

2.12.0

```
fashion_mnist = tf.keras.datasets.fashion_mnist

(train_images, train_labels), (test_images, test_labels) =
fashion_mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/train-labels-idx1-ubyte.gz
29515/29515 [=====] - 0s 0us/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/train-images-idx3-ubyte.gz
26421880/26421880 [=====] - 1s 0us/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/t10k-labels-idx1-ubyte.gz
5148/5148 [=====] - 0s 0us/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/t10k-images-idx3-ubyte.gz
4422102/4422102 [=====] - 1s 0us/step

class_names = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat',
               'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle boot']

train_images.shape

(60000, 28, 28)

len(train_labels)

60000

train_labels
```

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

```
test_images.shape
```

```
(10000, 28, 28)
```

```
len(test_labels)
```

```
10000
```

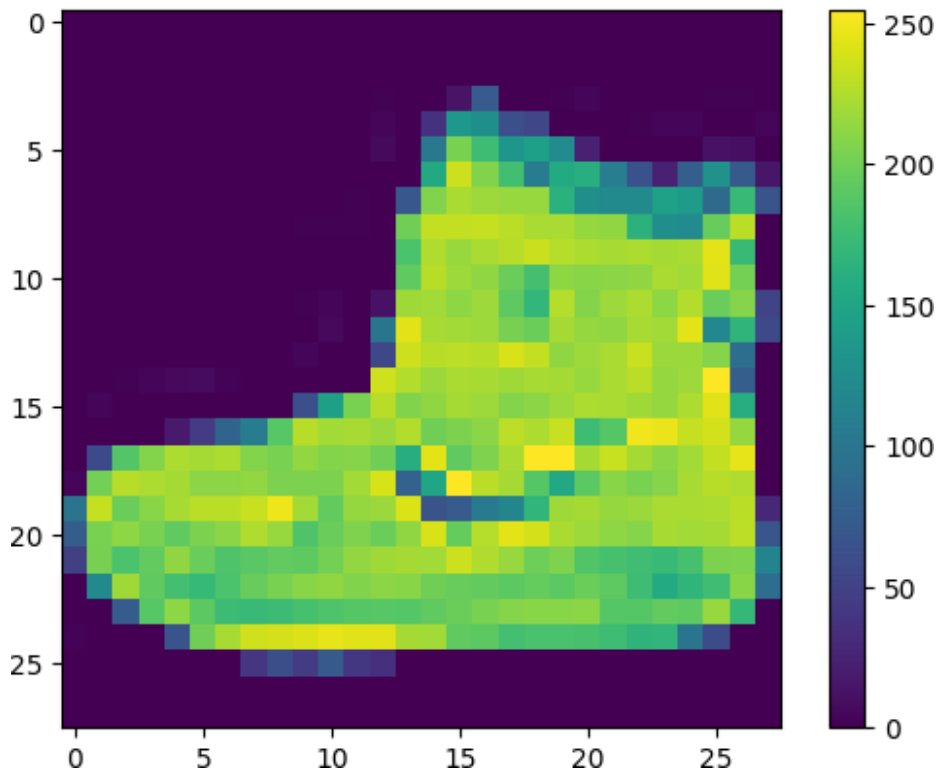
```
plt.figure()
```

```
plt.imshow(train_images[0])
```

```
plt.colorbar()
```

```
plt.grid(False)
```

```
plt.show()
```



```
train_images = train_images / 255.0
```

```
test_images = test_images / 255.0
```

```
plt.figure(figsize=(10,10))
```

```
for i in range(25):
```

```
    plt.subplot(5,5,i+1)
```

```
    plt.xticks([])
```

```
    plt.yticks([])
```

```
    plt.grid(False)
```

```
plt.imshow(train_images[i], cmap=plt.cm.binary)
plt.xlabel(class_names[train_labels[i]])
plt.show()
```



```
model = tf.keras.Sequential([
    tf.keras.layers.Flatten(input_shape=(28, 28)),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(10)
])
```

```
model.compile(optimizer='adam',
              loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
              metrics=['accuracy'])
```

```
model.fit(train_images, train_labels, epochs=10)
```

Epoch 1/10

```
1875/1875 [=====] - 11s 3ms/step - loss: 0.4998 - accuracy: 0.8235
```

Epoch 2/10

```
1875/1875 [=====] - 6s 3ms/step - loss:
0.3776 - accuracy: 0.8641
```

Epoch 3/10

```
1875/1875 [=====] - 5s 3ms/step - loss:
0.3363 - accuracy: 0.8761
```

Epoch 4/10

```
1875/1875 [=====] - 6s 3ms/step - loss:
0.3122 - accuracy: 0.8859
```

Epoch 5/10

```
1875/1875 [=====] - 5s 3ms/step - loss:
0.2930 - accuracy: 0.8922
```

Epoch 6/10

```
1875/1875 [=====] - 5s 3ms/step - loss:
0.2806 - accuracy: 0.8957
```

Epoch 7/10

```
1875/1875 [=====] - 5s 3ms/step - loss:
0.2663 - accuracy: 0.9007
```

Epoch 8/10

```
1875/1875 [=====] - 5s 3ms/step - loss:
0.2557 - accuracy: 0.9046
```

Epoch 9/10

```
1875/1875 [=====] - 6s 3ms/step - loss:
0.2458 - accuracy: 0.9087
```

Epoch 10/10

```
1875/1875 [=====] - 5s 3ms/step - loss: 0.2369 - accuracy: 0.9113
```

```
<keras.callbacks.History at 0x7a7ea01d91b0>
```

```
test_loss, test_acc = model.evaluate(test_images, test_labels,
                                     verbose=2)
```

```
print('\nTest accuracy:', test_acc)
```

```
313/313 - 1s - loss: 0.3358 - accuracy: 0.8806 - 1s/epoch - 4ms/step
```

Test accuracy: 0.8805999755859375

[illegible]

```

predictions = probability_model.predict(test_images)

313/313 [=====] - 1s 3ms/step

predictions[0]

array([6.9260415e-07, 7.0697497e-14, 5.6892759e-12, 1.6533005e-12,
       5.0840594e-09, 5.9169839e-04, 1.0581412e-07, 1.5398487e-02,
       5.5460696e-08, 9.8400903e-01], dtype=float32)

np.argmax(predictions[0])

9

test_labels[0]

9

def plot_image(i, predictions_array, true_label, img):
    true_label, img = true_label[i], img[i]
    plt.grid(False)
    plt.xticks([])
    plt.yticks([])

    plt.imshow(img, cmap=plt.cm.binary)

    predicted_label = np.argmax(predictions_array)
    if predicted_label == true_label:
        color = 'blue'
    else:
        color = 'red'

    plt.xlabel("{} {:2.0f}% ({})".format(class_names[predicted_label],
                                         100*np.max(predictions_array),
                                         class_names[true_label]),
              color=color)

def plot_value_array(i, predictions_array, true_label):
    true_label = true_label[i]
    plt.grid(False)
    plt.xticks(range(10))
    plt.yticks([])
    thisplot = plt.bar(range(10), predictions_array, color="#777777")
    plt.ylim([0, 1])
    predicted_label = np.argmax(predictions_array)

    thisplot[predicted_label].set_color('red')
    thisplot[true_label].set_color('blue')

i = 0
plt.figure(figsize=(6,3))
plt.subplot(1,2,1)

```

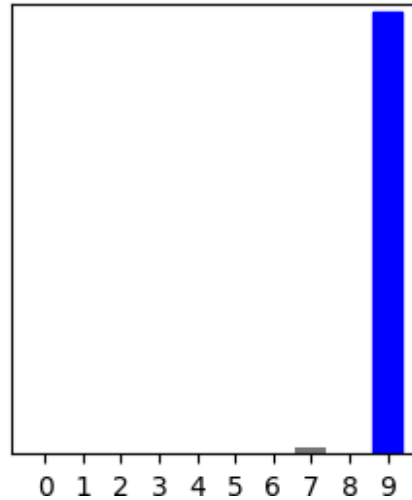
```

plot_image(i, predictions[i], test_labels, test_images)
plt.subplot(1,2,2)
plot_value_array(i, predictions[i], test_labels)
plt.show()

```



Ankle boot 98% (Ankle boot)



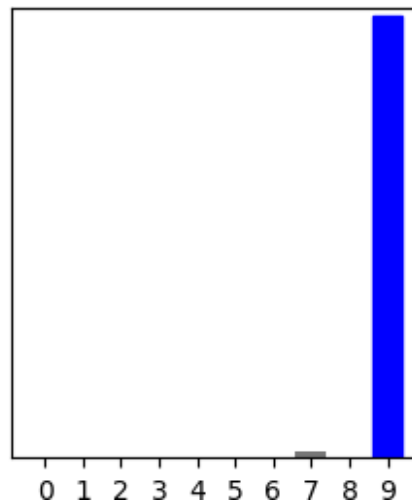
```

i = 0
plt.figure(figsize=(6,3))
plt.subplot(1,2,1)
plot_image(i, predictions[i], test_labels, test_images)
plt.subplot(1,2,2)
plot_value_array(i, predictions[i], test_labels)
plt.show()

```



Ankle boot 98% (Ankle boot)



```

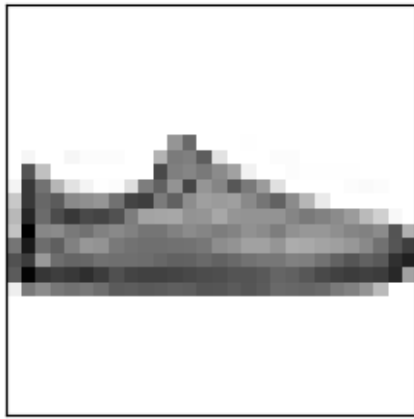
i = 12
plt.figure(figsize=(6,3))
plt.subplot(1,2,1)

```

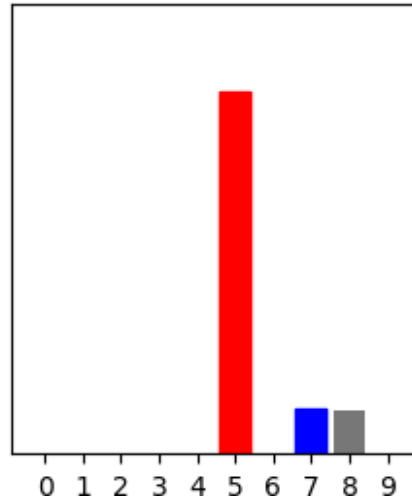
```

plot_image(i, predictions[i], test_labels, test_images)
plt.subplot(1,2,2)
plot_value_array(i, predictions[i], test_labels)
plt.show()

```



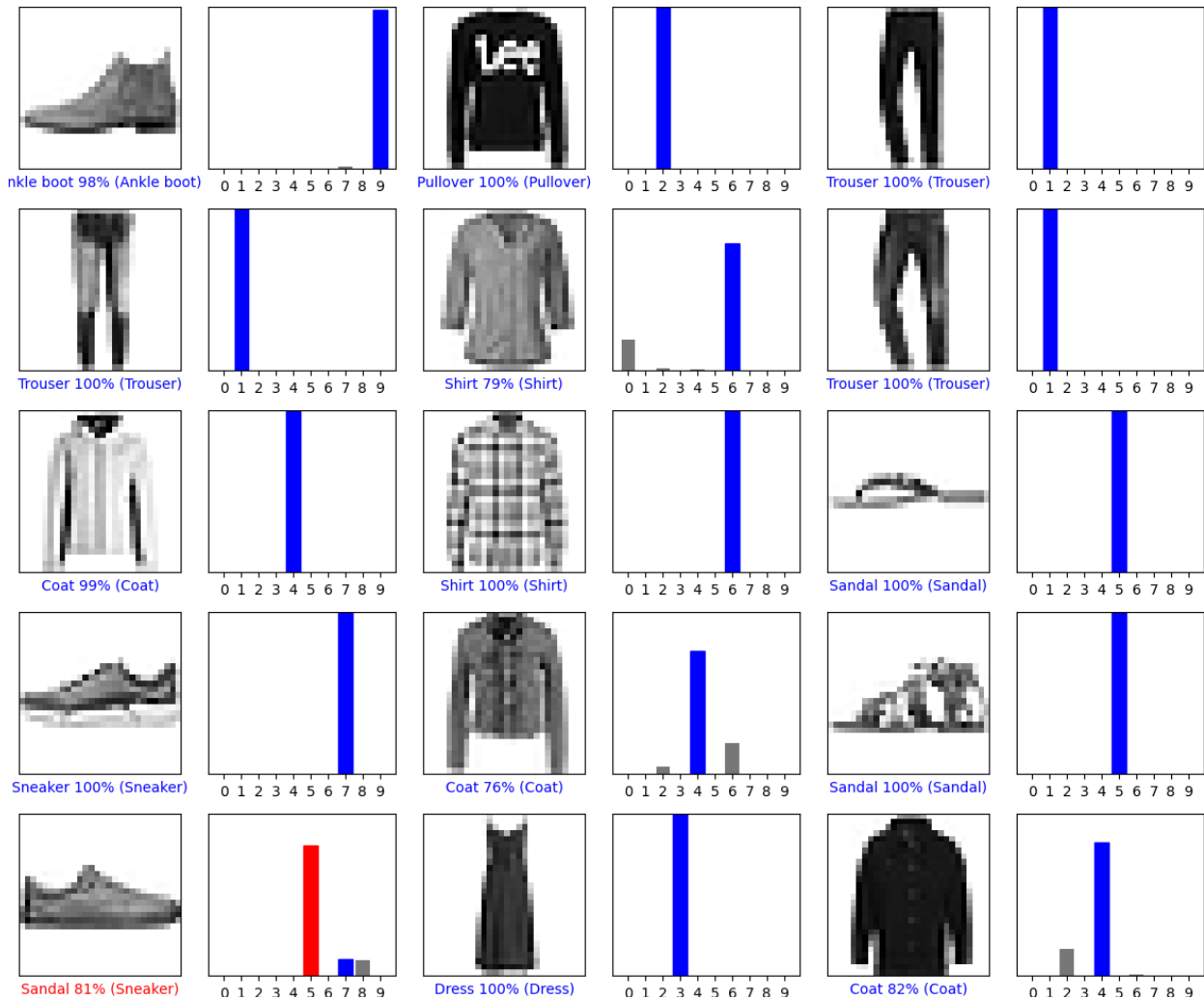
Sandal 81% (Sneaker)



```

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

```



```
# Grab an image from the test dataset.
img = test_images[1]

print(img.shape)

(28, 28)

# Add the image to a batch where it's the only member.
img = (np.expand_dims(img,0))

print(img.shape)

(1, 28, 28)

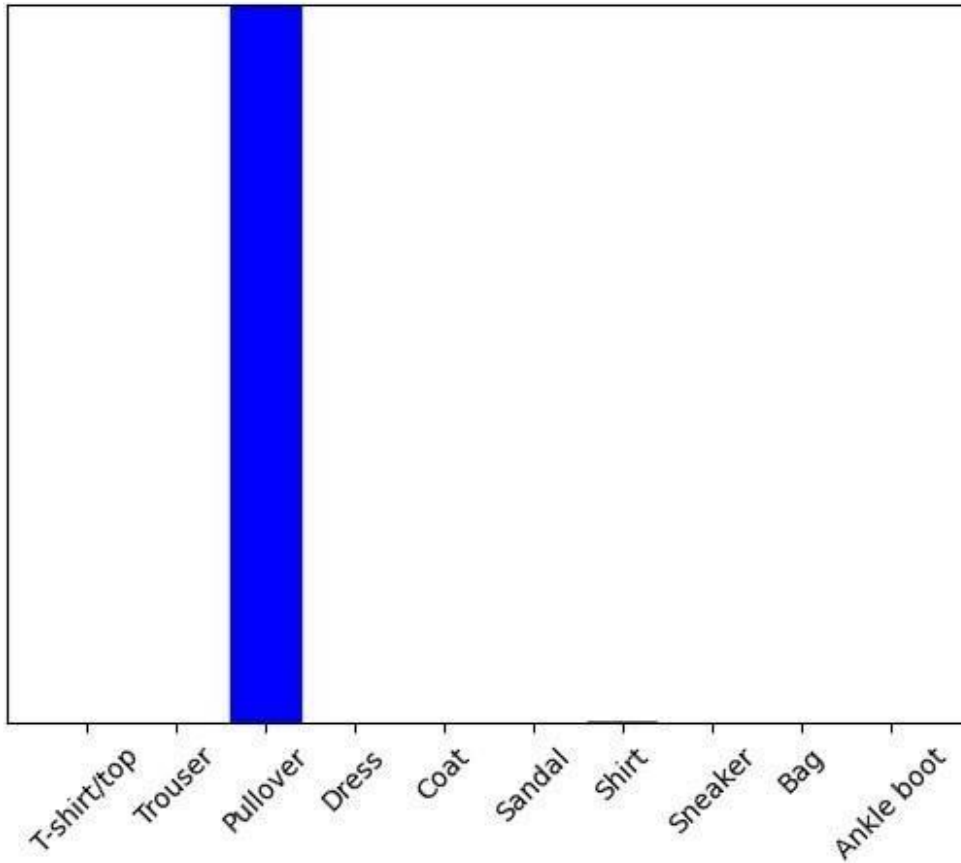
predictions_single = probability_model.predict(img)

print(predictions_single)
```



```
1/1 [=====] - 0s 24ms/step
[[7.1821800e-05 8.6609013e-16 9.9861825e-01 1.6916174e-10 2.3175428e-
04
 1.1837163e-12 1.0782134e-03 4.4021377e-11 9.3415310e-12 2.7698164e-
09]]

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



```
np.argmax(predictions_single[0])
2
```

**Results and Conclusion -** By following the above steps and learning about Keras, TensorFlow (or PyTorch), Google Colab, Kaggle, GitHub, and Git, you have gained essential skills for setting up environments, experimenting with deep learning libraries, utilizing GPU/TPU, collaborating on code projects, and version controlling your codebase. This knowledge empowers you to efficiently work on data science and machine learning projects, and you're now better equipped to dive into more complex tasks and explore advanced techniques.