### Import necessary packages

```
import tensorflow as tf
from tensorflow.keras import layers,models
from tensorflow import keras
import numpy as np
import matplotlib.pyplot as plt
```

### Loading dataset

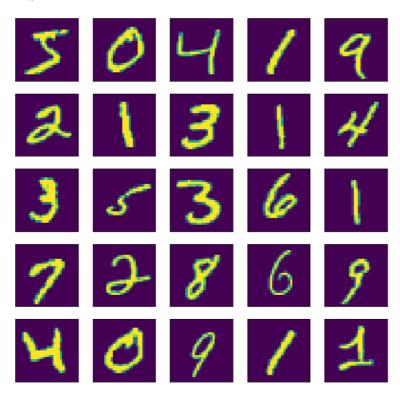
```
(X_train, y_train) , (X_test, y_test) = keras.datasets.mnist.load_data()
```

## Scaling the train and test dataset

```
X_{train} = X_{train} / 255
X_{\text{test}} = X_{\text{test}} / 255
X_train[0]
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             0.99215686, 0.99215686, 0.99215686, 0.88235294, 0.6745098,
             0.99215686, 0.94901961, 0.76470588, 0.25098039, 0.
            0. , 0. , 0. ],
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             0.99215686, 0.99215686, 0.98431373, 0.36470588, 0.32156863,
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             0.32156863, 0.21960784, 0.15294118, 0.
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```
a 16862715 a 60392157 a
```

```
X_train.shape
     (60000, 28, 28)
X_train = X_train.reshape(-1,28,28,1)
X_{train.shape}
     (60000, 28, 28, 1)
X_test = X_test.reshape(-1,28,28,1)
X test.shape
     (10000, 28, 28, 1)
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(X_train[i])
plt.show()
```



# Creating and Training a Convolutional Neural Network

```
convolutional_neural_network = models.Sequential([
    layers.Conv2D(filters=25, kernel_size=(3, 3), activation='relu', input_shape=(28,28,1)),
    layers.MaxPooling2D((2, 2)),

    layers.Conv2D(filters=64, kernel_size=(3, 3), activation='relu'),
    layers.MaxPooling2D((2, 2)),

    layers.MaxPooling2D((2, 2)),

    layers.Flatten(),
    layers.Platten(),
    layers.Dense(64, activation='relu'),
    layers.Dense(10, activation='softmax')
])

convolutional_neural_network.summary()

Model: "sequential_1"

Layer (type) Output Shape Param #
```

```
conv2d 3 (Conv2D)
                            (None, 26, 26, 25)
                                                        250
 max_pooling2d_3 (MaxPooling (None, 13, 13, 25)
 conv2d 4 (Conv2D)
                             (None, 11, 11, 64)
                                                       14464
 max_pooling2d_4 (MaxPooling (None, 5, 5, 64)
 conv2d 5 (Conv2D)
                                                        36928
                             (None, 3, 3, 64)
 max_pooling2d_5 (MaxPooling (None, 1, 1, 64)
 flatten 1 (Flatten)
                             (None, 64)
 dense_2 (Dense)
                             (None, 64)
                                                        4160
                                                        650
 dense_3 (Dense)
                             (None, 10)
Total params: 56,452
Trainable params: 56,452
Non-trainable params: 0
```

convolutional\_neural\_network.compile(optimizer='adam', loss='sparse\_categorical\_crossentropy', metrics=['accuracy'])
history=convolutional\_neural\_network.fit(X\_train, y\_train, epochs=10,validation\_data=(X\_test, y\_test))

```
Epoch 1/10
1875/1875 [===========] - 61s 32ms/step - loss: 0.0110 - accuracy: 0.9964 - val_loss: 0.0628 - val_accuracy: 0.9
Epoch 2/10
1875/1875 [===========] - 50s 27ms/step - loss: 0.0101 - accuracy: 0.9967 - val_loss: 0.0662 - val_accuracy: 0.9
Epoch 3/10
1875/1875 [=
           Epoch 4/10
1875/1875 [===========] - 51s 27ms/step - loss: 0.0075 - accuracy: 0.9975 - val_loss: 0.0680 - val_accuracy: 0.9
Epoch 5/10
1875/1875 [=
           Epoch 6/10
1875/1875 [===========] - 51s 27ms/step - loss: 0.0085 - accuracy: 0.9974 - val_loss: 0.0637 - val_accuracy: 0.9
Epoch 7/10
1875/1875 [===========] - 50s 27ms/step - loss: 0.0071 - accuracy: 0.9977 - val_loss: 0.0614 - val_accuracy: 0.9
Enoch 8/10
1875/1875 [============] - 51s 27ms/step - loss: 0.0057 - accuracy: 0.9982 - val_loss: 0.0772 - val_accuracy: 0.9
Epoch 9/10
1875/1875 [=
           Epoch 10/10
1875/1875 [===========] - 50s 27ms/step - loss: 0.0079 - accuracy: 0.9977 - val_loss: 0.0571 - val_accuracy: 0.9
```

## Evaluating the CNN model

#### Making Predictions

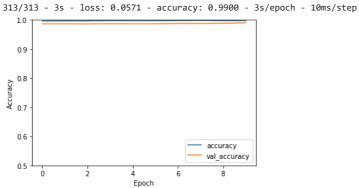
```
[7, 2, 1, 0, 4]
```

```
plt.plot(history.history['accuracy'],label='accuracy')
plt.plot(history.history['val_accuracy'],label = 'val_accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim([0.5, 1])
plt.legend(loc='lower right')

test_loss, test_acc = convolutional_neural_network.evaluate(X_test,
```

verbose=2)

y\_test,



print('Test Accuracy is',test\_acc)

Test Accuracy is 0.9900000095367432

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