

TEAM:PNT2022-TMID43750

#Performance Analysis (sprint-3)

(i).Performance Analysis

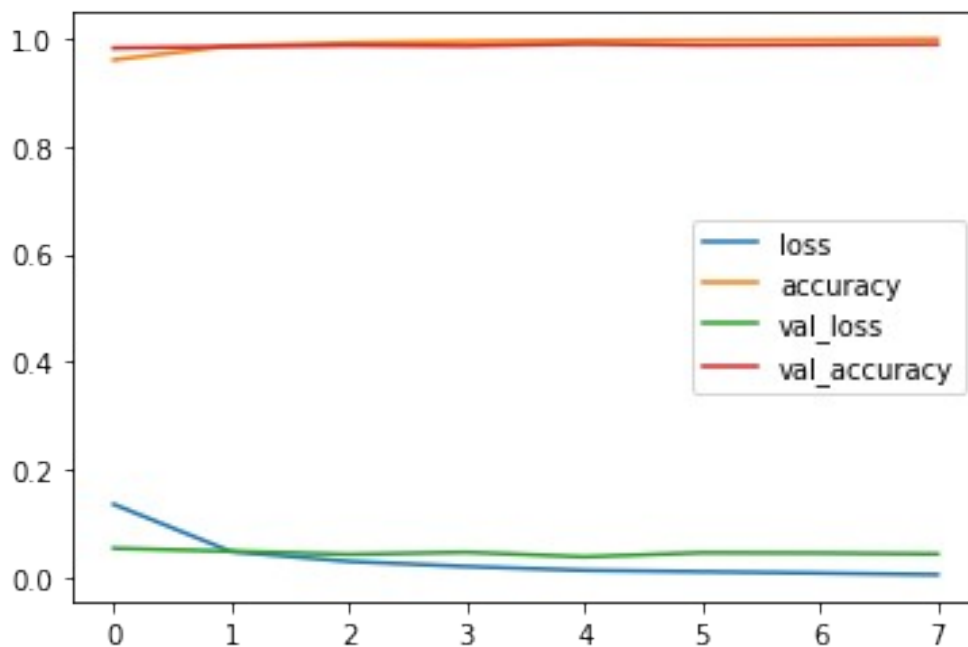
```
metrics = pd.DataFrame(model.history.history)
```

```
metrics
```

| | loss | accuracy | val_loss | val_accuracy |
|---|----------|----------|----------|--------------|
| 0 | 0.136240 | 0.959183 | 0.054753 | 0.9811 |
| 1 | 0.048557 | 0.985233 | 0.049157 | 0.9839 |
| 2 | 0.030406 | 0.990800 | 0.043443 | 0.9861 |
| 3 | 0.020990 | 0.993350 | 0.047409 | 0.9850 |
| 4 | 0.013883 | 0.995450 | 0.038858 | 0.9890 |
| 5 | 0.011308 | 0.996183 | 0.046504 | 0.9865 |
| 6 | 0.008813 | 0.996933 | 0.045933 | 0.9875 |
| 7 | 0.005928 | 0.997917 | 0.044267 | 0.9886 |

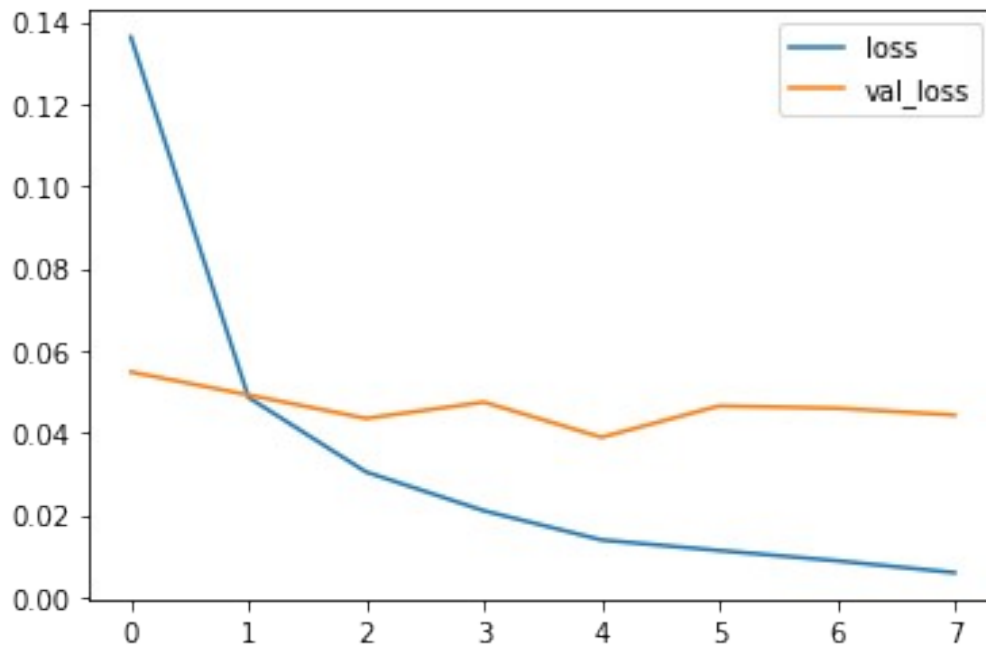
```
metrics.plot()
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f9be00620d0>



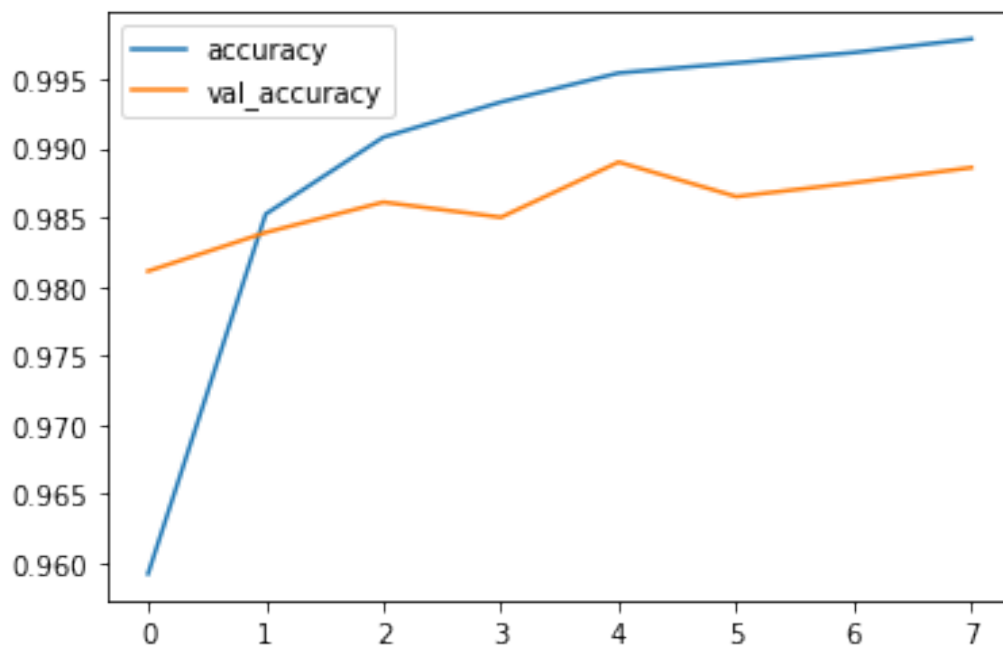
```
metrics[['loss', 'val_loss']].plot()
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f9b8a38eb90>



```
metrics[['accuracy', 'val_accuracy']].plot()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f9b8a2a36d0>
```



(ii).Evaluate the Model

```
model.evaluate(x_test,y_cat_test,verbose=0)
#loss | #accuracy
```

```
[0.04426722601056099, 0.9886000156402588]
```

```
from sklearn.metrics import classification_report, confusion_matrix
```

```
predict_x=model.predict(x_test)
classes_x=np.argmax(predict_x,axis=1)
```

```
313/313 [=====] - 1s 2ms/step
```

```
print(classification_report(y_test,classes_x))
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|------------|
| 0 | 0.99 | 1.00 | 0.99 | 980 |
| 1 | 0.99 | 1.00 | 1.00 | 1135 |
| 2 | 0.99 | 0.99 | 0.99 | 1032 |
| 3 | 0.98 | 1.00 | 0.99 | 1010 |
| 4 | 0.99 | 0.98 | 0.99 | 982 |
| 5 | 1.00 | 0.99 | 0.99 | 892 |
| 6 | 0.99 | 0.98 | 0.99 | 958 |
| 7 | 0.98 | 0.99 | 0.99 | 1028 |
| 8 | 0.99 | 0.98 | 0.99 | 974 |
| 9 | 0.98 | 0.98 | 0.98 | 1009 |
| accuracy | | | | 0.99 10000 |
| macro avg | 0.99 | 0.99 | 0.99 | 10000 |
| weighted avg | 0.99 | 0.99 | 0.99 | 10000 |

```
print(confusion_matrix(y_test,classes_x))
```

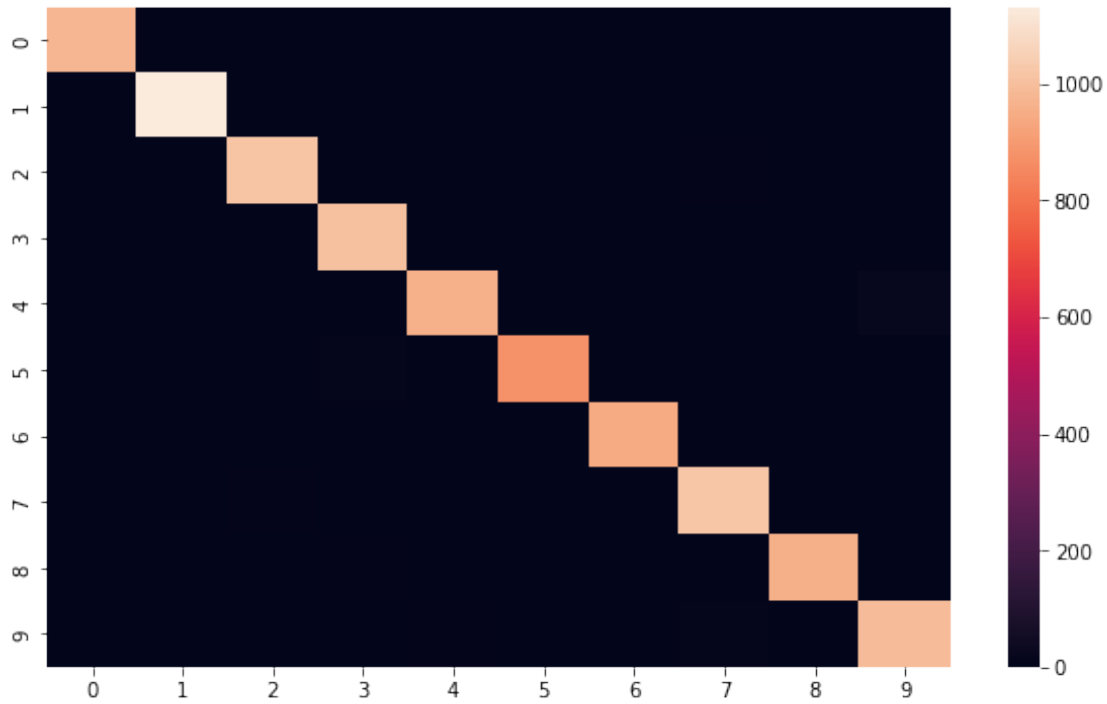
```
[ [ 977    0    0    1    0    0    1    0    1    0]
  [    0 1132    1    2    0    0    0    0    0    0]
  [    1    1 1017    2    0    0    2    7    2    0]
  [    0    0    1 1005    0    1    0    1    1    1]
  [    0    0    0    0  963    0    1    0    0   18]
  [    0    0    0   12    0  879    1    0    0    0]
  [    4    2    1    0    4    2  943    0    2    0]
  [    0    2    5    0    0    0    0 1019    1    1]
  [    4    0    1    6    0    0    0    1  959    3]
  [    0    1    0    1    5    1    0    9    0  992]]
```

```
import seaborn as sns
```

```
plt.figure(figsize=(10,6))
```

```
sns.heatmap(confusion_matrix(y_test,classes_x))
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f9b73f53750>
```



(iii).Make Prediction

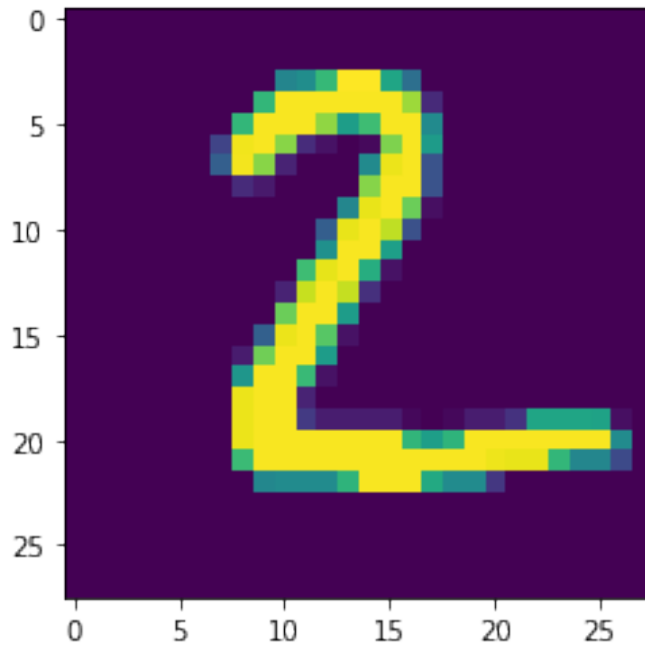
```
my_num = x_test[1]
```

```
classes_x
```

```
array([7, 2, 1, ..., 4, 5, 6])
```

```
plt.imshow(my_num.reshape(28,28))
```

```
<matplotlib.image.AxesImage at 0x7f9b73a95b10>
```



(iv).Save the Model

```
from tensorflow.keras.models import load_model
```

```
model.save('CNN.h5')  
print('Model Saved!')
```

```
savedModel=load_model('CNN.h5')  
savedModel.summary()
```

```
Model Saved!  
Model: "sequential"
```

| Layer (type) | Output Shape | Param # |
|------------------------------|--------------------|---------|
| conv2d (Conv2D) | (None, 25, 25, 32) | 544 |
| max_pooling2d (MaxPooling2D) | (None, 12, 12, 32) | 0 |
| flatten (Flatten) | (None, 4608) | 0 |
| dense (Dense) | (None, 128) | 589952 |
| dense_1 (Dense) | (None, 10) | 1290 |
| Total params: 591,786 | | |
| Trainable params: 591,786 | | |

Non-trainable params: 0
