

## #Performance Analysis (sprint-3)

### (i).Performance Analysis

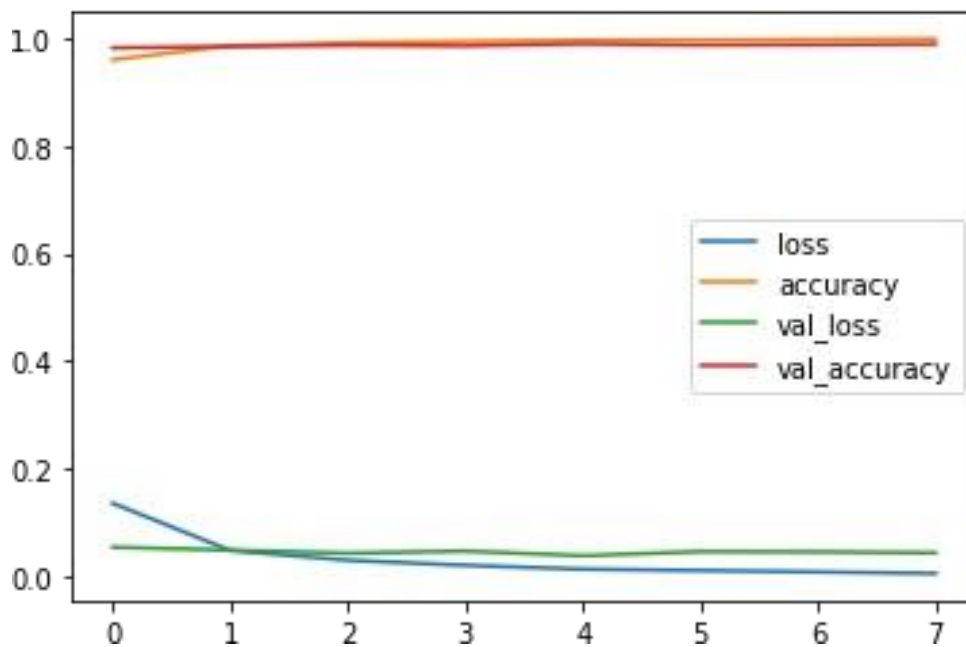
```
metrics = pd.DataFrame(model.history.history)
```

```
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

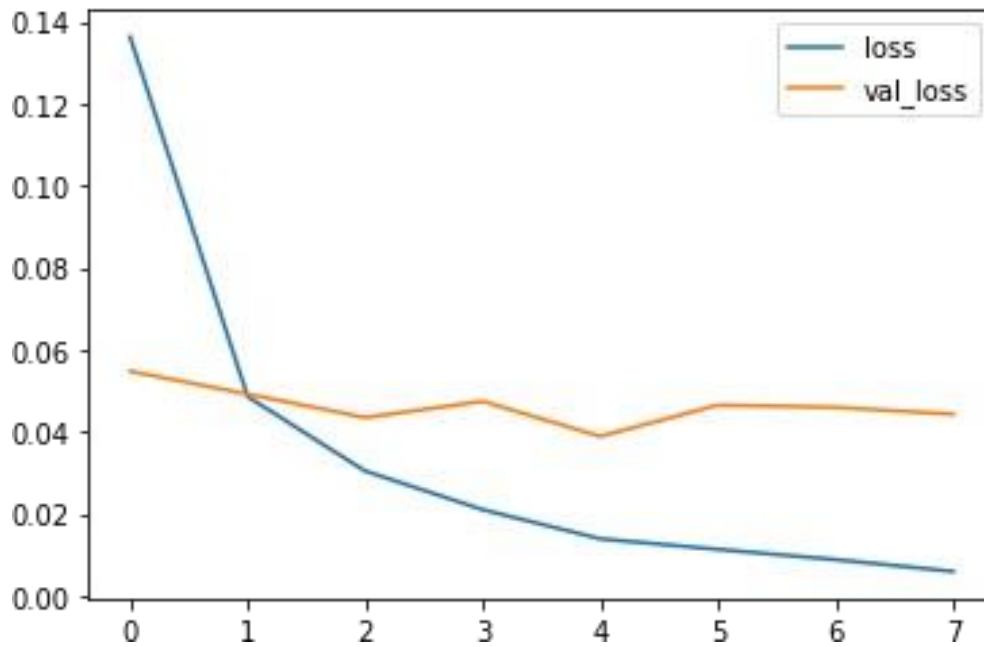
```
metrics.plot()
```

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



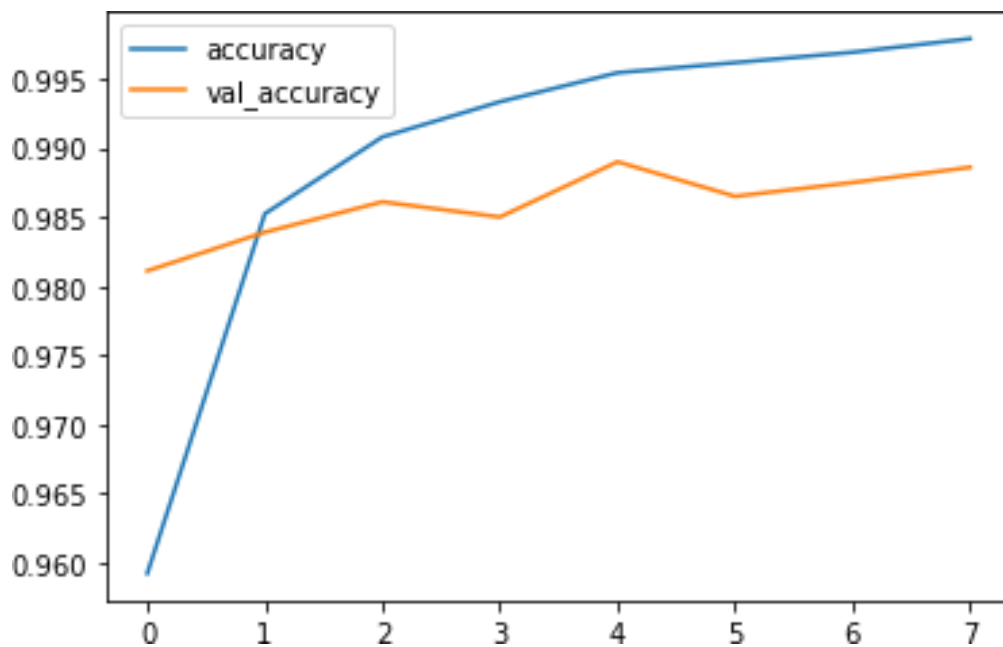
```
metrics[['loss', 'val_loss']].plot()
```

```
<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]]

```

```

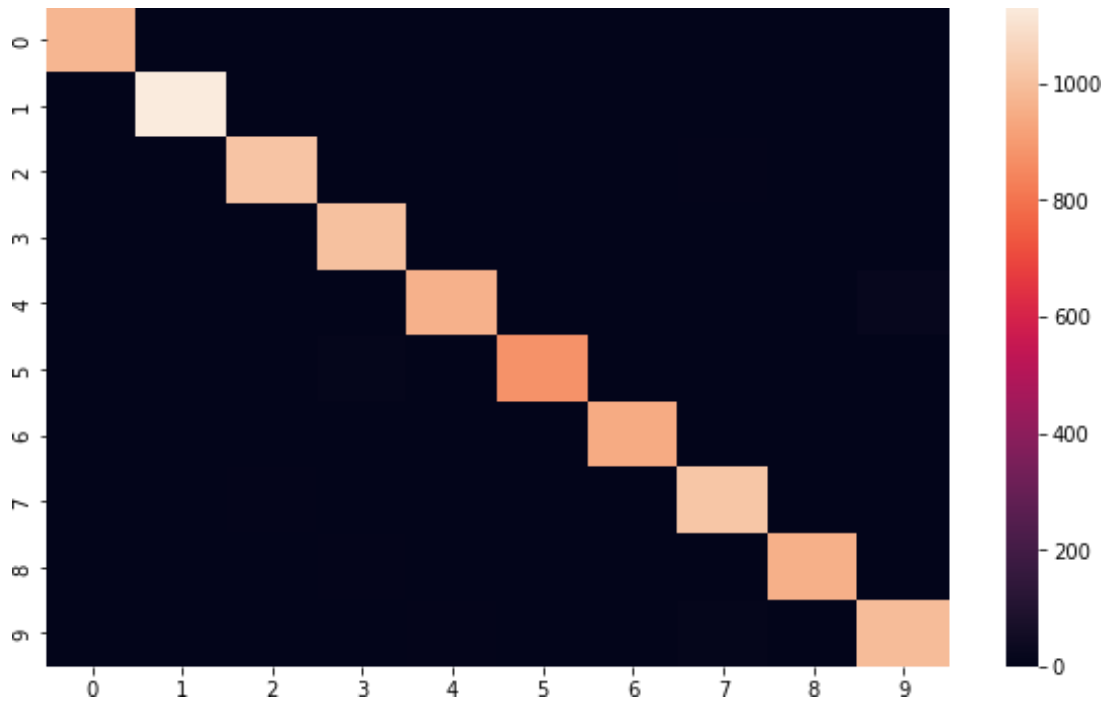
import seaborn as sns
plt.figure(figsize=(10,6))
sns.heatmap(confusion_matrix(y_test,classes_x))

```

```

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

```



(iii).Make Prediction

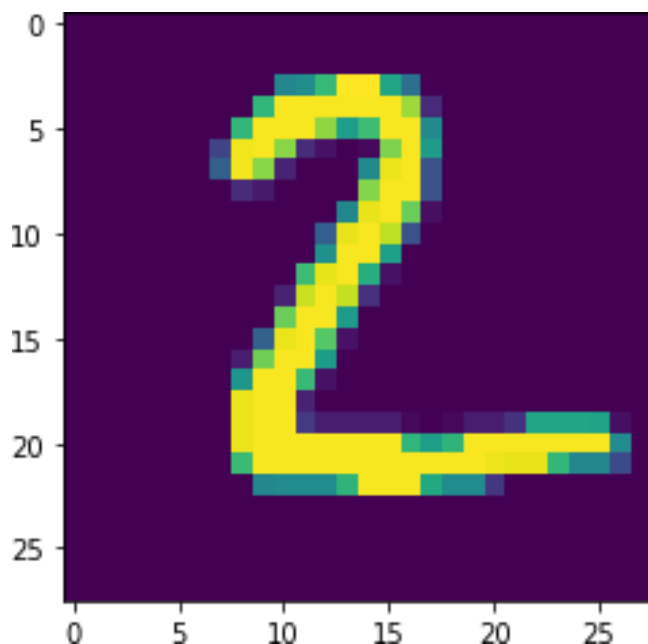
```
my_num = x_test[1]
```

```
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

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