

```
In [41]: #Assignment No: 3  
#Title: To Implement Image classification model using CNN Deep Learning Arch  
#Dataset: fashion_mnist
```

```
In [3]: from tensorflow import keras  
import matplotlib.pyplot as plt  
import numpy as np  
import pandas as pd  
from sklearn.metrics import classification_report
```

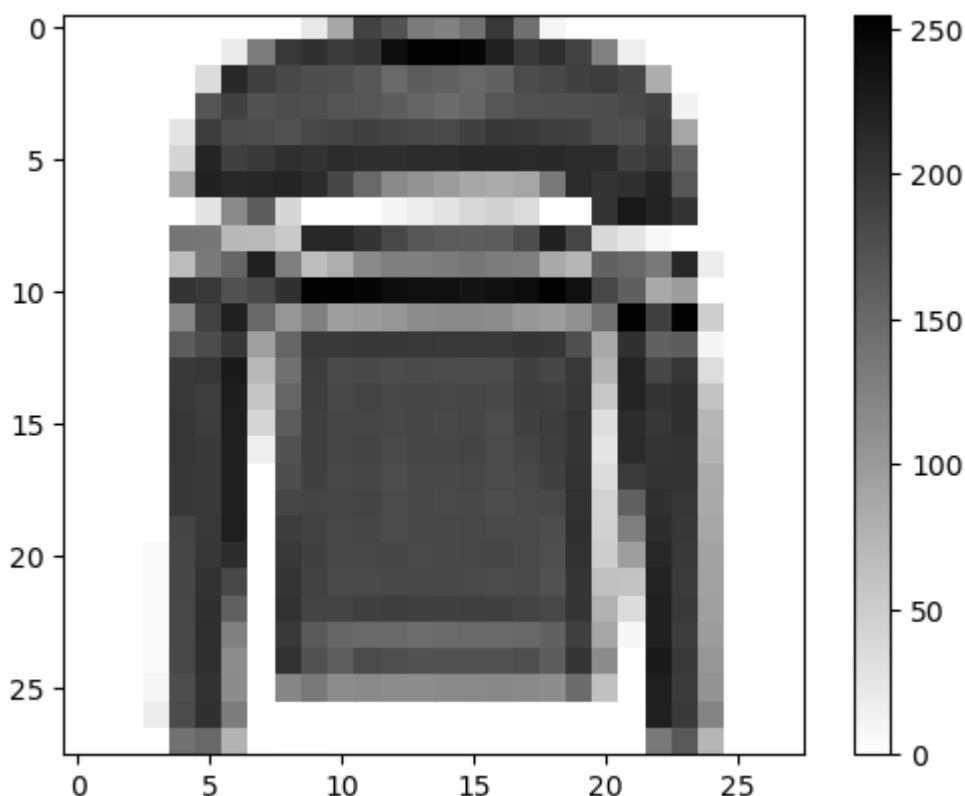
```
In [5]: fashion_mnist = keras.datasets.fashion_mnist  
(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()
```

```
In [6]: class_names = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat',  
                        'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle boot']
```

```
In [7]: print(train_images.shape)  
print(train_labels.shape)  
print(test_images.shape)  
print(test_labels.shape)
```

```
(60000, 28, 28)  
(60000,)  
(10000, 28, 28)  
(10000,)
```

```
In [8]: plt.figure()  
plt.imshow(train_images[5], cmap=plt.cm.binary)  
plt.colorbar()  
plt.grid(False)  
plt.show()
```



```
In [9]: 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()
```



```
In [10]: train_images = train_images / 255.0
test_images = test_images / 255.0
```

```
In [11]: print("train_images: ", train_images.shape)
print("test_images: ", test_images.shape)
```

```
train_images: (60000, 28, 28)
test_images: (10000, 28, 28)
```

```
In [12]: model = keras.Sequential([
    keras.layers.Conv2D(filters=64, kernel_size=3, activation='relu', padding='same'),
    keras.layers.MaxPool2D(pool_size=2),
    keras.layers.Conv2D(filters=128, kernel_size=3, activation='relu', padding='same'),
    keras.layers.MaxPool2D(pool_size=2),
    keras.layers.Flatten(),
    keras.layers.Dense(128, activation='relu'),
    keras.layers.Dropout(0.25),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dropout(0.25),
    keras.layers.Dense(10, activation='softmax')
])
```


C:\Users\yadav\anaconda3\Lib\site-packages\keras\src\layers\convolutional\base\_conv.py:107: UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```


```
In [13]: model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
```

In [14]: `model_history = model.fit(train_images, train_labels, batch_size=50, epochs=`

Epoch 1/10


**840/840**  **116s** 132ms/step - accuracy: 0.7107 - loss: 0.8085 - val\_accuracy: 0.8691 - val\_loss: 0.3588

Epoch 2/10


**840/840**  **54s** 113ms/step - accuracy: 0.8641 - loss: 0.3885

C:\Users\yadav\anaconda3\Lib\contextlib.py:158: UserWarning: Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps\_per\_epoch \* epochs` batches. You may need to use the `.repeat()` function when building your dataset.


self.gen.throw(typ, value, traceback)

**840/840**  **49s** 58ms/step - accuracy: 0.8654 - loss: 0.3773 - val\_accuracy: 0.8821 - val\_loss: 0.3264


Epoch 3/10

**840/840**  **147s** 124ms/step - accuracy: 0.8775 - loss: 0.3376 - val\_accuracy: 0.8980 - val\_loss: 0.2839


Epoch 4/10

**840/840**  **52s** 62ms/step - accuracy: 0.8943 - loss: 0.2879 - val\_accuracy: 0.8997 - val\_loss: 0.2792


Epoch 5/10

**840/840**  **102s** 122ms/step - accuracy: 0.9073 - loss: 0.2615 - val\_accuracy: 0.9091 - val\_loss: 0.2546


Epoch 6/10

**840/840**  **49s** 58ms/step - accuracy: 0.9047 - loss: 0.2595 - val\_accuracy: 0.9058 - val\_loss: 0.2526


Epoch 7/10

**840/840**  **155s** 124ms/step - accuracy: 0.9153 - loss: 0.2336 - val\_accuracy: 0.9089 - val\_loss: 0.2397


Epoch 8/10

**840/840**  **49s** 59ms/step - accuracy: 0.9190 - loss: 0.2177 - val\_accuracy: 0.9133 - val\_loss: 0.2464

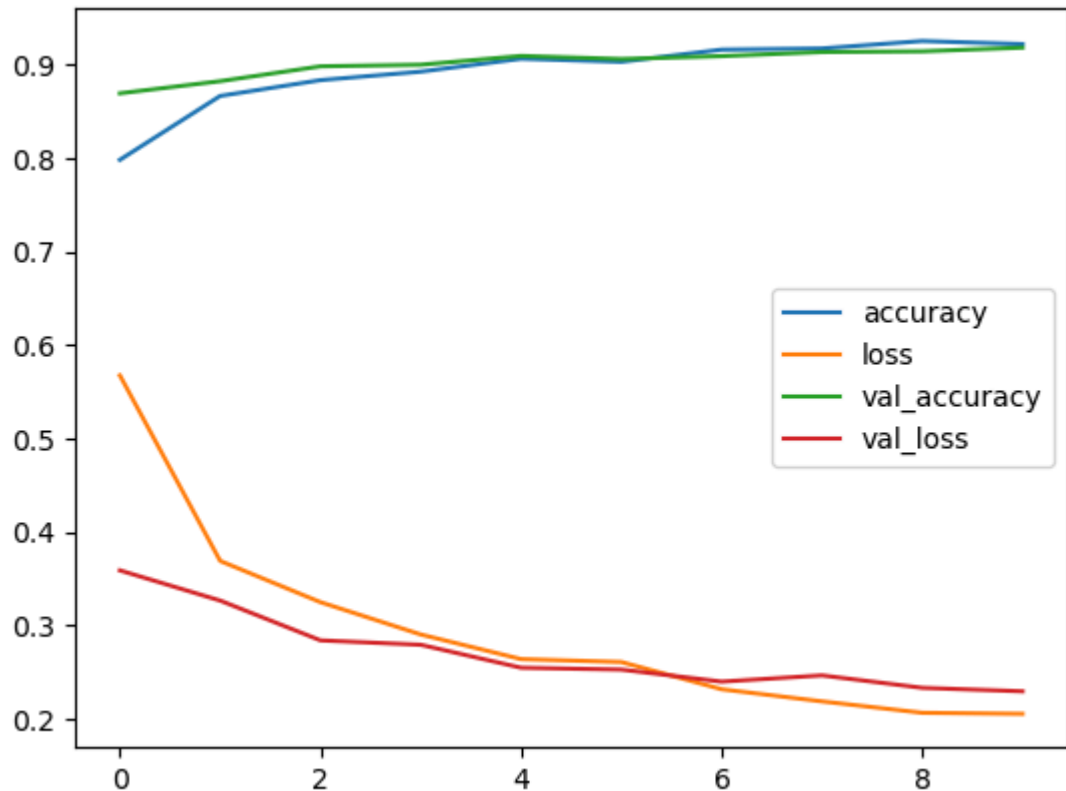
Epoch 9/10

**840/840**  **143s** 112ms/step - accuracy: 0.9248 - loss: 0.2095 - val\_accuracy: 0.9140 - val\_loss: 0.2331

Epoch 10/10

**840/840**  **49s** 58ms/step - accuracy: 0.9232 - loss: 0.2029 - val\_accuracy: 0.9180 - val\_loss: 0.2294

```
In [16]: pd.DataFrame(model_history.history).plot()
plt.show()
```



```
In [18]: test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)
print('\nTest accuracy:', test_acc)
```

313/313 - 7s - 24ms/step - accuracy: 0.9180 - loss: 0.2294

Test accuracy: 0.9179999828338623

```
In [19]: predictions = model.predict(test_images)
```

313/313 ————— 8s 25ms/step

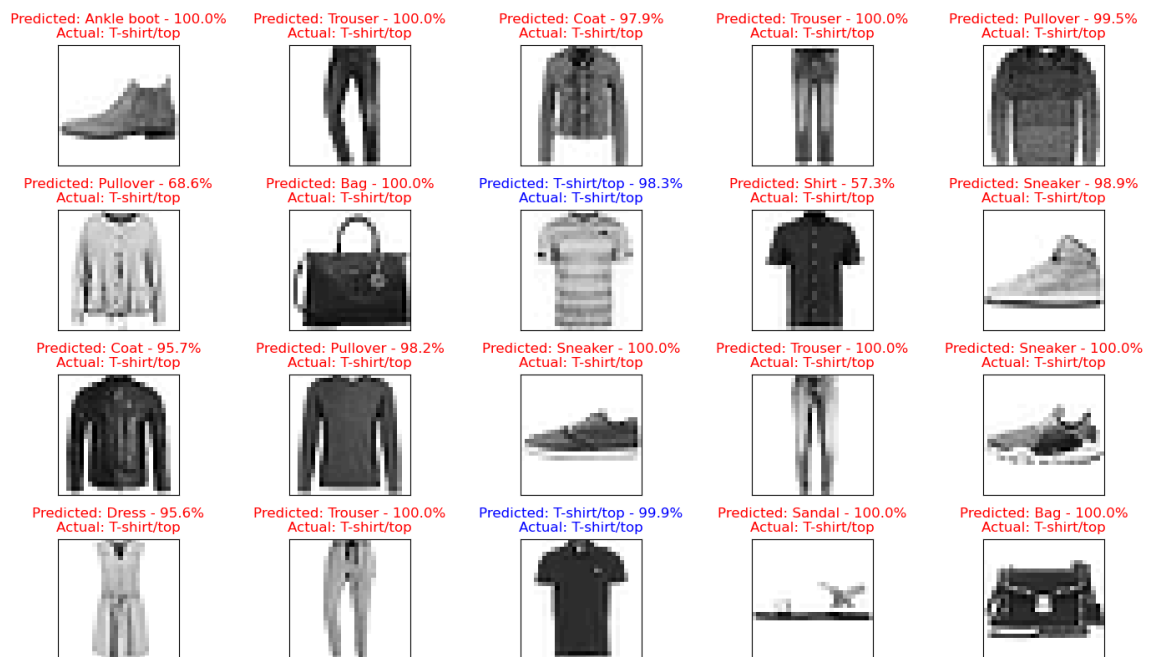
```
In [21]: def plot_img_label(img, pred_class, pred_percentage, true_class):
plt.imshow(img, cmap=plt.cm.binary)
if pred_class == true_class:
    color = 'blue'
else:
    color='red'
plt.title(label= f"Predicted: {pred_class} - {pred_percentage:2.1f}%\nAc
fontdict={'color': color})
```

```
In [23]: plt.figure(figsize=(14,10))

for i in range(20):
    plt.subplot(5,5,i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    i=i*5
    img = test_images[i].reshape(28,28)
    pred_class = class_names[np.argmax(predictions[i])]
    pred_percentage = np.max(predictions[i])*100
    true_class = class_names[np.argmax(test_labels[i])]

    plot_img_label(img, pred_class, pred_percentage, true_class)

plt.tight_layout()
plt.show()
```



```
In [32]: predicted_label = np.argmax(predictions, axis=1) if predictions.ndim > 1 else
true_label = np.argmax(test_labels, axis=1) if test_labels.ndim > 1 else tes
```

```
In [33]: print(classification_report(true_label, predicted_label, target_names=class_
```

	precision	recall	f1-score	support
T-shirt/top	0.89	0.85	0.87	1000
Trouser	0.99	0.98	0.99	1000
Pullover	0.89	0.84	0.86	1000
Dress	0.89	0.94	0.92	1000
Coat	0.83	0.90	0.87	1000
Sandal	0.99	0.98	0.99	1000
Shirt	0.77	0.75	0.76	1000
Sneaker	0.96	0.98	0.97	1000
Bag	0.99	0.98	0.99	1000
Ankle boot	0.98	0.97	0.97	1000
accuracy			0.92	10000
macro avg	0.92	0.92	0.92	10000
weighted avg	0.92	0.92	0.92	10000