## In [ ]:

```
import tensorflow as tf
import math
import numpy as np
import h5py
import matplotlib.pyplot as plt
import scipy
from PIL import Image
from scipy import ndimage

%matplotlib inline
np. random. seed(1)
```

## In [ ]:

```
def init_placeholders(n_H0, n_W0, n_C0, n_y):
    X = tf.placeholder(tf.float32, [None, n_H0, n_W0, n_C0])
    Y = tf.placeholder(tf.float32, [None, n_y])
    return X, Y

X, Y = create_placeholders(64, 64, 3, 6)
    print ("X = " + str(X))
    print ("Y = " + str(Y))
```

## In [ ]:

In [ ]:

```
def forward propagation(X, parameters):
    # Retrieve the parameters from the dictionary "parameters"
    W1 = parameters['W1']
    W2 = parameters['W2']
    # 1. cnn layer -----
    # CONV2D: stride of 1, padding 'SAME'
   Z1 = tf.nn.conv2d(X, W1, strides=[1, 1, 1, 1], padding='SAME')
    # RELU
   A1 = tf. nn. relu(Z1)
    # MAXPOOL: window 8x8, sride 8, padding 'SAME'
   P1 = tf.nn.max_pool(A1, ksize=[1, 8, 8, 1], strides=[1, 8, 8, 1], padding='SAME')
    # 2. cnn layer --
    # CONV2D: filters W2, stride 1, padding 'SAME'
   Z2 = tf.nn.conv2d(P1, W2, strides=[1, 1, 1, 1], padding='SAME')
    A2 = tf. nn. relu(Z2)
   P2 = tf.nn.max_pool(A2, ksize=[1, 4, 4, 1], strides=[1, 4, 4, 1], padding='SAME')
    # FLATTEN
   P2 = tf. contrib. layers. flatten(P2)
    # FULLY-CONNECTED without non-linear activation function (not not call softmax).
    # 6 neurons in output layer. Hint: one of the arguments should be "activation_fn=None"
   Z3 = tf.contrib.layers.fully_connected(P2, 6, activation_fn=None)
    return Z3
```

In [ ]:

```
def fit(learning rate=0.009,
        num_epochs=400,
        minibatch size=64,
        print cost=True):
    tf.reset default graph()
                                                      # to be able to rerun the model without overwrl
    tf. set random seed(1)
                                                       # to keep results consistent (tensorflow seed)
                                                       # to keep results consistent (numpy seed)
    seed = 3
    # get the dimensions by shape of training input
    (m, n HO, n WO, n CO) = mnist.train.images
    n y = Y train. shape[1]
   costs = []
                                                       # To keep track of the cost
    # Create Placeholders of the correct shape
               = init placeholders (n HO, n WO, n CO, n y)
    parameters = init_parameters(n_C0, 16)
    # Forward propagation: Build the forward propagation in the tensorflow graph
   Z3 = forward_propagation(X, parameters)
    # Cost function: Add cost function to tensorflow graph
    cost = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits_v2(logits=Z3, labels=Y))
    optimizer = tf. train. AdamOptimizer (learning rate=learning rate). minimize (cost)
    init = tf.global_variables_initializer()
    with tf. Session() as sess:
        sess.run(init)
        # Do the training loop
        for epoch in range (num_epochs):
            minibatch cost = 0.
            num minibatches = int(m / minibatch size) # number of minibatches of size minibatch size
            seed = seed + 1
            batch_xs, batch_ys = mnist.train.next_batch(minibatch_size)
            for in num minibatches:
                # Run the session to execute the optimizer and the cost, the feedict should contain
                _, temp_cost = sess.run([optimizer, cost], feed_dict={X: batch_xs.reshape(-1, 32, 3
                                                                        Y: batch ys. reshape (-1, 10))
                minibatch cost += temp cost / num minibatches
            # Print the cost every epoch
            if print cost == True and epoch % 100 == 0:
                print ("Cost after epoch %i: %f" % (epoch, minibatch_cost))
            if print cost == True and epoch % 1 == 0:
                costs.append(minibatch cost)
        # plot the cost
        plt. plot (np. squeeze (costs))
        plt.ylabel('cost')
        plt.xlabel('iterations (per tens)')
        plt. title("Learning rate =" + str(learning rate))
```

```
plt.show()

# Calculate the correct predictions
predict_op = tf.argmax(Z3, 1)
correct_prediction = tf.equal(predict_op, tf.argmax(Y, 1))

# Calculate accuracy on the test set
accuracy = tf.reduce_mean(tf.cast(correct_prediction, "float"))
print(accuracy)

train_accuracy = accuracy.eval({X: mnist.train.images, Y: mnist.train.labels})
# sess.run(accuracy, feed_dict={X: mnist.train.images, Y: mnist.train.labels})

test_accuracy = accuracy.eval({X: mnist.test.images, Y: mnist.test.images})

print("Train Accuracy:", train_accuracy)
print("Test Accuracy:", test_accuracy)
return train_accuracy, test_accuracy, parameters
```

## In [ ]:

fit()