```
#Performance Analysis (sprint-3)
```

## (i).Performance Analysis

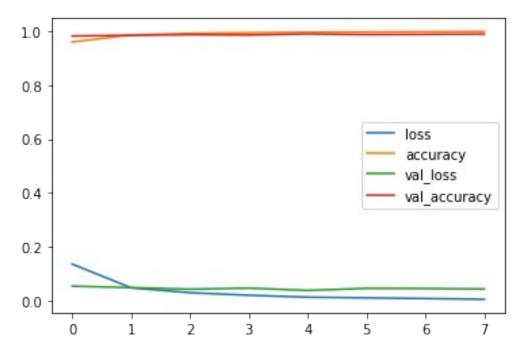
metrics = pd.DataFrame(model.history.history)

## 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<br>0.048557<br>0.030406<br>0.020990<br>0.013883<br>0.011308<br>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 |

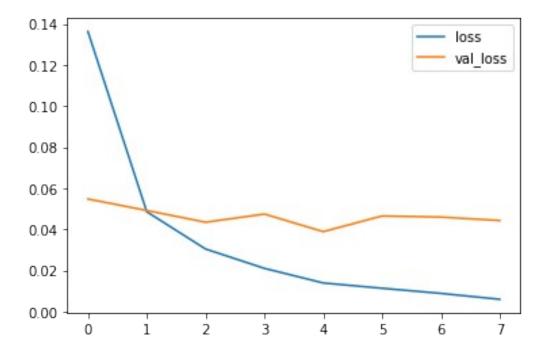
metrics.plot()

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



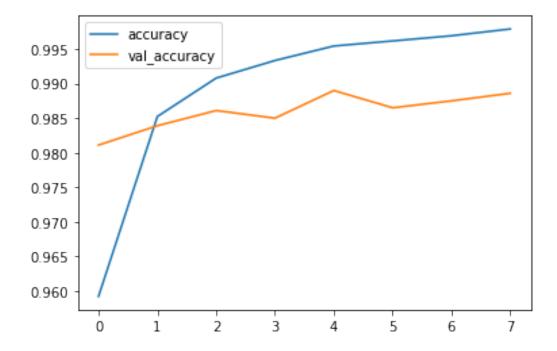
metrics[['loss','val\_loss']].plot()

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



metrics[['accuracy','val\_accuracy']].plot()

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



(ii).Evaluate the Model

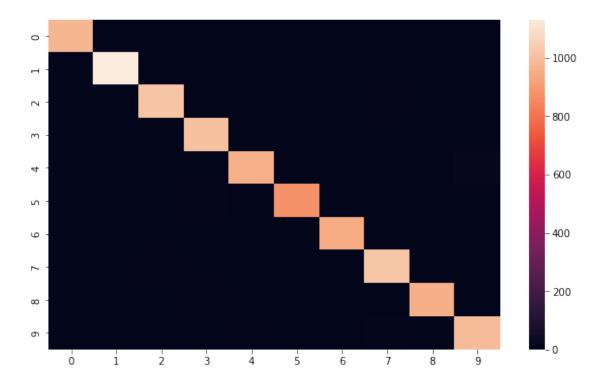
[0.04426722601056099, 0.9886000156402588]

```
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
 [
     0 1132
               1
                     2
                          0
                               0
                                          0
                                               0
                                                    01
     1
          1 1017
                     2
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                               0
                                    2
                                         7
                                               2
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 [
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                                    0
                                         1
                                               1
                                                    11
     0
          0
                          0
 [
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          0
                        963
                               0
                                    1
                                         0
                                               0
                                                   18]
               0
                    0
 [
     0
          0
               0
                    12
                             879
                                    1
                                         0
                                               0
                                                    0]
                          0
          2
 [
     4
               1
                               2
                                  943
                                               2
                                                    01
                    0
                          4
                                         0
     0
          2
               5
 [
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                               0
                                    0 1019
                                               1
                                                    1]
 [
     4
          0
               1
                                          1
                                             959
                                                    3]
                     6
                          0
                               0
                                    0
 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

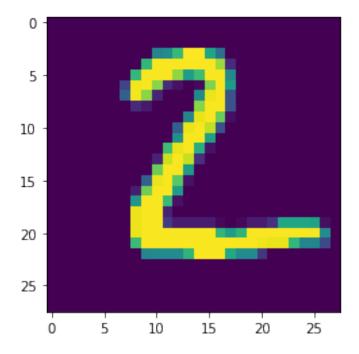
 $my_num = x_test[1]$ 

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