TEAM:PNT2022TMID31213

#Performance Analysis (sprint-3)

(i).Performance Analysis

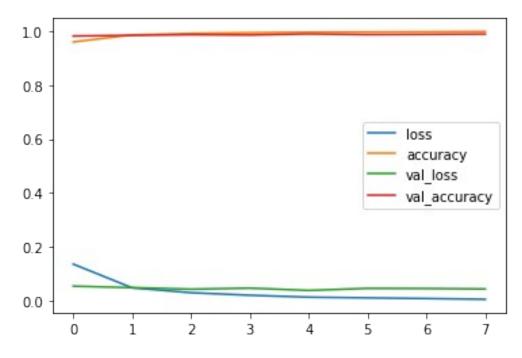
metrics = pd.DataFrame(model.history.history)

metrics

loss	accuracy	val_loss	val_accuracy
0.136240	0.959183	$0.0\overline{5}4753$	0.9811
0.048557	0.985233	0.049157	0.9839
0.030406	0.990800	0.043443	0.9861
0.020990	0.993350	0.047409	0.9850
0.013883	0.995450	0.038858	0.9890
0.011308	0.996183	0.046504	0.9865
0.008813	0.996933	0.045933	0.9875
0.005928	0.997917	0.044267	0.9886
	0.136240 0.048557 0.030406 0.020990 0.013883 0.011308 0.008813	0.1362400.9591830.0485570.9852330.0304060.9908000.0209900.9933500.0138830.9954500.0113080.9961830.0088130.996933	0.136240 0.959183 0.054753 0.048557 0.985233 0.049157 0.030406 0.990800 0.043443 0.020990 0.993350 0.047409 0.013883 0.995450 0.038858 0.011308 0.996183 0.046504 0.008813 0.996933 0.045933

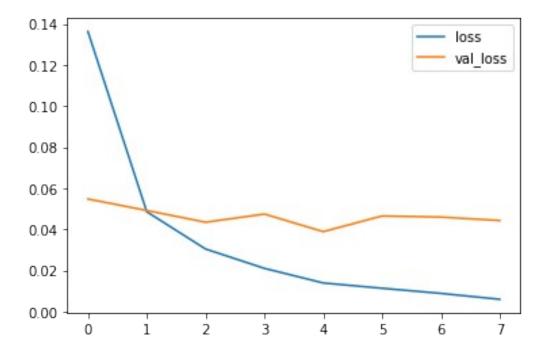
metrics.plot()

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



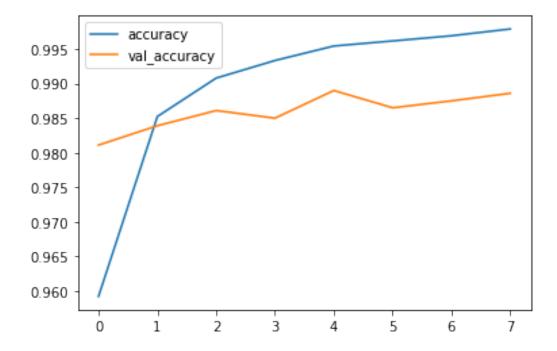
metrics[['loss','val_loss']].plot()

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



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

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



(ii).Evaluate the Model

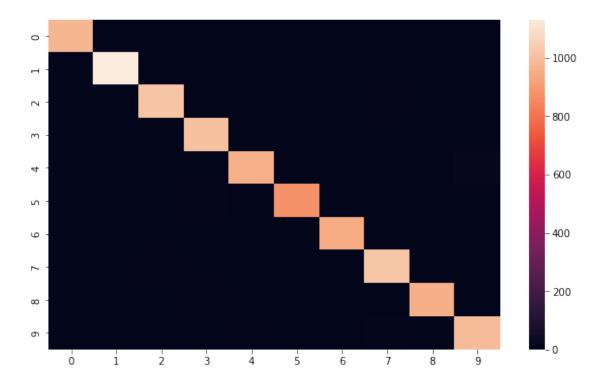
[0.04426722601056099, 0.9886000156402588]

```
from sklearn.metrics import classification report, confusion matrix
predict x=model.predict(x test)
classes x=np.argmax(predict x,axis=1)
print(classification report(y test, classes x))
                            recall
              precision
                                   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
                              0.99
                                         0.99
                                                    892
                    1.00
           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
                   0.99
                              0.99
                                         0.99
                                                  10000
   macro avq
                   0.99
                              0.99
                                         0.99
weighted avg
                                                  10000
print(confusion matrix(y test,classes x))
[[ 977
          0
               0
                     1
                          0
                               0
                                    1
                                               1
                                                    0]
                                         0
                                    0
 [
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               1
                     2
                          0
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                                         7
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                    0
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               0
                    12
                             879
                                    1
                                         0
                                               0
                                                    0]
                          0
          2
 [
     4
               1
                               2
                                  943
                                               2
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                                    0 1019
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 [
     4
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               1
                                          1
                                             959
                                                    3]
                     6
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 ſ
     0
          1
               0
                     1
                          5
                               1
                                    0
                                          9
                                                  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

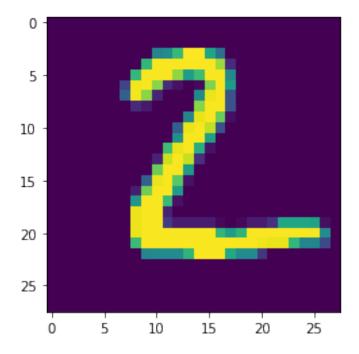
 $my_num = x_test[1]$

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
<pre>max_pooling2d (MaxPooling2D)</pre>	(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