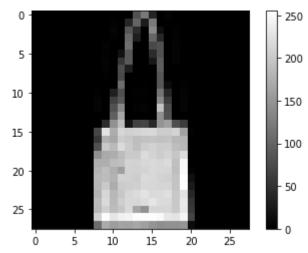
In [0]:

```
In [1]: #import tensorflow as tf
    import tensorflow.compat.v1 as tf
    tf.disable_v2_behavior()
    from tensorflow import keras
    import numpy as np
    import matplotlib.pyplot as plt
    from __future__ import print_function
    from tensorflow.keras.utils import to_categorical
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/pytho n/compat/v2_compat.py:96: disable_resource_variables (from tensorflow.python.op s.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

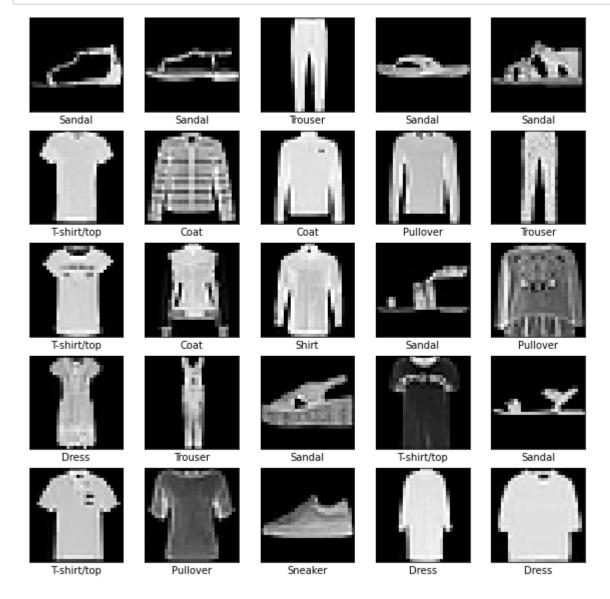
```
In [2]:
       fashion mnist = keras.datasets.fashion mnist
       (train_images, train_lables), (test_images, test_lables) = fashion_mnist.load_da
       Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datase
       ts/train-labels-idx1-ubyte.gz (https://storage.googleapis.com/tensorflow/tf-ker
       as-datasets/train-labels-idx1-ubyte.gz)
       Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datase
       ts/train-images-idx3-ubyte.gz (https://storage.googleapis.com/tensorflow/tf-ker
       as-datasets/train-images-idx3-ubyte.gz)
       26427392/26421880 [=========== ] - Os Ous/step
       Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datase
       ts/t10k-labels-idx1-ubyte.gz (https://storage.googleapis.com/tensorflow/tf-kera
       s-datasets/t10k-labels-idx1-ubyte.gz)
       8192/5148 [=======] - Os Ous/step
       Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datase
       ts/t10k-images-idx3-ubyte.gz (https://storage.googleapis.com/tensorflow/tf-kera
       s-datasets/t10k-images-idx3-ubyte.gz)
```

image_type = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat', 'Sandal', 'S



```
In [5]: plt.figure(figsize=(10,10))
for i in range(25):
    idx = np.random.randint(0,60000, dtype = int)
    plt.subplot(5,5,i+1)
    plt.grid(False)
    plt.xticks([])
    plt.yticks([])
    plt.imshow(train_images[idx], cmap = 'gray')
    plt.xlabel(image_type[train_lables[idx]])

plt.show()
```



```
In [6]: print("train images")
        print(train images.shape)
        train images = train images.reshape(train images.shape[0], 784)
        test images = test images.reshape(test images.shape[0], 784)
        print(train images.shape[0])
        print(train_images.shape[1])
        print("train lables")
        print(train lables.shape)
         ''' Normalizing the data '''
        train_images = train_images / 255
        test images = test images / 255
        train lables = to categorical(train lables)
        test_lables = to_categorical(test_lables)
        train_images
        (60000, 28, 28)
        60000
        784
        train lables
        (60000,)
In [7]: | display_step = 1
        learning rate = 0.001
        training epoch = 100
        cost_history = np.empty(shape = [1], dtype = float)
        n dim = train images.shape[1]
        n_{class} = 10
        n hidden 1 = 256
        n hidden 2 = 256
        n hidden 3 = 256
        n hidden 4 = 256
        n_{input} = 784
        batch size = 500
        x = tf.placeholder(tf.float32, [None, n input])
        # W = tf. Variable(tf.zeros([n dim, n class]))
        # b = tf.Variable(tf.zeros([n class]))
        y = tf.placeholder(tf.float32, [None, n class])
        print(x)
        print(y)
```

Tensor("Placeholder:0", shape=(?, 784), dtype=float32)
Tensor("Placeholder_1:0", shape=(?, 10), dtype=float32)

```
In [8]:
        weights = {
            "h1" : tf.Variable(tf.random normal([n input, n hidden 1])), #784 , 256
            "h2" : tf.Variable(tf.random normal([n hidden 1, n hidden 2])), #256 , 256
            "h3" : tf.Variable(tf.random normal([n hidden 2, n hidden 3])), #256 , 256
            "h4" : tf.Variable(tf.random normal([n hidden 3, n hidden 4])), #256 , 256
            "out" : tf.Variable(tf.random_normal([n_hidden_2, n_class])), #256 , 10
        }
        biases = {
            "b1"
                 : tf.Variable(tf.random normal([n hidden 1])),
            "b2" : tf.Variable(tf.random normal([n hidden 2])),
            "b3" : tf.Variable(tf.random normal([n hidden 3])),
            "b4" : tf.Variable(tf.random_normal([n_hidden_4])),
            "out" : tf.Variable(tf.random_normal([n_class]))
        }
        def MLP(x, weights, biases):
          # input layer
          layer 1 = tf.add(tf.matmul(x, weights["h1"]), biases["b1"])
          layer 1 =tf.nn.sigmoid(layer 1)
          # First hidden layer
          layer 2 = tf.add(tf.matmul(layer 1, weights["h2"]), biases["b2"])
          layer 2 = tf.nn.sigmoid(layer 2)
          #Second hidden Laver
          layer 3 = tf.add(tf.matmul(layer 2, weights["h3"]), biases["b3"])
          layer 3 = tf.nn.sigmoid(layer 3)
          # Third hidden Layer
          layer 4 = tf.add(tf.matmul(layer 3, weights["h4"]), biases["b4"])
          layer 4 = tf.nn.sigmoid(layer 4)
          # Output Layer
          out layer = tf.matmul(layer 4, weights["out"]) + biases["out"]
          return out layer
        pred = MLP(x, weights, biases)
        # Loss function and Optimizer
        cost = tf.reduce mean(tf.nn.softmax_cross_entropy_with_logits(logits= pred , label)
        optimizer = tf.train.AdamOptimizer(learning rate=learning rate).minimize(cost)
        init = tf.global variables initializer()
        cost history = []
        accuracy history = []
```

WARNING:tensorflow:From <ipython-input-8-53fa5ec0c00d>:39: softmax_cross_entrop y_with_logits (from tensorflow.python.ops.nn_ops) is deprecated and will be rem oved in a future version.

Instructions for updating:

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`.

```
In [9]: total batch = int(train images.shape[0]/batch size)
        with tf.Session() as sess:
          sess.run(init)
          for epoch in range(training epoch):
            avg cost = 0
            temp = 0
            for i in range(total_batch):
                   batch_x = train_images[temp:temp+batch_size]
                  batch_y = train_lables[temp:temp+batch_size]
                   _, c = sess.run([optimizer, cost], feed_dict={x: batch_x,y: batch_y})
                  temp = temp+batch size
                   # Compute average Loss
                   avg cost += c
            avg_cost = avg_cost/batch_size
            if epoch % display_step == 0:
              correct prediction = tf.equal(tf.argmax(pred,1), tf.argmax(y,1))
              accuracy = tf.reduce mean(tf.cast(correct prediction, "float"))
              acu_temp = accuracy.eval({x: test_images, y: test_lables})
              accuracy history.append(acu temp)
              cost history.append(avg cost)
              print("Epoch:", '%04d' % (epoch + 1), "- cost=", "{:.9f}".format(avg_cost)
          print("Optimization Finished!")
          #plot the cost history
          plt.plot(cost history)
          plt.show()
          #plot the accuracy history
          plt.plot(accuracy_history)
          plt.show()
          # Test model
          correct prediction = tf.equal(tf.argmax(pred, 1), tf.argmax(y, 1))
          # Calculate accuracy
          accuracy = tf.reduce mean(tf.cast(correct prediction, "float"))
          print("Accuracy:", accuracy.eval({x: test_images, y: test_lables}))
```

```
Epoch: 0001 - cost= 0.868805625 - Accuracy= 0.6575
Epoch: 0002 - cost= 0.213008063 - Accuracy= 0.7191
Epoch: 0003 - cost= 0.172829522 - Accuracy= 0.7478
Epoch: 0004 - cost= 0.151812314 - Accuracy= 0.7644
Epoch: 0005 - cost= 0.137984700 - Accuracy= 0.7765
Epoch: 0006 - cost= 0.127573230 - Accuracy= 0.7835
Epoch: 0007 - cost= 0.119343474 - Accuracy= 0.7899
Epoch: 0008 - cost= 0.112521329 - Accuracy= 0.7939
Epoch: 0009 - cost= 0.106693498 - Accuracy= 0.7981
Epoch: 0010 - cost= 0.101532109 - Accuracy= 0.8029
Epoch: 0011 - cost= 0.096933855 - Accuracy= 0.8049
Epoch: 0012 - cost= 0.092782219 - Accuracy= 0.8083
Epoch: 0013 - cost= 0.088988985 - Accuracy= 0.8108
Epoch: 0014 - cost= 0.085497362 - Accuracy= 0.8134
Epoch: 0015 - cost= 0.082261486 - Accuracy= 0.8159
Epoch: 0016 - cost= 0.079253022 - Accuracy= 0.8162
Epoch: 0017 - cost= 0.076453734 - Accuracy= 0.8173
Epoch: 0018 - cost= 0.073829636 - Accuracy= 0.8177
Epoch: 0019 - cost= 0.071358513 - Accuracy= 0.8185
```

In [0]: