In [1]:

```
import os
import cv2
import matplotlib.pyplot as plt
import numpy as np
from tqdm import tqdm
from random import shuffle
import tensorflow as tf
import pandas as pd
from sklearn.metrics import roc_auc_score
```

In [2]:

```
epochs = 50
step_size = 8
IMG_SIZE_ALEXNET = 227 # image size
validating_size = 20 # while cross validating, we are evaluating
batch by batch
nodes_fc1 = 4096 # no of nodes on fc layer 1
nodes_fc2 = 4096 # no of nodes on fc layer 2
output_classes = 3 # three classes: eggplant,
output_locations = 4 # minx, miny, maxx, maxy
TRAIN_DIR = os.getcwd()
```

In [3]:

```
data = np.load('object localization.npy')
train = data[:int(len(data)*0.8)]
cv = data[int(len(data)*0.8):]
#print(train)
# X for train input
X = np.array([i[0] for i in train]).reshape(-1,IMG SIZE ALEXNET,
IMG SIZE ALEXNET, 3)
\#X = np.array([i[0] for i in train])
#X = Xa.reshape(-1,IMG SIZE ALEXNET,IMG SIZE ALEXNET,3)
# Y1 for classification head
Y1 = np.array([i[1] for i in train])
# Y2 for regression head
Y2 = np.array([i[2] for i in train])
# cv x for train input
cv x = np.array([i[0] for i in cv]).reshape(-1,IMG SIZE ALEXNET,
IMG SIZE ALEXNET, 3)
# cv y1 for classification head
cv y1 = np.array([i[1] for i in cv])
# cv y2 for regression head
cv y2 = np.array([i[2] for i in cv])
print(cv_y1[:10])
print(X.shape)
print(Y1.shape)
print(cv x.shape)
print(cv y1.shape)
```

```
[[0 \ 0 \ 1]
 [0 \ 0 \ 1]
 [0 \ 0 \ 1]
 [0 0 1]
 [0 \ 0 \ 1]
 [0 \ 0 \ 1]
 [0 0 1]
 [0 \ 0 \ 1]
 [0 \ 0 \ 1]
 [0 0 1]]
(21196, 227, 227, 3)
(21196, 3)
(5300, 227, 227, 3)
(5300, 3)
In [4]:
steps = len(train)
print(steps)
remaining = steps % step size
#Resetting graph
tf.reset default graph()
#Defining Placeholders
x = tf.placeholder(tf.float32,shape=[None,IMG SIZE ALEXNET,IMG S
IZE ALEXNET, 3])
y_true_1 = tf.placeholder(tf.float32,shape=[None,output classes]
)
y true 2 = tf.placeholder(tf.float32,shape=[None,output location
s])
##CONVOLUTION LAYER 1
#Weights for layer 1
w 1 = tf.Variable(tf.truncated normal([11,11,3,96], stddev=0.01)
)
#Bias for layer 1
b 1 = tf. Variable(tf.constant(0.0, shape=[[11,11,3,96][3]]))
#Applying convolution
c 1 = tf.nn.conv2d(x, w 1, strides=[1, 4, 4, 1], padding='VALID')
#Adding bias
c 1 = c 1 + b 1
#Applying RELU
c 1 = tf.nn.relu(c 1)
```

```
print(c_1)
##POOLING LAYER1
p 1 = tf.nn.max pool(c 1, ksize=[1, 3, 3, 1], strides=[1, 2, 2, 1]
], padding='VALID')
print(p 1)
##CONVOLUTION LAYER 2
#Weights for layer 2
w 2 = tf.Variable(tf.truncated normal([5,5,96,256], stddev=0.01)
)
#Bias for layer 2
b = tf.Variable(tf.constant(1.0, shape=[[5,5,96,256][3]]))
#Applying convolution
c_2 = tf.nn.conv2d(p_1, w_2,strides=[1, 1, 1, 1], padding='SAME'
#Adding bias
c 2 = c 2 + b 2
#Applying RELU
c 2 = tf.nn.relu(c 2)
print(c 2)
21196
WARNING:tensorflow:From /anaconda3/lib/python3.7/sit
```

e-packages/tensorflow/python/framework/op_def_librar y.py:263: colocate_with (from tensorflow.python.fram ework.ops) is deprecated and will be removed in a fu ture version.

Instructions for updating:
Colocations handled automatically by placer.

Tensor("Relu:0", shape=(?, 55, 55, 96), dtype=float3 2)

Tensor("MaxPool:0", shape=(?, 27, 27, 96), dtype=float32)

Tensor("Relu 1:0", shape=(?, 27, 27, 256), dtype=float32)

at32)

In [5]:

```
p_2 = tf.nn.max_pool(c_2, ksize=[1, 3, 3, 1],strides=[1, 2, 2, 1
], padding='VALID')
print(p_2)
##CONVOLUTION LAYER 3
#Weights for layer 3
w 3 = tf.Variable(tf.truncated normal([3, 3, 256, 384], stddev=0
.01))
#Bias for layer 3
b 3 = tf. Variable(tf.constant(0.0, shape=[[3, 3, 256, 384][3]]))
#Applying convolution
c_3 = tf.nn.conv2d(p_2, w_3,strides=[1, 1, 1, 1], padding='SAME'
)
#Adding bias
c 3 = c 3 + b 3
#Applying RELU
c 3 = tf.nn.relu(c 3)
print(c 3)
```

```
Tensor("MaxPool_1:0", shape=(?, 13, 13, 256), dtype=
float32)
Tensor("Relu_2:0", shape=(?, 13, 13, 384), dtype=flo
at32)
```

```
In [6]:
```

```
w_4 = tf.Variable(tf.truncated_normal([3, 3, 384, 384], stddev=0
.01))
#Bias for layer 4
b_4 = tf.Variable(tf.constant(0.0, shape=[[3, 3, 384, 384][3]]))
#Applying convolution
c_4 = tf.nn.conv2d(c_3, w_4,strides=[1, 1, 1, 1], padding='SAME')

#Adding bias
c_4 = c_4 + b_4
#Applying RELU
c_4 = tf.nn.relu(c_4)

print(c_4)
```

Tensor("Relu_3:0", shape=(?, 13, 13, 384), dtype=flo at32)

In [7]:

```
w_5 = tf.Variable(tf.truncated_normal([3, 3, 384, 256], stddev=0
.01))
#Bias for layer 5
b_5 = tf.Variable(tf.constant(0.0, shape=[[3, 3, 384, 256][3]]))
#Applying convolution
c_5 = tf.nn.conv2d(c_4, w_5, strides=[1, 1, 1, 1], padding='SAME')

#Adding bias
c_5 = c_5 + b_5
#Applying RELU
c_5 = tf.nn.relu(c_5)
print(c_5)
```

Tensor("Relu_4:0", shape=(?, 13, 13, 256), dtype=flo at32)

```
In [8]:
p 3 = tf.nn.max pool(c 5, ksize=[1, 3, 3, 1], strides=[1, 2, 2, 1]
], padding='VALID')
print(p 3)
#Flattening
flattened = tf.reshape(p 3,[-1,6*6*256])
print(flattened)
Tensor("MaxPool 2:0", shape=(?, 6, 6, 256), dtype=fl
oat32)
Tensor("Reshape:0", shape=(?, 9216), dtype=float32)
In [9]:
input size = int( flattened.get shape()[1] )
#Weights for FC Layer 1
w1 fc = tf.Variable(tf.truncated normal([input size, nodes fc1],
stddev=0.01))
```

```
#Bias for FC Layer 1
b1 fc = tf.Variable( tf.constant(1.0, shape=[nodes fc1] ) )
#Summing Matrix calculations and bias
s fc1 = tf.matmul(flattened, w1 fc) + b1 fc
#Applying RELU
s fc1 = tf.nn.relu(s fc1)
#Dropout Layer 1
hold prob1 = tf.placeholder(tf.float32)
s fc1 = tf.nn.dropout(s fc1,keep prob=hold prob1)
print(s fc1)
```

```
WARNING:tensorflow:From <ipython-input-9-c7b7cb0b8f1
b>:13: calling dropout (from tensorflow.python.ops.n
n ops) with keep prob is deprecated and will be remo
ved in a future version.
Instructions for updating:
Please use `rate` instead of `keep prob`. Rate shoul
d be set to `rate = 1 - keep prob`.
Tensor("dropout/mul:0", shape=(?, 4096), dtype=float
32)
```

```
In [10]:
```

```
w2_fc = tf.Variable(tf.truncated_normal([nodes_fc1, nodes_fc2],
stddev=0.01))
#Bias for FC Layer 2
b2_fc = tf.Variable( tf.constant(1.0, shape=[nodes_fc2] ) )
#Summing Matrix calculations and bias
s_fc2 = tf.matmul(s_fc1, w2_fc) + b2_fc
#Applying RELU
s_fc2 = tf.nn.relu(s_fc2)
print(s_fc2)
```

Tensor("Relu_6:0", shape=(?, 4096), dtype=float32)

In [11]:

```
hold_prob2 = tf.placeholder(tf.float32)
s_fc2 = tf.nn.dropout(s_fc2,keep_prob=hold_prob1)

##Fully Connected Layer 3 -- CLASSIFICATION HEAD
#Weights for FC Layer 3
w3_fc_1 = tf.Variable(tf.truncated_normal([nodes_fc2,output_classes], stddev=0.01))
#Bias for FC Layer 3b3_fc = tf.Variable( tf.constant(1.0, shape=
[output_classes] ) )
b3_fc_1 = tf.Variable( tf.constant(1.0, shape=[output_classes] )
)
#Summing Matrix calculations and bias
y_pred_1 = tf.matmul(s_fc2, w3_fc_1) + b3_fc_1
#Applying RELU
print(y_pred_1)
```

Tensor("add 7:0", shape=(?, 3), dtype=float32)

In [12]:

```
w3_fc_2 = tf.Variable(tf.truncated_normal([nodes_fc2,output_loca
tions], stddev=0.01))
#Bias for FC Layer 3b3_fc = tf.Variable( tf.constant(1.0, shape=
[output_classes] ) )
b3_fc_2 = tf.Variable( tf.constant(1.0, shape=[output_locations]
) )
#Summing Matrix calculations and bias
y_pred_2 = tf.matmul(s_fc2, w3_fc_2) + b3_fc_2
#Applying RELU
print(y_pred_2)
```

Tensor("add_8:0", shape=(?, 4), dtype=float32)

```
In [13]:
```

```
cross entropy = tf.multiply(tf.reduce mean(tf.nn.softmax cross e
ntropy with logits v2(labels=y true 1,logits=y pred 1)),10)
#Defining Regression Loss
regression loss = tf.multiply(tf.reduce mean(tf.square(y pred 2
- y true 2)),1.0)
#Defining total loss
final loss = cross entropy + regression loss
#Defining objective
train = tf.train.AdamOptimizer(learning rate=0.00001).minimize(f
inal loss)
#Defining Accuracy
matches = tf.equal(tf.argmax(y_pred_1,1),tf.argmax(y_true_1,1))
acc = tf.reduce mean(tf.cast(matches,tf.float32))
#Initializing weights
init = tf.global variables initializer()
#Starting Empty lists to keep results
acc list = []
auc list = []
loss list = []
regression list = []
#In order to save, creating a tf.train.Saver() object.
saver = tf.train.Saver()
```

In [14]:

```
config = tf.ConfigProto(allow soft placement=True)
config.gpu options.allow growth = True
config.gpu options.allocator type = 'BFC'
tf.add to collection("classification head", y pred 1)
tf.add to collection("regression head", y pred 2)
```

In [15]:

```
def main():
   with tf.Session(config=config) as sess:
```

```
sess.run(init)
        for i in range(epochs):
            for j in range(0, steps-remaining, step size):
                #Feeding step size-amount data with 0.5 keeping
probabilities on DROPOUT LAYERS
                _,c = sess.run([train,final_loss],
                feed dict = {x:X[j:j+step size], y true 1:Y1[j:j
+step size],
                y true 2:Y2[j:j+step size],hold prob1:0.5,hold p
rob2:0.5})
            #Writing for loop to calculate test statistics. GTX
1050 isn't able to calculate all cv data.
            cv auc list = []
            cv acc list = []
            cv loss list = []
            cv regression list = []
            for v in range(0,len(cv_x)-int(len(cv_x) % validatin
g size), validating size):
                acc on cv,loss on cv,preds,coordinates = sess.ru
n([acc,cross_entropy,tf.nn.softmax(y_pred_1),y_pred_2],
                feed dict={x:cv x[v:v+validating size], y true 1
:cv y1[v:v+validating size], y true 2:cv y2[v:v+validating size]
                  hold prob1:1.0, hold prob2:1.0})
                auc on cv = roc auc score(cv y1[v:v+validating s
ize],preds,average='micro')
                regression loss = np.mean(pow(cv y2[v:v+validati
ng size] - coordinates , 2 ) )
                cv acc list.append(acc on cv)
                cv auc list.append(auc on cv)
                cv loss list.append(loss on cv)
                cv regression list.append(regression loss)
            acc_cv_ = round(np.mean(cv_acc list),5)
            auc cv = round(np.mean(cv auc list),5)
            loss_cv_ = round(np.mean(cv_loss_list),5)
            regression_loss_cv_ = round(np.mean(cv_loss list),5)
            acc list.append(acc cv )
            auc list.append(auc cv )
            loss list.append(loss cv )
            regression list.append(regression loss cv )
            print("Epoch:",i,"Accuracy:",acc_cv_,"Loss:",loss_cv
```