learning rate
epochs = 2
batch_size = 64
learning_rate = 0.001

# Output of model
logits = leNet(x)
model = tf.identity(logits, name='logits')

#Loss function & Optimization Algorithm
cost = tf.reduce_mean(tf.nn.sparse_softmax_cross_entropy_with_logits(logits=logits, labels=y))
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate).minimize(cost)

#Prediction and Accuracy
correct_pred = tf.equal(tf.argmax(logits, 1),tf.cast(y,tf.int64))
accuracy = tf.reduce_mean(tf.cast(correct_pred, tf.float32), name='accuracy')
```

```
In [ ]: # Shuffling Data
def batch_features_labels(features, labels, batch_size):

    rand_index=np.random.choice(len(features),size=len(features))
    for start in range(0,len(features),batch_size):
        end=min(start+batch_size,len(features))
        tmp=np.array(rand_index[start:end])
        yield features[tmp],labels[tmp]
```

```

In [ ]: #saving model
        model_path = './image_classification'

#Creating Session
print('Training...')
with tf.Session() as sess:
    #inistalizing Global_variables
    sess.run(tf.global_variables_initializer())
    #####

    # Data Augmentation
    data_tf=tf.convert_to_tensor(Xtrain,np.float32)
    big=tf.image.resize_images(Xtrain,(36,36))
    top_r=tf.image.crop_to_bounding_box(big,0,0,32,32)
    top_l=tf.image.crop_to_bounding_box(big,0,4,32,32)
    bot_r=tf.image.crop_to_bounding_box(big,4,0,32,32)
    bot_l=tf.image.crop_to_bounding_box(big, 4, 4, 32, 32)
    cen=tf.image.crop_to_bounding_box(big,2,2,32,32)
    flip=tf.image.flip_left_right(data_tf)

    fbig = tf.image.resize_images(flip, (36, 36))
    ftop_r = tf.image.crop_to_bounding_box(fbig, 0, 0, 32, 32)
    ftop_l = tf.image.crop_to_bounding_box(fbig, 0, 4, 32, 32)
    fbot_r = tf.image.crop_to_bounding_box(fbig, 4, 0, 32, 32)
    fbot_l = tf.image.crop_to_bounding_box(fbig, 4, 4, 32, 32)
    fcen = tf.image.crop_to_bounding_box(fbig, 2, 2, 32, 32)

    Xtrain1 = tf.concat([data_tf, top_r, top_l, bot_l, bot_r, cen, flip], axis
=0)
    Xtrain2=tf.concat([ftop_l,ftop_r,fbot_l,fbot_r,fcen],axis=0)

    # Xtrain=tf.concat([Xtrain1,Xtrain2],axis=0)

    sess.run(Xtrain1)
    sess.run(Xtrain2)
    Xtrain1=Xtrain1.eval()
    Xtrain2=Xtrain2.eval()
    Xtrain=np.concatenate((Xtrain1,Xtrain2))
    ytrain=np.concatenate((ytrain,ytrain,ytrain,ytrain,ytrain,ytrain,ytrain,yt
rain,ytrain,ytrain,ytrain,ytrain))
    # Xtrain=Xtrain.eval()
    print(type(Xtrain))
    print(Xtrain.shape)
    print(ytrain.shape)

    #Calculating mean of the dataSet
    sub = np.mean(Xtrain, axis=0)
    Xtrain = Xtrain - sub
    Xval=Xval-sub
    np.save("mean_vec",sub)

    # Traning and Validation
    for epoch in range(0,epochs):
        for batch_features, batch_labels in batch_features_labels(Xtrain, ytra

```

```

in, batch_size):
    sess.run(optimizer,
               feed_dict={
                   x: batch_features,
                   y: batch_labels
               })

    loss = sess.run(cost,
                    feed_dict={
                        x: batch_features,
                        y: batch_labels
                    })
    loss_graph_list.append(loss)

    if(epoch==epc):
        epc+=1
        print('Epoch {:>2}\n'.format(epoch), end='')
        print(sum(loss_graph_list)/len(loss_graph_list))

    if(epoch==count):
        count+=1
        valid_acc = sess.run(accuracy,
                             feed_dict={
                                 x: Xval,
                                 y: yval})

        print('Loss: {:>10.4f} Validation Accuracy: {:.6f}'.format(loss, valid_acc))

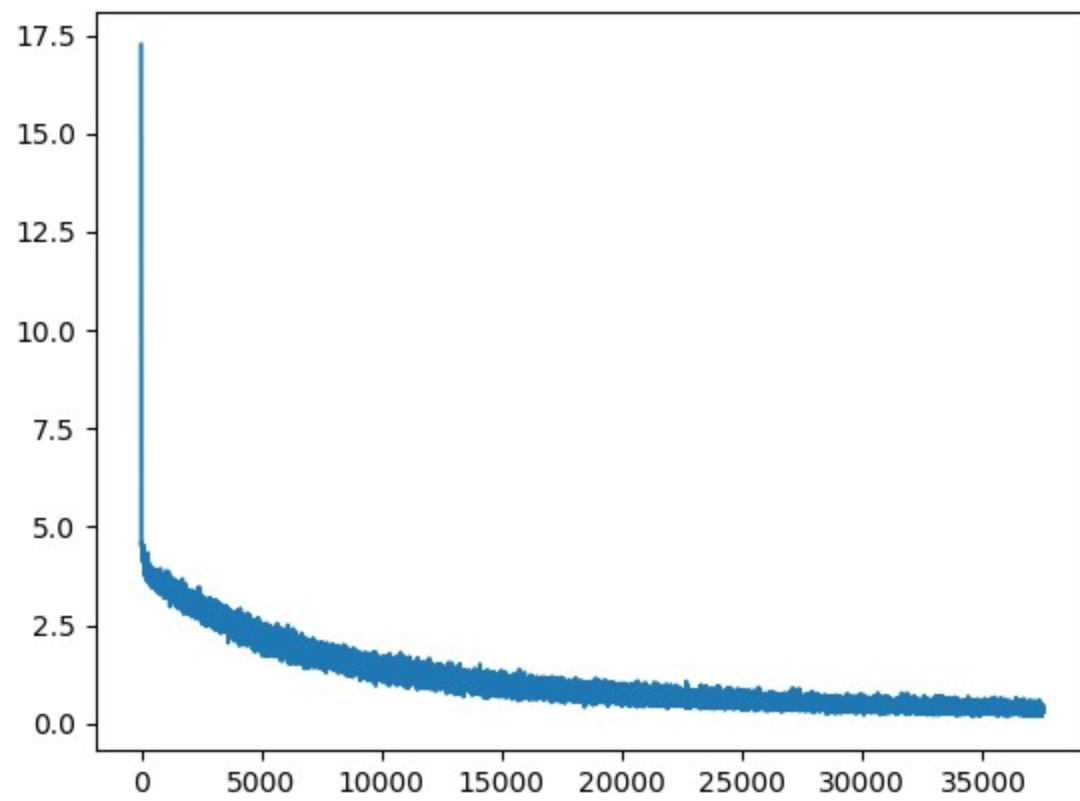
# Saving model and plotting Loss Graph
plt.plot(loss_graph_list)
plt.show()
plt.savefig("myfig")
model_saver = tf.train.Saver()
save_path_model = model_saver.save(sess, model_path)

```

```

Training...
2018-11-13 20:59:37.185457: I tensorflow/core/common_runtime
name: Tesla K80 major: 3 minor: 7 memoryClockRate(GHz): 0.
pciBusID: deb2:00:00.0
totalMemory: 11.92GiB freeMemory: 11.85GiB
2018-11-13 20:59:37.185501: I tensorflow/core/common_runtime
2018-11-13 20:59:37.462707: I tensorflow/core/common_runtime
2018-11-13 20:59:37.462757: I tensorflow/core/common_runtime
2018-11-13 20:59:37.462785: I tensorflow/core/common_runtime
2018-11-13 20:59:37.463060: I tensorflow/core/common_runtime
MB memory) -> physical GPU (device: 0, name: Tesla K80, po
<class 'numpy.ndarray'>
(480000, 32, 32, 3)
(480000,)
Epoch 1
Loss= 3.208299994945399
Loss: 3.2083 Validation Accuracy: 0.271600
Epoch 2
Loss= 2.639241218519217
Loss: 2.6392 Validation Accuracy: 0.352800
Epoch 3
Loss= 2.2469392018247825
Loss: 2.2469 Validation Accuracy: 0.375300
Epoch 4
Loss= 1.9602673269694555
Loss: 1.9603 Validation Accuracy: 0.386700
Epoch 5
Loss= 1.7422958605401098
Loss: 1.7423 Validation Accuracy: 0.391000
Epoch 6
Loss= 1.5704181075339836
Loss: 1.5704 Validation Accuracy: 0.381600
Epoch 7
Loss= 1.430973823993693
Loss: 1.4310 Validation Accuracy: 0.385100
Epoch 8
Loss= 1.3155695846680104
Loss: 1.3156 Validation Accuracy: 0.389100
Epoch 9
Loss= 1.218309359507156
Loss: 1.2183 Validation Accuracy: 0.384100

```



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```
In [1]: import tensorflow as tf
import pickle
import random
import pickle
import numpy as np
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix, accuracy_score
np.set_printoptions(threshold=np.nan)
```

```
/anaconda/envs/py35/lib/python3.5/site-packages/h5py/__init__.py:36: FutureWarning: Conversion of the second argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will be treated as `np.float64 == np.dtype(float).type`.
```

```
from ._conv import register_converters as _register_converters
```



```

In [2]: model_path = './image_classification'
        batch_size = 64
        n_samples = 10
        top_n_predictions = 5

        def unpickle(file):

            with open(file, 'rb') as fo:
                dict = pickle.load(fo, encoding='bytes')
            return dict

        train=unpickle("cifar-100-python/train")
        test=unpickle("cifar-100-python/test")
        meta=unpickle("cifar-100-python/meta")

        Xtest=test[b'data'].reshape((len(test[b'data']), 3, 32, 32)).transpose(0, 2, 3
        , 1)
        ytest=np.array(test[b'fine_labels'])

        Xtest=Xtest.astype(np.float32)

        sub=np.load("mean_vec.npy")
        Xtest=Xtest-sub

        ytestC=np.array(test[b'coarse_labels'])

        lab={}
        for i in range(0,20):
            lab[i]=set()
        for i in range(0,len(ytest)):
            lab[ytestC[i]].add(ytest[i])

        name_100_labels=[str(i)[2:-1] for i in meta[b'fine_label_names']]
        name_20_labels=[str(i)[2:-1] for i in meta[b'coarse_label_names']]

```

```

In [3]: def batch_features_labels(features, labels, batch_size):
        for start in range(0, len(features), batch_size):
            end = min(start + batch_size, len(features))
            yield features[start:end], labels[start:end]

```

```

In [4]: loaded_graph = tf.Graph()
        with tf.Session(graph=loaded_graph) as sess:

            loader = tf.train.import_meta_graph(model_path + '.meta')
            loader.restore(sess, model_path)
            loaded_x = loaded_graph.get_tensor_by_name('input_x:0')
            loaded_y = loaded_graph.get_tensor_by_name('output_y:0')
            loaded_logits = loaded_graph.get_tensor_by_name('logits:0')
            loaded_acc = loaded_graph.get_tensor_by_name('accuracy:0')

            label_predictor = sess.run(
                tf.nn.top_k(tf.nn.softmax(loaded_logits),1),
                feed_dict={loaded_x: Xtest, loaded_y: ytest})
            # print(label_predictor[1])

            matFine=confusion_matrix(ytest,label_predictor[1])
            accFine=accuracy_score(ytest,label_predictor[1])

            print("-----100 classes-----")

            print("\nAccuracy and confusion matrix for 100 labels")
            print("-----Accuracy=",100*accFine,"-----")
            print("\nConfusion-Matrix:\n",matFine)

            # print("\nPredicted_Labels:\n",label_predictor[1].reshape(len(label_predictor[1])).tolist())
            # print("\n Correct_Labels:\n",ytest)
            print("\n\n\n\n\n")

            coarse_pred=[]
            pred=np.reshape(label_predictor[1],(len(label_predictor[1]))).tolist()
            for i in range(0,len(pred)):
                for pos in lab.keys():
                    if(pred[i] in lab[pos]):
                        coarse_pred.append(pos)
            matCoa = confusion_matrix(ytestC,coarse_pred)
            accCoa = accuracy_score(ytestC, coarse_pred)
            print("-----20 classes-----")
            print("Accuracy and confusion matrix for 20 labels")
            print("-----Accuracy=",100*accCoa,"-----")
            print("\nConfusion-Matrix:\n",matCoa)

            # random_test_features, random_test_labels = tuple(zip(*random.sample(list
            (zip(Xtest, ytest)), n_samples)))

```

```

Top_3_pred= sess.run(
    tf.nn.top_k(tf.nn.softmax(loaded_logits), 1),
    feed_dict={loaded_x: Xtest, loaded_y: ytest})
print("-----Considering Top-3-----\n")
# print("Top 5 Classes:\n"Top_5_pred[1])
acc_top_3=0
for i in range(0,len(Top_3_pred[1])):
    if(ytest[i] in Top_3_pred[1][i]):
        acc_top_3+=1
print("-----Accuracy for 100 classes= ",100*acc_top_3/len(ytest))

cacc_top_3=0
for i in range(0,len(ytest)):
    temp=[]
    for j in lab.keys():
        if(ytest[i] in lab[j]):
            temp.append(lab[j])
            break
# print(temp)
# print(Top_5_pred[1][i])
if(len(temp)!=0):
    temp2=set(temp[0]).intersection(Top_3_pred[1][i])
# print(temp2)
if(len(temp2)!=0):
    cacc_top_3+=1
print("-----Accuracy for 20 classes= ",100*cacc_top_3/len(ytest))

Top_5_pred= sess.run(
    tf.nn.top_k(tf.nn.softmax(loaded_logits), top_n_predictions),
    feed_dict={loaded_x: Xtest, loaded_y: ytest})
print("-----Considering Top-5-----\n")
# print("Top 5 Classes:\n"Top_5_pred[1])
acc_top_5=0
for i in range(0,len(Top_5_pred[1])):
    if(ytest[i] in Top_5_pred[1][i]):
        acc_top_5+=1
print("-----Accuracy for 100 classes= ",100*acc_top_5/len(ytest))

cacc_top_5=0
for i in range(0,len(ytest)):
    temp=[]
    for j in lab.keys():
        if(ytest[i] in lab[j]):
            temp.append(list(lab[j]))
            break
    if(len(temp)!=0):
        temp2=set(temp[0]).intersection(Top_5_pred[1][i])

```

```
        if(len(temp2)!=0):
            cacc_top_5+=1
    print("-----Accuracy for 20 classes= ",100*cacc_top_5/len(ytest))
```

INFO:tensorflow:Restoring parameters from ./image\_classification  
-----100 classes-----

Accuracy and confusion matrix for 100 labels

-----Accuracy= 39.08 -----

Confusion-Matrix:

```
[[57  2  0  0  0  0  0  0  0  0  1  3  0  0  0  2  1  0  0  0  0  0  0  0
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  0  0  0  0  0  3  0  0  0  7  0  0  0  0  0  0  0  0  0  0  0  0  5  0
  0  0  0  0  0  0  2  0  0  0  0  9  0  0  0  0  0  0  0  0  4  0  0  0
  0  0  0  0]
[ 0 48  1  0  0  0  0  1  2  0  1  0  0  0  1  0  1  0  1  0  0  0  0  0
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  0  0  0  0]
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  0  0  0  0]
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  0  1  0  0]
```

```

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```

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```

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```



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[ 0 0 0 1 1 0 0 0 1 0 0 0 0 0 0 0 1 1 0 0 0 0 1
0 1 0 0 0 0 0 0 0 8 1 0 0 1 0 1 0 0 1 1 0 0 18
0 1 0 0 18 0 0 0 1 0 0 10 0 0 0 1 0 0 0 1 1 0 1 0
0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 0 0 0
25 0 0 0]
[ 0 1 0 0 2 0 0 1 0 0 0 0 0 0 0 1 2 0 0 2 0 1 1 1
0 0 1 0 1 1 0 0 0 4 1 0 0 1 4 1 0 0 0 1 0 0 1 0
0 0 2 0 0 0 0 0 2 0 0 0 0 0 0 2 2 0 5 0 0 0 0 0
2 0 4 0 0 0 2 0 2 0 0 0 1 0 0 0 1 1 0 0 0 0 0 0
0 44 1 1]
[ 0 1 5 1 0 1 2 0 1 0 1 4 0 0 1 1 0 0 0 0 0 1 2 0
1 0 5 0 0 0 0 0 1 0 0 18 1 1 0 0 1 0 0 0 1 8 0
0 0 0 3 0 2 1 1 1 1 0 0 0 0 0 3 1 1 0 0 0 1 0
0 0 0 1 0 0 1 0 0 0 0 0 1 1 3 0 1 1 0 0 0 0 0
0 0 17 1]
[ 2 0 1 0 1 0 0 2 1 0 2 0 2 1 0 0 0 0 0 0 0 6 1
0 2 3 0 1 0 1 0 0 0 1 0 0 0 1 4 2 1 1 0 1 1 0 0
0 0 1 1 0 0 0 0 0 0 0 0 0 3 0 0 1 0 1 2 0 3 0 0
0 1 1 1 0 0 7 1 0 0 0 0 0 0 1 1 0 1 0 0 0 0 0 1
0 0 0 35]]

```

```

-----20 classes-----
Accuracy and confusion matrix for 20 labels
-----Accuracy= 51.69000000000005 -----
-----
Confusion-Matrix:
[[202 40 1 3 7 12 5 8 31 7 9 20 21 21 12 27 35 6
 11 22]
 [ 37 241 9 10 8 11 5 11 6 3 12 6 19 20 10 49 16 4
 8 15]
 [ 0 4 339 14 50 2 3 21 0 1 2 1 5 16 12 10 5 2
 7 6]
 [ 7 7 11 268 15 61 15 10 3 5 2 7 8 11 24 12 7 2
 11 14]
 [ 6 13 51 17 293 5 2 19 11 0 0 13 6 18 14 18 5 3
 4 2]
 [ 2 6 7 32 8 263 29 12 8 12 9 8 11 15 12 20 4 3
 11 28]
 [ 1 5 5 28 3 46 253 5 5 16 10 14 10 13 23 8 9 6
 20 20]
 [ 10 10 17 11 13 4 3 283 15 3 4 7 12 42 8 22 18 5
 4 9]
 [ 31 7 2 5 16 8 4 18 226 1 10 39 47 11 10 19 27 3
 6 10]
 [ 4 3 0 10 0 11 8 2 6 333 23 8 1 4 6 6 3 18
 17 37]
 [ 16 13 5 6 7 11 4 8 3 21 324 3 7 5 2 7 4 24
 13 17]
 [ 26 6 1 6 10 7 6 26 46 12 11 217 34 13 15 21 9 6
 12 16]
 [ 28 14 3 1 3 8 2 18 52 5 2 32 197 24 12 33 48 4
 8 6]
 [ 15 10 10 14 13 22 9 50 13 4 6 7 17 185 15 49 28 9
 9 15]
 [ 7 12 13 13 12 19 14 14 21 4 0 13 17 27 259 6 17 5
 16 11]
 [ 39 28 2 3 8 16 4 32 26 9 15 24 21 48 15 156 21 6
 16 11]
 [ 27 15 4 11 12 12 3 21 37 6 8 29 48 36 22 39 156 3
 4 7]
 [ 6 4 7 1 1 6 1 5 10 19 18 3 5 2 1 5 1 388
 7 10]
 [ 5 8 7 8 0 10 7 13 2 25 7 4 8 4 8 12 0 3
 291 78]
 [ 4 10 3 3 2 12 10 11 7 35 7 7 4 12 3 11 3 5
 56 295]]
-----Considering Top-3-----
-----

-----Accuracy for 100 classes= 39.08
-----Accuracy for 20 classes= 51.69
-----Considering Top-5-----
-----

-----Accuracy for 100 classes= 67.66
-----Accuracy for 20 classes= 80.71

```