

Practical 5 - Dhruv Patel BDA 17162121014

```
In [2]: 1 import tensorflow as tf
        2 from tensorflow import keras
        3
        4
        5 import numpy as np
        6 import matplotlib.pyplot as plt
        7 import seaborn as sns
```

```
In [6]: 1 dataset = keras.datasets.fashion_mnist
        2
        3 (train_images, train_labels), (test_images, test_labels) = dataset.load_data()
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-labels-idx1-ubyte.gz> (<https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-labels-idx1-ubyte.gz>)

32768/29515 [=====] - 0s 1us/step

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-images-idx3-ubyte.gz> (<https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-images-idx3-ubyte.gz>)

26427392/26421880 [=====] - 3s 0us/step

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-labels-idx1-ubyte.gz> (<https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-labels-idx1-ubyte.gz>)

8192/5148 [=====] - 0s 0s/step

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-images-idx3-ubyte.gz> (<https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-images-idx3-ubyte.gz>)

4423680/4422102 [=====] - 1s 0us/step

Loading the dataset returns four NumPy arrays:

The train_images and train_labels arrays are the training set—the data the model uses to learn.

The model is tested against the test set, the test_images, and test_labels arrays. The images are 28x28 NumPy arrays, with pixel values ranging from 0 to 255. The labels are an array of integers, ranging from 0 to 9. These correspond to the class of clothing the image represents:

Label Class

0 T-shirt/top

1 Trouser

2 Pullover

3 Dress

4 Coat

5 Sandal

6 Shirt

7 Sneaker

8 Bag

9 Ankle boot

Each image is mapped to a single label. Since the class names are not included with the dataset, store them here to use later when plotting the images:

```
In [7]: 1 class_names = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat',  
2                  'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle boot']
```

```
In [8]: 1 train_images.shape
```

```
Out[8]: (60000, 28, 28)
```

```
In [9]: 1 test_images.shape
```

```
Out[9]: (10000, 28, 28)
```

```
In [10]: 1 train_labels.shape
```

```
Out[10]: (60000,)
```

```
In [11]: 1 test_labels.shape
```

```
Out[11]: (10000,)
```

```
In [12]: 1 train_labels
```

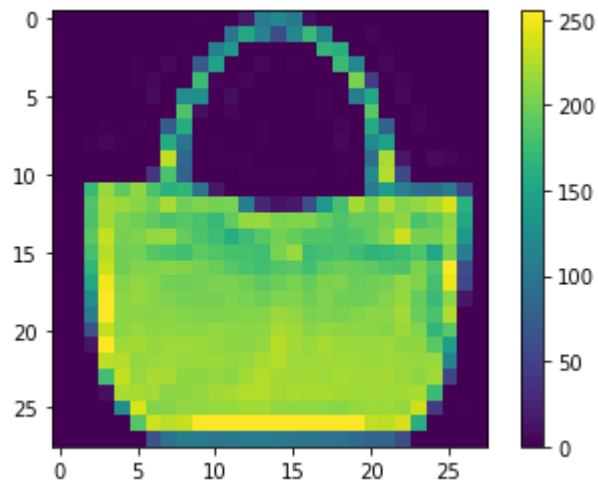
```
Out[12]: array([9, 0, 0, ..., 3, 0, 5], dtype=uint8)
```

```
In [20]: 1 len(train_images) , len(test_images) , len(test_labels) ,len(train_labels)
```

```
Out[20]: (60000, 10000, 10000, 60000)
```

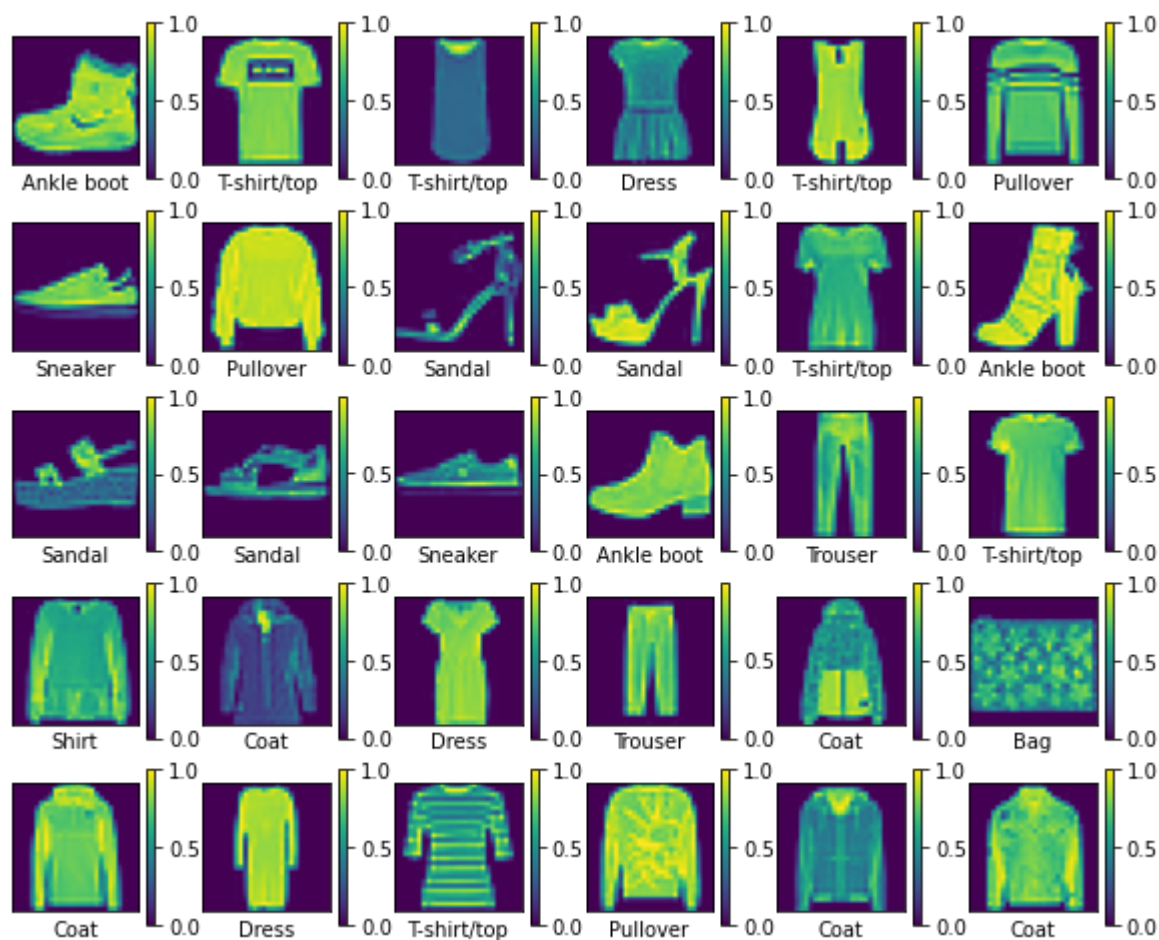
Preprocess the data

```
In [19]: 1 plt.figure()  
2 plt.imshow(train_images[100])  
3 plt.colorbar()  
4 plt.grid(False)  
5 plt.show()
```



```
In [21]: 1 train_images = train_images / 255.0  
2  
3 test_images = test_images / 255.0
```

```
In [31]: 1 plt.figure(figsize=(10,10))
2         for i in range(30):
3             plt.subplot(6,6,i+1)
4             plt.xticks([])
5             plt.yticks([])
6             plt.grid(False)
7             # plt.imshow(train_images[i], cmap=plt.cm.binary)
8             plt.imshow(train_images[i])
9             plt.colorbar()
10            plt.xlabel(class_names[train_labels[i]])
11            plt.show()
```



Build the model

```
In [33]: 1 #using keras we are now building model
2
3 model = keras.Sequential([
4     keras.layers.Flatten(input_shape=(28,28)),
5     keras.layers.Dense(128,activation='relu'),
6     keras.layers.Dense(128,activation='relu'),
7     keras.layers.Dense(10)
8 ])
```

```
In [34]: 1 model.compile(optimizer='adam',loss=tf.keras.losses.SparseCategoricalCrossen
2             metrics=['accuracy'])
```

```
In [36]: 1 model.fit(train_images,train_labels,epochs=20,use_multiprocessing=True) ## w
```

```
Epoch 1/20
1875/1875 [=====] - 1s 734us/step - loss: 0.2245 - acc
uracy: 0.9148
Epoch 2/20
1875/1875 [=====] - 1s 734us/step - loss: 0.2142 - acc
uracy: 0.9194
Epoch 3/20
1875/1875 [=====] - 1s 725us/step - loss: 0.2081 - acc
uracy: 0.9208
Epoch 4/20
1875/1875 [=====] - 1s 731us/step - loss: 0.2022 - acc
uracy: 0.9230
Epoch 5/20
1875/1875 [=====] - 1s 735us/step - loss: 0.1941 - acc
uracy: 0.9247
Epoch 6/20
1875/1875 [=====] - 1s 730us/step - loss: 0.1896 - acc
uracy: 0.9268
Epoch 7/20
1875/1875 [=====] - 1s 740us/step - loss: 0.1828 - acc
uracy: 0.9298
Epoch 8/20
1875/1875 [=====] - 1s 729us/step - loss: 0.1791 - acc
uracy: 0.9315
Epoch 9/20
1875/1875 [=====] - 1s 736us/step - loss: 0.1754 - acc
uracy: 0.9341
Epoch 10/20
1875/1875 [=====] - 1s 721us/step - loss: 0.1670 - acc
uracy: 0.9361
Epoch 11/20
1875/1875 [=====] - 1s 729us/step - loss: 0.1645 - acc
uracy: 0.9371
Epoch 12/20
1875/1875 [=====] - 1s 727us/step - loss: 0.1593 - acc
uracy: 0.9394
Epoch 13/20
1875/1875 [=====] - 1s 730us/step - loss: 0.1574 - acc
uracy: 0.9390
Epoch 14/20
1875/1875 [=====] - 1s 711us/step - loss: 0.1526 - acc
uracy: 0.9413
Epoch 15/20
1875/1875 [=====] - 1s 715us/step - loss: 0.1476 - acc
uracy: 0.9431
Epoch 16/20
1875/1875 [=====] - 1s 734us/step - loss: 0.1469 - acc
uracy: 0.9427
Epoch 17/20
1875/1875 [=====] - 1s 726us/step - loss: 0.1402 - acc
uracy: 0.9458
Epoch 18/20
1875/1875 [=====] - 1s 742us/step - loss: 0.1389 - acc
uracy: 0.9461
Epoch 19/20
```

```
1875/1875 [=====] - 1s 719us/step - loss: 0.1336 - acc
uracy: 0.9484
Epoch 20/20
1875/1875 [=====] - 1s 733us/step - loss: 0.1328 - acc
uracy: 0.9489
```

Out[36]: <tensorflow.python.keras.callbacks.History at 0x263f05fd130>

```
In [38]: 1 test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)
2
3 print('\nTest accuracy:', test_acc)
```

```
313/313 - 0s - loss: 0.4475 - accuracy: 0.8906
```

```
Test accuracy: 0.8906000256538391
```

```
In [39]: 1 model.save("fashion_mnist.h5")
```

Make Prediction

```
In [40]: 1 probability_model = tf.keras.Sequential([model,
2                                                  tf.keras.layers.Softmax()])
```

```
In [41]: 1 predictions = probability_model.predict(test_images)
```

```
In [42]: 1 predictions[0]
```

```
Out[42]: array([1.2450453e-10, 8.7283697e-10, 2.5722284e-11, 2.7918209e-13,
6.0904712e-09, 7.6097108e-06, 3.3953999e-11, 4.3500785e-04,
1.2557068e-12, 9.9955732e-01], dtype=float32)
```

```
In [59]: 1 for i in range(20):
2         print("Actual label for image ",i," : ",test_labels[i],end=" and ")
3         print("Predicted label for image ",i," : ",np.argmax(predictions[i]))
```

```
Actual label for image 0 : 9 and Predicted label for image 0 : 9
Actual label for image 1 : 2 and Predicted label for image 1 : 2
Actual label for image 2 : 1 and Predicted label for image 2 : 1
Actual label for image 3 : 1 and Predicted label for image 3 : 1
Actual label for image 4 : 6 and Predicted label for image 4 : 6
Actual label for image 5 : 1 and Predicted label for image 5 : 1
Actual label for image 6 : 4 and Predicted label for image 6 : 4
Actual label for image 7 : 6 and Predicted label for image 7 : 6
Actual label for image 8 : 5 and Predicted label for image 8 : 5
Actual label for image 9 : 7 and Predicted label for image 9 : 7
Actual label for image 10 : 4 and Predicted label for image 10 : 4
Actual label for image 11 : 5 and Predicted label for image 11 : 5
Actual label for image 12 : 7 and Predicted label for image 12 : 5
Actual label for image 13 : 3 and Predicted label for image 13 : 3
Actual label for image 14 : 4 and Predicted label for image 14 : 4
Actual label for image 15 : 1 and Predicted label for image 15 : 1
Actual label for image 16 : 2 and Predicted label for image 16 : 2
Actual label for image 17 : 4 and Predicted label for image 17 : 2
Actual label for image 18 : 8 and Predicted label for image 18 : 8
Actual label for image 19 : 0 and Predicted label for image 19 : 0
```

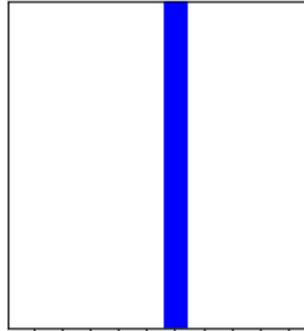
```
In [60]: 1 def plot_image(i, predictions_array, true_label, img):
2         true_label, img = true_label[i], img[i]
3         plt.grid(False)
4         plt.xticks([])
5         plt.yticks([])
6
7         plt.imshow(img, cmap=plt.cm.binary)
8
9         predicted_label = np.argmax(predictions_array)
10        if predicted_label == true_label:
11            color = 'blue'
12        else:
13            color = 'red'
14
15        plt.xlabel("{} {:.2f}% ({}).format(class_names[predicted_label],
16                                           100*np.max(predictions_array),
17                                           class_names[true_label]),
18                                           color=color)
19
20    def plot_value_array(i, predictions_array, true_label):
21        true_label = true_label[i]
22        plt.grid(False)
23        plt.xticks(range(10))
24        plt.yticks([])
25        thisplot = plt.bar(range(10), predictions_array, color="#777777")
26        plt.ylim([0, 1])
27        predicted_label = np.argmax(predictions_array)
28
29        thisplot[predicted_label].set_color('red')
30        thisplot[true_label].set_color('blue')
```

In [66]:

```
1 i = 11
2 plt.figure(figsize=(6,3))
3 plt.subplot(1,2,1)
4 plot_image(i, predictions[i], test_labels, test_images)
5 plt.subplot(1,2,2)
6 plot_value_array(i, predictions[i], test_labels)
7 plt.show()
```

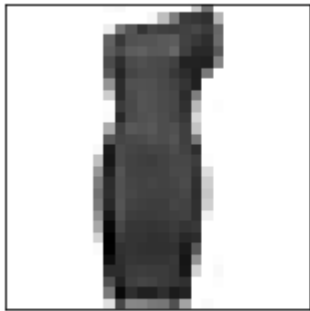


Sandal 100% (Sandal)

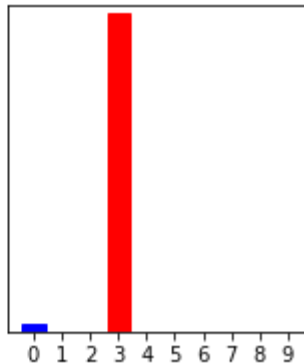


In [69]:

```
1 i = 1111
2 plt.figure(figsize=(6,3))
3 plt.subplot(1,2,1)
4 plot_image(i, predictions[i], test_labels, test_images)
5 plt.subplot(1,2,2)
6 plot_value_array(i, predictions[i], test_labels)
7 plt.show()
```



Dress 98% (T-shirt/top)

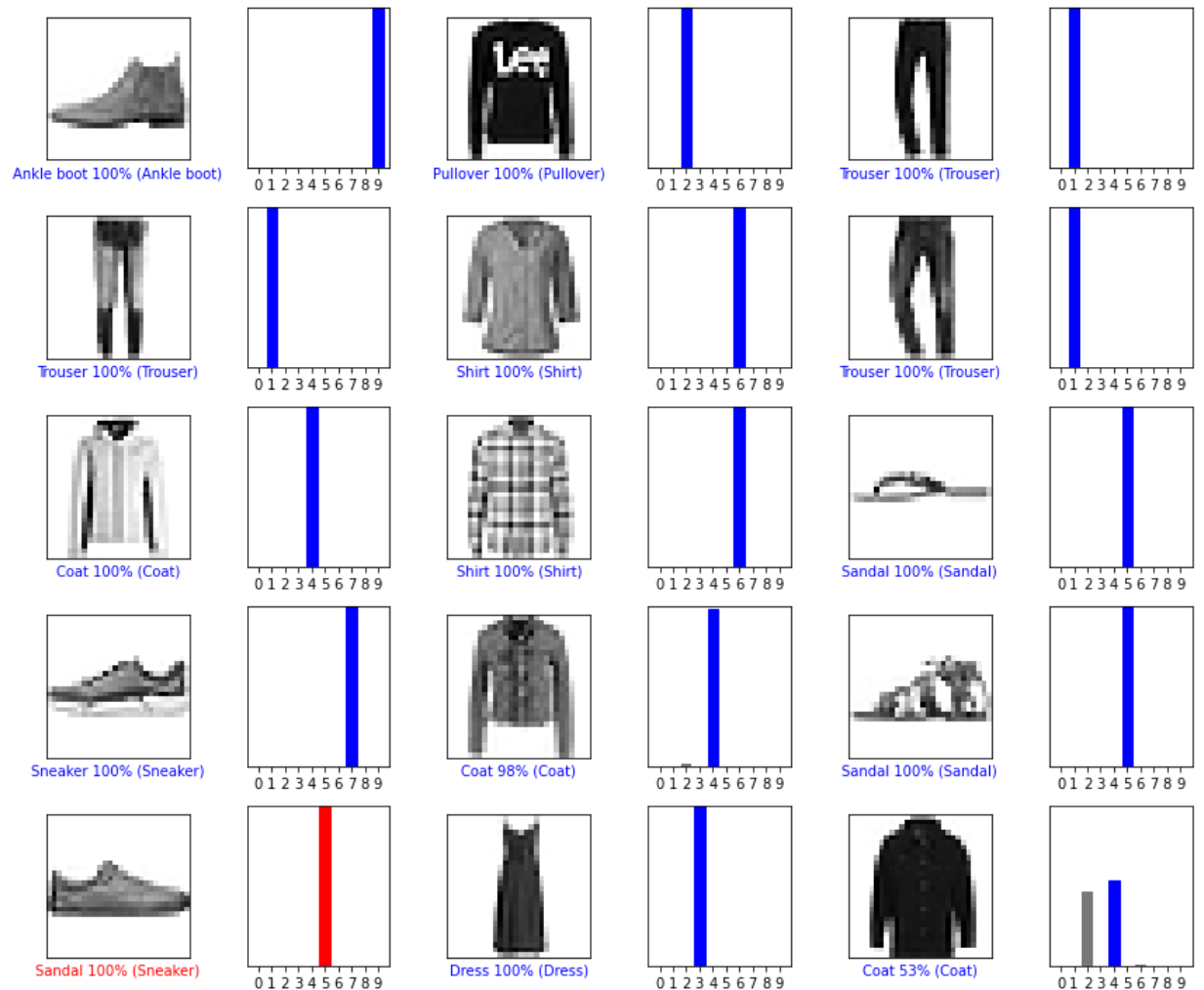


In [70]:

```

1  # Plot the first X test images, their predicted labels, and the true labels.
2  # Color correct predictions in blue and incorrect predictions in red.
3  num_rows = 5
4  num_cols = 3
5  num_images = num_rows*num_cols
6  plt.figure(figsize=(2*2*num_cols, 2*num_rows))
7  for i in range(num_images):
8      plt.subplot(num_rows, 2*num_cols, 2*i+1)
9      plot_image(i, predictions[i], test_labels, test_images)
10     plt.subplot(num_rows, 2*num_cols, 2*i+2)
11     plot_value_array(i, predictions[i], test_labels)
12 plt.tight_layout()
13 plt.show()

```



Testing on trained model

```
In [86]: 1 # Grab an image from the test dataset.  
2 img = test_images[1]  
3  
4 print(img.shape)
```

(28, 28)

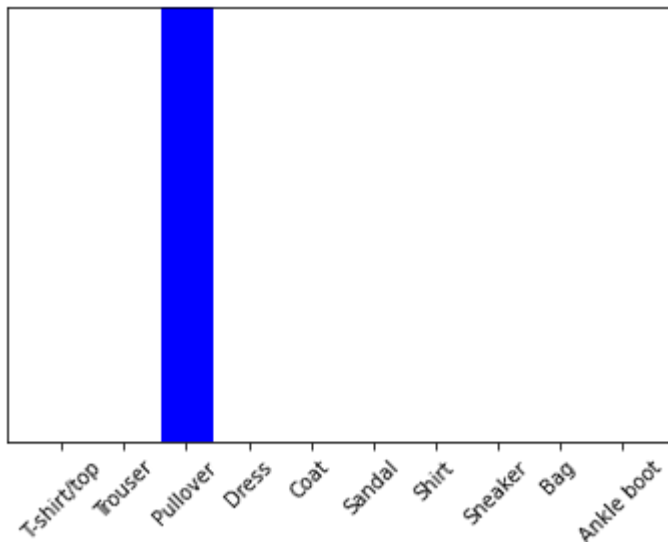
```
In [87]: 1 # Add the image to a batch where it's the only member.  
2 img = (np.expand_dims(img,0))  
3  
4 print(img.shape)
```

(1, 28, 28)

```
In [88]: 1 predictions_single = probability_model.predict(img)  
2  
3 print(predictions_single)
```

```
[[6.6959117e-08 1.5248392e-26 9.9999666e-01 1.4043638e-16 2.9703324e-06  
 1.8136180e-27 2.9681675e-07 3.4467158e-18 3.6841552e-19 3.8420584e-31]]
```

```
In [89]: 1 plot_value_array(1, predictions_single[0], test_labels)  
2 _ = plt.xticks(range(10), class_names, rotation=45)
```



```
In [90]: 1 ## thus we successfully trained an ANN for classification of fashion images
```

