```
In [1]: import tensorflow as tf
    from tensorflow import keras
    from tensorflow.keras import Sequential
    from tensorflow.keras.layers import Flatten, Dense, Dropout, BatchNormalizatic
    from tensorflow.keras.layers import Conv1D, MaxPooling1D
    from keras.layers import SpatialDropout1D
    from tensorflow.keras.optimizers import Adam
    print(tf.__version__)

2.11.0
```

In [41]: import pandas as pd
 import numpy as np
 import seaborn as sns
 import matplotlib.pyplot as plt
 from sklearn.model\_selection import train\_test\_split
 from sklearn.preprocessing import StandardScaler

In [42]: data = pd.read\_csv('creditcard.csv')
 data.head()

Out[42]:		Time	V1	V2	V3	V4	V5	V6	V7	V8	
	0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0
	1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0
	2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1
	3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1
	4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0

5 rows × 31 columns

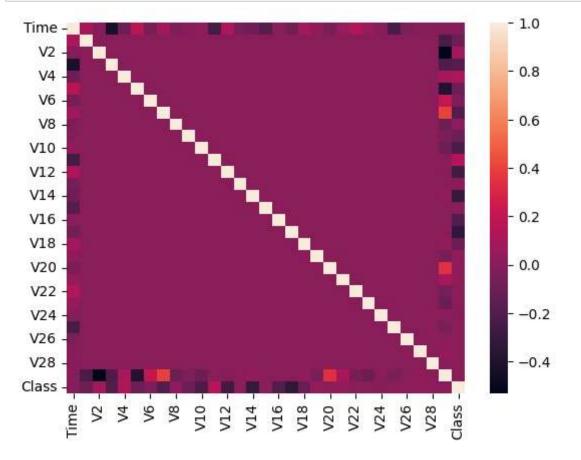
In [5]: data.shape

Out[5]: (284807, 31)

In [6]: data.isnull().sum() Out[6]: Time 0 V1 0 V2 0 ٧3 0 ٧4 0 0 V5 ۷6 0 ٧7 0 V8 0 ۷9 0 V10 0 0 V11 0 V12 V13 0 V14 0 V15 0 V16 0 V17 0 V18 0 V19 0 V20 0 V21 0 V22 0 V23 0 V24 0 V25 0 V26 0 V27 0 V28 0 Amount Class

dtype: int64

```
In [8]: dataplot=sns.heatmap(data.corr())
plt.show()
```



```
In [48]: data = fraud.append(non_fraud, ignore_index=True)
    data
```

C:\Users\Administrator\AppData\Local\Temp\ipykernel\_7648\96343452.py:1: Futur eWarning: The frame.append method is deprecated and will be removed from pand as in a future version. Use pandas.concat instead.

	data = fraud.append(non_fraud, ignore_index=True)									
Out[48]:		Time	V1	V2	V3	V4	V5	V6	<b>V</b> 7	
	0	406.0	-2.312227	1.951992	-1.609851	3.997906	-0.522188	-1.426545	-2.537387	1.391(
	1	472.0	-3.043541	-3.157307	1.088463	2.288644	1.359805	-1.064823	0.325574	-0.067
	2	4462.0	-2.303350	1.759247	-0.359745	2.330243	-0.821628	-0.075788	0.562320	<b>-</b> 0.399
	3	6986.0	-4.397974	1.358367	-2.592844	2.679787	-1.128131	-1.706536	-3.496197	-0.248
	4	7519.0	1.234235	3.019740	<b>-</b> 4.304597	4.732795	3.624201	-1.357746	1.713445	-0.496
	979	125205.0	0.080544	0.744861	-0.029200	-0.743787	0.664837	-0.666655	0.905852	-0.0678
	980	79826.0	1.242763	0.056145	0.045949	0.109549	-0.435421	-1.330599	0.307449	<b>-</b> 0.283
	981	73745.0	-0.670076	1.037461	-0.678171	-1.056501	2.244517	3.200812	-0.241885	1.363
	982	132947.0	-1.143580	1.173592	1.107021	-0.453706	-0.646007	-0.120689	0.002402	0.470
	983	75904.0	1.104687	0.043815	0.970525	1.418752	-0.849291	-0.659503	-0.127395	-0.0428
	984 r	ows × 31	columns							
	<b>←</b>								•	
In [49]:	data	['Class'	].value_c	ounts()						
Out[49]:	1 0 Name	492 492 : Class,	dtype: i	.nt64						
In [50]:	<pre>X = data.drop('Class', axis = 1) y = data['Class']</pre>									
In [51]:	X_tr	ain, X_t	est, y_tr	ain, y_t	est = tra	in_test_s	split(X,	y, test_s	size = 0.	2, rar

```
In [52]: X_train.shape, X_test.shape
```

Out[52]: ((787, 30), (197, 30))

```
In [53]: scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

#### **CNN MODEL**

```
In [68]: epochs = 20
    model = Sequential()
    model.add(Conv1D(32, 2, activation='relu', input_shape = X_train[0].shape))
    model.add(BatchNormalization())
    model.add(Dropout(0.2))

model.add(BatchNormalization())
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

model.add(Flatten())
    model.add(Dense(64, activation='relu'))
    model.add(Dropout(0.5))

model.add(Dense(1, activation='sigmoid'))
```

#### In [69]: model.summary()

Model: "sequential\_18"

Layer (type) Output Shape Param	# ====
conv1d 14 (Conv1D) (None, 29, 32) 96	====
<pre>batch_normalization_5 (Batc (None, 29, 32)</pre>	
dropout_16 (Dropout) (None, 29, 32) 0	
conv1d_15 (Conv1D) (None, 28, 64) 4160	
batch_normalization_6 (Batc (None, 28, 64) 256 hNormalization)	
dropout_17 (Dropout) (None, 28, 64) 0	
flatten_2 (Flatten) (None, 1792) 0	
dense_24 (Dense) (None, 64) 114752	2
dropout_18 (Dropout) (None, 64) 0	
dense_25 (Dense) (None, 1) 65	

Total params: 119,457 Trainable params: 119,265 Non-trainable params: 192

In [70]: model.compile(optimizer=Adam(lr=0.0001), loss = 'binary\_crossentropy', metrics

WARNING:absl:`lr` is deprecated, please use `learning\_rate` instead, or use the legacy optimizer, e.g.,tf.keras.optimizers.legacy.Adam.

In [71]: history = model.fit(X\_train, y\_train, epochs=epochs, validation\_data=(X\_test,

```
Epoch 1/20
cy: 0.8475 - val_loss: 0.5386 - val_accuracy: 0.5330
25/25 [============== ] - 0s 8ms/step - loss: 0.2522 - accurac
y: 0.9199 - val_loss: 0.5274 - val_accuracy: 0.5025
Epoch 3/20
25/25 [============ ] - 0s 8ms/step - loss: 0.2320 - accurac
y: 0.9276 - val_loss: 0.5684 - val_accuracy: 0.4975
Epoch 4/20
25/25 [============== ] - 0s 8ms/step - loss: 0.1492 - accurac
y: 0.9339 - val_loss: 0.6860 - val_accuracy: 0.4975
Epoch 5/20
25/25 [============== ] - 0s 8ms/step - loss: 0.1779 - accurac
y: 0.9352 - val_loss: 0.7443 - val_accuracy: 0.4975
Epoch 6/20
25/25 [============= ] - 0s 8ms/step - loss: 0.1560 - accurac
y: 0.9504 - val_loss: 0.6206 - val_accuracy: 0.4975
Epoch 7/20
y: 0.9466 - val loss: 0.7656 - val accuracy: 0.4975
y: 0.9555 - val_loss: 0.6463 - val_accuracy: 0.5076
Epoch 9/20
y: 0.9479 - val_loss: 0.7070 - val_accuracy: 0.5127
Epoch 10/20
25/25 [============= ] - 0s 7ms/step - loss: 0.1302 - accurac
y: 0.9555 - val loss: 0.4579 - val accuracy: 0.7005
Epoch 11/20
25/25 [============= ] - 0s 7ms/step - loss: 0.1426 - accurac
y: 0.9504 - val_loss: 0.5595 - val_accuracy: 0.6041
Epoch 12/20
25/25 [============= ] - 0s 7ms/step - loss: 0.1171 - accurac
y: 0.9517 - val loss: 0.3684 - val accuracy: 0.8173
Epoch 13/20
25/25 [============= ] - 0s 8ms/step - loss: 0.1274 - accurac
y: 0.9581 - val_loss: 0.2945 - val_accuracy: 0.8680
Epoch 14/20
25/25 [============= ] - 0s 10ms/step - loss: 0.1134 - accura
cy: 0.9492 - val_loss: 0.2776 - val_accuracy: 0.8883
Epoch 15/20
25/25 [============= ] - 0s 8ms/step - loss: 0.1178 - accurac
y: 0.9670 - val_loss: 0.2099 - val_accuracy: 0.9289
Epoch 16/20
25/25 [============== ] - 0s 8ms/step - loss: 0.1209 - accurac
y: 0.9606 - val_loss: 0.1825 - val_accuracy: 0.9239
Epoch 17/20
25/25 [============ ] - 0s 8ms/step - loss: 0.1097 - accurac
y: 0.9606 - val_loss: 0.1928 - val_accuracy: 0.9289
Epoch 18/20
y: 0.9606 - val_loss: 0.1852 - val_accuracy: 0.9340
Epoch 19/20
25/25 [============== ] - 0s 8ms/step - loss: 0.0978 - accurac
y: 0.9657 - val_loss: 0.1660 - val_accuracy: 0.9239
```

```
In [72]: model = Sequential()
  model.add(Conv1D(16, 2, activation='relu', input_shape = X_train[0].shape))
  model.add(Dense(16, activation='relu'))
  model.add(Dropout(0.5))

model.add(Dense(16, activation='sigmoid'))
  model.summary()
```

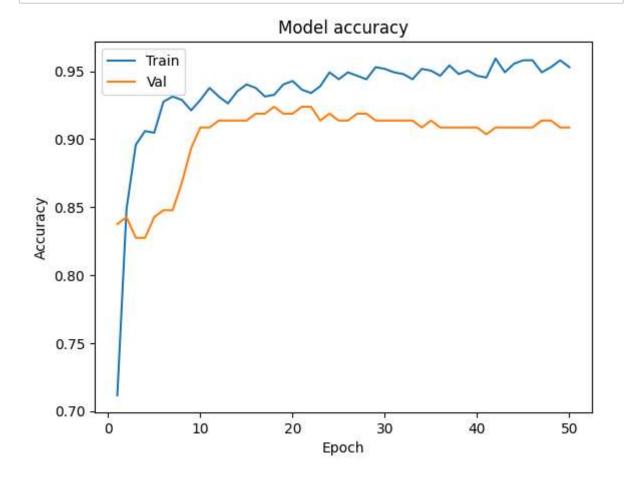
Model: "sequential\_19"

Layer (type)	Output Shape	Param #
conv1d_16 (Conv1D)	(None, 29, 16)	48
dense_26 (Dense)	(None, 29, 16)	272
dropout_19 (Dropout)	(None, 29, 16)	0
dense_27 (Dense)	(None, 29, 16)	272
Total params: 592	=======================================	=======

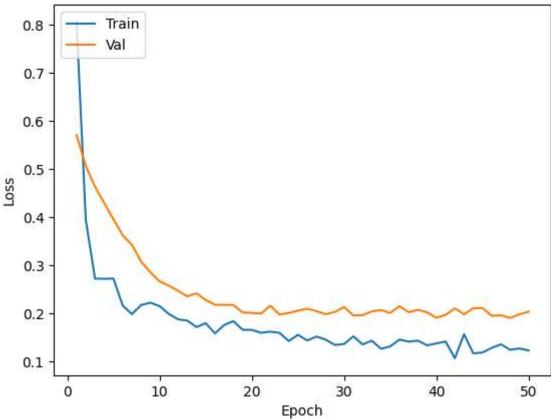
Trainable params: 592
Non-trainable params: 0

```
In [24]: | def plot_learningCurve(history, epoch):
           # Plot training & validation accuracy values
           epoch range = range(1, epoch+1)
           plt.plot(epoch range, history.history['accuracy'])
           plt.plot(epoch_range, history.history['val_accuracy'])
           plt.title('Model accuracy')
           plt.ylabel('Accuracy')
           plt.xlabel('Epoch')
           plt.legend(['Train', 'Val'], loc='upper left')
           plt.show()
           # Plot training & validation loss values
           plt.plot(epoch_range, history.history['loss'])
           plt.plot(epoch_range, history.history['val_loss'])
           plt.title('Model loss')
           plt.ylabel('Loss')
           plt.xlabel('Epoch')
           plt.legend(['Train', 'Val'], loc='upper left')
           plt.show()
```

In [25]: plot\_learningCurve(history, epochs)



# Model loss



**Adding MaxPool Layer** 

```
In [26]: epochs = 50
        model = Sequential()
        model.add(Conv1D(32, 2, activation='relu', input_shape = X_train[0].shape))
        model.add(BatchNormalization())
        model.add(MaxPool1D(2))
        model.add(Dropout(0.2))
        model.add(Conv1D(64, 2, activation='relu'))
        model.add(BatchNormalization())
        model.add(MaxPool1D(2))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(64, activation='relu'))
        model.add(Dropout(0.5))
        model.add(Dense(1, activation='sigmoid'))
        model.compile(optimizer=Adam(lr=0.0001), loss = 'binary_crossentropy', metrics
        history = model.fit(X_train, y_train, epochs=epochs, validation_data=(X_test,
        plot_learningCurve(history, epochs)
        WARNING:absl:`lr` is deprecated, please use `learning rate` instead, or us
        e the legacy optimizer, e.g.,tf.keras.optimizers.legacy.Adam.
        Epoch 1/50
        uracy: 0.7408 - val loss: 0.5620 - val accuracy: 0.8477
        Epoch 2/50
        25/25 [============ ] - 0s 11ms/step - loss: 0.3311 - acc
        uracy: 0.8704 - val loss: 0.5066 - val accuracy: 0.8629
        25/25 [=============== ] - 0s 13ms/step - loss: 0.3050 - acc
        uracy: 0.8844 - val loss: 0.4616 - val accuracy: 0.8883
        Epoch 4/50
        25/25 [============ ] - 0s 13ms/step - loss: 0.2693 - acc
        uracy: 0.9136 - val_loss: 0.4165 - val_accuracy: 0.9036
        Epoch 5/50
        25/25 [=========== ] - 0s 13ms/step - loss: 0.2259 - acc
        uracy: 0.9174 - val loss: 0.3729 - val accuracy: 0.9086
        Epoch 6/50
```

## **Architecture of 14 Layers**

```
In [27]:
    epochs = 100
    model = Sequential()
    model.add(Conv1D(32, 2, activation='relu', input_shape = X_train[0].shape))
    model.add(BatchNormalization())
    model.add(Dropout(0.2))

model.add(Conv1D(64, 2, activation='relu'))
    model.add(BatchNormalization())
    model.add(Dropout(0.5))

model.add(Dense(64, activation='relu'))
    model.add(Dense(64, activation='relu'))
    model.add(Dense(100, activation='relu'))
    model.add(Dense(50, activation='relu'))
    model.add(Dense(25, activation='relu'))

model.add(Dense(1, activation='relu'))
```

#### In [28]: model.summary()

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
conv1d_6 (Conv1D)	(None, 29, 32)	96
<pre>batch_normalization_6 (Batc hNormalization)</pre>	(None, 29, 32)	128
dropout_9 (Dropout)	(None, 29, 32)	0
conv1d_7 (Conv1D)	(None, 28, 64)	4160
<pre>batch_normalization_7 (Batc hNormalization)</pre>	(None, 28, 64)	256
dropout_10 (Dropout)	(None, 28, 64)	0
flatten_3 (Flatten)	(None, 1792)	0
dense_6 (Dense)	(None, 64)	114752
dropout_11 (Dropout)	(None, 64)	0
dense_7 (Dense)	(None, 100)	6500
dense_8 (Dense)	(None, 50)	5050
dense_9 (Dense)	(None, 25)	1275
dense_10 (Dense)	(None, 1)	26

.------

Total params: 132,243 Trainable params: 132,051 Non-trainable params: 192

In [29]: model.compile(optimizer=Adam(lr=0.0001), loss = 'binary\_crossentropy', metrics

WARNING:absl:`lr` is deprecated, please use `learning\_rate` instead, or use the legacy optimizer, e.g.,tf.keras.optimizers.legacy.Adam.

```
history = model.fit(X_train, y_train, epochs=epochs, validation_data=(X_test,
Epoch 1/100
uracy: 0.8272 - val_loss: 0.5518 - val_accuracy: 0.8985
Epoch 2/100
25/25 [============== ] - Os 14ms/step - loss: 0.2461 - acc
uracy: 0.9123 - val_loss: 0.4877 - val_accuracy: 0.9340
Epoch 3/100
uracy: 0.9327 - val_loss: 0.4258 - val_accuracy: 0.9188
Epoch 4/100
uracy: 0.9288 - val_loss: 0.4292 - val_accuracy: 0.7614
Epoch 5/100
25/25 [============ ] - 0s 16ms/step - loss: 0.1903 - acc
uracy: 0.9314 - val_loss: 0.3944 - val_accuracy: 0.8883
Epoch 6/100
25/25 [============ ] - 0s 15ms/step - loss: 0.1675 - acc
uracy: 0.9403 - val_loss: 0.3480 - val_accuracy: 0.9188
Epoch 7/100
```

### **Architecture of 17 Layers**

```
In [47]: epochs = 100
         model = Sequential()
         model.add(Conv1D(32, 2, activation='relu', input_shape = X_train[0].shape))
         model.add(BatchNormalization())
         model.add(Dropout(0.2))
         model.add(Conv1D(64, 2, activation='relu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.5))
         model.add(Conv1D(64, 2, activation='relu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.25))
         model.add(Flatten())
         model.add(Dense(64, activation='relu'))
         model.add(Dropout(0.5))
         model.add(Dense(100, activation='relu'))
         model.add(Dense(50,activation='relu'))
         model.add(Dense(25,activation='relu'))
         model.add(Dense(1, activation='sigmoid'))
         model.summary()
```

Model: "sequential\_10"

Layer (type)	Output Shape	Param #
conv1d_22 (Conv1D)	(None, 29, 32)	96
<pre>batch_normalization_22 (Bat chNormalization)</pre>	(None, 29, 32)	128
dropout_29 (Dropout)	(None, 29, 32)	0
conv1d_23 (Conv1D)	(None, 28, 64)	4160
<pre>batch_normalization_23 (Bat chNormalization)</pre>	(None, 28, 64)	256
dropout_30 (Dropout)	(None, 28, 64)	0
conv1d_24 (Conv1D)	(None, 27, 64)	8256
<pre>batch_normalization_24 (Bat chNormalization)</pre>	(None, 27, 64)	256
dropout_31 (Dropout)	(None, 27, 64)	0
flatten_7 (Flatten)	(None, 1728)	0
dense_28 (Dense)	(None, 64)	110656
dropout_32 (Dropout)	(None, 64)	0
dense_29 (Dense)	(None, 3)	195
dense_30 (Dense)	(None, 1)	4

\_\_\_\_\_

Total params: 124,007 Trainable params: 123,687 Non-trainable params: 320

In [50]: model.compile(loss = 'binary\_crossentropy', metrics=['accuracy'])

```
In [51]: history = model.fit(X_train, y_train, epochs=epochs, validation_data=(X_test,
       Epoch 1/100
       25/25 [============== ] - 2s 20ms/step - loss: 0.1232 - acc
       uracy: 0.9733 - val_loss: 0.5547 - val_accuracy: 0.9188
       Epoch 2/100
       25/25 [============== ] - Os 10ms/step - loss: 0.1233 - acc
       uracy: 0.9809 - val_loss: 0.5359 - val_accuracy: 0.9239
       Epoch 3/100
       uracy: 0.9822 - val_loss: 0.5835 - val_accuracy: 0.9137
       Epoch 4/100
       uracy: 0.9746 - val_loss: 0.5435 - val_accuracy: 0.9239
       Epoch 5/100
       25/25 [============ ] - 0s 11ms/step - loss: 0.1063 - acc
       uracy: 0.9848 - val_loss: 0.6482 - val_accuracy: 0.9137
       Epoch 6/100
       25/25 [============ ] - 0s 10ms/step - loss: 0.1308 - acc
       uracy: 0.9771 - val_loss: 0.5573 - val_accuracy: 0.9239
       Epoch 7/100
```

### **Architecture of 20 Layers**

```
In [31]: epochs = 100
         model = Sequential()
         model.add(Conv1D(32, 2, activation='relu', input_shape = X_train[0].shape))
         model.add(BatchNormalization())
         model.add(Dropout(0.2))
         model.add(Conv1D(64, 2, activation='relu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.5))
         model.add(Conv1D(64, 2, activation='relu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.5))
         model.add(Conv1D(64, 2, activation='relu'))
         model.add(BatchNormalization())
         model.add(Dropout(0.25))
         model.add(Flatten())
         model.add(Dense(64, activation='relu'))
         model.add(Dropout(0.5))
         model.add(Dense(100, activation='relu'))
         model.add(Dense(50, activation='relu'))
         model.add(Dense(25, activation='relu'))
         model.add(Dense(1, activation='sigmoid'))
         model.summary()
```

Model: "sequential\_4"

Layer (type)	Output Shape	Param #
conv1d_8 (Conv1D)		96
<pre>batch_normalization_8 (Batc hNormalization)</pre>	(None, 29, 32)	128
dropout_12 (Dropout)	(None, 29, 32)	0
conv1d_9 (Conv1D)	(None, 28, 64)	4160
<pre>batch_normalization_9 (Batc hNormalization)</pre>	(None, 28, 64)	256
dropout_13 (Dropout)	(None, 28, 64)	0
conv1d_10 (Conv1D)	(None, 27, 64)	8256
<pre>batch_normalization_10 (Bat chNormalization)</pre>	(None, 27, 64)	256
dropout_14 (Dropout)	(None, 27, 64)	0
conv1d_11 (Conv1D)	(None, 26, 64)	8256
<pre>batch_normalization_11 (Bat chNormalization)</pre>	(None, 26, 64)	256
dropout_15 (Dropout)	(None, 26, 64)	0
flatten_4 (Flatten)	(None, 1664)	0
dense_11 (Dense)	(None, 64)	106560
dropout_16 (Dropout)	(None, 64)	0
dense_12 (Dense)	(None, 100)	6500
dense_13 (Dense)	(None, 50)	5050
dense_14 (Dense)	(None, 25)	1275
dense_15 (Dense)	(None, 1)	26
		=======

Total params: 141,075 Trainable params: 140,627 Non-trainable params: 448

In [32]: model.compile(optimizer=Adam(lr=0.0001), loss = 'binary\_crossentropy', metrics

WARNING:absl:`lr` is deprecated, please use `learning\_rate` instead, or use the legacy optimizer, e.g.,tf.keras.optimizers.legacy.Adam.

```
history = model.fit(X train, y train, epochs=epochs, validation data=(X test,
In [33]:
      Epoch 1/100
      uracy: 0.7421 - val_loss: 0.6665 - val_accuracy: 0.5228
      Epoch 2/100
      25/25 [============= ] - 1s 24ms/step - loss: 0.2991 - acc
      uracy: 0.8971 - val_loss: 0.5551 - val_accuracy: 0.8071
      Epoch 3/100
      uracy: 0.9098 - val_loss: 0.4415 - val_accuracy: 0.9188
      Epoch 4/100
      uracy: 0.9136 - val_loss: 0.3537 - val_accuracy: 0.9137
      Epoch 5/100
      uracy: 0.9314 - val_loss: 0.2989 - val_accuracy: 0.9188
      Epoch 6/100
      25/25 [============ ] - 1s 21ms/step - loss: 0.2117 - acc
      uracy: 0.9288 - val_loss: 0.2739 - val_accuracy: 0.9137
      Epoch 7/100
      ar /ar r
                                               A 4047
In [ ]:
```

localhost:8888/notebooks/CNN cred2.ipynb