

# ASSIGNMENT: 1

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## Assignment Document

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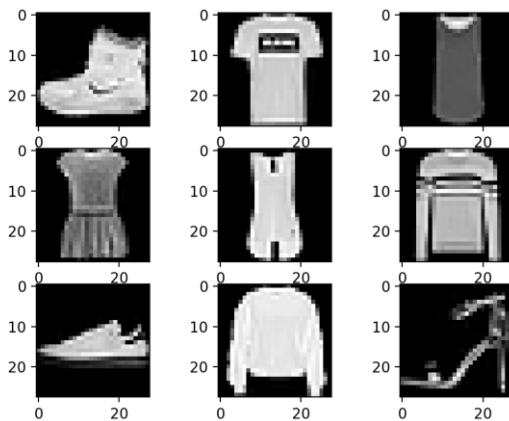
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## Scenario 1

### DESCRIPTION

#### Problem Statement:

The dataset is similar to MNIST, but includes images of certain clothing and accessory. The objective is to classify images into specific classes using single layer perceptron and multilayer perceptron.



#### Dataset:

Total Images :- 70,000

Train Images :- 60,000

Test Images :- 10,000

Image Size :- 28 X 28

#### Different Classes:

Classes: 'T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat', 'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle Boot'

**Note:** Please use google collab to work on this project. Also, make sure to select GPU backend while selecting a runtime.

<https://medium.com/deep-learning-turkey/google-colab-free-gpu-tutorial-e113627b9f5d>

**Question 1:**

A) Reshape and Normalize the data:

- reshape the input image
  - train; 60000, 28 \* 28 \* 1
  - test; 10000, 28 \* 28 \* 1
- change the type of data to float32
- normalize the data by dividing with 255

B) Convert the y\_train and y\_test to categorical by using keras to\_categorical function and define num\_classes=10

**HINT:-**

```
# Loading data from Keras library
fashion_mnist = keras.datasets.fashion_mnist
(X_train, y_train), (X_test, y_test) = tf.keras.datasets.fashion_mnist.load_data()
print ("The shape of data for train", X_train.shape)
```

The shape of data for train (60000, 28, 28)

**Question 2:**

Write a multi layer perceptron using tensorflow for fashion MNIST data to classify all 10 classes. Solve for each questions:

**Hint:**

```
#define the important parameters and variable to work with the tensors
learning_rate = 0.001
training_epochs = 100
cost_history = np.empty(shape=[1], dtype=float)
n_dim = X_train.shape[1]
n_class = 10

#define the number of hidden layers and number of neurons for each layer
n_hidden_1 = 256
n_hidden_2 = 256
n_hidden_3 = 256
n_hidden_4 = 256

x = tf.placeholder(tf.float32, [None, n_dim])
W = tf.Variable(tf.zeros([n_dim, n_class]))
b = tf.Variable(tf.zeros([n_class]))
y_ = tf.placeholder(tf.float32, [None, n_class])
```

Take epoch= 2000

A) Define a function that creates 4 layers using TensorFlow layers:

**HINT: -**

```
# Hidden layer with RELU activationsd
layer_1 = tf.add(tf.matmul(x, weights['h1']), biases['b1'])
layer_1 = tf.nn.sigmoid(layer_1)
```

B) Define a function that creates 4 layers using TensorFlow layers:

**HINT:-**

- i. 'h1': tf.Variable(tf.truncated\_normal([n\_dim, n\_hidden\_1])),
- ii. 'b1': tf.Variable(tf.truncated\_normal([n\_hidden\_1])),

C) Call the function:

```
y = multilayer_perceptron(x, weights, biases)
```

D) Define Cost function and Optimizer:

- i. Cost: tf.nn.softmax\_cross\_entropy\_with\_logits
- ii. Optimizer: tf.train.GradientDescentOptimizer

## Initiate the training:

```
for epoch in range(training_epochs):
    sess.run(training_step, feed_dict={x: X_train, y_: y_train})
    cost = sess.run(cost_function, feed_dict={x: X_train, y_: y_train})
    cost_history = np.append(cost_history, cost)
    correct_prediction = tf.equal(tf.argmax(y, 1), tf.argmax(y_, 1))
    accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
    accuracy = sess.run(accuracy, feed_dict={x: X_test, y_: y_test})
    if epoch % 200 == 0:
        print('epoch : ', epoch, ' - ', 'cost: ', cost, "- Accuracy: ", accuracy)
```

### Expected Output:

```
epoch: 0 - cost: 192.05133 - Accuracy: 0.1558
epoch: 200 - cost: 17.27994 - Accuracy: 0.3904
epoch: 400 - cost: 11.775045 - Accuracy: 0.488
epoch: 600 - cost: 9.359716 - Accuracy: 0.5298
epoch: 800 - cost: 7.8768625 - Accuracy: 0.5556
```

### Question 3:

Similarly like MLP, define a function that can create single layer network called perceptron on Fashion MNIST.

- Learning rate: 0.0001
- Epochs: 5000

Hint: Use the hints given in MLP questions. Also, make sure to correctly check for layers that are going to be input to output layer.

### Question 4:

Now, initialize variables, cost function and optimizer to initiate the training.

- Adam Optimizer

```

for epoch in range(training_epochs):
    sess.run(training_step, feed_dict={x: X_train, y_: y_train})
    cost = sess.run(cost_function, feed_dict={x: X_train, y_: y_train})
    cost_history = np.append(cost_history, cost)
    correct_prediction = tf.equal(tf.argmax(y, 1), tf.argmax(y_, 1))
    accuracy = tf.reduce_mean(tf.cast(correct_prediction, tf.float32))
    accuracy = sess.run(accuracy, feed_dict={x: X_test, y_: y_test})
    if epoch % 200 == 0:
        print('epoch : ', epoch, ' - ', 'cost: ', cost, "- Accuracy: ", accuracy)

```

## Expected Output:-

Future major versions of TensorFlow will allow gradients to flow into the labels input on backprop by default.

See ``tf.nn.softmax_cross_entropy_with_logits_v2``.

```
epoch : 0 - cost: 12.307046 - Accuracy: 0.0841
```

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