Assignment 3: TENSORFLOW – COMPUTATIONAL GRAPH

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Submit before MONDAY (08/09) 11:30PM for both CQ & CLC classes

```
When running these examples, students must deal with errors. This is a way to study!
```

```
3.1.
     Running and explain (Constant trong tensorflow)
3.1.1 import tensorflow as tf
h = tf.constant("Hello")
w = tf.constant(" World!")
hw = h + w
with tf.Session() as
ans = sess.run(hw)
print (ans)
3.1.2
     import tensorflow.compat.v1 as tf
     x = tf.constant(5, tf.float32)
     y = tf.constant([5], tf.float32)
     z = tf.constant([5,3,4], tf.float32)
     t = tf.constant([[5,3,4,6],[2,3,4,7]], tf.float[32)
     u = tf.constant([[[5,3,4,6],[2,3,4,0]]], tf.float32)
     v = tf.constant([[[5,3,4,6],[2,3,4,0]],
                       [[5,3,4,6],[2,3,4,0]],
                       [[5,3,4,6],[2,3,4,0]]
                      ], tf.float32)
     print(v)
```

3.2. Running and explain (Variable in tensorflow)

3.2.1

```
import tensorflow.compat.v1 as tf
tf.compat.v1.disable_eager_execution()

x1 = tf.Variable(5.3, tf.float32)
x2 = tf.Variable(4.3, tf.float32)
x = tf.multiply(x1,x2)

init = tf.global_variables_initializer()
with tf.Session() as sess:
```

```
sess.run(init)
          t = sess.run(x)
         print(t)
     _____
3.2.2 import tensorflow.compat.v1 as tf
     tf.compat.v1.disable eager execution()
     x1 = tf.Variable([[5.3, 4.5, 6.0],
                      [4.3,4.3,7.0]
                     1, tf.float32)
     x2 = tf.Variable([[4.3, 4.3, 7.0],
                      [5.3, 4.5, 6.0]
                      ], tf.float32)
     x = tf.multiply(x1, x2)
     init = tf.global variables initializer()
     with tf.Session() as sess:
        sess.run(init)
        t = sess.run(x)
       print(t)
3.2.3 import tensorflow.compat.v1 as tf
     # creating nodes in computation graph
     node = tf.Variable(tf.zeros([2,2]))
     # running computation graph
     with tf.Session() as sess:
         # initialize all global variables
         sess.run(tf.global variables initializer())
         # evaluating node
        print("Tensor value before addition:\n", sess.run(node))
        # elementwise addition to tensor
        node = node.assign(node + tf.ones([2,2]))
         # evaluate node again
        print("Tensor value after addition:\n", sess.run(node))
         sess.close()
3.3. Running and explain (Placeholder)
```

```
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32, None)
y = tf.add(x, x)
with tf.Session() as sess:
    x data= 5
    result = sess.run(y, feed dict={x:x data})
    print(result)
3.3.2
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32,[None,3])
y = tf.add(x,x)
with tf.Session() as sess:
    x data = [[1.5, 2.0, 3.3]]
    result = sess.run(y, feed dict={x:x data})
    print(result)
3.3.3
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32,[None,None,3])
y = tf.add(x, x)
with tf.Session() as sess:
    x data = [[[1,2,3]]]
    result = sess.run(y, feed dict={x:x data})
    print(result)
3.3.4
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32, [None, 4, 3])
y = tf.add(x, x)
with tf.Session() as sess:
    x data = [[[1,2,3],
              [2,3,4],
              [2,3,5],
              [0,1,2]
            ]]
    result = sess.run(y, feed dict={x:x data})
    print(result)
```

```
3.3.5
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32, [2,4,3])
y = tf.add(x,x)
with tf.Session() as sess:
    x data = [[[1,2,3],
               [2,3,4],
               [2,3,5],
               [0,1,2]
            ],
             [[1,2,3],
               [2,3,4],
               [2,3,5],
               [0,1,2]
            ]]
    result = sess.run(y, feed dict={x:x data})
    print(result)
3.3.6
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x = tf.placeholder(tf.float32,[2,4,3])
y = tf.placeholder(tf.float32, [2,4,3])
z = tf.add(x,y)
u = tf.multiply(x,y)
with tf.Session() as sess:
    x data = [[[1,2,3],
               [2,3,4],
               [2,3,5],
               [0,1,2]
            ],
             [[1,2,3],
               [2,3,4],
              [2,3,5],
               [0,1,2]
            ]]
    y data = [[[1,2,3],
              [2,3,4],
         [2,3,5],
               [0,1,2]
            ],
             [[1,2,3],
```

```
[2,3,4],
          [2,3,5],
          [0,1,2]
        11
result1 = sess.run(z, feed dict={x:x data, y:y data})
result2 = sess.run(u,feed dict={x:x data, y:y data})
print("result1 = ", result1)
print("result2 =", result2)
```

3.4. Operation. Run and explain

```
3.4.1
     import tensorflow.compat.v1 as tf
     tf.compat.v1.disable eager execution()
     x1 = tf.constant(5.3, tf.float32)
     x2 = tf.constant(1.5, tf.float32)
     w1 = tf.Variable(0.7, tf.float32)
     w2 = tf.Variable(0.5, tf.float32)
     u = tf.multiply(x1, w1)
     v = tf.multiply(x2, w2)
     z = tf.add(u,v)
     result = tf.sigmoid(z)
     init = tf.global variables initializer()
     with tf.Session() as sess:
         sess.run(init)
         print(sess.run(result))
3.4.2
import numpy as np
import matplotlib.pyplot as plt
number of points = 500
x point = []
y point = []
a = 0.22
b = 0.78
for i in range (number of points):
    x = np.random.normal(0.0, 0.5)
    y = a*x + b + np.random.normal(0.0,0.1)
    x point.append([x])
    y point.append([y])
```

```
plt.plot(x point, y point, 'o', label = 'Input Data')
plt.legend()
plt.show()
3.4.3
import tensorflow.compat.v1 as tf
tf.compat.v1.disable eager execution()
x1 = tf.placeholder(tf.float32,[None,3])
x2 = tf.placeholder(tf.float32,[None,3])
w1 = tf.Variable([0.5, 0.4, 0.7], tf.float32)
w2 = tf.Variable([0.8, 0.5, 0.6], tf.float32)
u1 = tf.multiply(w1, x1)
u2 = tf.multiply(w2, x2)
v = tf.add(u1,u2)
z = tf.sigmoid(v)
init = tf.qlobal variables initializer()
with tf.Session() as sess:
    x1 data = [[1, 2, 3]]
    x2 data = [[1, 2, 3]]
    sess.run(init)
    result = sess.run(z,feed dict={x1:x1 data, x2:x2 data})
    print(result)
3.4.4
import tensorflow as tf
import numpy as np
matrix1 = np.array([(2,2,2),(2,2,2),(2,2,2)],dtype = 'int32')
matrix2 = np.array([(1,1,1),(1,1,1),(1,1,1)],dtype = 'int32')
print (matrix1)
print (matrix2)
matrix1 = tf.constant(matrix1)
matrix2 = tf.constant(matrix2)
matrix product = tf.matmul(matrix1, matrix2)
matrix sum = tf.add(matrix1, matrix2)
matrix 3 = \text{np.array}([(2,7,2),(1,4,2),(9,0,2)],\text{dtype} = 'float32')
print (matrix 3)
matrix det = tf.matrix determinant(matrix 3)
with tf.Session() as sess:
   result1 = sess.run(matrix product)
```

```
result2 = sess.run(matrix_sum)
result3 = sess.run(matrix_det)

print (result1)
print (result2)
print (result3)
```

3.5. Running and explain **Linear Regression model** using TensorFlow Core API.

```
# importing the dependencies
import tensorflow.compat.v1 as tf
import numpy as np
import matplotlib.pyplot as plt
# Model Parameters
learning rate = 0.01
training epochs = 2000
display step = 200
# Training Data
train X =
np.asarray([3.3,4.4,5.5,6.71,6.93,4.168,9.779,6.182,7.59,2.167,
7.042,10.791,5.313,7.997,5.654,9.27,3.1])
train y =
np.asarray([1.7,2.76,2.09,3.19,1.694,1.573,3.366,2.596,2.53,1.22
1,
                         2.827, 3.465, 1.65, 2.904, 2.42, 2.94, 1.3])
n samples = train X.shape[0]
# Test Data
test X = np.asarray([6.83, 4.668, 8.9, 7.91, 5.7, 8.7, 3.1,
2.1])
# Set placeholders for feature and target vectors
X = tf.placeholder(tf.float32)
y = tf.placeholder(tf.float32)
# Set model weights and bias)
test y = np.asarray([1.84, 2.273, 3.2, 2.831, 2.92, 3.24, 1.35,
1.03])
W = tf.Variable(np.random.randn(), name="weight")
b = tf.Variable(np.random.randn(), name="bias")
# Construct a linear model
linear model = W*X + b
# Mean squared error
```

```
cost = tf.reduce sum(tf.square(linear model - y)) /
(2*n samples)
# Gradient descent
optimizer =
tf.train.GradientDescentOptimizer(learning rate).minimize(cost)
# Initializing the variables
init = tf.qlobal variables initializer()
# Launch the graph
with tf.Session() as sess:
    # Load initialized variables in current session
   sess.run(init)
 # Fit all training data
    for epoch in range (training epochs):
        # perform gradient descent step
        sess.run(optimizer, feed dict={X: train X, y: train y})
        # Display logs per epoch step
        if (epoch+1) % display step == 0:
            c = sess.run(cost, feed dict={X: train X, y:
train y})
            print("Epoch:{0:6} \t Cost:{1:10.4} \t W:{2:6.4} \t
b:{3:6.4}".
                  format(epoch+1, c, sess.run(W), sess.run(b)))
    # Print final parameter values
   print("Optimization Finished!")
   training cost = sess.run(cost, feed dict={X: train X, y:
train y})
   print("Final training cost:", training cost, "W:",
sess.run(W), "b:",
          sess.run(b), '\n')
    # Graphic display
   plt.plot(train X, train y, 'ro', label='Original data')
   plt.plot(train X, sess.run(W) * train X + sess.run(b),
label='Fitted line')
   plt.legend()
   plt.show()
    # Testing the model
   testing cost = sess.run(tf.reduce sum(tf.square(linear model
- y)) / (2 * test X.shape[0]),
                            feed dict={X: test X, y: test y})
```

```
print("Final testing cost:", testing_cost)
  print("Absolute mean square loss difference:",
abs(training_cost - testing_cost))

# Display fitted line on test data
  plt.plot(test_X, test_y, 'bo', label='Testing data')
  plt.plot(train_X, sess.run(W) * train_X + sess.run(b),
label='Fitted line')
  plt.legend()
  plt.show()
```

3.6. Model Neuron Network

a. View Fig1.1 in the diagram. Code and explain

Demonstration of Activation Function

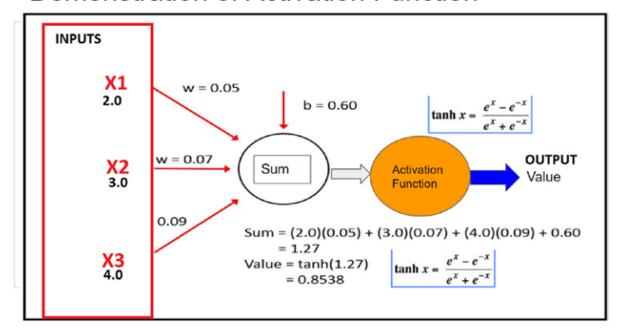
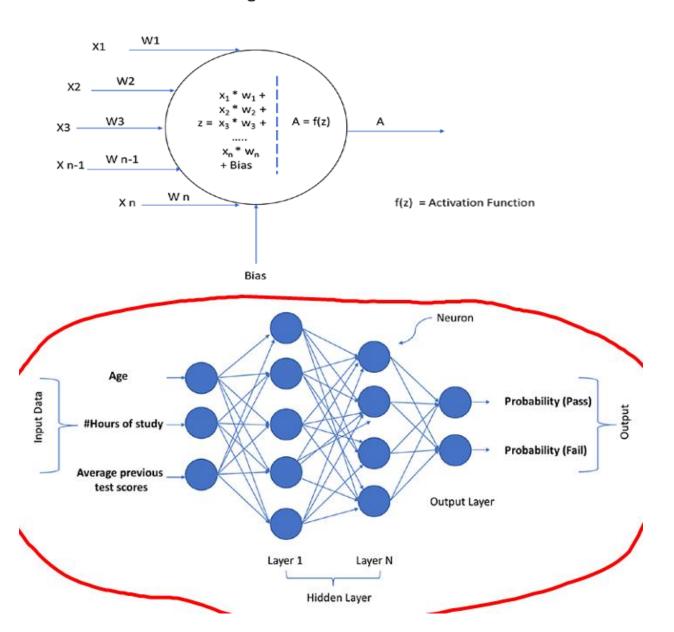


Figure 1-1. An activation function

- b. Design a simple neuron network that s given in the digram. Student replaces tanh various sigmoid, relu... Visualize all versions and explain. Copy code and images of running program
- 3.7 Desing code with tensorflow for the network given in the following Figure. Copy code and running image. Explain code.

A Single Neuron



- 3.8 Refer to textbook 2 and Explain with simple examples: model, layer, batch, epoch, structure, loss, optimization, gradient. Explain Figures 2.21, 2.22, 2.23, 2.24, 2.25. Explain your understand in section 3.6
- 3.9 Running code processing MNIST with keras and WITHOUT keras (pure tensorflow). Refer Chapter 2 and 4
- 3.10 Present your understand of Chap 4. Running code with datasets MNIST and IMDB. Explain.

TensorFlow operator	Shortcut	Description
tf.add()	a + b	Adds a and b, element-wise.
<pre>tf.multiply()</pre>	a * b	Multiplies a and b, element-wise.
tf.subtract()	a - b	Subtracts a from b, element-wise.
tf.divide()	a / b	Computes Python-style division of a by b.
tf.pow()	a ** b	Returns the result of raising each element in a to its corresponding element b, element-wise.
tf.mod()	a % b	Returns the element-wise modulo.
<pre>tf.logical_and()</pre>	a & b	Returns the truth table of a & b, element-wise. dtype must be tf.bool.
tf.greater()	a > b	Returns the truth table of a > b, element-wise.
tf.greater_equal()	a >= b	Returns the truth table of a >= b, element-wise.
<pre>tf.less_equal()</pre>	a <= b	Returns the truth table of a <= b, element-wise.
tf.less()	a < b	Returns the truth table of a < b, element-wise.
<pre>tf.negative()</pre>	-a	Returns the negative value of each element in a.
tf.logical_not()	~a	Returns the logical NOT of each element in a. Only compatible with Tensor objects with dtype of tf.bool.
tf.abs()	abs(a)	Returns the absolute value of each element in a.
<pre>tf.logical_or()</pre>	a b	Returns the truth table of a b, element-wise. dtype must be tf.bool.

Data type	Python type	Description
DT_FLOAT	tf.float32	32-bit floating point.
DT_DOUBLE	tf.float64	64-bit floating point.
DT_INT8	tf.int8	8-bit signed integer.
DT_INT16	tf.int16	16-bit signed integer.
DT_INT32	tf.int32	32-bit signed integer.
DT_INT64	tf.int64	64-bit signed integer.
DT_UINT8	tf.uint8	8-bit unsigned integer.
DT_UINT16	tf.uint16	16-bit unsigned integer.
DT_STRING	tf.string	Variable-length byte array. Each element of a Tensor is a byte array.
DT_B00L	tf.bool	Boolean.
DT_COMPLEX64	tf.complex64	Complex number made of two 32-bit floating points: real and imaginary parts.
DT_COMPLEX128	tf.complex128	Complex number made of two 64-bit floating points: real and imaginary parts.
DT_QINT8	tf.qint8	8-bit signed integer used in quantized ops.
DT_QINT32	tf.qint32	32-bit signed integer used in quantized ops.
DT_QUINT8	tf.quint8	8-bit unsigned integer used in quantized ops.