

```
In [17]: import tensorflow.compat.v1 as tf
         tf.disable_v2_behavior()
```

```
In [18]: sess = tf.InteractiveSession()
```

```
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py:1761: UserWarning: An interactive session is already active. This can cause out-of-memory errors in some cases. You must explicitly call `InteractiveSession.close()` to release resources held by the other session(s).
  warnings.warn('An interactive session is already active. This can '
```

```
In [19]: import numpy as np
         import math
```

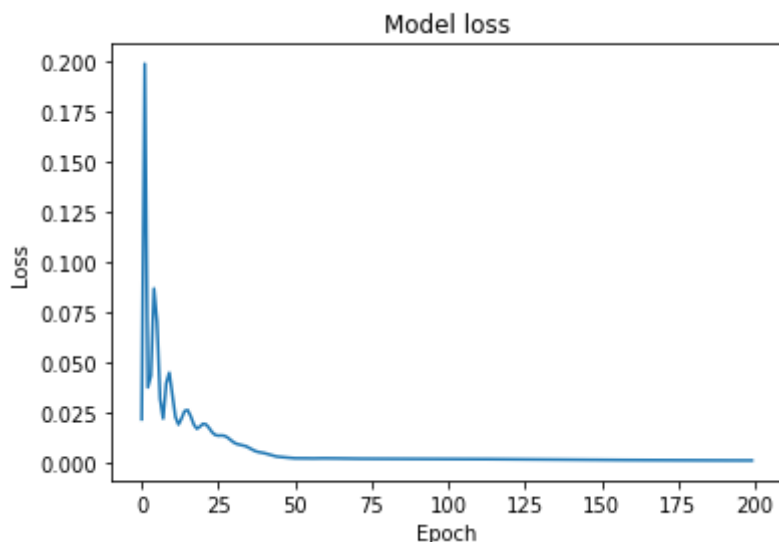
```
In [20]: X = np.expand_dims(np.arange(0.0, 5.0, 0.01), 1)
         Y = np.sinc(5*X)
         x = tf.placeholder(tf.float64, [500, 1], name='x')
         y = tf.placeholder(tf.float64, [500, 1], name='y')
```

insert the data type for your tensor model 1 is $200+200400+400200+200*200+200=200400$

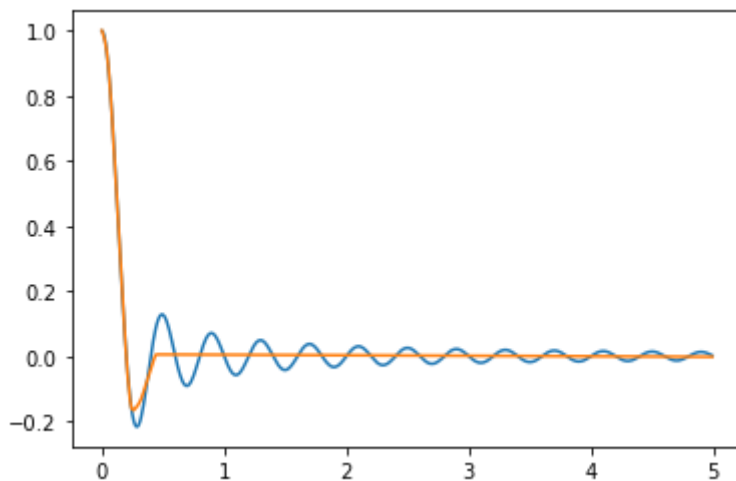
```
In [22]: input_layer = tf.layers.dense(x, 200, activation= tf.nn.relu)
         hidden_layer1 = tf.layers.dropout(input_layer, 0.2)
         hidden_layer2 = tf.layers.dense(hidden_layer1, 400, activation=tf.nn.relu)
         hidden_layer3 = tf.layers.dense(hidden_layer2, 200, activation=tf.nn.relu)
         output_layer = tf.layers.dense(hidden_layer3, 1)
         Loss = tf.losses.mean_squared_error(y, output_layer)
         Optimizer = tf.train.AdamOptimizer(learning_rate= 0.001).minimize(Loss)
         init = tf.global_variables_initializer()
```

```
In [23]: loss_list=[]
         sess.run(init)
         for i in range(0, 200):
             fd = {x:X, y:Y}
             _, loss_val = sess.run([Optimizer, Loss], feed_dict=fd)
             #print ('loss = %s' % loss_val)
             loss_list.append(loss_val)
         YP = sess.run(output_layer, feed_dict={x:X})
```

```
In [24]: # Plot training loss values
         import matplotlib.pyplot as plt
         plt.plot(loss_list)
         plt.title('Model loss')
         plt.ylabel('Loss')
         plt.xlabel('Epoch')
         plt.show()
```



```
In [25]: plt.plot(X,Y)
plt.plot(X,YP)
plt.show()
```



insert the data type for your tensor model 1 is $200+200500+500200+200=200400$ variables

```
In [26]: input_layer_m2 = tf.layers.dense(x, 200, activation= tf.nn.relu)
hidden_layer1_m2 = tf.layers.dropout(input_layer_m2,0.2)
hidden_layer2_m2 = tf.layers.dense(hidden_layer1_m2,500,activation=tf.nn.relu)

output_layer_m2 = tf.layers.dense(hidden_layer2_m2,1)
Loss_m2 =tf.losses.mean_squared_error(y , output_layer_m2)
Optimizer_m2 = tf.train.AdamOptimizer(learning_rate= 0.001).minimize(Loss_m2)
init_m2 = tf.global_variables_initializer()
```

/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.

warnings.warn("`tf.layers.dense` is deprecated and "

/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be removed in a future version. Please use `layer.__call__` method instead.

warnings.warn("`layer.apply` is deprecated and "

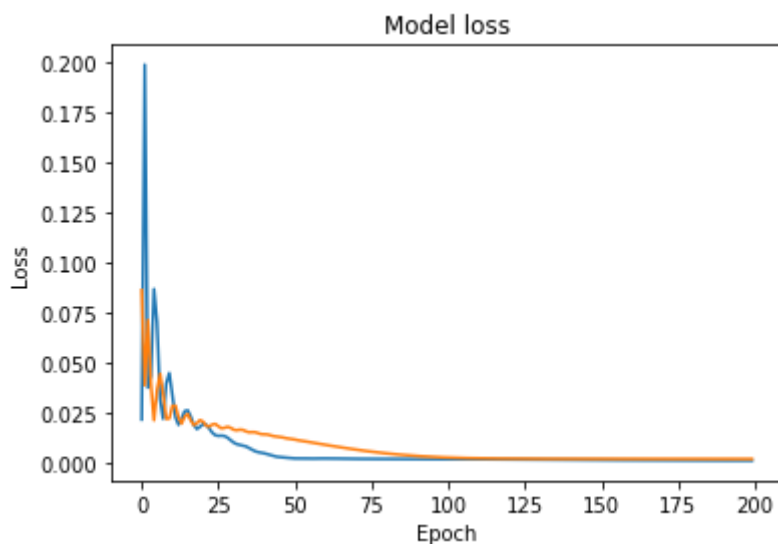
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:268: UserWarning: `tf.layers.dropout` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dropout` instead.

warnings.warn("`tf.layers.dropout` is deprecated and "

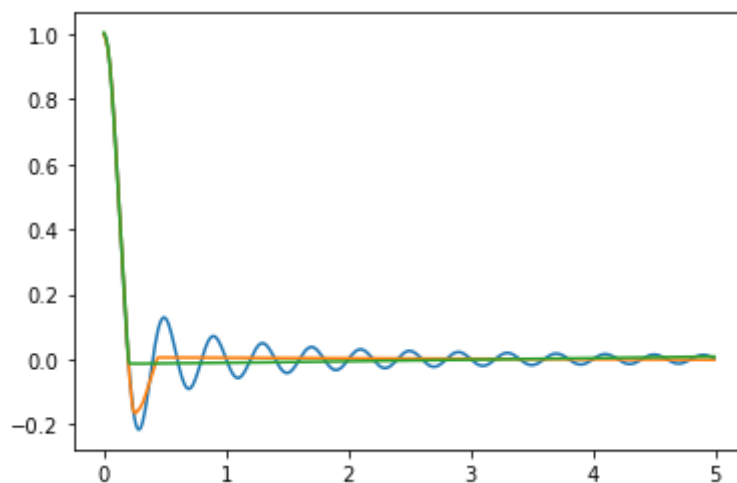
```
In [27]: loss_list_m2=[]
sess.run(init_m2)
for i in range(0,200):
    fd = {x:X, y:Y}
    _, loss_val_m2 = sess.run([Optimizer_m2, Loss_m2], feed_dict=fd)
    #print ('loss = %s' % loss_val)
    loss_list_m2.append(loss_val_m2)

YP_m2 = sess.run(output_layer_m2,feed_dict={x:X})
```

```
In [28]: # Plot training loss values
import matplotlib.pyplot as plt
plt.plot(loss_list)
plt.plot(loss_list_m2)
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.show()
```



```
In [29]: plt.plot(X,Y)
plt.plot(X,YP)
plt.plot(X,YP_m2)
plt.show()
```



Describe the models you use, including the number of parameters (at least two models) and the

function you use.

In one graph, plot the predicted function curve of all models and the ground-truth function curve.

In []:

```
In [2]: %matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
import numpy as np
import math
import absl
import logging
logger = tf.get_logger()
logger.setLevel(logging.ERROR)
```

WARNING:tensorflow:From /Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

Instructions for updating:

non-resource variables are not supported in the long term

```
In [3]: tf.__version__
```

```
Out[3]: '2.4.1'
```

```
In [4]: from tensorflow.examples.tutorials.mnist import input_data
#data = input_data.read_data_sets('data/MNIST/', one_hot=True)
mnist = input_data.read_data_sets('./mnist', one_hot=True) # they has been norm
test_x = mnist.test.images[:10000]
test_y = mnist.test.labels[:10000]
```

Extracting ./mnist/train-images-idx3-ubyte.gz

Extracting ./mnist/train-labels-idx1-ubyte.gz

Extracting ./mnist/t10k-images-idx3-ubyte.gz

Extracting ./mnist/t10k-labels-idx1-ubyte.gz

```
In [5]: print("Size of:")
print("- Training-set:\t\t{}".format(len(mnist.train.labels)))
print("- Test-set:\t\t{}".format(len(mnist.test.labels)))
print("- Validation-set:\t\t{}".format(len(mnist.validation.labels)))
```

Size of:

- Training-set: 55000

- Test-set: 10000

- Validation-set: 5000

```
In [6]: mnist.test.cls = np.argmax(mnist.test.labels, axis=1)
```

```
In [7]: # We know that MNIST images are 28 pixels in each dimension.
img_size = 28

# Images are stored in one-dimensional arrays of this length.
img_size_flat = img_size * img_size

# Tuple with height and width of images used to reshape arrays.
img_shape = (img_size, img_size)

# Number of colour channels for the images: 1 channel for gray-scale.
num_channels = 1

# Number of classes, one class for each of 10 digits.
num_classes = 10
```

```
In [8]: def plot_images(images, cls_true, cls_pred=None):
    assert len(images) == len(cls_true) == 9

    # Create figure with 3x3 sub-plots.
    fig, axes = plt.subplots(3, 3)
    fig.subplots_adjust(hspace=0.3, wspace=0.3)

    for i, ax in enumerate(axes.flat):
        # Plot image.
        ax.imshow(images[i].reshape(img_shape), cmap='binary')

        # Show true and predicted classes.
        if cls_pred is None:
            xlabel = "True: {0}".format(cls_true[i])
        else:
            xlabel = "True: {0}, Pred: {1}".format(cls_true[i], cls_pred[i])

        # Show the classes as the label on the x-axis.
        ax.set_xlabel(xlabel)

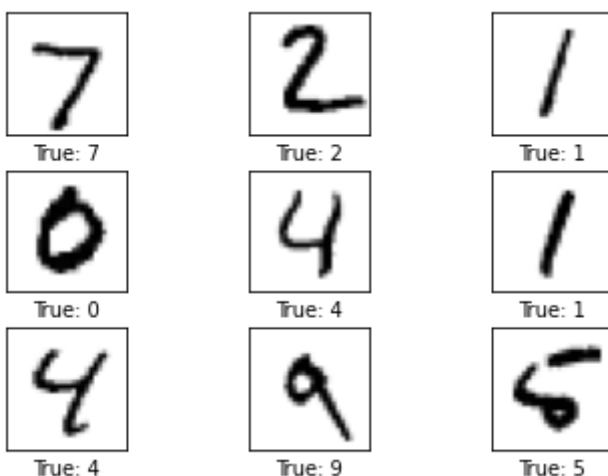
        # Remove ticks from the plot.
        ax.set_xticks([])
        ax.set_yticks([])

    # Ensure the plot is shown correctly with multiple plots
    # in a single Notebook cell.
    plt.show()
```

```
In [9]: # Get the first images from the test-set.
images = mnist.test.images[0:9]

# Get the true classes for those images.
cls_true = mnist.test.cls[0:9]

# Plot the images and labels using our helper-function above.
plot_images(images=images, cls_true=cls_true)
```



```
In [10]: x = tf.placeholder(tf.float32, shape=[None, img_size_flat], name='x') / 255
```

```
In [11]: x_image = tf.reshape(x, [-1, img_size, img_size, num_channels])
```

```
In [12]: y_true = tf.placeholder(tf.float32, shape=[None, num_classes], name='y_true')
```

```
In [13]: y_true_cls = tf.argmax(y_true, dimension=1)
```

```
In [14]: conv1 = tf.layers.conv2d(    # shape (28, 28, 1)
        inputs=x_image,
        filters=16,
        kernel_size=5,
        strides=1,
        padding='same',
        activation=tf.nn.relu
    )    # -> (28, 28, 16)
pool1 = tf.layers.max_pooling2d(
    conv1,
    pool_size=2,
    strides=2,
)    # -> (14, 14, 16)
conv2 = tf.layers.conv2d(pool1, 36, 5, 1, 'same', activation=tf.nn.relu)    # ->
pool2 = tf.layers.max_pooling2d(conv2, 2, 2)    # -> (7, 7, 36)
flat = tf.reshape(pool2, [-1, 7*7*36])    # -> (7*7*36, )
output = tf.layers.dense(flat, 10)    # output layer

loss = tf.losses.softmax_cross_entropy(onehot_labels=y_true, logits=output)
train_op = tf.train.AdamOptimizer(0.001).minimize(loss) #learning rate is 0.1

accuracy = tf.metrics.accuracy(    # return (acc, update_op), and create 2
    labels=tf.argmax(y_true, axis=1), predictions=tf.argmax(output, axis=1),)[1]
```

/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/convolutional.py:414: UserWarning: `tf.layers.conv2d` is deprecated and will be removed in a future version. Please Use `tf.keras.layers.Conv2D` instead.

warnings.warn("`tf.layers.conv2d` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be removed in a future version. Please use `layer.__call__` method instead.

warnings.warn("`layer.apply` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/pooling.py:310: UserWarning: `tf.layers.max_pooling2d` is deprecated and will be removed in a future version. Please use `tf.keras.layers.MaxPoolin
g2D` instead.

warnings.warn("`tf.layers.max_pooling2d` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.

warnings.warn("`tf.layers.dense` is deprecated and '

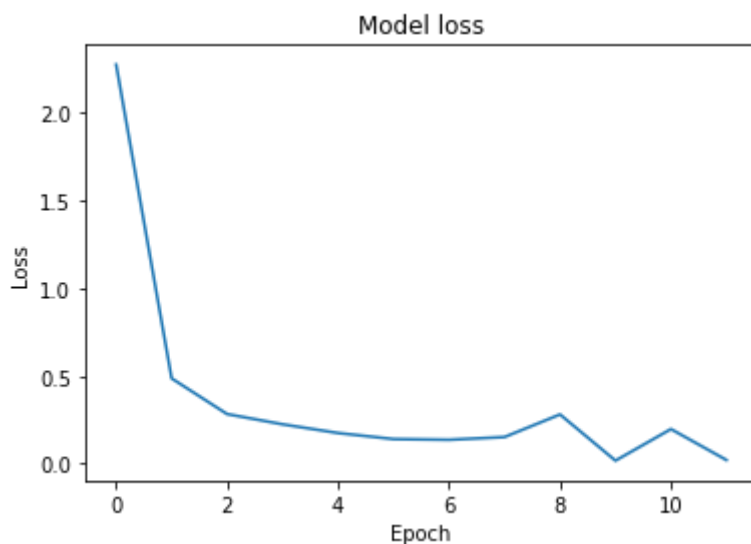
```
In [15]: session=tf.Session()
```

```
In [16]: init_op = tf.group(tf.global_variables_initializer(), tf.local_variables_initializer())
        session.run(init_op)    # initialize var in graph
```

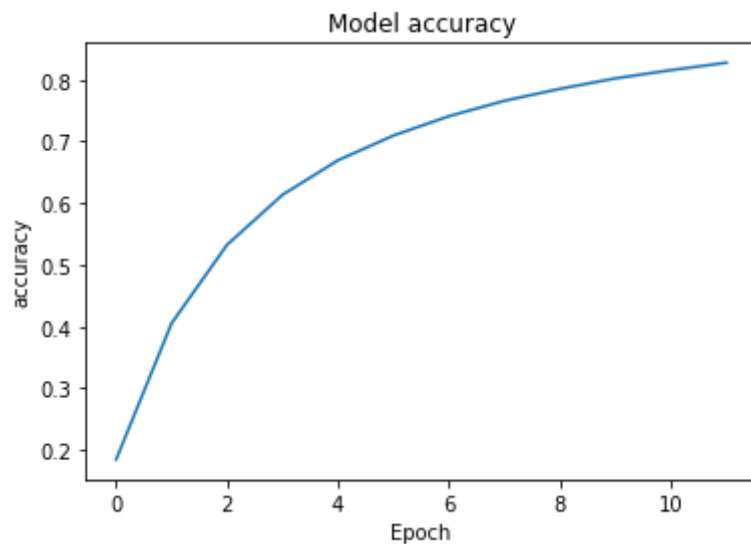
```
In [18]: loss_list=[]
        accuracy_list=[]
        for step in range(600):
            b_x, b_y = mnist.train.next_batch(50) #Batch_size is 50
            _, loss_ = session.run([train_op, loss], {x: b_x, y_true: b_y})
            if step % 50 == 0:
                accuracy_, flat_representation = session.run([accuracy, flat], {x: test_
                print('Step:', step, '| train loss: %.4f' % loss_, '| test accuracy: %.2
                loss_list.append(loss_)
                accuracy_list.append(accuracy_)
```

Step: 0	train loss: 2.2764	test accuracy: 0.18
Step: 50	train loss: 0.4868	test accuracy: 0.41
Step: 100	train loss: 0.2832	test accuracy: 0.53
Step: 150	train loss: 0.2249	test accuracy: 0.61
Step: 200	train loss: 0.1751	test accuracy: 0.67
Step: 250	train loss: 0.1406	test accuracy: 0.71
Step: 300	train loss: 0.1367	test accuracy: 0.74
Step: 350	train loss: 0.1521	test accuracy: 0.77
Step: 400	train loss: 0.2811	test accuracy: 0.79
Step: 450	train loss: 0.0176	test accuracy: 0.80
Step: 500	train loss: 0.1976	test accuracy: 0.82
Step: 550	train loss: 0.0212	test accuracy: 0.83

```
In [19]: plt.plot(loss_list)
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.show()
```



```
In [20]: plt.plot(accuracy_list)
plt.title('Model accuracy')
plt.ylabel('accuracy')
plt.xlabel('Epoch')
plt.show()
```




```
In [21]: test_output = session.run(output, {x: test_x[:10]})
pred_y = np.argmax(test_output, 1)
print(pred_y, 'prediction number')
print(np.argmax(test_y[:10], 1), 'real number')

[7 2 1 0 4 1 4 9 5 9] prediction number
[7 2 1 0 4 1 4 9 5 9] real number
```

```
In [22]: session.close()
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
import numpy as np
import math
import absl
import logging
logger = tf.get_logger()
logger.setLevel(logging.ERROR)
```

WARNING:tensorflow:From /Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

Instructions for updating:

non-resource variables are not supported in the long term

```
In [2]: from tensorflow.examples.tutorials.mnist import input_data
#data = input_data.read_data_sets('data/MNIST/', one_hot=True)
mnist = input_data.read_data_sets('./mnist', one_hot=True) # they has been norm
test_x = mnist.test.images[:10000]
test_y = mnist.test.labels[:10000]
```

Extracting ./mnist/train-images-idx3-ubyte.gz

Extracting ./mnist/train-labels-idx1-ubyte.gz

Extracting ./mnist/t10k-images-idx3-ubyte.gz

Extracting ./mnist/t10k-labels-idx1-ubyte.gz

```
In [3]: mnist.test.cls = np.argmax(mnist.test.labels, axis=1)
```

```
In [4]: # We know that MNIST images are 28 pixels in each dimension.
img_size = 28

# Images are stored in one-dimensional arrays of this length.
img_size_flat = img_size * img_size

# Tuple with height and width of images used to reshape arrays.
img_shape = (img_size, img_size)

# Number of colour channels for the images: 1 channel for gray-scale.
num_channels = 1

# Number of classes, one class for each of 10 digits.
num_classes = 10
```

```
In [5]: x = tf.placeholder(tf.float32, shape=[None, img_size_flat], name='x') / 255
x_image = tf.reshape(x, [-1, img_size, img_size, num_channels])
y_true = tf.placeholder(tf.float32, shape=[None, num_classes], name='y_true')
y_true_cls = tf.argmax(y_true, dimension=1)
```

create the CNN

```
In [6]: net = tf.layers.conv2d(inputs=x_image, name='layer_conv1', padding='same',
                             filters=16, kernel_size=5, activation=tf.nn.relu)
net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)

# layer_conv2
net = tf.layers.conv2d(inputs=net, name='layer_conv2', padding='same',
```

```

        filters=36, kernel_size=5, activation=tf.nn.relu)
net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)

print(net)
net = tf.layers.flatten(net)

print(net)
net = tf.layers.dense(inputs=net, name='layer_fc1',
                      units=128, activation=tf.nn.relu)
logits = tf.layers.dense(inputs=net, name='layer_fc_out',
                        units=num_classes, activation=None)
print(logits)

```

```

Tensor("max_pooling2d_1/MaxPool:0", shape=(?, 7, 7, 36), dtype=float32)
Tensor("flatten/Reshape:0", shape=(?, 1764), dtype=float32)
Tensor("layer_fc_out/BiasAdd:0", shape=(?, 10), dtype=float32)
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/convolutional.py:414: UserWarning: `tf.layers.conv2d` is deprecated and will be removed in a future version. Please Use `tf.keras.layers.Conv2D` instead.
  warnings.warn("`tf.layers.conv2d` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be removed in a future version. Please use `layer.__call__` method instead.
  warnings.warn("`layer.apply` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/pooling.py:310: UserWarning: `tf.layers.max_pooling2d` is deprecated and will be removed in a future version. Please use `tf.keras.layers.MaxPooling2D` instead.
  warnings.warn("`tf.layers.max_pooling2d` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:329: UserWarning: `tf.layers.flatten` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Flatten` instead.
  warnings.warn("`tf.layers.flatten` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.
  warnings.warn("`tf.layers.dense` is deprecated and ")

```

```

In [7]: y_pred = tf.nn.softmax(logits=logits)
        y_pred_cls = tf.argmax(y_pred, dimension=1)

```

```

In [8]: cross_entropy = tf.nn.softmax_cross_entropy_with_logits(labels=y_true, logits=logits)
        loss = tf.reduce_mean(cross_entropy)

```

Optimization

```

In [9]: opt = tf.train.AdamOptimizer(learning_rate=1e-4)
        optimizer = opt.minimize(loss)

```

Classification Accuracy

```

In [10]: correct_prediction = tf.equal(y_pred_cls, y_true_cls)
         accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))

```

Getting the weight

```

In [12]: trainable_var_list = tf.trainable_variables()
        def get_weights_variable(layer_name):
            with tf.variable_scope(layer_name, reuse=True):
                variable = tf.get_variable('kernel')

```

```
return variable
```

```
weights_conv1 = get_weights_variable(layer_name='layer_conv1')
weights_conv2 = get_weights_variable(layer_name='layer_conv2')
print(weights_conv1)
print(weights_conv2)

weights_fc1 = get_weights_variable(layer_name='layer_fc1')
weights_fc_out = get_weights_variable(layer_name='layer_fc_out')
print(weights_fc1)
print(weights_fc_out)
```

```
<tf.Variable 'layer_conv1/kernel:0' shape=(5, 5, 1, 16) dtype=float32_ref>
<tf.Variable 'layer_conv2/kernel:0' shape=(5, 5, 16, 36) dtype=float32_ref>
<tf.Variable 'layer_fc1/kernel:0' shape=(1764, 128) dtype=float32_ref>
<tf.Variable 'layer_fc_out/kernel:0' shape=(128, 10) dtype=float32_ref>
```

start one session, and initialize the variables

```
In [13]: session = tf.Session()
        session.run(tf.global_variables_initializer())
```

get gradient during training; compute hessian

```
In [14]: #grad = opt.compute_gradients(loss, weights_fc_out)[0]
        grads = tf.gradients(loss, weights_fc_out)[0]
        print(grads)
        hessian = tf.reduce_sum(tf.hessians(loss, weights_fc_out)[0], axis = 2)
        print(hessian)
        train_batch_size = 64
```

```
Tensor("gradients_1/layer_fc_out/MatMul_grad/MatMul_1:0", shape=(128, 10), dtype=
=float32)
Tensor("Sum:0", shape=(128, 10, 10), dtype=float32)
```

start the optimization

```
In [34]: total_iterations = 0
        def optimize(num_iterations):
            global total_iterations

            for i in range(total_iterations,
                           total_iterations + num_iterations):

                x_batch, y_true_batch = mnist.train.next_batch(train_batch_size)

                feed_dict_train = {x: x_batch,
                                   y_true: y_true_batch}

                session.run(optimizer, feed_dict=feed_dict_train)
                if i % 10 == 0:
                    # Calculate the accuracy on the training-set.
                    los, acc = session.run([loss, accuracy], feed_dict=feed_dict_train)

                    grads_vals, hess_vals = session.run([grads, hessian], feed_dict=feed
                    print(grads_vals.shape)
                    print(hess_vals.shape)

                    # Message for printing.
                    msg = "Iteration: {0:>6}, Training Loss: {1:>1.6}, Training Accuracy
```

```

        # Print it.
        print(msg.format(i + 1, los, acc))

    # Update the total number of iterations performed.
    total_iterations += num_iterations

```

```

In [18]: optimize(num_iterations=99)

(128, 10)
(128, 10, 10)
Iteration:      1, Training Loss: 2.29611, Training Accuracy:  20.3%
(128, 10)
(128, 10, 10)
Iteration:     11, Training Loss: 2.22555, Training Accuracy:  29.7%
(128, 10)
(128, 10, 10)
Iteration:     21, Training Loss: 2.17334, Training Accuracy:  29.7%
(128, 10)
(128, 10, 10)
Iteration:     31, Training Loss: 2.16627, Training Accuracy:  43.8%
(128, 10)
(128, 10, 10)
Iteration:     41, Training Loss: 2.03829, Training Accuracy:  68.8%
(128, 10)
(128, 10, 10)
Iteration:     51, Training Loss: 1.99218, Training Accuracy:  64.1%
(128, 10)
(128, 10, 10)
Iteration:     61, Training Loss: 1.75646, Training Accuracy:  73.4%
(128, 10)
(128, 10, 10)
Iteration:     71, Training Loss: 1.53628, Training Accuracy:  84.4%
(128, 10)
(128, 10, 10)
Iteration:     81, Training Loss: 1.32736, Training Accuracy:  87.5%
(128, 10)
(128, 10, 10)
Iteration:     91, Training Loss: 1.35038, Training Accuracy:  67.2%

```

```

In [27]: def plot_conv_weights(weights, input_channel=0):

    w = session.run(weights)

    w_min = np.min(w)
    w_max = np.max(w)

    num_filters = w.shape[3]

    num_grids = math.ceil(math.sqrt(num_filters))

    fig, axes = plt.subplots(num_grids, num_grids)

    for i, ax in enumerate(axes.flat):
        if i < num_filters:

            img = w[:, :, input_channel, i]

            ax.imshow(img, vmin=w_min, vmax=w_max,
                      interpolation='nearest', cmap='seismic')

            ax.set_xticks([])
            ax.set_yticks([])

```

```
plt.show()
```

```
In [28]: from sklearn.decomposition import PCA
def plot_fc_weights(weights_list):

    w_list = session.run(weights_list)

    pca = PCA(n_components=2)

    fig = plt.figure(figsize = (8,8))
    ax = fig.add_subplot(1,1,1)
    ax.set_xlabel('Principal Component 1', fontsize = 15)
    ax.set_ylabel('Principal Component 2', fontsize = 15)
    ax.set_title('2 component PCA', fontsize = 20)

    for w in w_list:

        print(w.shape)

        principalComponents = pca.fit_transform(w)

        ax.scatter(principalComponents[:,0], principalComponents[:,1], label=w.s

    ax.legend()
    plt.show()
```

```
In [39]: weights_list=[]
for step in range(100):
    if step % 3 == 0:
        weights=session.run(trainable_var_list)
        weights=PCA(weights)
        weights_list.append(weights)
```

```
In [40]: print(weights_list)

[PCA(n_components=[array([[[-7.02866837e-02,  1.08631827e-01,  4.99007441e-02,
-4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
-0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671]),
      dtype=float32))], PCA(n_components=[array([[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
-4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],
```

```

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
 0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
 -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
 0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
 -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
 0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
 -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
 0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
 -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
 4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
 6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
 1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
 9.53982845e-02]]],

```

```

array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

      [[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
      -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
      -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
      -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
      -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

      [[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
      -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
      -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
      -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
      -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

      [[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
      -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
      -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
      -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
      -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

```



```

1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
 -0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
 -0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
 -0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

```

```

-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

```

4.86401096e021.26112536e013.08717322e02
4.23055924e025.49924783e028.90975073e02
6.02630228e029.03390869e021.13589466e01
1.15379550e011.19940154e011.08178295e01
9.53982845e02

1.21127687e019.15692896e022.99458336e02
7.25981370e028.97659659...
0.178886530.137434140.0086686...0.16559619
0.163273410.07722243
0.169259220.131860210.05882641...0.06325659
0.0978770.01333893
0.105820940.092717860.00425618...0.17365101
0.133021860.01772023dt eload32
arra0.000257080.005007240.00162470.002117080.
00095305
0.000592950.002238830.001162890.002907410.00169671
dt eload32 nom onentsarra7.02866837e021.08631827
e014.99007441e02
4.86401096e021.26112536e013.08717322e02
4.23055924e025.49924783e028.90975073e02
6.02630228e029.03390869e021.13589466e01
1.15379550e011.19940154e011.08178295e01
9.53982845e02

1.21127687e019.15692896e022.99458336e02
7.25981370e028.97659659...
0.178886530.137434140.0086686...0.16559619
0.163273410.07722243
0.169259220.131860210.05882641...0.06325659
0.0978770.01333893
0.105820940.092717860.00425618...0.17365101
0.133021860.01772023dt eload32
arra0.000257080.005007240.00162470.002117080.
00095305
0.000592950.002238830.001162890.002907410.00169671
dt eload32 nom onentsarra7.02866837e021.08631827
e014.99007441e02
4.86401096e021.26112536e013.08717322e02
4.23055924e025.49924783e028.90975073e02
6.02630228e029.03390869e021.13589466e01
1.15379550e011.19940154e011.08178295e01
9.53982845e02

1.21127687e019.15692896e022.99458336e02
7.25981370e028.97659659...
0.178886530.137434140.0086686...0.16559619
0.163273410.07722243
0.169259220.131860210.05882641...0.06325659
0.0978770.01333893
0.105820940.092717860.00425618...0.17365101
0.133021860.01772023dt eload32
arra0.000257080.005007240.00162470.002117080.
00095305
0.000592950.002238830.001162890.002907410.00169671
dt eload32 nom onentsarra7.02866837e021.08631827
e014.99007441e02
4.86401096e021.26112536e013.08717322e02
4.23055924e025.49924783e028.90975073e02
6.02630228e029.03390869e021.13589466e01
1.15379550e011.19940154e011.08178295e01
9.53982845e02

1.21127687e019.15692896e022.99458336e02
7.25981370e028.97659659...

00095305,

```

0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
-7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
-4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
9.53982845e-02]]],

```

```

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

[[ 1.21127687e-01,  9.15692896e-02,  2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653,  0.13743414,  0.0086686 , ...,  0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922,  0.13186021,  0.05882641, ..., -0.06325659,
 -0.097877 ,  0.01333893],
[-0.10582094,  0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186,  0.01772023]], dtype=float32),
      array([-0.00025708,  0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
      0.00059295, -0.00223883,  0.00116289, -0.00290741,  0.00169671],
      dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02,  1.08631827
e-01,  4.99007441e-02,
      -4.86401096e-02,  1.26112536e-01, -3.08717322e-02,
      4.23055924e-02, -5.49924783e-02,  8.90975073e-02,
      6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
      1.15379550e-01,  1.19940154e-01,  1.08178295e-01,
      9.53982845e-02]]],

```

```

-0.13302186, 0.01772023]], dtype=float32),
    array([-0.00025708, 0.00500724, -0.0016247, -0.00211708, -0.
00095305,
    0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
    dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
    -4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
    4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
    6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
    1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
    9.53982845e-02]],

    [[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
    -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686, ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877, 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
    array([-0.00025708, 0.00500724, -0.0016247, -0.00211708, -0.
00095305,
    0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
    dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
    -4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
    4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
    6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
    1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
    9.53982845e-02]],

    [[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
    -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686, ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877, 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
    array([-0.00025708, 0.00500724, -0.0016247, -0.00211708, -0.
00095305,
    0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
    dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
    -4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
    4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
    6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
    1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
    9.53982845e-02]],

    [[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
    -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686, ..., 0.16559619,
-0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
-0.097877, 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
-0.13302186, 0.01772023]], dtype=float32),
    array([-0.00025708, 0.00500724, -0.0016247, -0.00211708, -0.
00095305,
    0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
    dtype=float32))), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
    -4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
    4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
    6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
    1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
    9.53982845e-02]]],

```

```

        6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
        1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
        9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
 -0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186, 0.01772023]], dtype=float32),
        array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
        0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
        dtype=float32)), PCA(n_components=[array([[[[-7.02866837e-02, 1.08631827
e-01, 4.99007441e-02,
        -4.86401096e-02, 1.26112536e-01, -3.08717322e-02,
        4.23055924e-02, -5.49924783e-02, 8.90975073e-02,
        6.02630228e-02, -9.03390869e-02, -1.13589466e-01,
        1.15379550e-01, 1.19940154e-01, 1.08178295e-01,
        9.53982845e-02]],

[[ 1.21127687e-01, 9.15692896e-02, 2.99458336e-02,
   -7.25981370e-02, -8.97659659...
[-0.17888653, 0.13743414, 0.0086686 , ..., 0.16559619,
 -0.16327341, -0.07722243],
[ 0.16925922, 0.13186021, 0.05882641, ..., -0.06325659,
 -0.097877 , 0.01333893],
[-0.10582094, 0.09271786, -0.00425618, ..., -0.17365101,
 -0.13302186, 0.01772023]], dtype=float32),
        array([-0.00025708, 0.00500724, -0.0016247 , -0.00211708, -0.
00095305,
        0.00059295, -0.00223883, 0.00116289, -0.00290741, 0.00169671],
        dtype=float32)))]

```

In []:

Github see: <https://github.com/Gloriabhsfer/DeeplearningCPS843>


```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
import numpy as np
import math
import abs1
import logging
logger = tf.get_logger()
logger.setLevel(logging.ERROR)
```

WARNING:tensorflow:From /Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

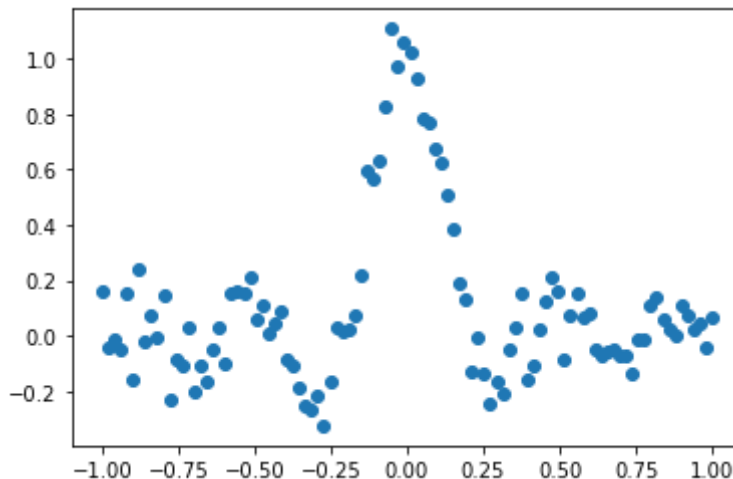
Instructions for updating:

non-resource variables are not supported in the long term

```
In [2]: tf.set_random_seed(1)
np.random.seed(1)
```

```
In [3]: # fake data
x = np.linspace(-1, 1, 100)[: , np.newaxis]           # shape (100, 1)
noise = np.random.normal(0, 0.1, size=x.shape)
y = np.sinc(5*x) + noise                               # shape (100, 1) + some noise
```

```
In [4]: plt.scatter(x, y)
plt.show()
```



```
In [5]: input_x = tf.placeholder(tf.float32, [None, 1])    # input x
#input_x = tf.placeholder(tf.float32, [100, 1])
output_y = tf.placeholder(tf.float32, [None, 1])         # input y
#output_y = tf.placeholder(tf.float32, [100, 1])
```

```
In [6]: # neural network layers
h1 = tf.layers.dense(inputs=input_x, units=10, activation=tf.nn.relu, name='h1')
h2 = tf.layers.dense(inputs=h1, units=10, activation=tf.nn.relu, name='h2')
output = tf.layers.dense(inputs=h2, units=1, name='output')
```

/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.

```
warnings.warn(`tf.layers.dense` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engi
ne/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be r
emoved in a future version. Please use `layer.__call__` method instead.
warnings.warn(`layer.apply` is deprecated and '

```

```
In [7]: loss = tf.losses.mean_squared_error(output_y, output)    # compute cost
```

```
In [8]: optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.5)
train_op = optimizer.minimize(loss)
```

```
In [9]: def get_weights_variable(layer_name):
        with tf.variable_scope(layer_name, reuse=True):
            variable = tf.get_variable('kernel')
        return variable
```

```
In [10]: weights_h1 = get_weights_variable(layer_name='h1')
weights_h2 = get_weights_variable(layer_name='h2')
print(weights_h1)
print(weights_h2)

weights_out = get_weights_variable(layer_name='output')
print(weights_out)

<tf.Variable 'h1/kernel:0' shape=(1, 10) dtype=float32_ref>
<tf.Variable 'h2/kernel:0' shape=(10, 10) dtype=float32_ref>
<tf.Variable 'output/kernel:0' shape=(10, 1) dtype=float32_ref>
```

```
In [11]: session = tf.Session()
session.run(tf.global_variables_initializer())
```

```
In [12]: grads = tf.gradients(loss, weights_out)[0]
print(grads)

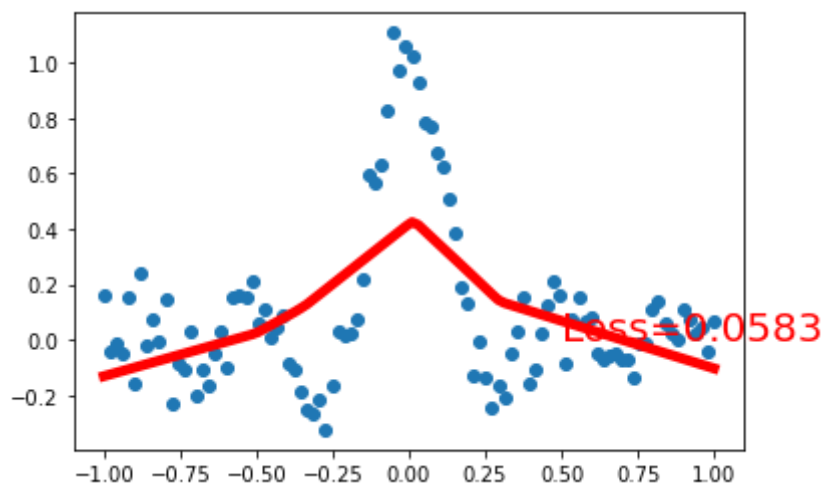
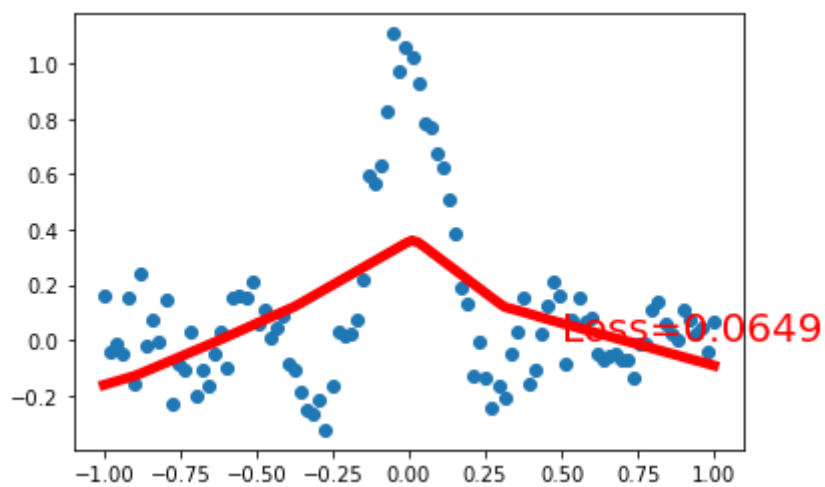
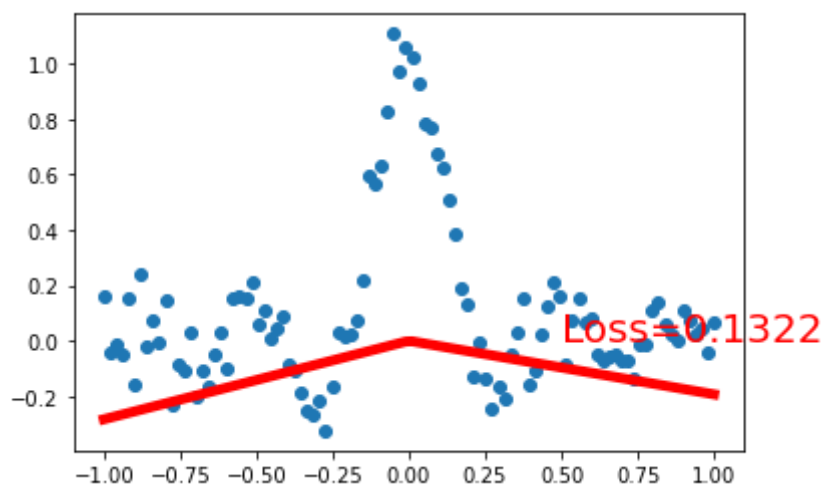
Tensor("gradients_1/output/MatMul_grad/MatMul_1:0", shape=(10, 1), dtype=float32)
```

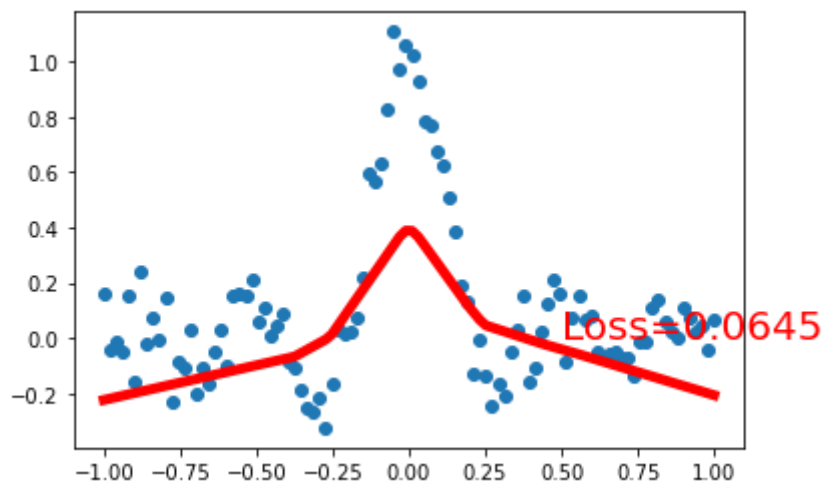
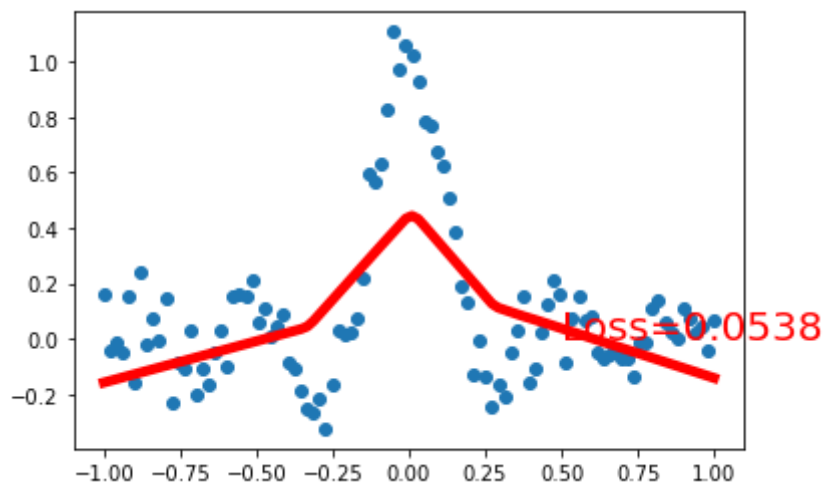
```
In [13]: hessian = tf.reduce_sum(tf.hessians(loss, weights_out)[0], axis = 2)
print(hessian)

Tensor("Sum:0", shape=(10, 1, 1), dtype=float32)
```

```
In [14]: train_op = optimizer.minimize(loss)
loss = tf.losses.mean_squared_error(output_y, output)
grad_list=[]
for iteration in range(100):
    # train and net output
    _, l, pred = session.run([train_op, loss, output], feed_dict={input_x: x, ou
grads_vals, hess_vals = session.run([grads, hessian], feed_dict={input_x: x,
grad_list.append(grads_vals)
    if iteration % 20 == 0:
        # plot and show learning process
        plt.cla()
        plt.scatter(x, y)
        plt.plot(x, pred, 'r-', lw=5)
        plt.text(0.5, 0, 'Loss=%.4f' % l, fontdict={'size': 20, 'color': 'red'})
        plt.pause(0.1)

plt.ioff()
plt.show()
```





```
In [15]: print(np.array(grad_list)[: ,1])
```

```
[[ 0.00000000e+00]
 [-7.8980060e-04]
 [-1.0695765e-03]
 [-4.8301509e-03]
 [-4.7677797e-03]
 [-9.3810521e-03]
 [-7.3561757e-03]
 [-1.2528453e-02]
 [-7.7160024e-03]
 [-1.4058841e-02]
 [-7.0987376e-03]
 [-1.4022825e-02]
 [-6.0905674e-03]
 [-1.4796937e-02]
 [-5.7100896e-03]
 [-1.4008171e-02]
 [-4.9671270e-03]
 [-1.3931324e-02]
 [-5.4148789e-03]
 [-1.3930539e-02]
 [-5.8548488e-03]
 [-1.3511472e-02]
 [-5.9597297e-03]
 [-1.4357337e-02]
 [-7.2026704e-03]
 [-1.4265257e-02]
 [-8.1384927e-03]
 [-1.3989154e-02]]
```

[-8.6637046e-03]
[-1.3730988e-02]
[-8.7621585e-03]
[-1.3933657e-02]
[-9.1672121e-03]
[-1.3594082e-02]
[-8.9389831e-03]
[-1.4210852e-02]
[-9.4519332e-03]
[-1.3906778e-02]
[-9.7884713e-03]
[-1.4076617e-02]
[-1.0410661e-02]
[-1.4455598e-02]
[-1.0436121e-02]
[-1.5210116e-02]
[-1.0600964e-02]
[-1.5431061e-02]
[-1.0328175e-02]
[-1.6042177e-02]
[-1.0618213e-02]
[-1.6053556e-02]
[-1.0719976e-02]
[-1.6368756e-02]
[-1.0473526e-02]
[-1.6791299e-02]
[-1.0394309e-02]
[-1.6872868e-02]
[-9.8758899e-03]
[-1.7407937e-02]
[-9.6798949e-03]
[-1.7772447e-02]
[-8.8173887e-03]
[-1.8435292e-02]
[-8.3800722e-03]
[-1.8802870e-02]
[-6.5830722e-03]
[-1.9527199e-02]
[-5.4314313e-03]
[-2.0260584e-02]
[-3.2129863e-03]
[-2.0847376e-02]
[-2.2784125e-03]
[-2.1440562e-02]
[3.4779657e-04]
[-2.1615600e-02]
[9.1947493e-04]
[-2.1912297e-02]
[1.3826289e-03]
[-2.2122620e-02]
[1.5806267e-03]
[-2.2686904e-02]
[1.2445450e-03]
[-2.2885662e-02]
[9.6987031e-04]
[-2.3141067e-02]
[6.2804917e-05]
[-2.3551786e-02]
[-5.3269105e-05]
[-2.3642881e-02]
[1.2487029e-04]
[-2.3946270e-02]
[-5.6667515e-04]
[-2.4217928e-02]
[4.6457927e-04]

```
[-2.4391733e-02]  
[ 1.3372998e-03]  
[-2.4842920e-02]  
[ 7.8088709e-04]  
[-2.4964593e-02]  
[ 1.9344395e-03]  
[-2.5296777e-02]]
```

```
In [16]: min(np.array(grad_list)[: ,1])
```

```
Out[16]: array([-0.02529678], dtype=float32)
```

```
In [ ]:
```

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
import numpy as np
import math
import absl
import logging
logger = tf.get_logger()
logger.setLevel(logging.ERROR)
```

WARNING:tensorflow:From /Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

Instructions for updating:

non-resource variables are not supported in the long term

```
In [2]: from tensorflow.examples.tutorials.mnist import input_data
#data = input_data.read_data_sets('data/MNIST/', one_hot=True)
mnist = input_data.read_data_sets('./mnist', one_hot=True) # they has been norm
test_x = mnist.test.images[:10000]
test_y = mnist.test.labels[:10000]
```

Extracting ./mnist/train-images-idx3-ubyte.gz

Extracting ./mnist/train-labels-idx1-ubyte.gz

Extracting ./mnist/t10k-images-idx3-ubyte.gz

Extracting ./mnist/t10k-labels-idx1-ubyte.gz

```
In [3]: mnist.test.cls = np.argmax(mnist.test.labels, axis=1)
```

```
In [4]: # We know that MNIST images are 28 pixels in each dimension.
img_size = 28

# Images are stored in one-dimensional arrays of this length.
img_size_flat = img_size * img_size

# Tuple with height and width of images used to reshape arrays.
img_shape = (img_size, img_size)

# Number of colour channels for the images: 1 channel for gray-scale.
num_channels = 1

# Number of classes, one class for each of 10 digits.
num_classes = 10
x = tf.placeholder(tf.float32, shape=[None, img_size_flat], name='x') / 255
x_image = tf.reshape(x, [-1, img_size, img_size, num_channels])
y_true = tf.placeholder(tf.float32, shape=[None, num_classes], name='y_true')
y_true_cls = tf.argmax(y_true, dimension=1)
```

```
In [5]: np.random.shuffle(mnist.train.labels)
```

```
In [6]: #network1
net = tf.layers.conv2d(inputs=x_image, name='layer_conv1_1', padding='same',
                       filters=16, kernel_size=5, activation=tf.nn.relu)
net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)

# layer_conv2
```

```

net = tf.layers.conv2d(inputs=net, name='layer_conv2_1', padding='same',
                        filters=36, kernel_size=5, activation=tf.nn.relu)
net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)

print(net)
net = tf.layers.flatten(net)

print(net)
net = tf.layers.dense(inputs=net, name='layer_fc1_1',
                       units=128, activation=tf.nn.relu)
logits = tf.layers.dense(inputs=net, name='layer_fc_out_1',
                           units=num_classes, activation=None)

print(logits)
y_pred = tf.nn.softmax(logits=logits)
y_pred_cls = tf.argmax(y_pred, dimension=1)
cross_entropy = tf.nn.softmax_cross_entropy_with_logits(labels=y_true, logits=logits)
loss = tf.reduce_mean(cross_entropy)

```

```

Tensor("max_pooling2d_1/MaxPool:0", shape=(?, 7, 7, 36), dtype=float32)
Tensor("flatten/Reshape:0", shape=(?, 1764), dtype=float32)
Tensor("layer_fc_out_1/BiasAdd:0", shape=(?, 10), dtype=float32)
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/convolutional.py:414: UserWarning: `tf.layers.conv2d` is deprecated and will be removed in a future version. Please Use `tf.keras.layers.Conv2D` instead.
  warnings.warn("`tf.layers.conv2d` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be removed in a future version. Please use `layer.__call__` method instead.
  warnings.warn("`layer.apply` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/pooling.py:310: UserWarning: `tf.layers.max_pooling2d` is deprecated and will be removed in a future version. Please use `tf.keras.layers.MaxPooling2D` instead.
  warnings.warn("`tf.layers.max_pooling2d` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:329: UserWarning: `tf.layers.flatten` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Flatten` instead.
  warnings.warn("`tf.layers.flatten` is deprecated and ")
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.
  warnings.warn("`tf.layers.dense` is deprecated and ")

```

```

In [7]: def get_weights_variable(layer_name):

        with tf.variable_scope(layer_name, reuse=True):
            variable = tf.get_variable('kernel')

        return variable

total_weights = []
weights_conv1 = get_weights_variable(layer_name='layer_fc_out_1')

print(weights_conv1)

```

```
<tf.Variable 'layer_fc_out_1/kernel:0' shape=(128, 10) dtype=float32_ref>
```

```

In [11]: opt = tf.train.AdamOptimizer(learning_rate=1e-4)
optimizer = opt.minimize(loss)
correct_prediction = tf.equal(y_pred_cls, y_true_cls)
accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))

```



```
In [13]: train_batch_size = 64
loss_list=[]
loss_list_test=[]
accuracy_list=[]
accuracy_list_test=[]
with tf.Session() as sess:
    sess.run(tf.global_variables_initializer())
    for iteration in range(200000):
        x_batch, y_true_batch = mnist.train.next_batch(train_batch_size)
        feed_dict_train = {x: x_batch, y_true: y_true_batch}
        sess.run(optimizer, feed_dict = feed_dict_train)
        if iteration % 100 == 0:
            feed_dict_test = {x: mnist.test.images, y_true: mnist.test.labels}
            y_true_cls_batch = np.argmax(y_true_batch, axis=1)
            loss_, accuracy_ = sess.run([loss, accuracy], feed_dict = feed_dict_train)
            loss_test_, accuracy_test_ = sess.run([loss, accuracy], feed_dict = feed_dict_test)
            loss_list.append(loss_)
            loss_list_test.append(loss_test_)
            accuracy_list.append(accuracy_)
            accuracy_list_test.append(accuracy_test_)
```

```
-----
KeyboardInterrupt                                Traceback (most recent call last)
<ipython-input-13-c29c6df58936> in <module>
      9         x_batch, y_true_batch = mnist.train.next_batch(train_batch_size)
     10         feed_dict_train = {x: x_batch, y_true: y_true_batch}
----> 11         sess.run(optimizer, feed_dict = feed_dict_train)
     12         if iteration % 100 == 0:
     13             feed_dict_test = {x: mnist.test.images, y_true: mnist.test.l
abels}

~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in r
un(self, fetches, feed_dict, options, run_metadata)
     965
     966     try:
--> 967         result = self._run(None, fetches, feed_dict, options_ptr,
     968                             run_metadata_ptr)
     969         if run_metadata:

~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in _
run(self, handle, fetches, feed_dict, options, run_metadata)
    1188     # or if the call is a partial run that specifies feeds.
    1189     if final_fetches or final_targets or (handle and feed_dict_tensor):
-> 1190         results = self._do_run(handle, final_targets, final_fetches,
    1191                                 feed_dict_tensor, options, run_metadata)
    1192     else:

~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in _
do_run(self, handle, target_list, fetch_list, feed_dict, options, run_metadata)
    1366
    1367     if handle is None:
-> 1368         return self._do_call(_run_fn, feeds, fetches, targets, options,
    1369                               run_metadata)
    1370     else:

~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in _
do_call(self, fn, *args)
    1373     def _do_call(self, fn, *args):
    1374         try:
-> 1375             return fn(*args)
    1376         except errors.OpError as e:
    1377             message = compat.as_text(e.message)
```

```
~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in _
run_fn(feed_dict, fetch_list, target_list, options, run_metadata)
    1357         # Ensure any changes to the graph are reflected in the runtime.
    1358         self._extend_graph()
-> 1359         return self._call_tf_sessionrun(options, feed_dict, fetch_list,
    1360                                         target_list, run_metadata)
    1361

~/anaconda3/lib/python3.8/site-packages/tensorflow/python/client/session.py in _
call_tf_sessionrun(self, options, feed_dict, fetch_list, target_list, run_metada
ta)
    1449     def _call_tf_sessionrun(self, options, feed_dict, fetch_list, target_l
ist,
    1450                               run_metadata):
-> 1451         return tf_session.TF_SessionRun_wrapper(self._session, options, feed
_dict,
    1452                                                  fetch_list, target_list,
    1453                                                  run_metadata)
```

KeyboardInterrupt:

In []:

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
import numpy as np
import math
import abs1
import logging
logger = tf.get_logger()
logger.setLevel(logging.ERROR)
```

WARNING:tensorflow:From /Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.ops.variable_scope) is deprecated and will be removed in a future version.

Instructions for updating:

non-resource variables are not supported in the long term

```
In [2]: from tensorflow.examples.tutorials.mnist import input_data
#data = input_data.read_data_sets('data/MNIST/', one_hot=True)
mnist = input_data.read_data_sets('./mnist', one_hot=True) # they has been norm
test_x = mnist.test.images[:10000]
test_y = mnist.test.labels[:10000]
```

Extracting ./mnist/train-images-idx3-ubyte.gz

Extracting ./mnist/train-labels-idx1-ubyte.gz

Extracting ./mnist/t10k-images-idx3-ubyte.gz

Extracting ./mnist/t10k-labels-idx1-ubyte.gz

```
In [3]: mnist.test.cls = np.argmax(mnist.test.labels, axis=1)
```

```
In [5]: # We know that MNIST images are 28 pixels in each dimension.
img_size = 28

# Images are stored in one-dimensional arrays of this length.
img_size_flat = img_size * img_size

# Tuple with height and width of images used to reshape arrays.
img_shape = (img_size, img_size)

# Number of colour channels for the images: 1 channel for gray-scale.
num_channels = 1

# Number of classes, one class for each of 10 digits.
num_classes = 10
```

```
In [6]: x = tf.placeholder(tf.float32, shape=[None, img_size_flat], name='x') / 255
x_image = tf.reshape(x, [-1, img_size, img_size, num_channels])
y_true = tf.placeholder(tf.float32, shape=[None, num_classes], name='y_true')
y_true_cls = tf.argmax(y_true, dimension=1)
```

create the CNN

```
In [10]: conv1 = tf.layers.conv2d( # shape (28, 28, 1)
    inputs=x_image,
    filters=16,
    kernel_size=5,
    strides=1,
    padding='same',
```

```

        activation=tf.nn.relu
    )
    # -> (28, 28, 16)
    pool1 = tf.layers.max_pooling2d(
        conv1,
        pool_size=2,
        strides=2,
    )
    # -> (14, 14, 16)
    conv2 = tf.layers.conv2d(pool1, 36, 5, 1, 'same', activation=tf.nn.relu) # ->
    pool2 = tf.layers.max_pooling2d(conv2, 2, 2) # -> (7, 7, 36)
    flat = tf.reshape(pool2, [-1, 7*7*36]) # -> (7*7*36, )
    output = tf.layers.dense(flat, 10) # output layer

    loss = tf.losses.softmax_cross_entropy(onehot_labels=y_true, logits=output)
    train_op = tf.train.AdamOptimizer(0.001).minimize(loss) #learning rate is 0.1

    accuracy = tf.metrics.accuracy( # return (acc, update_op), and create 2
        labels=tf.argmax(y_true, axis=1), predictions=tf.argmax(output, axis=1),)[1]

```

/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/convolutional.py:414: UserWarning: `tf.layers.conv2d` is deprecated and will be removed in a future version. Please Use `tf.keras.layers.Conv2D` instead.

warnings.warn("`tf.layers.conv2d` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/base_layer_v1.py:1719: UserWarning: `layer.apply` is deprecated and will be removed in a future version. Please use `layer.__call__` method instead.

warnings.warn("`layer.apply` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/pooling.py:310: UserWarning: `tf.layers.max_pooling2d` is deprecated and will be removed in a future version. Please use `tf.keras.layers.MaxPooling2D` instead.

warnings.warn("`tf.layers.max_pooling2d` is deprecated and '
/Users/gloria/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/legacy_tf_layers/core.py:171: UserWarning: `tf.layers.dense` is deprecated and will be removed in a future version. Please use `tf.keras.layers.Dense` instead.

warnings.warn("`tf.layers.dense` is deprecated and '

```

In [11]: from matplotlib import cm
try: from sklearn.manifold import TSNE; HAS_SK = True
except: HAS_SK = False; print('\nPlease install sklearn for layer visualization')
def plot_with_labels(lowDWeights, labels):
    plt.cla(); X, Y = lowDWeights[:, 0], lowDWeights[:, 1]
    for x, y, s in zip(X, Y, labels):
        c = cm.rainbow(int(255 * s / 8)); plt.text(x, y, s, backgroundcolor=c, f
    plt.xlim(X.min(), X.max()); plt.ylim(Y.min(), Y.max()); plt.title('Visualize
plt.ion()

```

```

In [12]: session=tf.Session()
init_op = tf.group(tf.global_variables_initializer(), tf.local_variables_initializer())
session.run(init_op) # initialize var in graph

```

```

In [13]: loss_list=[]
accuracy_list=[]
for step in range(600):
    b_x, b_y = mnist.train.next_batch(50) #Batch_size is 50
    _, loss_ = session.run([train_op, loss], {x: b_x, y_true: b_y})
    if step % 50 == 0:
        accuracy_, flat_representation = session.run([accuracy, flat], {x: test_
        print('Step:', step, '| train loss: %.4f' % loss_, '| test accuracy: %.2
        loss_list.append(loss_)

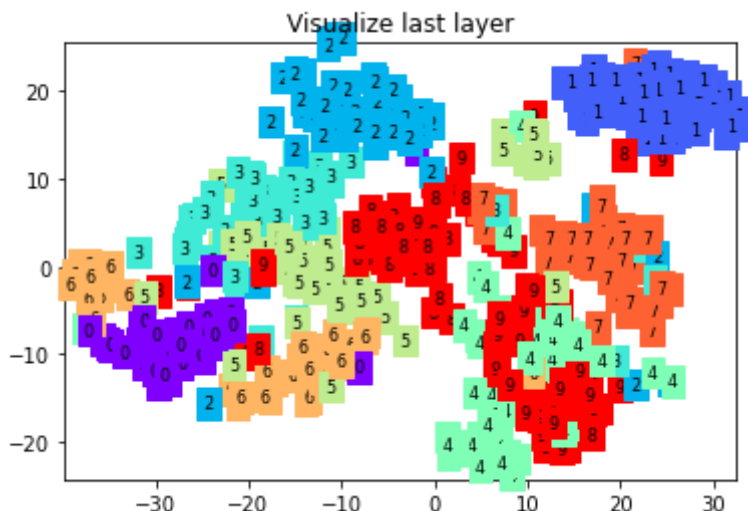
```

```

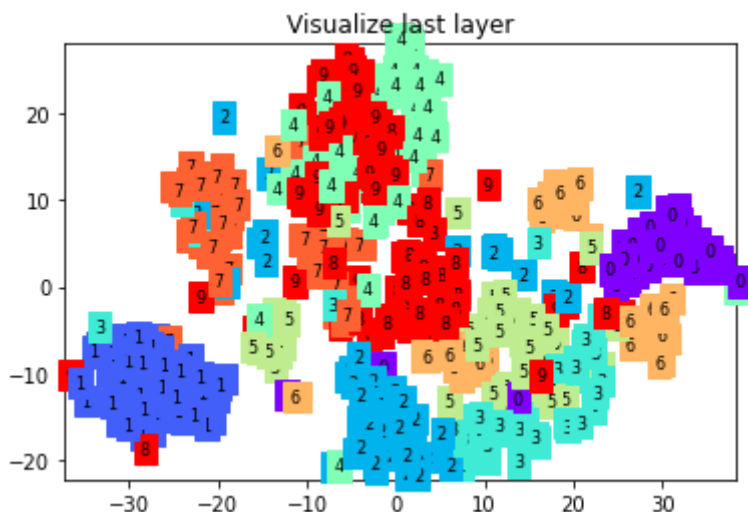
accuracy_list.append(accuracy_)
if HAS_SK:
    # Visualization of trained flatten layer (T-SNE)
    tsne = TSNE(perplexity=30, n_components=2, init='pca', n_iter=5000);
    low_dim_embs = tsne.fit_transform(flat_representation[:plot_only, :])
    labels = np.argmax(test_y, axis=1)[:plot_only]; plot_with_labels(low
plt.ioff()

```

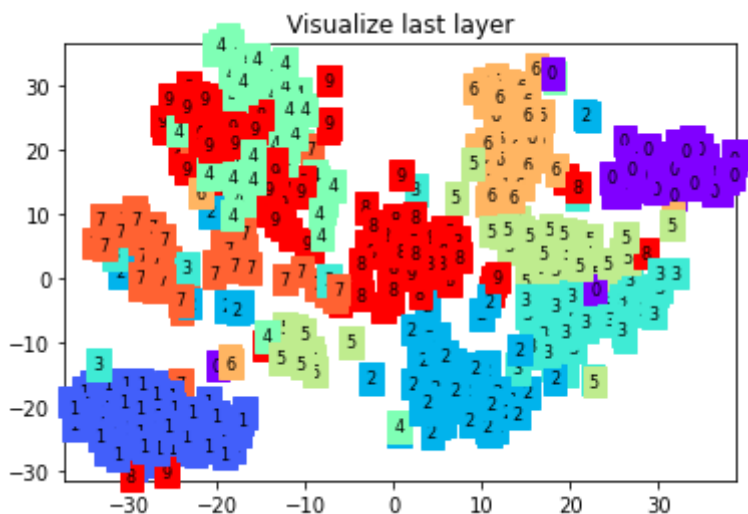
Step: 0 | train loss: 2.2945 | test accuracy: 0.18



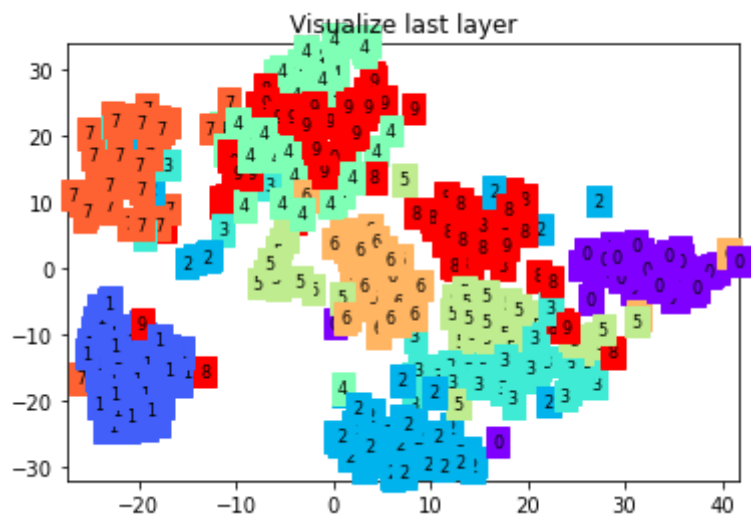
Step: 50 | train loss: 0.4907 | test accuracy: 0.53



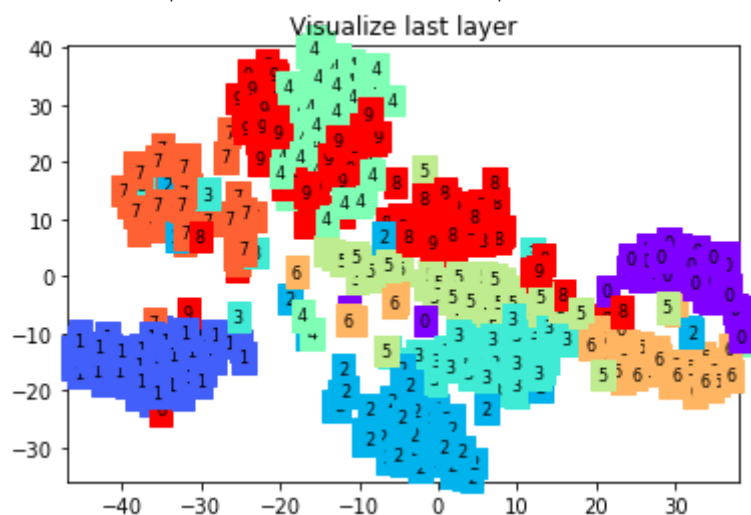
Step: 100 | train loss: 0.3523 | test accuracy: 0.66



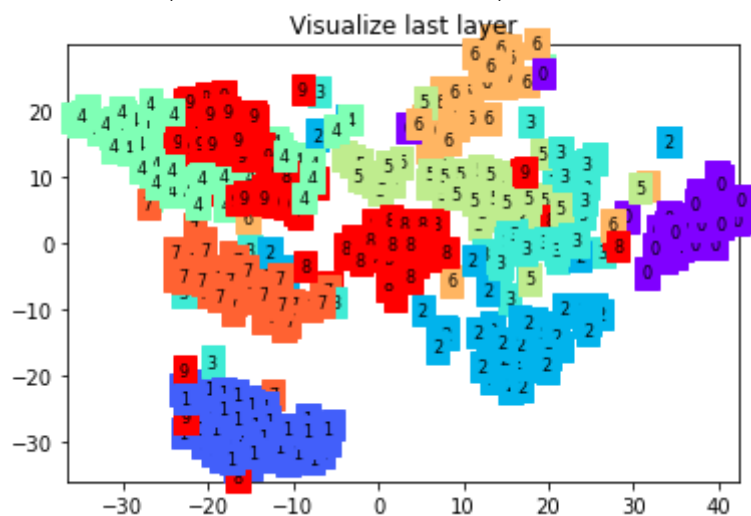
Step: 150 | train loss: 0.2431 | test accuracy: 0.73



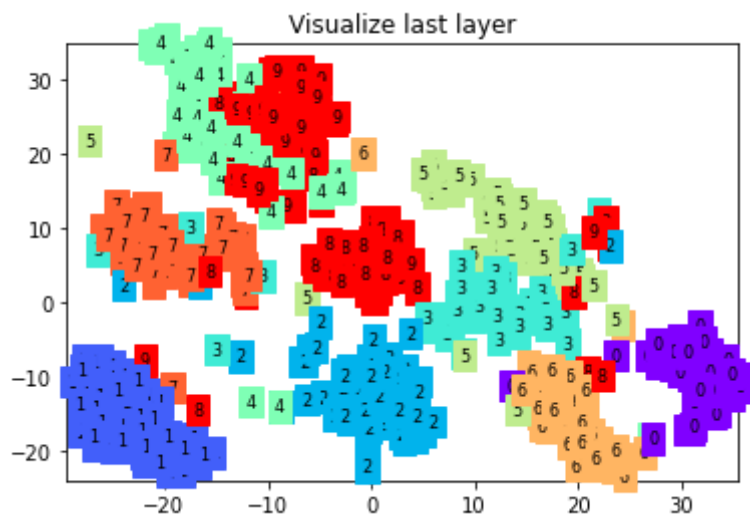
Step: 200 | train loss: 0.1333 | test accuracy: 0.77



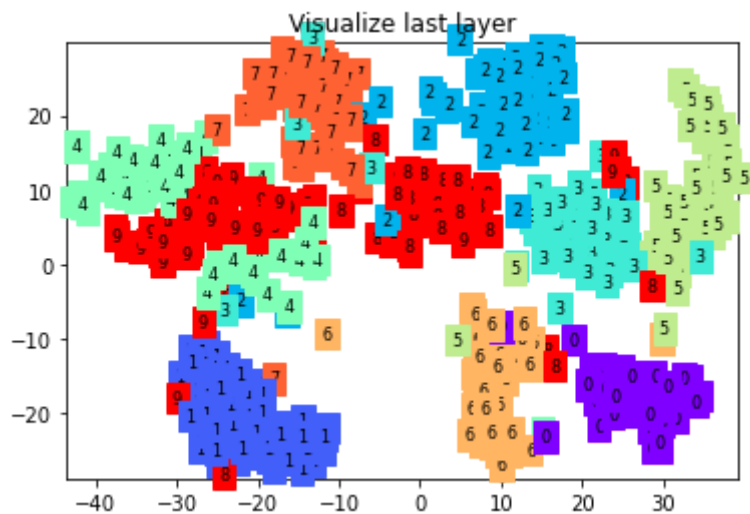
Step: 250 | train loss: 0.1475 | test accuracy: 0.80



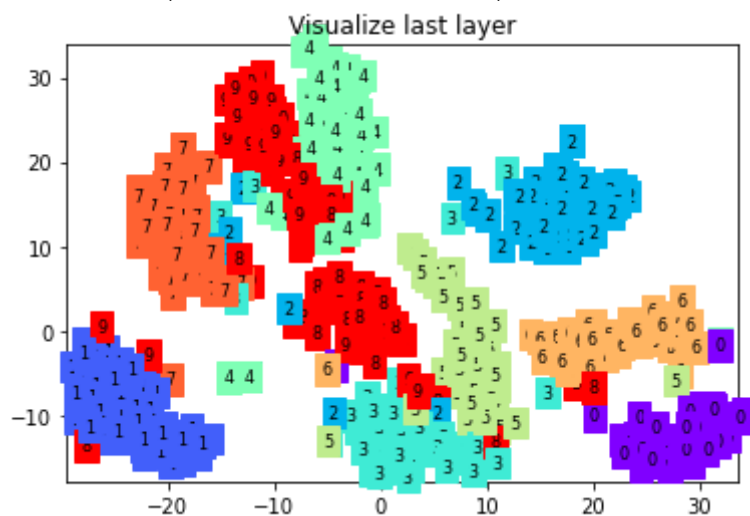
Step: 300 | train loss: 0.1584 | test accuracy: 0.83



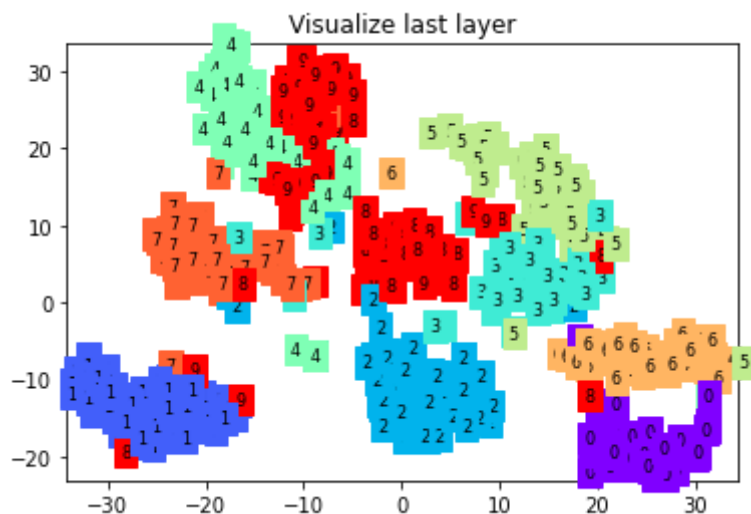
Step: 350 | train loss: 0.0986 | test accuracy: 0.84



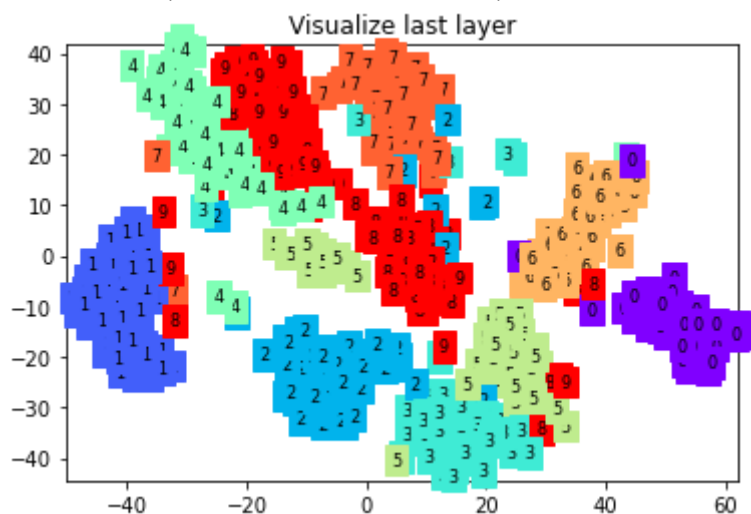
Step: 400 | train loss: 0.0751 | test accuracy: 0.86



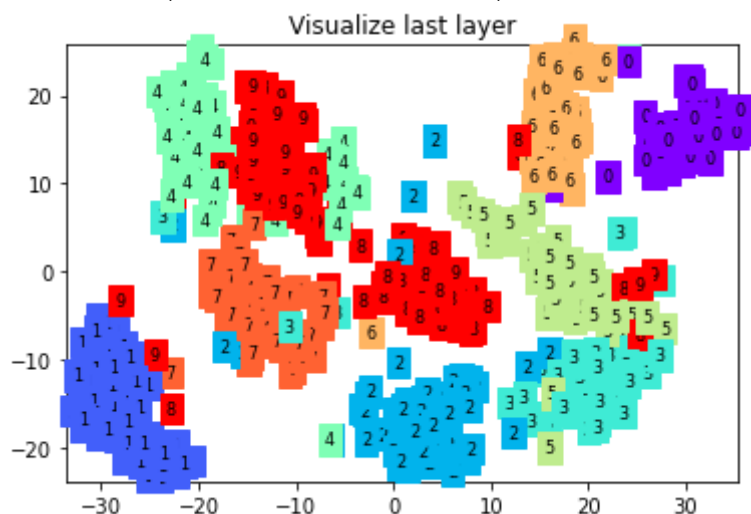
Step: 450 | train loss: 0.1182 | test accuracy: 0.87



Step: 500 | train loss: 0.1666 | test accuracy: 0.88



Step: 550 | train loss: 0.1771 | test accuracy: 0.88



```
In [14]: test_output = session.run(output, {x: test_x[:10]})
pred_y = np.argmax(test_output, 1)
print(pred_y, 'prediction number')
print(np.argmax(test_y[:10], 1), 'real number')
```

```
[7 2 1 0 4 1 4 9 5 9] prediction number
[7 2 1 0 4 1 4 9 5 9] real number
```


Github see: <https://github.com/Gloriabhsfer/DeeplearningCPS843>

In []: