```
In [99]:
            import numpy as np
            import pandas as pd
            import matplotlib.pyplot as plt
            import seaborn as sns
            df=pd.read_csv('/stroke_data.csv')
In [13]:
In [16]:
            df.head()
                           hypertension heart_disease ever_married work_type Residence_type avg_glucose_lev
Out[16]:
              gender
                       age
                      58.0
                                       1
                                                     0
                                                                                           Urban
                                                                                                              87.9
                Male
                                                                  Yes
                                                                          Private
              Female
                      70.0
                                       0
                                                                  Yes
                                                                          Private
                                                                                            Rural
                                                                                                              69.0
              Female
                      52.0
                                       0
                                                     0
                                                                                           Urban
                                                                                                             77.5
           2
                                                                  Yes
                                                                          Private
                                                                            Self-
              Female
                      75.0
                                                                  Yes
                                                                                            Rural
                                                                                                             243.5
                                                                       employed
              Female 32.0
                                                     0
                                                                  Yes
                                                                          Private
                                                                                            Rural
                                                                                                             77.6
 In [ ]:
```

label encoding the categorical column

```
In [22]:
           from sklearn.preprocessing import LabelEncoder
           df['gender'].unique()
In [23]:
          array(['Male', 'Female'], dtype=object)
Out[23]:
In [24]:
           label_encoder=LabelEncoder()
In [25]:
           df['gender label'] = label encoder.fit transform(df['gender'])
In [26]:
           df.head()
Out[26]:
                                                                               Residence_type avg_glucose_lev
              gender
                      age hypertension heart_disease ever_married work_type
           0
                      58.0
                                                    0
                                                                                        Urban
                                                                                                          87.9
                Male
                                                                Yes
                                                                        Private
              Female
                      70.0
                                                                        Private
                                                                                         Rural
                                                                                                          69.0
                                                                Yes
                     52.0
                                                    0
                                                                                        Urban
                                                                                                          77.5
              Female
                                                                Yes
                                                                        Private
           2
                                                                          Self-
              Female
                     75.0
                                                                                         Rural
                                                                                                         243.5
                                                                Yes
                                                                     employed
              Female 32.0
                                                                        Private
                                                                                         Rural
                                                                                                          77.6
                                                                Yes
```

```
df['smoking_label'] = label_encoder.fit_transform(df['smoking_status'])
In [27]:
In [28]:
           df.head()
Out[28]:
                          hypertension heart_disease ever_married work_type
                                                                               Residence_type avg_glucose_lev
             gender
                      age
          0
                Male
                      58.0
                                                                Yes
                                                                        Private
                                                                                        Urban
                                                                                                          87.9
              Female
                      70.0
                                      0
                                                    0
                                                                Yes
                                                                        Private
                                                                                         Rural
                                                                                                          69.0
           1
                      52.0
                                      0
                                                   0
                                                                                        Urban
                                                                                                          77.5
          2
              Female
                                                               Yes
                                                                        Private
                                                                         Self-
                                                               Yes
          3
              Female
                     75.0
                                      0
                                                                                         Rural
                                                                                                         243.5
                                                                     employed
              Female 32.0
                                      0
                                                    0
                                                                                                          77.6
                                                               Yes
                                                                        Private
                                                                                         Rural
In [29]:
           df['marriage label'] = label encoder.fit transform(df['ever married'])
In [30]:
           df['residence_label'] = label_encoder.fit_transform(df['Residence_type'])
           df['work_label'] = label_encoder.fit_transform(df['work_type'])
In [31]:
In [35]:
           df.head()
Out[35]:
              age hypertension heart_disease ever_married work_type Residence_type avg_glucose_level
                                                                                                        bmi
          0
             58.0
                              1
                                           0
                                                       Yes
                                                                                Urban
                                                                                                  87.96
                                                                Private
                                                                                                        39.2
             70.0
                              0
                                           0
                                                       Yes
                                                                Private
                                                                                 Rural
                                                                                                  69.04
                                                                                                        35.9
          2
             52.0
                              0
                                           0
                                                       Yes
                                                                Private
                                                                                Urban
                                                                                                  77.59
                                                                                                        17.7
                                                                 Self-
             75.0
                                           1
                                                       Yes
                                                                                 Rural
                                                                                                 243.53
                                                                                                        27.0
                                                             employed
             32.0
                              0
                                           0
                                                       Yes
                                                                Private
                                                                                 Rural
                                                                                                  77.67
                                                                                                        32.3
                                                                                                           DIVIDING THE DATA INTO TRAINING SET AND TESTING SET
In [69]:
           df.drop('ever married', axis=1, inplace=True)
In [70]:
           df.drop('work_type', axis=1, inplace=True)
In [71]:
           df.drop('Residence_type', axis=1, inplace=True)
In [74]:
           df.head()
```

```
Out[74]:
                                              bmi stroke gender_label smoking_label
          age hypertension heart_disease avg_glucose_level
       0 58.0
                     1
                               0
                                          87.96
                                              39.2
                                                     0
                                                               1
                                                                          1
         70.0
                     0
                               0
                                          69.04
                                              35.9
                                                               0
                                                                          0
       1
                                                     0
       2
          52.0
                     0
                               0
                                          77.59
                                              17.7
                                                               0
                                                                          0
                                                     0
       3
         75.0
                     0
                               1
                                         243.53
                                              27.0
                                                     0
                                                               0
                                                                          1
         32.0
                     0
                               0
                                         77.67 32.3
                                                     0
                                                               0
                                                                          2
In [75]:
        x=df.drop('stroke', axis=1)
In [76]:
        y=df['stroke']
        from sklearn.model selection import train test split
In [77]:
        x test,x train,y test,y train=train test split(x,y,test size=0.3,random state=2021)
In [78]:
In [79]:
        import tensorflow as tf
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Dropout
        ann_model = Sequential()
In [80]:
        ann model.add(Dense(units=30, activation='relu'))
                                                   #input layer
In [81]:
In [82]:
        ann model.add(Dense(units=15, activation='relu'))
                                                    #hidden layer
        ann model.add(Dense(units=1, activation='sigmoid'))
In [83]:
                                                    #output layer
        ann model.compile(loss='binary crossentropy', optimizer='adam', metrics=['accuracy'])
In [84]:
In [85]:
        ann model.fit(
        x=x_train,
        y=y_train,
        epochs=600,
        validation_data=(x_test, y_test),
        verbose=1
        )
       Epoch 1/600
       - val_loss: 0.0969 - val_accuracy: 0.9816
       Epoch 2/600
       - val_loss: 0.1057 - val_accuracy: 0.9816
       Epoch 3/600
       - val_loss: 0.0952 - val_accuracy: 0.9811
       Epoch 4/600
       - val_loss: 0.0949 - val_accuracy: 0.9811
       Epoch 5/600
```

```
- val loss: 0.0883 - val accuracy: 0.9816
Epoch 6/600
- val loss: 0.0937 - val accuracy: 0.9816
Epoch 7/600
- val loss: 0.0888 - val accuracy: 0.9816
Epoch 8/600
- val loss: 0.0886 - val accuracy: 0.9816
Epoch 9/600
- val loss: 0.0867 - val accuracy: 0.9816
Epoch 10/600
- val_loss: 0.1034 - val_accuracy: 0.9816
Epoch 11/600
- val_loss: 0.1017 - val_accuracy: 0.9795
Epoch 12/600
- val loss: 0.0861 - val accuracy: 0.9816
Epoch 13/600
- val loss: 0.0872 - val accuracy: 0.9816
Epoch 14/600
- val loss: 0.1118 - val accuracy: 0.9816
Epoch 15/600
- val loss: 0.0851 - val accuracy: 0.9816
Epoch 16/600
- val loss: 0.0885 - val accuracy: 0.9816
Epoch 17/600
- val loss: 0.0846 - val accuracy: 0.9816
Epoch 18/600
- val loss: 0.0856 - val accuracy: 0.9816
Epoch 19/600
val loss: 0.0950 - val accuracy: 0.9812
Epoch 20/600
- val loss: 0.0956 - val accuracy: 0.9811
Epoch 21/600
- val loss: 0.0907 - val accuracy: 0.9810
Epoch 22/600
- val loss: 0.0869 - val accuracy: 0.9816
Epoch 23/600
- val_loss: 0.0858 - val_accuracy: 0.9816
Epoch 24/600
- val_loss: 0.0892 - val_accuracy: 0.9816
Epoch 25/600
- val_loss: 0.0910 - val_accuracy: 0.9816
Epoch 26/600
- val loss: 0.0852 - val accuracy: 0.9816
```

```
Epoch 27/600
- val loss: 0.0953 - val accuracy: 0.9816
Epoch 28/600
- val loss: 0.0838 - val accuracy: 0.9816
Epoch 29/600
- val loss: 0.0850 - val accuracy: 0.9816
Epoch 30/600
- val_loss: 0.0846 - val_accuracy: 0.9816
Epoch 31/600
- val loss: 0.0866 - val accuracy: 0.9816
Epoch 32/600
- val_loss: 0.0960 - val_accuracy: 0.9816
Epoch 33/600
- val loss: 0.0869 - val accuracy: 0.9816
Epoch 34/600
- val loss: 0.0846 - val accuracy: 0.9816
Epoch 35/600
- val_loss: 0.0871 - val_accuracy: 0.9815
Epoch 36/600
- val loss: 0.0842 - val accuracy: 0.9816
Epoch 37/600
- val loss: 0.0846 - val accuracy: 0.9816
Epoch 38/600
- val_loss: 0.0863 - val_accuracy: 0.9816
Epoch 39/600
- val_loss: 0.0843 - val_accuracy: 0.9816
Epoch 40/600

    val loss: 0.0835 - val accuracy: 0.9816

Epoch 41/600
- val loss: 0.0884 - val accuracy: 0.9816
Epoch 42/600
- val loss: 0.0838 - val accuracy: 0.9816
Epoch 43/600
- val loss: 0.0849 - val accuracy: 0.9816
Epoch 44/600
- val loss: 0.0868 - val accuracy: 0.9816
Epoch 45/600
- val_loss: 0.0896 - val_accuracy: 0.9814
Epoch 46/600
- val_loss: 0.0840 - val_accuracy: 0.9816
Epoch 47/600
- val_loss: 0.0863 - val_accuracy: 0.9816
Epoch 48/600
```

```
- val loss: 0.0869 - val accuracy: 0.9816
Epoch 49/600
- val loss: 0.0836 - val accuracy: 0.9816
Epoch 50/600
- val loss: 0.0896 - val accuracy: 0.9816
Epoch 51/600
- val_loss: 0.0826 - val_accuracy: 0.9816
Epoch 52/600
- val loss: 0.0870 - val accuracy: 0.9816
Epoch 53/600
- val loss: 0.0877 - val accuracy: 0.9816
Epoch 54/600
- val loss: 0.0843 - val accuracy: 0.9816
Epoch 55/600
- val loss: 0.0888 - val accuracy: 0.9813
Epoch 56/600
- val_loss: 0.0832 - val_accuracy: 0.9816
Epoch 57/600
- val loss: 0.0942 - val accuracy: 0.9813
Epoch 58/600
- val loss: 0.0823 - val accuracy: 0.9816
Epoch 59/600
- val_loss: 0.0875 - val_accuracy: 0.9816
Epoch 60/600
- val_loss: 0.0832 - val_accuracy: 0.9816
Epoch 61/600
- val_loss: 0.0840 - val_accuracy: 0.9816
Epoch 62/600
- val_loss: 0.0834 - val_accuracy: 0.9816
Epoch 63/600
- val loss: 0.0854 - val accuracy: 0.9816
Epoch 64/600
- val loss: 0.0828 - val accuracy: 0.9816
Epoch 65/600
- val loss: 0.0862 - val accuracy: 0.9816
Epoch 66/600
- val_loss: 0.0871 - val_accuracy: 0.9816
Epoch 67/600
- val_loss: 0.0827 - val_accuracy: 0.9816
Epoch 68/600
- val loss: 0.0833 - val accuracy: 0.9816
Epoch 69/600
- val loss: 0.0827 - val accuracy: 0.9816
Epoch 70/600
```

```
- val loss: 0.0834 - val accuracy: 0.9816
Epoch 71/600
- val loss: 0.0865 - val accuracy: 0.9816
Epoch 72/600
- val loss: 0.0833 - val accuracy: 0.9816
Epoch 73/600
- val loss: 0.0831 - val accuracy: 0.9816
Epoch 74/600
- val loss: 0.0866 - val accuracy: 0.9816
Epoch 75/600
- val_loss: 0.0840 - val_accuracy: 0.9816
Epoch 76/600
- val_loss: 0.0852 - val_accuracy: 0.9816
Epoch 77/600
- val loss: 0.0888 - val accuracy: 0.9816
Epoch 78/600
- val loss: 0.0826 - val accuracy: 0.9816
Epoch 79/600
- val loss: 0.0833 - val accuracy: 0.9816
Epoch 80/600
- val loss: 0.0843 - val accuracy: 0.9815
Epoch 81/600
- val loss: 0.0826 - val accuracy: 0.9816
Epoch 82/600
- val loss: 0.0842 - val accuracy: 0.9816
Epoch 83/600
- val loss: 0.0832 - val accuracy: 0.9816
Epoch 84/600
- val loss: 0.0829 - val accuracy: 0.9816
Epoch 85/600
- val loss: 0.0840 - val accuracy: 0.9816
Epoch 86/600
- val loss: 0.0828 - val accuracy: 0.9816
Epoch 87/600
- val loss: 0.0843 - val accuracy: 0.9816
Epoch 88/600
- val_loss: 0.0828 - val_accuracy: 0.9816
Epoch 89/600
- val_loss: 0.0826 - val_accuracy: 0.9816
Epoch 90/600
- val_loss: 0.0832 - val_accuracy: 0.9816
Epoch 91/600
273/273 [============] - 1s 3ms/step - loss: 0.0829 - accuracy: 0.9803
- val loss: 0.0851 - val accuracy: 0.9816
```

```
Epoch 92/600
- val loss: 0.0843 - val accuracy: 0.9816
Epoch 93/600
- val loss: 0.0861 - val accuracy: 0.9816
Epoch 94/600
- val_loss: 0.0831 - val_accuracy: 0.9816
Epoch 95/600
- val_loss: 0.0822 - val_accuracy: 0.9816
Epoch 96/600
- val loss: 0.0829 - val accuracy: 0.9816
Epoch 97/600
- val_loss: 0.0821 - val_accuracy: 0.9816
Epoch 98/600
- val loss: 0.0830 - val accuracy: 0.9816
Epoch 99/600
- val loss: 0.0847 - val accuracy: 0.9816
Epoch 100/600
- val_loss: 0.0825 - val_accuracy: 0.9816
Epoch 101/600
- val loss: 0.0837 - val accuracy: 0.9816
Epoch 102/600
- val loss: 0.0830 - val accuracy: 0.9816
Epoch 103/600
- val_loss: 0.0823 - val_accuracy: 0.9816
Epoch 104/600
- val_loss: 0.0825 - val_accuracy: 0.9816
Epoch 105/600
- val loss: 0.0825 - val accuracy: 0.9816
Epoch 106/600
- val loss: 0.0825 - val accuracy: 0.9816
Epoch 107/600
- val loss: 0.0827 - val accuracy: 0.9816
Epoch 108/600
- val loss: 0.0823 - val accuracy: 0.9816
Epoch 109/600
- val loss: 0.0821 - val accuracy: 0.9816
Epoch 110/600
- val_loss: 0.0833 - val_accuracy: 0.9815
Epoch 111/600
- val_loss: 0.0831 - val_accuracy: 0.9816
Epoch 112/600
- val_loss: 0.0822 - val_accuracy: 0.9816
Epoch 113/600
```

```
- val loss: 0.0911 - val accuracy: 0.9816
Epoch 114/600
- val loss: 0.0821 - val accuracy: 0.9816
Epoch 115/600
- val loss: 0.0815 - val accuracy: 0.9816
Epoch 116/600
- val_loss: 0.0832 - val_accuracy: 0.9816
Epoch 117/600
- val_loss: 0.0837 - val_accuracy: 0.9816
Epoch 118/600
- val loss: 0.0865 - val accuracy: 0.9816
Epoch 119/600
- val loss: 0.0837 - val accuracy: 0.9816
Epoch 120/600
- val loss: 0.0827 - val accuracy: 0.9816
Epoch 121/600
- val_loss: 0.0829 - val_accuracy: 0.9816
Epoch 122/600
- val loss: 0.0822 - val accuracy: 0.9816
Epoch 123/600
- val loss: 0.0833 - val accuracy: 0.9816
Epoch 124/600
- val_loss: 0.0823 - val_accuracy: 0.9816
Epoch 125/600
- val_loss: 0.0835 - val_accuracy: 0.9816
Epoch 126/600
- val_loss: 0.0837 - val_accuracy: 0.9816
Epoch 127/600
- val_loss: 0.0843 - val_accuracy: 0.9816
Epoch 128/600
- val loss: 0.0830 - val accuracy: 0.9816
Epoch 129/600
- val loss: 0.0823 - val accuracy: 0.9816
Epoch 130/600
- val loss: 0.0834 - val accuracy: 0.9816
Epoch 131/600
- val_loss: 0.0878 - val_accuracy: 0.9816
Epoch 132/600
- val_loss: 0.0831 - val_accuracy: 0.9816
Epoch 133/600
- val loss: 0.0828 - val accuracy: 0.9816
Epoch 134/600
- val loss: 0.0836 - val accuracy: 0.9816
Epoch 135/600
```

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- val loss: 0.0841 - val accuracy: 0.9816
Epoch 136/600
- val loss: 0.0846 - val accuracy: 0.9816
Epoch 137/600
- val loss: 0.0822 - val accuracy: 0.9816
Epoch 138/600
- val loss: 0.0897 - val accuracy: 0.9816
Epoch 139/600
- val_loss: 0.0827 - val_accuracy: 0.9816
Epoch 140/600
- val_loss: 0.0831 - val_accuracy: 0.9816
Epoch 141/600
- val loss: 0.0848 - val accuracy: 0.9816
Epoch 142/600
- val loss: 0.0830 - val accuracy: 0.9816
Epoch 143/600
- val loss: 0.0847 - val accuracy: 0.9816
Epoch 144/600
- val loss: 0.0841 - val accuracy: 0.9816
Epoch 145/600
- val loss: 0.0831 - val accuracy: 0.9816
Epoch 146/600
- val_loss: 0.0841 - val_accuracy: 0.9816
Epoch 147/600
- val loss: 0.0869 - val accuracy: 0.9816
Epoch 148/600
- val loss: 0.0843 - val accuracy: 0.9816
Epoch 149/600
- val loss: 0.0840 - val accuracy: 0.9815
Epoch 150/600
- val loss: 0.0839 - val accuracy: 0.9816
Epoch 151/600
- val loss: 0.0847 - val accuracy: 0.9815
Epoch 152/600
- val loss: 0.0825 - val accuracy: 0.9816
Epoch 153/600
- val_loss: 0.0825 - val_accuracy: 0.9816
Epoch 154/600
- val_loss: 0.0840 - val_accuracy: 0.9816
Epoch 155/600
- val_loss: 0.0839 - val_accuracy: 0.9816
Epoch 156/600
- val loss: 0.0833 - val accuracy: 0.9816
```

```
Epoch 157/600
- val loss: 0.0832 - val accuracy: 0.9811
Epoch 158/600
- val loss: 0.0878 - val accuracy: 0.9815
Epoch 159/600
- val_loss: 0.0856 - val_accuracy: 0.9815
Epoch 160/600
- val_loss: 0.0836 - val_accuracy: 0.9815
Epoch 161/600
- val loss: 0.0833 - val accuracy: 0.9816
Epoch 162/600
- val loss: 0.0850 - val accuracy: 0.9816
Epoch 163/600
- val loss: 0.0843 - val accuracy: 0.9816
Epoch 164/600
- val loss: 0.0841 - val accuracy: 0.9815
Epoch 165/600
- val_loss: 0.0831 - val_accuracy: 0.9814
Epoch 166/600
- val loss: 0.0890 - val accuracy: 0.9812
Epoch 167/600
- val loss: 0.0861 - val accuracy: 0.9816
Epoch 168/600
- val_loss: 0.0831 - val_accuracy: 0.9815
Epoch 169/600
- val_loss: 0.0846 - val_accuracy: 0.9816
Epoch 170/600
- val loss: 0.0866 - val accuracy: 0.9816
Epoch 171/600
- val loss: 0.0838 - val accuracy: 0.9815
Epoch 172/600
- val loss: 0.0832 - val accuracy: 0.9815
Epoch 173/600
- val loss: 0.0840 - val accuracy: 0.9816
Epoch 174/600
- val loss: 0.0835 - val accuracy: 0.9816
Epoch 175/600
- val_loss: 0.0846 - val_accuracy: 0.9810
Epoch 176/600
- val_loss: 0.0836 - val_accuracy: 0.9815
Epoch 177/600
- val_loss: 0.0865 - val_accuracy: 0.9814
Epoch 178/600
```

```
- val loss: 0.0870 - val accuracy: 0.9816
Epoch 179/600
- val loss: 0.0838 - val accuracy: 0.9814
Epoch 180/600
- val loss: 0.0844 - val accuracy: 0.9815
Epoch 181/600
- val_loss: 0.0831 - val_accuracy: 0.9813
Epoch 182/600
- val_loss: 0.0856 - val_accuracy: 0.9816
Epoch 183/600
- val loss: 0.0843 - val accuracy: 0.9810
Epoch 184/600
- val loss: 0.0873 - val accuracy: 0.9811
Epoch 185/600
- val loss: 0.0839 - val accuracy: 0.9816
Epoch 186/600
- val_loss: 0.0836 - val_accuracy: 0.9815
Epoch 187/600
- val loss: 0.0850 - val accuracy: 0.9815
Epoch 188/600
- val loss: 0.0838 - val accuracy: 0.9816
Epoch 189/600
- val loss: 0.0843 - val accuracy: 0.9815
Epoch 190/600
- val loss: 0.0845 - val accuracy: 0.9811
Epoch 191/600
- val_loss: 0.0846 - val_accuracy: 0.9816
Epoch 192/600
- val_loss: 0.0854 - val_accuracy: 0.9816
Epoch 193/600
- val loss: 0.0856 - val accuracy: 0.9815
Epoch 194/600
- val loss: 0.0852 - val accuracy: 0.9816
Epoch 195/600
- val loss: 0.0862 - val accuracy: 0.9815
Epoch 196/600
- val_loss: 0.0841 - val_accuracy: 0.9816
Epoch 197/600
- val_loss: 0.0843 - val_accuracy: 0.9813
Epoch 198/600
- val loss: 0.0848 - val accuracy: 0.9811
Epoch 199/600
- val_loss: 0.0844 - val_accuracy: 0.9814
Epoch 200/600
```

```
- val loss: 0.0846 - val accuracy: 0.9816
Epoch 201/600
- val loss: 0.0839 - val accuracy: 0.9816
Epoch 202/600
val loss: 0.0853 - val accuracy: 0.9814
Epoch 203/600
- val loss: 0.0867 - val accuracy: 0.9816
Epoch 204/600
- val loss: 0.0844 - val accuracy: 0.9815
Epoch 205/600
- val_loss: 0.0854 - val_accuracy: 0.9816
Epoch 206/600
- val loss: 0.0853 - val accuracy: 0.9814
Epoch 207/600
- val loss: 0.0862 - val accuracy: 0.9810
Epoch 208/600
val loss: 0.0855 - val accuracy: 0.9809
Epoch 209/600
- val loss: 0.0846 - val accuracy: 0.9815
Epoch 210/600
- val loss: 0.0928 - val accuracy: 0.9816
Epoch 211/600
- val_loss: 0.0882 - val_accuracy: 0.9815
Epoch 212/600
- val loss: 0.0872 - val accuracy: 0.9815
Epoch 213/600
- val loss: 0.0861 - val accuracy: 0.9811
Epoch 214/600
- val loss: 0.0848 - val accuracy: 0.9815
Epoch 215/600
- val loss: 0.0868 - val accuracy: 0.9814
Epoch 216/600
- val loss: 0.0865 - val accuracy: 0.9815
Epoch 217/600
- val loss: 0.0886 - val accuracy: 0.9816
Epoch 218/600
- val_loss: 0.0873 - val_accuracy: 0.9816
Epoch 219/600
- val_loss: 0.0884 - val_accuracy: 0.9816
Epoch 220/600
- val_loss: 0.0867 - val_accuracy: 0.9815
Epoch 221/600
- val loss: 0.0875 - val accuracy: 0.9816
```

```
Epoch 222/600
- val loss: 0.0853 - val accuracy: 0.9815
Epoch 223/600
- val loss: 0.0858 - val accuracy: 0.9814
Epoch 224/600
- val_loss: 0.0868 - val_accuracy: 0.9815
Epoch 225/600
- val_loss: 0.0871 - val_accuracy: 0.9815
Epoch 226/600
- val loss: 0.0864 - val accuracy: 0.9815
Epoch 227/600
- val_loss: 0.0871 - val_accuracy: 0.9815
Epoch 228/600
- val loss: 0.0883 - val accuracy: 0.9810
Epoch 229/600
- val loss: 0.0853 - val accuracy: 0.9815
Epoch 230/600
- val_loss: 0.0869 - val_accuracy: 0.9815
Epoch 231/600
- val loss: 0.0874 - val accuracy: 0.9814
Epoch 232/600
- val loss: 0.0863 - val accuracy: 0.9815
Epoch 233/600
- val_loss: 0.0867 - val_accuracy: 0.9815
Epoch 234/600
- val_loss: 0.0880 - val_accuracy: 0.9813
Epoch 235/600
273/273 [============] - 1s 3ms/step - loss: 0.0711 - accuracy: 0.9825
val loss: 0.0937 - val accuracy: 0.9799
Epoch 236/600
- val loss: 0.0906 - val accuracy: 0.9816
Epoch 237/600
- val loss: 0.0852 - val accuracy: 0.9814
Epoch 238/600
- val_loss: 0.0868 - val_accuracy: 0.9814
Epoch 239/600
val loss: 0.0871 - val accuracy: 0.9812
Epoch 240/600
- val_loss: 0.0871 - val_accuracy: 0.9814
Epoch 241/600
- val_loss: 0.0867 - val_accuracy: 0.9815
Epoch 242/600
- val loss: 0.0894 - val accuracy: 0.9810
Epoch 243/600
```

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- val loss: 0.0887 - val accuracy: 0.9814
Epoch 244/600
- val loss: 0.0871 - val accuracy: 0.9814
Epoch 245/600
- val loss: 0.0894 - val accuracy: 0.9813
Epoch 246/600
- val_loss: 0.0879 - val_accuracy: 0.9814
Epoch 247/600
- val loss: 0.0884 - val accuracy: 0.9812
Epoch 248/600
- val loss: 0.0872 - val accuracy: 0.9813
Epoch 249/600
- val_loss: 0.0887 - val_accuracy: 0.9815
Epoch 250/600
- val loss: 0.0970 - val accuracy: 0.9793
Epoch 251/600
- val_loss: 0.0892 - val_accuracy: 0.9816
Epoch 252/600
- val loss: 0.0954 - val accuracy: 0.9801
Epoch 253/600
- val loss: 0.0865 - val accuracy: 0.9815
Epoch 254/600
- val loss: 0.0866 - val accuracy: 0.9813
Epoch 255/600
- val_loss: 0.0879 - val_accuracy: 0.9815
Epoch 256/600
- val_loss: 0.0884 - val_accuracy: 0.9813
Epoch 257/600
- val_loss: 0.0907 - val_accuracy: 0.9815
Epoch 258/600
- val loss: 0.0898 - val accuracy: 0.9815
Epoch 259/600
- val loss: 0.0897 - val accuracy: 0.9813
Epoch 260/600
- val loss: 0.0900 - val accuracy: 0.9812
Epoch 261/600
- val_loss: 0.0901 - val_accuracy: 0.9812
Epoch 262/600
- val_loss: 0.0882 - val_accuracy: 0.9814
Epoch 263/600
- val loss: 0.0882 - val accuracy: 0.9815
Epoch 264/600
- val_loss: 0.0927 - val_accuracy: 0.9814
Epoch 265/600
```

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- val loss: 0.0867 - val accuracy: 0.9814
Epoch 266/600
- val loss: 0.0881 - val accuracy: 0.9813
Epoch 267/600
- val loss: 0.0884 - val accuracy: 0.9814
Epoch 268/600
- val loss: 0.0877 - val accuracy: 0.9813
Epoch 269/600
- val loss: 0.0909 - val accuracy: 0.9815
Epoch 270/600
- val_loss: 0.0883 - val_accuracy: 0.9815
Epoch 271/600
- val loss: 0.0917 - val accuracy: 0.9815
Epoch 272/600
- val loss: 0.0939 - val accuracy: 0.9814
Epoch 273/600
val loss: 0.0894 - val accuracy: 0.9814
Epoch 274/600
- val loss: 0.0935 - val accuracy: 0.9813
Epoch 275/600
- val loss: 0.0863 - val accuracy: 0.9813
Epoch 276/600
- val loss: 0.0905 - val accuracy: 0.9815
Epoch 277/600
- val loss: 0.0884 - val accuracy: 0.9816
Epoch 278/600
- val loss: 0.0891 - val accuracy: 0.9815
Epoch 279/600
- val loss: 0.0908 - val accuracy: 0.9816
Epoch 280/600
- val loss: 0.0904 - val accuracy: 0.9815
Epoch 281/600
- val loss: 0.0877 - val accuracy: 0.9814
Epoch 282/600
- val loss: 0.0905 - val accuracy: 0.9816
Epoch 283/600
- val_loss: 0.0880 - val_accuracy: 0.9816
Epoch 284/600
- val_loss: 0.0932 - val_accuracy: 0.9815
Epoch 285/600
- val_loss: 0.0870 - val_accuracy: 0.9814
Epoch 286/600
- val loss: 0.0902 - val accuracy: 0.9814
```

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Epoch 287/600
- val loss: 0.0930 - val accuracy: 0.9816
Epoch 288/600
- val loss: 0.0900 - val accuracy: 0.9814
Epoch 289/600
- val_loss: 0.0931 - val_accuracy: 0.9815
Epoch 290/600
- val_loss: 0.0894 - val_accuracy: 0.9814
Epoch 291/600
- val loss: 0.0903 - val accuracy: 0.9815
Epoch 292/600
- val_loss: 0.0938 - val_accuracy: 0.9815
Epoch 293/600
- val loss: 0.0904 - val accuracy: 0.9816
Epoch 294/600
val loss: 0.0883 - val accuracy: 0.9814
Epoch 295/600
- val_loss: 0.0882 - val_accuracy: 0.9815
Epoch 296/600
- val loss: 0.0936 - val accuracy: 0.9808
Epoch 297/600
- val loss: 0.0901 - val accuracy: 0.9815
Epoch 298/600
- val_loss: 0.0938 - val_accuracy: 0.9801
Epoch 299/600
- val_loss: 0.0932 - val_accuracy: 0.9815
Epoch 300/600
273/273 [============] - 1s 4ms/step - loss: 0.0778 - accuracy: 0.9795
val loss: 0.0926 - val accuracy: 0.9814
Epoch 301/600
- val loss: 0.0910 - val accuracy: 0.9814
Epoch 302/600
- val loss: 0.0899 - val accuracy: 0.9815
Epoch 303/600
- val loss: 0.0940 - val accuracy: 0.9816
Epoch 304/600
val loss: 0.0935 - val accuracy: 0.9814
Epoch 305/600
- val_loss: 0.0902 - val_accuracy: 0.9815
Epoch 306/600
- val_loss: 0.0920 - val_accuracy: 0.9815
Epoch 307/600
- val_loss: 0.0931 - val_accuracy: 0.9815
Epoch 308/600
```

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- val loss: 0.0884 - val accuracy: 0.9814
Epoch 309/600
- val loss: 0.0924 - val accuracy: 0.9813
Epoch 310/600
- val loss: 0.0918 - val accuracy: 0.9815
Epoch 311/600
- val_loss: 0.0904 - val_accuracy: 0.9814
Epoch 312/600
- val_loss: 0.0897 - val_accuracy: 0.9814
Epoch 313/600
- val loss: 0.0887 - val accuracy: 0.9815
Epoch 314/600
- val loss: 0.0947 - val accuracy: 0.9815
Epoch 315/600
- val loss: 0.0901 - val accuracy: 0.9811
Epoch 316/600
- val_loss: 0.0898 - val_accuracy: 0.9810
Epoch 317/600
- val loss: 0.0990 - val accuracy: 0.9797
Epoch 318/600
- val loss: 0.0931 - val accuracy: 0.9810
Epoch 319/600
- val_loss: 0.0913 - val_accuracy: 0.9813
Epoch 320/600
- val_loss: 0.0965 - val_accuracy: 0.9815
Epoch 321/600
- val_loss: 0.0949 - val_accuracy: 0.9814
Epoch 322/600
- val loss: 0.0946 - val accuracy: 0.9815
Epoch 323/600
val loss: 0.0917 - val accuracy: 0.9814
Epoch 324/600
- val loss: 0.1034 - val accuracy: 0.9812
Epoch 325/600
- val loss: 0.0970 - val accuracy: 0.9810
Epoch 326/600
- val_loss: 0.0943 - val_accuracy: 0.9815
Epoch 327/600
- val_loss: 0.0942 - val_accuracy: 0.9811
Epoch 328/600
- val loss: 0.0921 - val accuracy: 0.9815
Epoch 329/600
- val loss: 0.0902 - val_accuracy: 0.9814
Epoch 330/600
```

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- val loss: 0.0906 - val accuracy: 0.9812
Epoch 331/600
- val loss: 0.0924 - val accuracy: 0.9814
Epoch 332/600
- val loss: 0.0922 - val accuracy: 0.9810
Epoch 333/600
val loss: 0.0983 - val accuracy: 0.9812
Epoch 334/600
- val_loss: 0.0918 - val_accuracy: 0.9812
Epoch 335/600
- val_loss: 0.0916 - val_accuracy: 0.9811
Epoch 336/600
- val loss: 0.0910 - val accuracy: 0.9811
Epoch 337/600
- val loss: 0.0914 - val accuracy: 0.9810
Epoch 338/600
val loss: 0.0933 - val accuracy: 0.9814
Epoch 339/600
- val loss: 0.0915 - val accuracy: 0.9811
Epoch 340/600
- val loss: 0.0925 - val accuracy: 0.9810
Epoch 341/600
- val loss: 0.0919 - val accuracy: 0.9814
Epoch 342/600
- val loss: 0.0927 - val accuracy: 0.9807
Epoch 343/600
- val loss: 0.0952 - val accuracy: 0.9811
Epoch 344/600
- val loss: 0.0911 - val accuracy: 0.9815
Epoch 345/600
- val loss: 0.0988 - val accuracy: 0.9815
Epoch 346/600
- val loss: 0.1001 - val accuracy: 0.9815
Epoch 347/600
- val loss: 0.0972 - val accuracy: 0.9814
Epoch 348/600
- val_loss: 0.0940 - val_accuracy: 0.9812
Epoch 349/600
- val_loss: 0.0975 - val_accuracy: 0.9815
Epoch 350/600
- val_loss: 0.0972 - val_accuracy: 0.9804
Epoch 351/600
- val loss: 0.0923 - val accuracy: 0.9813
```

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Epoch 352/600
- val loss: 0.0950 - val accuracy: 0.9809
Epoch 353/600
- val loss: 0.0914 - val accuracy: 0.9815
Epoch 354/600
- val_loss: 0.0933 - val_accuracy: 0.9815
Epoch 355/600
- val_loss: 0.0955 - val_accuracy: 0.9814
Epoch 356/600
- val loss: 0.0982 - val accuracy: 0.9799
Epoch 357/600
- val_loss: 0.0978 - val_accuracy: 0.9808
Epoch 358/600
- val loss: 0.0946 - val accuracy: 0.9813
Epoch 359/600
val loss: 0.0943 - val accuracy: 0.9798
Epoch 360/600
- val_loss: 0.0973 - val_accuracy: 0.9810
Epoch 361/600
- val loss: 0.1040 - val accuracy: 0.9810
Epoch 362/600
- val loss: 0.0942 - val accuracy: 0.9808
Epoch 363/600
- val_loss: 0.0953 - val_accuracy: 0.9812
Epoch 364/600
- val_loss: 0.0986 - val_accuracy: 0.9809
Epoch 365/600
273/273 [============] - 1s 3ms/step - loss: 0.0602 - accuracy: 0.9848
- val_loss: 0.1022 - val_accuracy: 0.9807
Epoch 366/600
- val loss: 0.0973 - val accuracy: 0.9801
Epoch 367/600
- val loss: 0.0952 - val accuracy: 0.9811
Epoch 368/600
- val_loss: 0.0951 - val_accuracy: 0.9811
Epoch 369/600
val loss: 0.0899 - val accuracy: 0.9813
Epoch 370/600
- val_loss: 0.0959 - val_accuracy: 0.9810
Epoch 371/600
- val_loss: 0.1006 - val_accuracy: 0.9814
Epoch 372/600
273/273 [============] - 1s 3ms/step - loss: 0.0701 - accuracy: 0.9829
- val_loss: 0.0970 - val_accuracy: 0.9813
Epoch 373/600
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- val loss: 0.0970 - val accuracy: 0.9815
Epoch 374/600
- val loss: 0.1010 - val accuracy: 0.9812
Epoch 375/600
- val loss: 0.0944 - val accuracy: 0.9815
Epoch 376/600
- val_loss: 0.0974 - val_accuracy: 0.9815
Epoch 377/600
- val loss: 0.0998 - val accuracy: 0.9809
Epoch 378/600
- val loss: 0.1066 - val accuracy: 0.9812
Epoch 379/600
- val loss: 0.0960 - val accuracy: 0.9806
Epoch 380/600
- val loss: 0.0975 - val accuracy: 0.9811
Epoch 381/600
- val_loss: 0.0938 - val_accuracy: 0.9804
Epoch 382/600
- val loss: 0.0948 - val accuracy: 0.9810
Epoch 383/600
- val loss: 0.0954 - val accuracy: 0.9808
Epoch 384/600
- val_loss: 0.0971 - val_accuracy: 0.9790
Epoch 385/600
- val loss: 0.0964 - val accuracy: 0.9809
Epoch 386/600
- val_loss: 0.0951 - val_accuracy: 0.9811
Epoch 387/600
- val_loss: 0.0935 - val_accuracy: 0.9815
Epoch 388/600
- val loss: 0.0980 - val accuracy: 0.9810
Epoch 389/600
- val loss: 0.0956 - val accuracy: 0.9814
Epoch 390/600
- val loss: 0.0948 - val accuracy: 0.9803
Epoch 391/600
- val_loss: 0.0991 - val_accuracy: 0.9808
Epoch 392/600
- val_loss: 0.0959 - val_accuracy: 0.9803
Epoch 393/600
- val loss: 0.0983 - val accuracy: 0.9810
Epoch 394/600
- val_loss: 0.0985 - val_accuracy: 0.9814
Epoch 395/600
```

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- val loss: 0.0977 - val accuracy: 0.9810
Epoch 396/600
- val loss: 0.0964 - val accuracy: 0.9807
Epoch 397/600
- val loss: 0.0893 - val accuracy: 0.9807
Epoch 398/600
- val loss: 0.0896 - val accuracy: 0.9813
Epoch 399/600
- val_loss: 0.0912 - val_accuracy: 0.9810
Epoch 400/600
- val_loss: 0.0963 - val_accuracy: 0.9804
Epoch 401/600
- val_loss: 0.0930 - val_accuracy: 0.9809
Epoch 402/600
- val loss: 0.0989 - val accuracy: 0.9782
Epoch 403/600
val loss: 0.0940 - val accuracy: 0.9804
Epoch 404/600
- val loss: 0.0961 - val accuracy: 0.9802
Epoch 405/600
- val loss: 0.0911 - val accuracy: 0.9805
Epoch 406/600
- val_loss: 0.0947 - val_accuracy: 0.9804
Epoch 407/600
- val loss: 0.0933 - val accuracy: 0.9814
Epoch 408/600
- val loss: 0.0953 - val accuracy: 0.9806
Epoch 409/600
- val loss: 0.0993 - val accuracy: 0.9803
Epoch 410/600
- val loss: 0.0980 - val accuracy: 0.9809
Epoch 411/600
- val loss: 0.0974 - val accuracy: 0.9803
Epoch 412/600
- val loss: 0.0969 - val accuracy: 0.9812
Epoch 413/600
- val_loss: 0.0997 - val_accuracy: 0.9809
Epoch 414/600
- val_loss: 0.1090 - val_accuracy: 0.9805
Epoch 415/600
- val_loss: 0.0961 - val_accuracy: 0.9797
Epoch 416/600
- val loss: 0.0994 - val accuracy: 0.9806
```

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Epoch 417/600
- val loss: 0.1002 - val accuracy: 0.9806
Epoch 418/600
- val loss: 0.1003 - val accuracy: 0.9809
Epoch 419/600
- val_loss: 0.0985 - val_accuracy: 0.9803
Epoch 420/600
- val_loss: 0.1030 - val_accuracy: 0.9810
Epoch 421/600
- val loss: 0.0993 - val accuracy: 0.9809
Epoch 422/600
- val_loss: 0.1039 - val_accuracy: 0.9809
Epoch 423/600
- val loss: 0.1002 - val accuracy: 0.9808
Epoch 424/600
- val loss: 0.1070 - val accuracy: 0.9795
Epoch 425/600
- val_loss: 0.1197 - val_accuracy: 0.9811
Epoch 426/600
- val loss: 0.1016 - val accuracy: 0.9811
Epoch 427/600
- val loss: 0.1006 - val accuracy: 0.9812
Epoch 428/600
- val_loss: 0.1060 - val_accuracy: 0.9809
Epoch 429/600
- val_loss: 0.1035 - val_accuracy: 0.9809
Epoch 430/600
val loss: 0.1118 - val accuracy: 0.9802
Epoch 431/600
- val loss: 0.1050 - val accuracy: 0.9808
Epoch 432/600
- val loss: 0.1004 - val accuracy: 0.9792
Epoch 433/600
- val_loss: 0.0981 - val_accuracy: 0.9802
Epoch 434/600
val loss: 0.0992 - val accuracy: 0.9809
Epoch 435/600
- val_loss: 0.1056 - val_accuracy: 0.9805
Epoch 436/600
- val_loss: 0.1072 - val_accuracy: 0.9797
Epoch 437/600
- val loss: 0.1089 - val accuracy: 0.9803
Epoch 438/600
```

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- val loss: 0.0960 - val accuracy: 0.9809
Epoch 439/600
- val loss: 0.1032 - val accuracy: 0.9809
Epoch 440/600
- val loss: 0.1037 - val accuracy: 0.9814
Epoch 441/600
- val_loss: 0.1005 - val_accuracy: 0.9808
Epoch 442/600
- val_loss: 0.1019 - val_accuracy: 0.9812
Epoch 443/600
- val loss: 0.1025 - val accuracy: 0.9812
Epoch 444/600
- val loss: 0.0986 - val accuracy: 0.9808
Epoch 445/600
- val loss: 0.1010 - val accuracy: 0.9792
Epoch 446/600
- val_loss: 0.1062 - val_accuracy: 0.9813
Epoch 447/600
- val loss: 0.1039 - val accuracy: 0.9797
Epoch 448/600
- val loss: 0.1054 - val accuracy: 0.9808
Epoch 449/600
- val loss: 0.0994 - val accuracy: 0.9806
Epoch 450/600
- val_loss: 0.1074 - val_accuracy: 0.9808
Epoch 451/600
- val loss: 0.0963 - val accuracy: 0.9810
Epoch 452/600
- val_loss: 0.1005 - val_accuracy: 0.9810
Epoch 453/600
val loss: 0.1000 - val accuracy: 0.9808
Epoch 454/600
- val_loss: 0.1152 - val_accuracy: 0.9812
Epoch 455/600
- val loss: 0.1021 - val accuracy: 0.9812
Epoch 456/600
- val_loss: 0.1080 - val_accuracy: 0.9812
Epoch 457/600
- val_loss: 0.1092 - val_accuracy: 0.9800
Epoch 458/600
- val loss: 0.1042 - val accuracy: 0.9803
Epoch 459/600
- val loss: 0.1018 - val accuracy: 0.9810
Epoch 460/600
```

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- val loss: 0.1060 - val accuracy: 0.9811
Epoch 461/600
- val loss: 0.1045 - val accuracy: 0.9803
Epoch 462/600
- val loss: 0.1040 - val accuracy: 0.9810
Epoch 463/600
val loss: 0.1093 - val accuracy: 0.9790
Epoch 464/600
- val loss: 0.0942 - val accuracy: 0.9804
Epoch 465/600
- val_loss: 0.1080 - val_accuracy: 0.9814
Epoch 466/600
- val loss: 0.1173 - val accuracy: 0.9811
Epoch 467/600
- val loss: 0.1030 - val accuracy: 0.9800
Epoch 468/600
val loss: 0.1041 - val accuracy: 0.9804
Epoch 469/600
- val loss: 0.1040 - val accuracy: 0.9809
Epoch 470/600
- val loss: 0.1064 - val accuracy: 0.9790
Epoch 471/600
- val_loss: 0.1038 - val_accuracy: 0.9809
Epoch 472/600
- val loss: 0.1079 - val accuracy: 0.9809
Epoch 473/600
val loss: 0.0991 - val accuracy: 0.9807
Epoch 474/600
- val loss: 0.1046 - val accuracy: 0.9800
Epoch 475/600
- val loss: 0.1064 - val accuracy: 0.9804
Epoch 476/600
- val loss: 0.1059 - val accuracy: 0.9812
Epoch 477/600
- val loss: 0.1066 - val accuracy: 0.9809
Epoch 478/600
- val_loss: 0.1068 - val_accuracy: 0.9805
Epoch 479/600
- val_loss: 0.1035 - val_accuracy: 0.9810
Epoch 480/600
- val_loss: 0.1170 - val_accuracy: 0.9798
Epoch 481/600
- val loss: 0.1105 - val accuracy: 0.9811
```

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Epoch 482/600
- val loss: 0.1055 - val accuracy: 0.9806
Epoch 483/600
- val loss: 0.1038 - val accuracy: 0.9803
Epoch 484/600
- val_loss: 0.1056 - val_accuracy: 0.9809
Epoch 485/600
- val_loss: 0.1102 - val_accuracy: 0.9810
Epoch 486/600
- val loss: 0.1015 - val accuracy: 0.9805
Epoch 487/600
- val_loss: 0.1104 - val_accuracy: 0.9810
Epoch 488/600
- val loss: 0.1084 - val accuracy: 0.9812
Epoch 489/600
- val loss: 0.1093 - val accuracy: 0.9806
Epoch 490/600
- val_loss: 0.1007 - val_accuracy: 0.9799
Epoch 491/600
- val loss: 0.1109 - val accuracy: 0.9806
Epoch 492/600
- val loss: 0.1040 - val accuracy: 0.9812
Epoch 493/600
- val_loss: 0.1125 - val_accuracy: 0.9807
Epoch 494/600
- val_loss: 0.0978 - val_accuracy: 0.9808
Epoch 495/600
val loss: 0.1043 - val accuracy: 0.9800
Epoch 496/600
- val loss: 0.1043 - val accuracy: 0.9806
Epoch 497/600
- val loss: 0.1034 - val accuracy: 0.9803
Epoch 498/600
- val_loss: 0.1047 - val_accuracy: 0.9809
Epoch 499/600
val loss: 0.1074 - val accuracy: 0.9802
Epoch 500/600
- val_loss: 0.1130 - val_accuracy: 0.9814
Epoch 501/600
- val_loss: 0.1017 - val_accuracy: 0.9801
Epoch 502/600
- val_loss: 0.1013 - val_accuracy: 0.9805
Epoch 503/600
```

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- val loss: 0.1064 - val accuracy: 0.9800
Epoch 504/600
- val loss: 0.1031 - val accuracy: 0.9799
Epoch 505/600
- val loss: 0.1021 - val accuracy: 0.9800
Epoch 506/600
- val_loss: 0.1073 - val_accuracy: 0.9812
Epoch 507/600
- val_loss: 0.1052 - val_accuracy: 0.9802
Epoch 508/600
- val loss: 0.1039 - val accuracy: 0.9810
Epoch 509/600
- val loss: 0.1014 - val accuracy: 0.9813
Epoch 510/600
- val loss: 0.1045 - val accuracy: 0.9810
Epoch 511/600
- val_loss: 0.1072 - val_accuracy: 0.9805
Epoch 512/600
- val loss: 0.1244 - val accuracy: 0.9802
Epoch 513/600
- val loss: 0.1024 - val accuracy: 0.9800
Epoch 514/600
- val loss: 0.1024 - val accuracy: 0.9809
Epoch 515/600
- val_loss: 0.1113 - val_accuracy: 0.9811
Epoch 516/600
- val_loss: 0.1056 - val_accuracy: 0.9811
Epoch 517/600
- val_loss: 0.0974 - val_accuracy: 0.9806
Epoch 518/600
val loss: 0.1061 - val accuracy: 0.9798
Epoch 519/600
- val loss: 0.1020 - val accuracy: 0.9803
Epoch 520/600
- val loss: 0.1063 - val accuracy: 0.9805
Epoch 521/600
- val_loss: 0.1024 - val_accuracy: 0.9807
Epoch 522/600
- val_loss: 0.1073 - val_accuracy: 0.9802
Epoch 523/600
- val loss: 0.1052 - val accuracy: 0.9807
Epoch 524/600
- val_loss: 0.1117 - val_accuracy: 0.9804
Epoch 525/600
```

```
- val loss: 0.1063 - val accuracy: 0.9798
Epoch 526/600
- val loss: 0.1150 - val accuracy: 0.9807
Epoch 527/600
- val loss: 0.1016 - val accuracy: 0.9811
Epoch 528/600
val loss: 0.1024 - val accuracy: 0.9807
Epoch 529/600
- val_loss: 0.1137 - val_accuracy: 0.9804
Epoch 530/600
- val_loss: 0.1122 - val_accuracy: 0.9810
Epoch 531/600
- val_loss: 0.1114 - val_accuracy: 0.9804
Epoch 532/600
- val loss: 0.1137 - val accuracy: 0.9806
Epoch 533/600
val loss: 0.1173 - val accuracy: 0.9771
Epoch 534/600
- val loss: 0.1176 - val accuracy: 0.9802
Epoch 535/600
- val loss: 0.1084 - val accuracy: 0.9805
Epoch 536/600
- val_loss: 0.1131 - val_accuracy: 0.9803
Epoch 537/600
- val loss: 0.1131 - val accuracy: 0.9803
Epoch 538/600
val loss: 0.1082 - val accuracy: 0.9799
Epoch 539/600
- val loss: 0.1057 - val accuracy: 0.9803
Epoch 540/600
- val loss: 0.1093 - val accuracy: 0.9801
Epoch 541/600
- val loss: 0.1106 - val accuracy: 0.9798
Epoch 542/600
- val loss: 0.1074 - val accuracy: 0.9803
Epoch 543/600
- val_loss: 0.1136 - val_accuracy: 0.9800
Epoch 544/600
- val_loss: 0.1079 - val_accuracy: 0.9807
Epoch 545/600
- val_loss: 0.1130 - val_accuracy: 0.9803
Epoch 546/600
- val loss: 0.1021 - val accuracy: 0.9804
```

```
Epoch 547/600
- val loss: 0.1029 - val accuracy: 0.9805
Epoch 548/600
- val loss: 0.1080 - val accuracy: 0.9805
Epoch 549/600
- val_loss: 0.1054 - val_accuracy: 0.9803
Epoch 550/600
- val_loss: 0.1093 - val_accuracy: 0.9800
Epoch 551/600
- val loss: 0.1084 - val accuracy: 0.9808
Epoch 552/600
- val_loss: 0.1084 - val_accuracy: 0.9800
Epoch 553/600
- val loss: 0.1053 - val accuracy: 0.9812
Epoch 554/600
- val loss: 0.1048 - val accuracy: 0.9806
Epoch 555/600
- val_loss: 0.1114 - val_accuracy: 0.9807
Epoch 556/600
- val loss: 0.1088 - val accuracy: 0.9801
Epoch 557/600
- val loss: 0.1113 - val accuracy: 0.9805
Epoch 558/600
- val_loss: 0.1123 - val_accuracy: 0.9805
Epoch 559/600
- val_loss: 0.1051 - val_accuracy: 0.9812
Epoch 560/600
273/273 [============] - 1s 4ms/step - loss: 0.0681 - accuracy: 0.9817
- val_loss: 0.1072 - val_accuracy: 0.9807
Epoch 561/600
- val loss: 0.1026 - val accuracy: 0.9807
Epoch 562/600
- val loss: 0.1126 - val accuracy: 0.9808
Epoch 563/600
- val loss: 0.1109 - val accuracy: 0.9809
Epoch 564/600
val loss: 0.1080 - val accuracy: 0.9812
Epoch 565/600
- val_loss: 0.1162 - val_accuracy: 0.9801
Epoch 566/600
- val_loss: 0.1212 - val_accuracy: 0.9799
Epoch 567/600
- val_loss: 0.1201 - val_accuracy: 0.9808
Epoch 568/600
```

```
- val loss: 0.1113 - val accuracy: 0.9801
Epoch 569/600
- val loss: 0.1077 - val accuracy: 0.9809
Epoch 570/600
- val loss: 0.1112 - val accuracy: 0.9803
Epoch 571/600
- val_loss: 0.1112 - val_accuracy: 0.9803
Epoch 572/600
- val_loss: 0.1125 - val_accuracy: 0.9805
Epoch 573/600
- val loss: 0.1223 - val accuracy: 0.9804
Epoch 574/600
- val loss: 0.1176 - val accuracy: 0.9802
Epoch 575/600
- val loss: 0.1206 - val accuracy: 0.9806
Epoch 576/600
- val_loss: 0.1207 - val_accuracy: 0.9808
Epoch 577/600
- val loss: 0.1159 - val accuracy: 0.9808
Epoch 578/600
- val loss: 0.1182 - val accuracy: 0.9800
Epoch 579/600
- val_loss: 0.1196 - val_accuracy: 0.9772
Epoch 580/600
- val_loss: 0.1120 - val_accuracy: 0.9792
Epoch 581/600
- val_loss: 0.1069 - val_accuracy: 0.9806
Epoch 582/600
- val_loss: 0.1088 - val_accuracy: 0.9805
Epoch 583/600
- val loss: 0.1109 - val accuracy: 0.9799
Epoch 584/600
- val loss: 0.1114 - val accuracy: 0.9810
Epoch 585/600
- val loss: 0.1099 - val accuracy: 0.9806
Epoch 586/600
- val_loss: 0.1076 - val_accuracy: 0.9810
Epoch 587/600
- val_loss: 0.1209 - val_accuracy: 0.9808
Epoch 588/600
- val loss: 0.1175 - val accuracy: 0.9791
Epoch 589/600
- val_loss: 0.1180 - val_accuracy: 0.9808
Epoch 590/600
```

```
- val loss: 0.1176 - val accuracy: 0.9791
     Epoch 591/600
     - val loss: 0.1309 - val accuracy: 0.9802
     Epoch 592/600
     val loss: 0.0995 - val accuracy: 0.9797
     Epoch 593/600
     - val loss: 0.1063 - val accuracy: 0.9800
     Epoch 594/600
     - val loss: 0.1166 - val accuracy: 0.9808
     Epoch 595/600
     - val_loss: 0.1175 - val_accuracy: 0.9804
     Epoch 596/600
     - val_loss: 0.1158 - val_accuracy: 0.9805
     Epoch 597/600
     - val loss: 0.1141 - val accuracy: 0.9805
     Epoch 598/600
     val loss: 0.1136 - val accuracy: 0.9803
     Epoch 599/600
     - val loss: 0.1322 - val accuracy: 0.9806
     Epoch 600/600
     - val loss: 0.1083 - val accuracy: 0.9808
Out[85]: <tensorflow.python.keras.callbacks.History at 0x7f721f276750>
      predictions =ann_model.predict(x_test)[:,0]
In [94]:
      predictions[0:10]
Out[94]: array([5.1400346e-01, 6.3963707e-10, 2.3623914e-02, 8.3117070e-29,
          5.2477092e-02, 9.6591539e-06, 9.7921491e-04, 2.3538500e-02,
         8.6560249e-03, 6.8458521e-06], dtype=float32)
In [100...
      predictions = np.round(ann model.predict(x test)[:,0])
      predictions[0:10]
Out[100... array([1., 0., 0., 0., 0., 0., 0., 0., 0.], dtype=float32)
     CONFUSION MATRIX AND CLASIFICATION REPORT TO EVALUATE PERFORMANCE
In [89]:
      from sklearn.metrics import classification report , confusion matrix
In [102...
      print(classification report(y test,predictions))
              precision
                      recall f1-score
                                 support
            0
                 0.98
                       1.00
                             0.99
                                  19970
            1
                 0.11
                       0.01
                             0.01
                                   375
                             0.98
                                  20345
        accuracy
                             0.50
                                  20345
                 0.54
                       0.50
       macro avg
                             0.97
                                  20345
     weighted avg
                 0.97
                       0.98
```