## DL Lab Exp No.3

```
[1]: import tensorflow as tf
     from sklearn.model_selection import train_test_split
     from mlxtend.plotting import plot_confusion_matrix
     from sklearn import metrics
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     %matplotlib inline
     from tqdm.notebook import tqdm
     import random
     import warnings
     warnings.filterwarnings("ignore")
[2]: (trainX, trainY), (testX, testY) = tf.keras.datasets.fashion_mnist.load_data()
     trainX = trainX.reshape((trainX.shape[0], 28, 28, 1))
     testX = testX.reshape((testX.shape[0], 28, 28, 1))
     trainY_cat = tf.keras.utils.to_categorical(trainY)
     testY_cat = tf.keras.utils.to_categorical(testY)
[3]: train_norm = trainX.astype('float32')
     test_norm = testX.astype('float32')
     train_norm = train_norm / 255.0
     test_norm = test_norm / 255.0
[4]: class_names = ['T-shirt/top', 'Trouser', 'Pullover', 'Dress', 'Coat',
                    'Sandal', 'Shirt', 'Sneaker', 'Bag', 'Ankle boot']
[5]: plt.figure(figsize=(10,10))
     for i in range(25):
         plt.subplot(5,5,i+1)
         plt.xticks([])
         plt.yticks([])
         plt.grid(False)
```

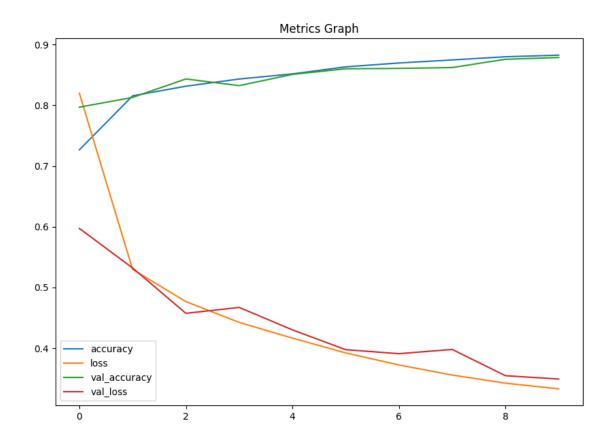
```
plt.imshow(trainX[i], cmap=plt.cm.binary)
   plt.xlabel(class_names[trainY[i]])
plt.show()
```



```
tf.keras.layers.Dense(len(class_names), activation="softmax",_
       ⇔name="output-layer")
      ])
 [7]: model.compile(loss="categorical_crossentropy", optimizer="adam", ___
       →metrics=["accuracy"])
 [8]: history = model.fit(trainX, trainY_cat, epochs=10, validation_data=(testX,__
       →testY_cat))
     Epoch 1/10
     1875/1875
                           80s 41ms/step -
     accuracy: 0.6166 - loss: 1.2695 - val_accuracy: 0.7968 - val_loss: 0.5972
     Epoch 2/10
     1875/1875
                           48s 25ms/step -
     accuracy: 0.8084 - loss: 0.5460 - val accuracy: 0.8129 - val loss: 0.5320
     Epoch 3/10
     1875/1875
                           46s 25ms/step -
     accuracy: 0.8274 - loss: 0.4839 - val_accuracy: 0.8433 - val_loss: 0.4574
     Epoch 4/10
     1875/1875
                           47s 25ms/step -
     accuracy: 0.8415 - loss: 0.4441 - val_accuracy: 0.8323 - val_loss: 0.4671
     Epoch 5/10
     1875/1875
                           47s 25ms/step -
     accuracy: 0.8473 - loss: 0.4259 - val_accuracy: 0.8508 - val_loss: 0.4301
     Epoch 6/10
                           56s 30ms/step -
     1875/1875
     accuracy: 0.8624 - loss: 0.3952 - val_accuracy: 0.8601 - val_loss: 0.3975
     Epoch 7/10
     1875/1875
                           49s 26ms/step -
     accuracy: 0.8679 - loss: 0.3760 - val accuracy: 0.8607 - val loss: 0.3909
     Epoch 8/10
     1875/1875
                           51s 27ms/step -
     accuracy: 0.8735 - loss: 0.3594 - val_accuracy: 0.8621 - val_loss: 0.3979
     Epoch 9/10
     1875/1875
                           53s 28ms/step -
     accuracy: 0.8766 - loss: 0.3504 - val_accuracy: 0.8757 - val_loss: 0.3546
     Epoch 10/10
     1875/1875
                           45s 24ms/step -
     accuracy: 0.8834 - loss: 0.3331 - val_accuracy: 0.8786 - val_loss: 0.3491
 [9]: tf.keras.utils.plot_model(model, show_shapes=True)
     You must install graphviz (see instructions at
     https://graphviz.gitlab.io/download/) for `plot model` to work.
[10]: model.summary()
```

## Model: "sequential"

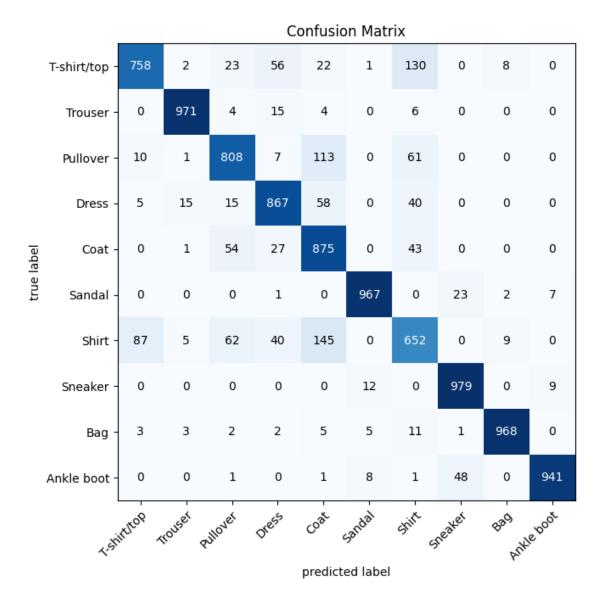
```
Layer (type)
                                              Output Shape
                                                                                   Ш
      →Param #
      conv-layer-1 (Conv2D)
                                              (None, 28, 28, 64)
                                                                                       Ш
      ⇔640
      pooling-layer-1 (AveragePooling2D)
                                             (None, 14, 14, 64)
                                                                                       ш
      conv-layer-2 (Conv2D)
                                              (None, 14, 14, 32)
                                                                                    ш
      ⊶18,464
      pooling-layer-2 (AveragePooling2D)
                                              (None, 7, 7, 32)
      pooling-layer-3
                                              (None, 32)
                                                                                       Ш
      → 0
       (GlobalAveragePooling2D)
                                                                                       ш
      output-layer (Dense)
                                              (None, 10)
                                                                                       Ш
      ⇔330
      Total params: 58,304 (227.75 KB)
      Trainable params: 19,434 (75.91 KB)
      Non-trainable params: 0 (0.00 B)
      Optimizer params: 38,870 (151.84 KB)
[11]: pd.DataFrame(history.history).plot(figsize=(10,7))
      plt.title("Metrics Graph")
      plt.show()
```



```
[12]: model.evaluate(testX, testY_cat)
     313/313
                         3s 9ms/step -
     accuracy: 0.8810 - loss: 0.3418
[12]: [0.3491450846195221, 0.878600001335144]
[13]: predictions = model.predict(testX)
     313/313
                         3s 7ms/step
[14]: predictions = tf.argmax(predictions, axis=1)
[15]: tf.keras.utils.plot_model(model, show_shapes=True)
     You must install graphviz (see instructions at
     https://graphviz.gitlab.io/download/) for `plot_model` to work.
[16]: model.evaluate(testX, testY_cat)
     313/313
                         3s 10ms/step -
     accuracy: 0.8810 - loss: 0.3418
```

```
[16]: [0.3491450846195221, 0.878600001335144]
[17]: predictions = model.predict(testX)
      predictions = tf.argmax(predictions, axis=1)
      y_test = tf.argmax(testY_cat, axis=1)
      y_test = tf.Variable(y_test)
     313/313
                         2s 8ms/step
[18]: print("Accuracy: ",metrics.accuracy_score(y_test, predictions))
     Accuracy: 0.8786
[19]: print(metrics.classification_report(y_test, predictions))
                   precision
                                 recall f1-score
                                                    support
                0
                         0.88
                                   0.76
                                                        1000
                                             0.81
                                   0.97
                                                        1000
                1
                         0.97
                                             0.97
                2
                         0.83
                                   0.81
                                             0.82
                                                        1000
                3
                         0.85
                                   0.87
                                             0.86
                                                        1000
                4
                        0.72
                                   0.88
                                             0.79
                                                        1000
                5
                        0.97
                                   0.97
                                             0.97
                                                        1000
                6
                                   0.65
                                                        1000
                        0.69
                                             0.67
                7
                        0.93
                                   0.98
                                             0.95
                                                        1000
                8
                         0.98
                                   0.97
                                             0.97
                                                        1000
                9
                         0.98
                                   0.94
                                             0.96
                                                        1000
                                             0.88
                                                       10000
         accuracy
        macro avg
                         0.88
                                   0.88
                                             0.88
                                                       10000
                                   0.88
                                             0.88
                                                       10000
     weighted avg
                         0.88
[20]: cm = metrics.confusion_matrix(y_test, predictions)
      plot_confusion_matrix(cm, figsize=(10,7), class_names=class_names)
      plt.title("Confusion Matrix")
```

plt.show()



```
[21]: images = []
labels = []
random_indices = random.sample(range(len(testX)), 10)
for idx in random_indices:
    images.append(testX[idx])
    labels.append(testY_cat[idx])
images = np.array(images)
labels = np.array(labels)

fig = plt.figure(figsize=(20, 8))
rows = 2
cols = 5
```

```
x = 1
for image, label in zip(images, labels):
    fig.add_subplot(rows, cols, x)
    prediction = model.predict(tf.expand_dims(image, axis=0))
    prediction = class_names[tf.argmax(prediction.flatten())]
    label = class_names[tf.argmax(label)]
    plt.title(f"Label: {label}, Prediction: {prediction}")
    plt.imshow(image/255.)
    plt.axis("off")
    x += 1
```

```
1/1
                0s 108ms/step
1/1
                Os 101ms/step
1/1
                Os 99ms/step
1/1
                Os 160ms/step
1/1
                Os 124ms/step
1/1
                Os 136ms/step
1/1
                Os 101ms/step
                0s 106ms/step
1/1
1/1
                Os 94ms/step
1/1
                Os 93ms/step
```

