

Lab Assignment No. 12

Code:

```
import tensorflow as tf
from tensorflow import keras
from keras import Sequential
from keras.layers import Conv2D, Dense, MaxPooling2D, Flatten, Dropout, Input
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from keras.utils import plot_model
from keras.datasets import mnist

num_classes = 10
input_shape = (28, 28, 1)
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train = x_train.astype("float32") / 255
x_test = x_test.astype("float32") / 255
x_train = np.expand_dims(x_train, -1)
x_test = np.expand_dims(x_test, -1)
print("x_train shape:", x_train.shape)
print(x_train.shape[0], "train samples")
print(x_test.shape[0], "test samples")
# convert class vectors to binary class matrices
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-
datasets/mnist.npz
11490434/11490434 [=====] - 1s 0us/step
x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples

print("len of x_train: ", len(x_train), "Len of y_train: ", len(y_train), "
len of x_test: ", len(x_test), " len y_test: ", len(y_test))

len of x_train: 60000 Len of y_train: 60000 len of x_test: 10000 len
y_test: 10000

model = keras.Sequential(
    [
        keras.Input(shape=input_shape),
        Conv2D(32, kernel_size=(3, 3), activation="relu"),
        MaxPooling2D(pool_size=(2, 2)),
        Conv2D(64, kernel_size=(3, 3), activation="relu"),
```

```

        MaxPooling2D(pool_size=(2, 2)),
        Flatten(),
        Dropout(0.5),
        Dense(num_classes, activation="softmax"),
    ]
)

model.summary()

model.compile(optimizer='adam', loss="categorical_crossentropy",
metrics=['accuracy'])

Model: "sequential"

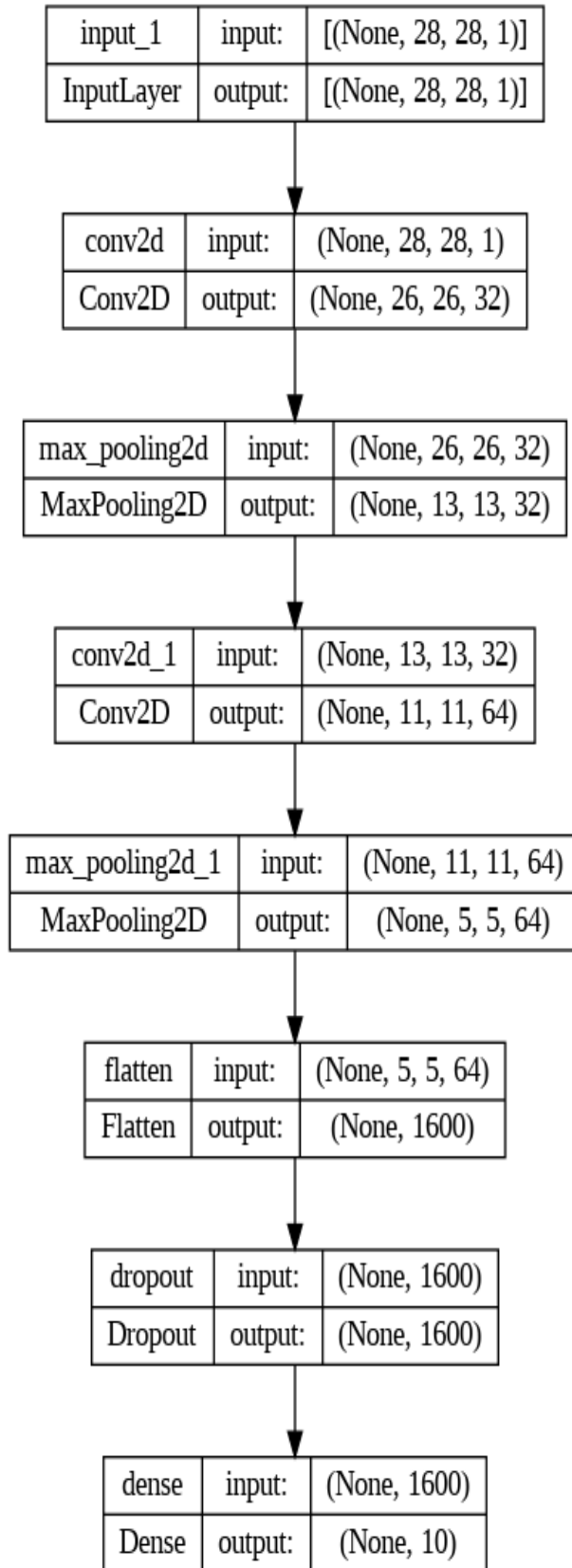
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
conv2d_1 (Conv2D)	(None, 11, 11, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
flatten (Flatten)	(None, 1600)	0
dropout (Dropout)	(None, 1600)	0
dense (Dense)	(None, 10)	16010

```

=====
Total params: 34,826
Trainable params: 34,826
Non-trainable params: 0
=====
plot_model(model, show_shapes=True)

```



```
x_train.shape  
(60000, 28, 28, 1)  
  
y_train.shape  
(60000, 10)  
  
epochs=10  
batch_size = 128  
history = model.fit(x_train, y_train, batch_size=batch_size, epochs=epochs,  
validation_split=0.1)
```

Output:

```
Epoch 1/10  
422/422 [=====] - 57s 133ms/step - loss: 0.3786 -  
accuracy: 0.8882 - val_loss: 0.0825 - val_accuracy: 0.9783  
Epoch 2/10  
422/422 [=====] - 49s 117ms/step - loss: 0.1133 -  
accuracy: 0.9661 - val_loss: 0.0569 - val_accuracy: 0.9850  
Epoch 3/10  
422/422 [=====] - 50s 117ms/step - loss: 0.0847 -  
accuracy: 0.9744 - val_loss: 0.0498 - val_accuracy: 0.9865  
Epoch 4/10  
422/422 [=====] - 48s 114ms/step - loss: 0.0713 -  
accuracy: 0.9777 - val_loss: 0.0415 - val_accuracy: 0.9890  
Epoch 5/10  
422/422 [=====] - 47s 111ms/step - loss: 0.0638 -  
accuracy: 0.9797 - val_loss: 0.0436 - val_accuracy: 0.9885  
Epoch 6/10  
422/422 [=====] - 48s 114ms/step - loss: 0.0574 -  
accuracy: 0.9824 - val_loss: 0.0360 - val_accuracy: 0.9910  
Epoch 7/10  
422/422 [=====] - 51s 120ms/step - loss: 0.0512 -  
accuracy: 0.9841 - val_loss: 0.0384 - val_accuracy: 0.9883  
Epoch 8/10  
422/422 [=====] - 48s 114ms/step - loss: 0.0476 -  
accuracy: 0.9846 - val_loss: 0.0321 - val_accuracy: 0.9905  
Epoch 9/10  
422/422 [=====] - 53s 125ms/step - loss: 0.0439 -  
accuracy: 0.9861 - val_loss: 0.0302 - val_accuracy: 0.9918  
Epoch 10/10  
422/422 [=====] - 50s 117ms/step - loss: 0.0434 -  
accuracy: 0.9862 - val_loss: 0.0315 - val_accuracy: 0.9908  
  
prediction = model.predict(x_test)  
  
313/313 [=====] - 3s 9ms/step
```

```
prediction.shape
```

```
(10000, 10)
```

```
y_test.shape
```

```
(10000, 10)
```

```
loss, accuracy = model.evaluate(x_test, y_test)
```

```
313/313 [=====] - 3s 9ms/step - loss: 0.0276 -  
accuracy: 0.9908
```

```
print(f"Loss of model is on testing data: {loss} and accuracy of model is on  
testing data: {accuracy}")
```

```
Loss of model is on testing data: 0.027647219598293304 and accuracy of model  
is on testing data: 0.9908000230789185
```

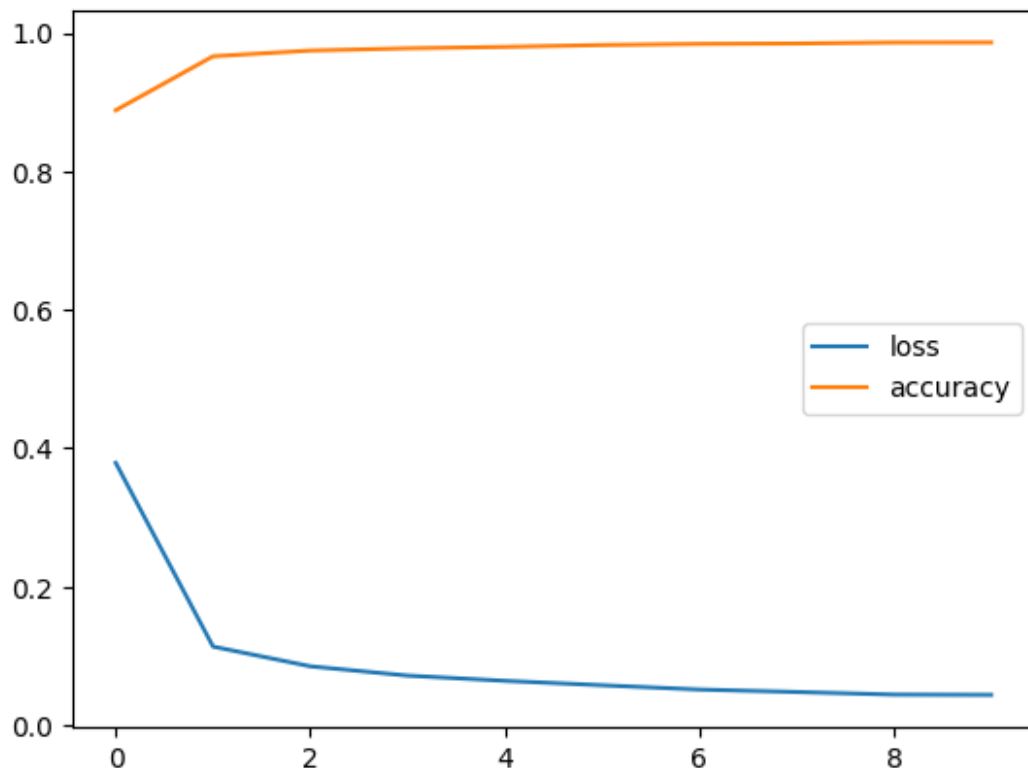
```
import matplotlib.pyplot as plt
```

```
plt.plot(history.history['loss'], label='loss')
```

```
plt.plot(history.history['accuracy'], label='accuracy')
```

```
plt.legend()
```

```
plt.show()
```



```
plt.plot(history.history['val_loss'], label='Val loss')  
plt.plot(history.history['val_accuracy'], label='Val accuracy')  
plt.legend()  
plt.show()
```

