

# Project Presentation

## Architecture used in the Paper::

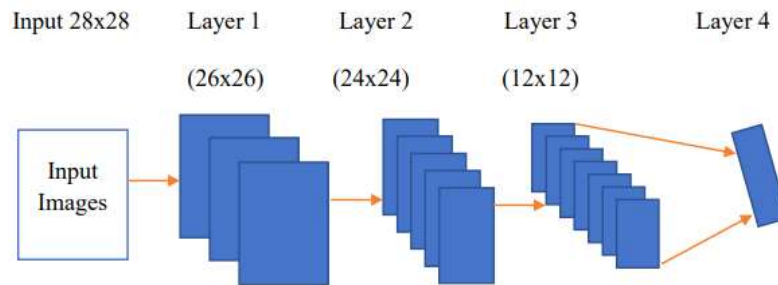


Figure 1: Architecture of Convolutional Neural Network (CNN)

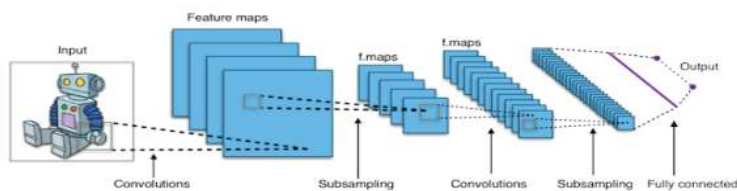


Figure. 2 Typical CNN Architecture

## Datasets for it : **MNIST dataset**

### dataset link :

<https://www.tensorflow.org/datasets/catalog/mnist>

It includes 70000 image for (Training & Testing)

**number of classes : 10**

dimension of images = 28\*28

number of traning : 60000

number of testing : 10000

## **Our Model :**

```
[69]: model = Sequential()

model.add(Conv2D (32 ,(3,3) ,kernel_regularizer=regularizers.l2(0.0001), input_shape = x_train.shape [1: ]))
model.add(Activation ( "relu" ))
model.add(MaxPooling2D (pool_size = (2,2)))

model.add(Conv2D (64 ,(3,3),kernel_regularizer=regularizers.l2(0.0001)))
model.add(Activation ( "relu" ))
model.add(MaxPooling2D (pool_size = (2,2)))

model.add(Conv2D (64 ,(3,3),kernel_regularizer=regularizers.l2(0.0001)))
model.add(Activation ( "relu" ))
model.add(MaxPooling2D (pool_size = (2,2)))

model.add (Flatten())
model.add (Dense (64))
model.add (Activation ( "relu" ))

model.add (Dense (32))
model.add (Activation ( "relu" ))

model.add (Dense (10))
model.add (Activation ( "softmax" ))
```

## model.summary:

```
In [70]: model.summary()
```

```
Model: "sequential_7"

```

Layer (type)	Output Shape	Param #
conv2d_14 (Conv2D)	(None, 26, 26, 32)	320
activation_21 (Activation)	(None, 26, 26, 32)	0
max_pooling2d_11 (MaxPooling)	(None, 13, 13, 32)	0
conv2d_15 (Conv2D)	(None, 11, 11, 64)	18496
activation_22 (Activation)	(None, 11, 11, 64)	0
max_pooling2d_12 (MaxPooling)	(None, 5, 5, 64)	0
conv2d_16 (Conv2D)	(None, 3, 3, 64)	36928
activation_23 (Activation)	(None, 3, 3, 64)	0
max_pooling2d_13 (MaxPooling)	(None, 1, 1, 64)	0
flatten_3 (Flatten)	(None, 64)	0
dense_9 (Dense)	(None, 64)	4160
activation_24 (Activation)	(None, 64)	0
dense_10 (Dense)	(None, 32)	2080
activation_25 (Activation)	(None, 32)	0
dense_11 (Dense)	(None, 10)	330
activation_26 (Activation)	(None, 10)	0

## Hyperparameters:

optimizer = 'adam'

loss = 'sparse\_categorical\_crossentropy'

epochs = 5

batch\_size = 128

learning rate = 0.001

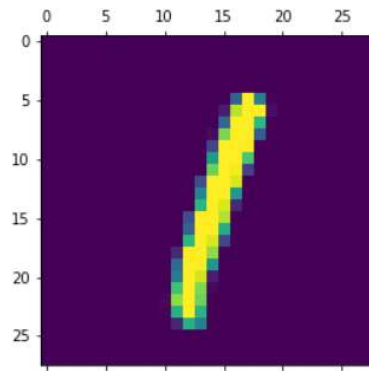
regularization = 0.0001

## Result:::

**this image is number 1 and he is predicted as num 1**

```
In [115]: plt.matshow(x_test[5])
```

```
Out[115]: <matplotlib.image.AxesImage at 0x1bdb751c940>
```



```
In [116]: y_predicted = model.predict(x_test)
```

```
In [117]: np.argmax(y_predicted[5])
```

```
Out[117]: 1
```

```
In [118]: y_test[5]
```

```
Out[118]: 1
```

## Accuracy::: 98%

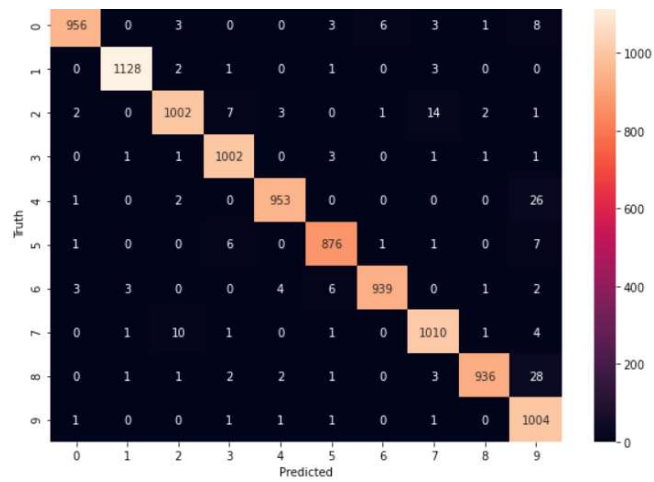
```
In [72]: hist = model.fit(x_train, y_train, batch_size = 128, epochs = 5)
```

```
Epoch 1/5  
469/469 [=====] - 13s 28ms/step - loss: 0.4025 - accuracy: 0.8778  
Epoch 2/5  
469/469 [=====] - 13s 29ms/step - loss: 0.1301 - accuracy: 0.9648  
Epoch 3/5  
469/469 [=====] - 13s 28ms/step - loss: 0.1005 - accuracy: 0.9745  
Epoch 4/5  
469/469 [=====] - 13s 28ms/step - loss: 0.0847 - accuracy: 0.9791  
Epoch 5/5  
469/469 [=====] - 13s 29ms/step - loss: 0.0755 - accuracy: 0.9825
```

```
In [73]: model.evaluate(x_test, y_test)
```

```
313/313 [=====] - 1s 3ms/step - loss: 0.0810 - accuracy: 0.9806  
Out[73]: [0.08100221306085587, 0.9805999994277954]
```

**Confusion matrix:**



## loss Curve:

```

In [82]: plt.figure(figsize = (10,7))
plt.plot(xc , train_loss)
plt.xlabel('num of epochs')
plt.ylabel('loss')

```

```

Out[82]: Text(0, 0.5, 'loss')

```

