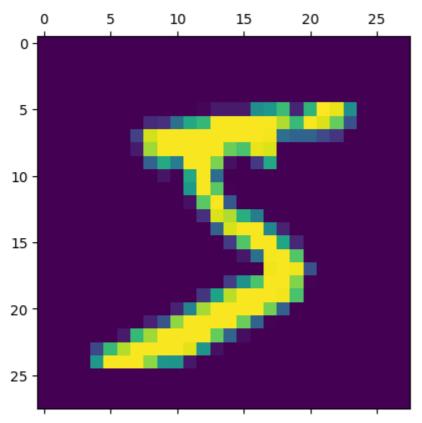
Experiment-12

12.consider mnist dataset preprocess it and build a simple sequantial Neural Network model on it, test and evaluate the model using different metrics

Handwritten digits classification using neural network

In this notebook we will classify handwritten digits using a simple neural network which has only input and output layers. We will than add a hidden layer and see how the performance of the model improves



```
In [5]: X_train = X_train / 255
X_test = X_test / 255

In [6]: X_train.shape,X_test.shape
Out[6]: ((60000, 28, 28), (10000, 28, 28))

In [7]: X_train_flattened = X_train.reshape(len(X_train), 28*28)
X_test_flattened = X_test.reshape(len(X_test), 28*28)

In [8]: X_train_flattened.shape,X_test_flattened.shape
Out[8]: ((60000, 784), (10000, 784))
```

Very simple neural network with no hidden layers

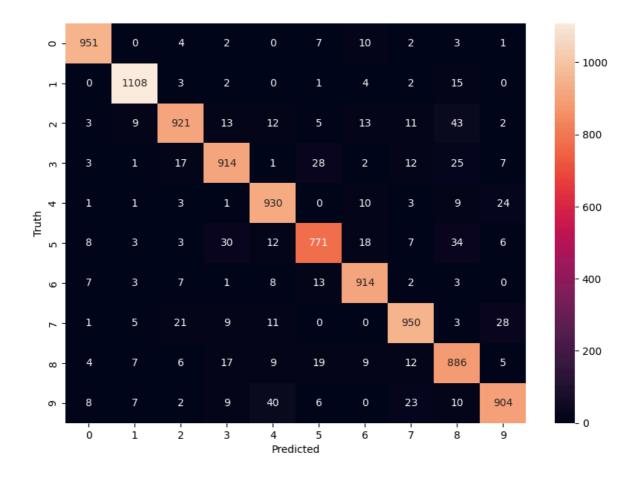
No description has been provided for this image

```
C:\Users\Guru Kiran\AppData\Local\Programs\Python\Python313\Lib\site-packages\ker
        as\src\layers\core\dense.py:92: UserWarning: Do not pass an `input_shape`/`input_
        dim` argument to a layer. When using Sequential models, prefer using an `Input(sh
        ape)` object as the first layer in the model instead.
         super().__init__(activity_regularizer=activity_regularizer, **kwargs)
        Epoch 1/5
        1875/1875
                                     — 8s 4ms/step - accuracy: 0.8766 - loss: 0.4714
        Epoch 2/5
                                     - 6s 3ms/step - accuracy: 0.9146 - loss: 0.3044
        1875/1875
        Epoch 3/5
        1875/1875
                                      - 10s 3ms/step - accuracy: 0.9204 - loss: 0.2836
        Epoch 4/5
        1875/1875
                                      - 11s 3ms/step - accuracy: 0.9240 - loss: 0.2732
        Epoch 5/5
        1875/1875
                                     - 10s 3ms/step - accuracy: 0.9251 - loss: 0.2670
Out[9]: <keras.src.callbacks.history.History at 0x1ac5004c050>
In [10]: model.evaluate(X_test_flattened, y_test)
        313/313 -
                                  - 1s 3ms/step - accuracy: 0.9249 - loss: 0.2674
Out[10]: [0.26744475960731506, 0.9248999953269958]
In [11]: y_predicted = model.predict(X_test_flattened)
         y_predicted[0]
        313/313 -
                                   - 1s 2ms/step
Out[11]: array([2.31749155e-02, 4.47567004e-07, 4.52204160e-02, 9.54810023e-01,
                 3.92512884e-03, 9.99573916e-02, 2.21967275e-06, 9.99806225e-01,
                 1.19874306e-01, 6.88449085e-01], dtype=float32)
In [12]: plt.matshow(X_test[0])
         plt.show()
                                                           25
             0
                      5
                               10
                                        15
                                                  20
          0
         5 ·
        10 -
        15 -
        20 -
```

25 -

np.argmax finds a maximum element from an array and returns the index of it

```
In [13]: np.argmax(y_predicted[0])
Out[13]: np.int64(7)
In [14]: y_predicted_labels = [np.argmax(i) for i in y_predicted]
In [15]: y_predicted_labels[:5]
Out[15]: [np.int64(7), np.int64(2), np.int64(1), np.int64(0), np.int64(4)]
In [16]: cm = tf.math.confusion_matrix(labels=y_test,predictions=y_predicted_labels)
        \mathsf{cm}
Out[16]: <tf.Tensor: shape=(10, 10), dtype=int32, numpy=
                                                     2,
                                                          3,
        array([[ 951,
                      0, 4, 2,
                                      0, 7, 10,
                                                               1],
                 0, 1108,
                          3,
                               2,
                                    0, 1, 4,
                                                    2,
                                                         15,
                                                               0],
              [
                 3,
                     9, 921,
                                         5, 13,
              13, 12,
                                                    11,
                                                         43,
                                                               2],
                      1, 17, 914,
                                          28,
                                    1,
              3,
                                               2,
                                                    12,
                                                         25,
                                                               7],
              1,
                     1, 3, 1, 930, 0, 10, 3, 9,
                                                              24],
                 8,
                     3, 3, 30, 12, 771, 18,
                                                    7,
                                                         34,
                                                              6],
              7,
                     3,
                          7, 1,
                                    8, 13, 914,
                                                    2,
                                                          3,
                                                               0],
              9, 11,
                                          0, 0, 950,
                      5, 21,
              1,
                                                          3,
                                                              28],
                      7, 6, 17, 9, 19, 9, 12, 886,
                                                               5],
                 4,
                     7,
                          2, 9, 40, 6, 0,
                                                    23, 10, 904]],
              8,
             dtype=int32)>
In [17]: import seaborn as sn
        plt.figure(figsize = (10,7))
        sn.heatmap(cm, annot=True, fmt='d')
        plt.xlabel('Predicted')
        plt.ylabel('Truth')
        plt.show()
```



Using hidden layer

```
In [18]:
        model = keras.Sequential([
             keras.layers.Dense(100, input_shape=(784,), activation='relu'),
             keras.layers.Dense(10, activation='sigmoid')
         1)
         model.compile(optimizer='adam',
                       loss='sparse_categorical_crossentropy',
                       metrics=['accuracy'])
         model.fit(X_train_flattened, y_train, epochs=5)
        C:\Users\Guru Kiran\AppData\Local\Programs\Python\Python313\Lib\site-packages\ker
        as\src\layers\core\dense.py:92: UserWarning: Do not pass an `input_shape`/`input_
        dim` argument to a layer. When using Sequential models, prefer using an `Input(sh
        ape)` object as the first layer in the model instead.
          super().__init__(activity_regularizer=activity_regularizer, **kwargs)
        Epoch 1/5
        1875/1875
                                      - 13s 6ms/step - accuracy: 0.9232 - loss: 0.2710
        Epoch 2/5
                                       11s 6ms/step - accuracy: 0.9636 - loss: 0.1229
        1875/1875
        Epoch 3/5
        1875/1875
                                       21s 6ms/step - accuracy: 0.9731 - loss: 0.0869
        Epoch 4/5
        1875/1875
                                       21s 6ms/step - accuracy: 0.9794 - loss: 0.0656
        Epoch 5/5
                                      - 20s 6ms/step - accuracy: 0.9844 - loss: 0.0515
        1875/1875
Out[18]: <keras.src.callbacks.history.History at 0x1ac53205090>
In [19]: model.evaluate(X_test_flattened,y_test)
```

```
- 1s 4ms/step - accuracy: 0.9770 - loss: 0.0763
Out[19]: [0.07625898718833923, 0.9769999980926514]
In [20]:
          y_predicted = model.predict(X_test_flattened)
          y_predicted_labels = [np.argmax(i) for i in y_predicted]
          cm = tf.math.confusion_matrix(labels=y_test,predictions=y_predicted_labels)
          plt.figure(figsize = (10,7))
          sn.heatmap(cm, annot=True, fmt='d')
          plt.xlabel('Predicted')
          plt.ylabel('Truth')
          plt.show()
         313/313
                                        1s 3ms/step
                               1
                                             0
                                                    1
                974
                        0
                                                                                            - 1000
                      1123
                               4
                                                    0
                 0
                                      0
                                             0
                                                                  0
                              1005
                 3
                        1
                                             3
                                                    0
                                                           2
                                                                  5
                                                                         8
                                                                                 1
                                                                                            - 800
                        0
                                     978
                                             2
                        0
                                      0
                                            959
                                                    0
                                                           3
                                                                  1
                                                                         1
                                                                                12
                                                                                            - 600
                                                           5
                 3
                        0
                               0
                                             2
                                                                                 2
                                                                                            - 400
                                                    4
                                                          940
                                                                  0
           9
                                                    0
                                                                         2
                              11
                                      2
                                             3
                                                           0
                                                                 997
                                                                                 9
                                                                                            - 200
                        0
                               3
                                      5
                                             3
                                                    3
                                                           0
                                                                  1
                                                                                 3
                 1
                                                    3
                                                                                974
```

Using Flatten layer so that we don't have to call .reshape on input dataset

Predicted

ò

C:\Users\Guru Kiran\AppData\Local\Programs\Python\Python313\Lib\site-packages\ker as\src\layers\reshaping\flatten.py:37: UserWarning: Do not pass an `input_shape`/ `input_dim` argument to a layer. When using Sequential models, prefer using an `I nput(shape)` object as the first layer in the model instead.

```
super().__init__(**kwargs)
```

```
Epoch 1/10
                            -- 13s 6ms/step - accuracy: 0.9218 - loss: 0.2747
1875/1875 •
Epoch 2/10
                             - 11s 6ms/step - accuracy: 0.9639 - loss: 0.1244
1875/1875 -
Epoch 3/10
                             - 20s 6ms/step - accuracy: 0.9739 - loss: 0.0865
1875/1875
Epoch 4/10
1875/1875 -
                             - 13s 7ms/step - accuracy: 0.9806 - loss: 0.0651
Epoch 5/10
1875/1875 -
                             - 19s 6ms/step - accuracy: 0.9840 - loss: 0.0519
Epoch 6/10
                           21s 6ms/step - accuracy: 0.9866 - loss: 0.0420
1875/1875 -
Epoch 7/10
                           20s 6ms/step - accuracy: 0.9887 - loss: 0.0351
1875/1875 -
Epoch 8/10
1875/1875 -
                             - 20s 6ms/step - accuracy: 0.9912 - loss: 0.0283
Epoch 9/10
                             - 21s 6ms/step - accuracy: 0.9926 - loss: 0.0239
1875/1875 -
Epoch 10/10
1875/1875
                           21s 6ms/step - accuracy: 0.9938 - loss: 0.0202
```

Out[21]: <keras.src.callbacks.history.History at 0x1ac71ef0910>

```
In [22]: model.evaluate(X_test,y_test)
```

Out[22]: [0.0869947299361229, 0.9753999710083008]