

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import confusion_matrix,accuracy_score,classification_report,ConfusionMatrixDisplay
df=pd.read_csv('/content/drive/MyDrive/creditcard.csv')
df
```

	Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	...	V21	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0.363787	...	-0.018307	0
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0.255425	...	-0.225775	-0
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1.514654	...	0.247998	0
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1.387024	...	-0.108300	0
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0.817739	...	-0.009431	0
...	
284802	172786.0	-11.881118	10.071785	-9.834783	-2.066656	-5.364473	-2.606837	-4.918215	7.305334	1.914428	...	0.213454	0
284803	172787.0	-0.732789	-0.055080	2.035030	-0.738589	0.868229	1.058415	0.024330	0.294869	0.584800	...	0.214205	0
284804	172788.0	1.919565	-0.301254	-3.249640	-0.557828	2.630515	3.031260	-0.296827	0.708417	0.432454	...	0.232045	0
284805	172788.0	-0.240440	0.530483	0.702510	0.689799	-0.377961	0.623708	-0.686180	0.679145	0.392087	...	0.265245	0
284806	172792.0	-0.533413	-0.189733	0.703337	-0.506271	-0.012546	-0.649617	1.577006	-0.414650	0.486180	...	0.261057	0

284807 rows × 31 columns

```
df.head()
```

	Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	...	V21	V22	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0.363787	...	-0.018307	0.277838	-0.
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0.255425	...	-0.225775	-0.638672	0.
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1.514654	...	0.247998	0.771679	0.
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1.387024	...	-0.108300	0.005274	-0.
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0.817739	...	-0.009431	0.798278	-0.

5 rows × 31 columns

```
df.tail()
```

	Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	...	V21	
284802	172786.0	-11.881118	10.071785	-9.834783	-2.066656	-5.364473	-2.606837	-4.918215	7.305334	1.914428	...	0.213454	0.1
284803	172787.0	-0.732789	-0.055080	2.035030	-0.738589	0.868229	1.058415	0.024330	0.294869	0.584800	...	0.214205	0.9
284804	172788.0	1.919565	-0.301254	-3.249640	-0.557828	2.630515	3.031260	-0.296827	0.708417	0.432454	...	0.232045	0.5
284805	172788.0	-0.240440	0.530483	0.702510	0.689799	-0.377961	0.623708	-0.686180	0.679145	0.392087	...	0.265245	0.8
284806	172792.0	-0.533413	-0.189733	0.703337	-0.506271	-0.012546	-0.649617	1.577006	-0.414650	0.486180	...	0.261057	0.6

5 rows × 31 columns


```
df.columns
```

```
Index(['Time', 'V1', 'V2', 'V3', 'V4', 'V5', 'V6', 'V7', 'V8', 'V9', 'V10',
      'V11', 'V12', 'V13', 'V14', 'V15', 'V16', 'V17', 'V18', 'V19', 'V20',
      'V21', 'V22', 'V23', 'V24', 'V25', 'V26', 'V27', 'V28', 'Amount',
      'Class'],
      dtype='object')
```

```
df.shape
```


```
(284807, 31)
```

```
df.isna().sum()
```




	0
Time	0
V1	0
V2	0
V3	0
V4	0
V5	0
V6	0
V7	0
V8	0
V9	0
V10	0
V11	0
V12	0
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	0
Class	0

df.dtypes



	0
Time	float64
V1	float64
V2	float64
V3	float64
V4	float64
V5	float64
V6	float64
V7	float64
V8	float64
V9	float64
V10	float64
V11	float64
V12	float64
V13	float64
V14	float64
V15	float64
V16	float64
V17	float64
V18	float64
V19	float64
V20	float64
V21	float64
V22	float64
V23	float64
V24	float64
V25	float64
V26	float64
V27	float64
V28	float64
Amount	float64
Class	int64

```
df['Class'].value_counts()
```

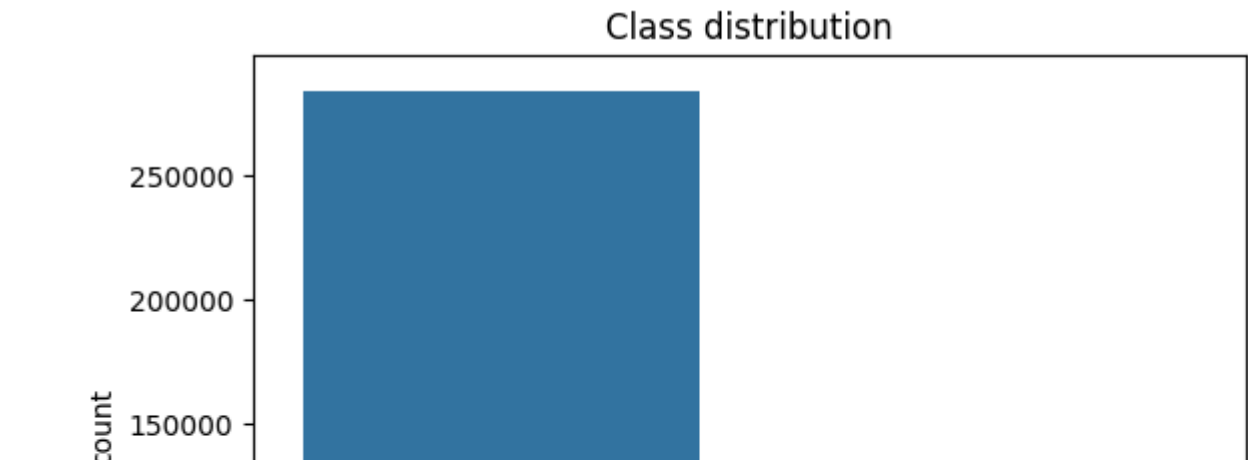


	count
Class	
0	284315
1	492

dtype: int64

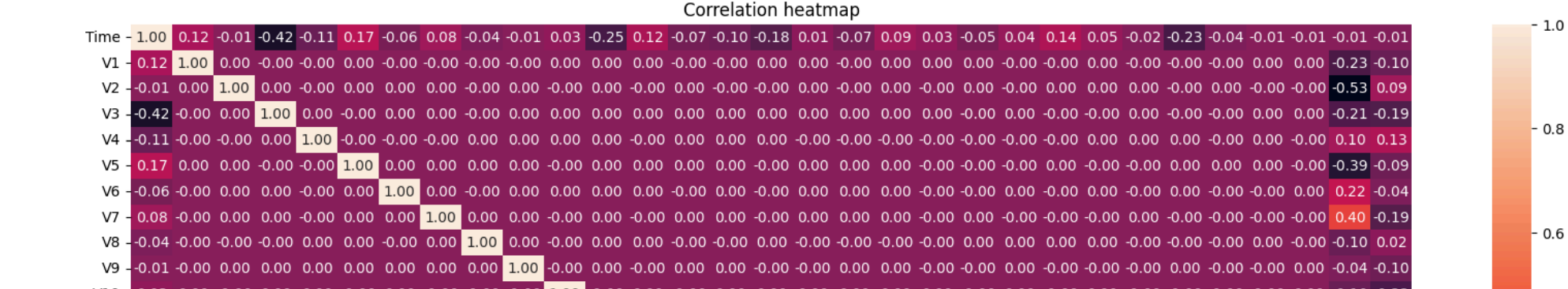
```
sns.countplot(x='Class',data=df)
plt.title("Class distribution")
```

```
Text(0.5, 1.0, 'Class distribution')
```



```
plt.figure(figsize=(20,10))
sns.heatmap(df.corr(),annot=True,fmt='0.2f')
plt.title("Correlation heatmap")
```

```
Text(0.5, 1.0, 'Correlation heatmap')
```



```
x=df.drop(columns=['Class'])
y=df['Class']
```



```
119879  0.161405 -0.006140  0.091444  0.109235 -0.020922  0.003967    1.98
259178  0.241830  0.682820 -1.635109 -0.770941  0.066006  0.137056    89.23
131932 -0.291198  0.402849  0.237383 -0.398467 -0.121139 -0.196195     3.94
146867  0.143326  0.678869  0.319710  0.426309  0.496912  0.335822     1.00
121958 -0.059544 -0.109073  0.290326 -0.393074  0.001217  0.038588   113.00
```

```
[199364 rows x 30 columns]
```

```
print(y_train)
```

```
⇒ 2557      0
   247823    0
   152342    0
   103385    0
    8771     0
      ..
   119879    0
   259178    0
   131932    0
   146867    0
   121958    0
Name: Class, Length: 199364, dtype: int64
```

```
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
scaler.fit(x_train)
x_train=scaler.transform(x_train)
x_test=scaler.transform(x_test)
print(x_train)
```

```
⇒ [[-1.95144063 -1.16681856 -0.28654908 ...  0.96676   -0.4486209
    -0.33974783]
   [ 1.24181171 -0.15922175 -2.43543667 ... -0.71311931  0.33519183
    4.27773998]
   [ 0.05339701 -0.92214029 -0.33881865 ... -0.15212107  0.26753092
    -0.0533992 ]
   ...
   [-0.31497174 -0.07400401  0.59673261 ... -0.29672056 -0.59571272
    -0.32837128]
   [-0.14359433 -1.50291696  1.41328113 ...  1.2243654   1.01984233
    -0.33974783]
   [-0.38688453  0.62962541 -0.46917078 ...  0.0044092   0.11724428
    0.09364469]]
```

```
print(x_test)
```

```
⇒ [[-1.12151562 -8.42703903  5.1486402   ... -4.96650013 -3.16657662
    1.06564352]
   [-1.06346299  0.17404952 -1.64365414 ...  0.10230985  0.3099202
    1.66902653]
   [-1.24834248  0.71449029 -0.35276585 ...  0.02949291  0.01413826
    -0.22366055]
   ...
   [-1.32360446  0.38159558 -0.61498614 ...  0.1450263   0.177531
    0.41327168]
   [ 0.97516144  1.06174851  0.1019923   ... -0.15089508 -0.09379868
    -0.34017349]
   [-0.70684801  0.55744777 -0.03940111 ...  0.19863631  0.10769086
    -0.30492165]]
```

```
dt=DecisionTreeClassifier()
dt.fit(x_train,y_train)
y_pred=dt.predict(x_test)
y_pred
```

```
⇒ array([1, 0, 0, ..., 0, 0, 0])
```

```
print(y_test)
```

```
⇒ 43428      1
   49906      0
   29474      0
   276481     0
   278846     0
      ..
   180795     0
   259979     0
```

```
21885      0
217812     0
86348      0
Name: Class, Length: 85443, dtype: int64
```

```
accuracy_score(y_test,y_pred)
```

```
0.9991222218320986
```

```
cm=confusion_matrix(y_test,y_pred)
cm
```

```
array([[85261,  46],
       [ 29,  107]])
```

```
print(classification_report(y_test,y_pred))
```

		precision	recall	f1-score	support
	0	1.00	1.00	1.00	85307
	1	0.70	0.79	0.74	136
accuracy				1.00	85443
macro avg		0.85	0.89	0.87	85443
weighted avg		1.00	1.00	1.00	85443

```
cmd=ConfusionMatrixDisplay(cm,display_labels=[0,1])
cmd.plot()
```

```
<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x7f71f3c7add0>
```

