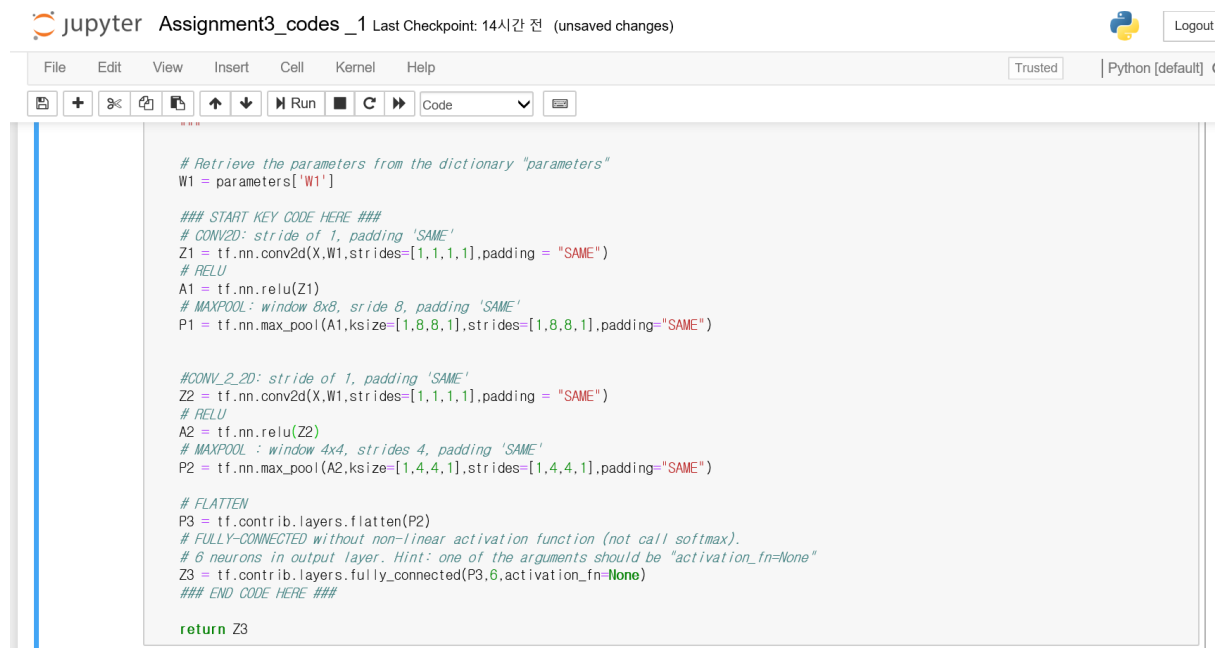


1.

1) Briefly describe how you added the layer.



```
# Retrieve the parameters from the dictionary "parameters"
W1 = parameters['W1']

### START KEY CODE HERE ###
# 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, stride 8, padding 'SAME'
P1 = tf.nn.max_pool(A1, ksize=[1, 8, 8, 1], strides=[1, 8, 8, 1], padding="SAME")

# CONV_2_2D: stride of 1, padding 'SAME'
Z2 = tf.nn.conv2d(X, W1, strides=[1, 1, 1, 1], padding = "SAME")
# RELU
A2 = tf.nn.relu(Z2)
# MAXPOOL: window 4x4, strides 4, padding 'SAME'
P2 = tf.nn.max_pool(A2, ksize=[1, 4, 4, 1], strides=[1, 4, 4, 1], padding="SAME")

# FLATTEN
P3 = tf.contrib.layers.flatten(P2)
# FULLY-CONNECTED without non-linear activation function (not call softmax).
# 6 neurons in output layer. Hint: one of the arguments should be "activation_fn=None"
Z3 = tf.contrib.layers.fully_connected(P3, 6, activation_fn=None)
### END CODE HERE ###

return Z3
```

We added layer2 between layer1 and flatten. The followings are what we added.

#CONV\_2\_2D: stride of 1, padding 'SAME'

```
Z2 = tf.nn.conv2d(X, W1, strides=[1, 1, 1, 1], padding = "SAME")
```

```
# RELU
```

```
A2 = tf.nn.relu(Z2)
```

```
# MAXPOOL : window 4x4, strides 4, padding 'SAME'
```

```
P2 = tf.nn.max_pool(A2, ksize=[1, 4, 4, 1], strides=[1, 4, 4, 1], padding="SAME")
```

As we added P2, We changed original P2 to P3.

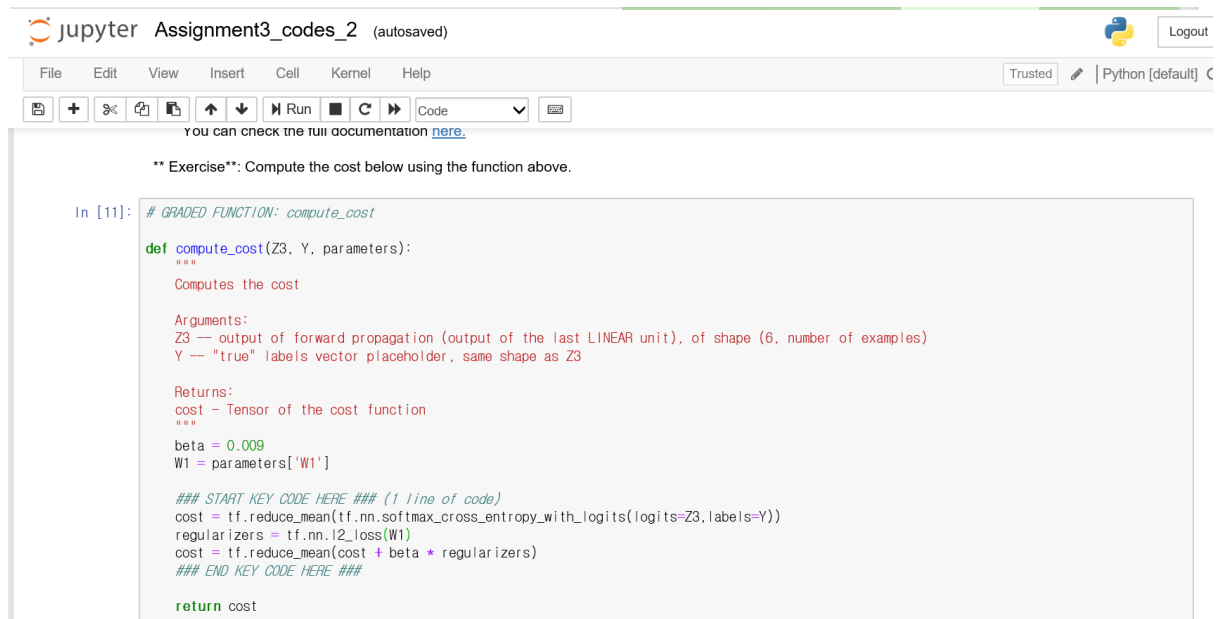
**2) Report your train and test accuracy (%) by running the following commend. Write the average accuracy of the total of 5 runs (e.g., 80% = (79+80+81+78+82)/5). Each run has 150 epochs (iteration).**

Train : 0.166667

Test : 0.166667

2.

## 1) Briefly describe how you implemented the regularization term.



The image shows a Jupyter Notebook interface with a code cell. The notebook title is "Assignment3\_codes\_2 (autosaved)". The code cell contains a function `compute_cost` that calculates the cost with L2 regularization. The function takes `Z3` and `Y` as inputs and returns the cost. The regularization term is implemented as `tf.nn.l2_loss(W1)` multiplied by a regularization parameter `beta`.

```
In [11]: # GRADED FUNCTION: compute_cost

def compute_cost(Z3, Y, parameters):
    """
    Computes the cost

    Arguments:
    Z3 -- output of forward propagation (output of the last LINEAR unit), of shape (6, number of examples)
    Y -- "true" labels vector placeholder, same shape as Z3

    Returns:
    cost - Tensor of the cost function
    """
    beta = 0.009
    W1 = parameters['W1']

    ### START KEY CODE HERE ### (1 line of code)
    cost = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(logits=Z3, labels=Y))
    regularizers = tf.nn.l2_loss(W1)
    cost = tf.reduce_mean(cost + beta * regularizers)
    ### END KEY CODE HERE ###

    return cost
```

We wanted to regularize  $W1$ , so we added the following codes

Regularizes = `tf.nn.l2_loss(W1)`

Cost=`tf.reduce_mean(cost + beta * regularizers)`

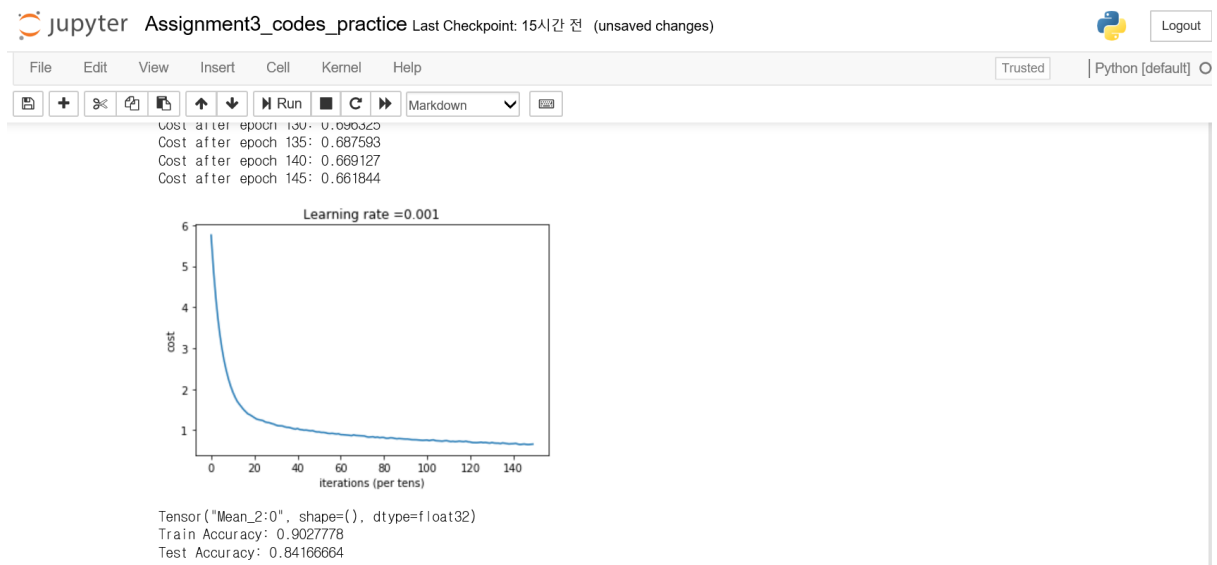
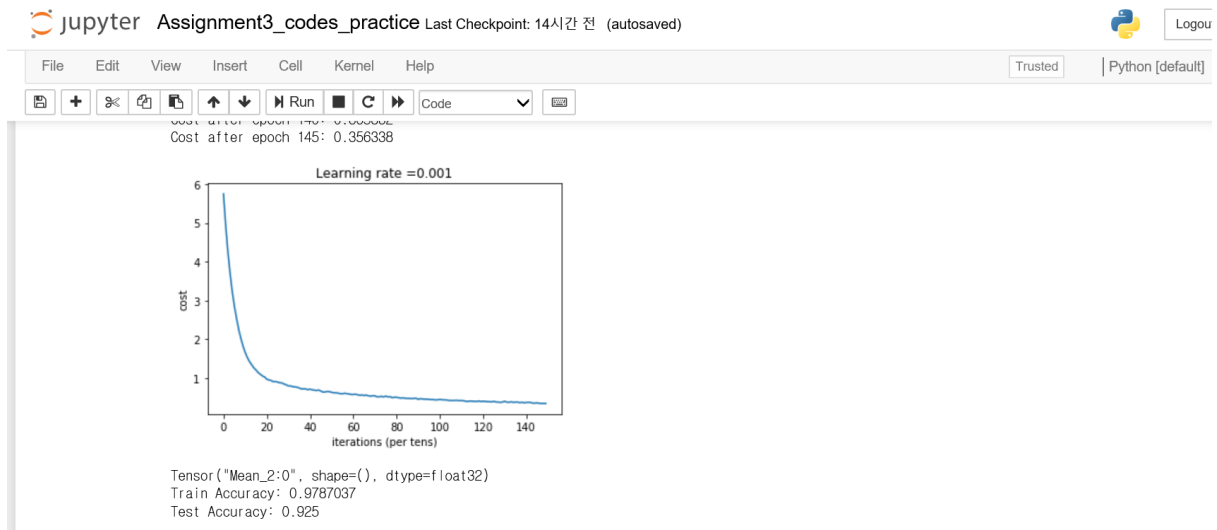
## 2) Report your train and test accuracy (%) by running the following command. Write the average accuracy of the total of 5 runs. Each run has 150 epochs (iteration).

Train : 0.982407

Test : 0.808333

## 3.

### 1) Describe how you improved its "test" accuracy (e.g., tuning the hyperparameters, changing the structure of layers).



**2) Report your test accuracy (%) by running the following commend. Write the average accuracy of the total of 5 runs. You may choose the number of epochs in your discretion, but it should not exceed 1000 epochs per run. ### START KEY CODE HERE ### (1 line of code)**

Train : 1.0

Test : 0.775

Train : 0.910185

Test : 0.858333

