

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")

df=pd.read_csv("/content/drive/MyDrive/Colab
Notebooks/creditcard.csv")
```

```
df.shape
```

```
(284807, 31)
```

```
df.head()
```

	Time	V1	V2	V3	...	V27	V28	Amount
Class								
0	0.0	-1.359807	-0.072781	2.536347	...	0.133558	-0.021053	149.62
0								
1	0.0	1.191857	0.266151	0.166480	...	-0.008983	0.014724	2.69
0								
2	1.0	-1.358354	-1.340163	1.773209	...	-0.055353	-0.059752	378.66
0								
3	1.0	-0.966272	-0.185226	1.792993	...	0.062723	0.061458	123.50
0								
4	2.0	-1.158233	0.877737	1.548718	...	0.219422	0.215153	69.99
0								

```
[5 rows x 31 columns]
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 284807 entries, 0 to 284806
Data columns (total 31 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0   Time    284807 non-null  float64
 1   V1       284807 non-null  float64
 2   V2       284807 non-null  float64
 3   V3       284807 non-null  float64
 4   V4       284807 non-null  float64
 5   V5       284807 non-null  float64
 6   V6       284807 non-null  float64
 7   V7       284807 non-null  float64
 8   V8       284807 non-null  float64
 9   V9       284807 non-null  float64
10  V10      284807 non-null  float64
11  V11      284807 non-null  float64
12  V12      284807 non-null  float64
13  V13      284807 non-null  float64
```

```

14 V14      284807 non-null float64
15 V15      284807 non-null float64
16 V16      284807 non-null float64
17 V17      284807 non-null float64
18 V18      284807 non-null float64
19 V19      284807 non-null float64
20 V20      284807 non-null float64
21 V21      284807 non-null float64
22 V22      284807 non-null float64
23 V23      284807 non-null float64
24 V24      284807 non-null float64
25 V25      284807 non-null float64
26 V26      284807 non-null float64
27 V27      284807 non-null float64
28 V28      284807 non-null float64
29 Amount    284807 non-null float64
30 Class     284807 non-null int64

```

```
dtypes: float64(30), int64(1)
```

```
memory usage: 67.4 MB
```

```
df.describe()
```

	Time	V1	...	Amount	Class
count	284807.000000	2.848070e+05	...	284807.000000	284807.000000
mean	94813.859575	3.919560e-15	...	88.349619	0.001727
std	47488.145955	1.958696e+00	...	250.120109	0.041527
min	0.000000	-5.640751e+01	...	0.000000	0.000000
25%	54201.500000	-9.203734e-01	...	5.600000	0.000000
50%	84692.000000	1.810880e-02	...	22.000000	0.000000
75%	139320.500000	1.315642e+00	...	77.165000	0.000000
max	172792.000000	2.454930e+00	...	25691.160000	1.000000

```
[8 rows x 31 columns]
```

```
df.isnull().sum()
```

```

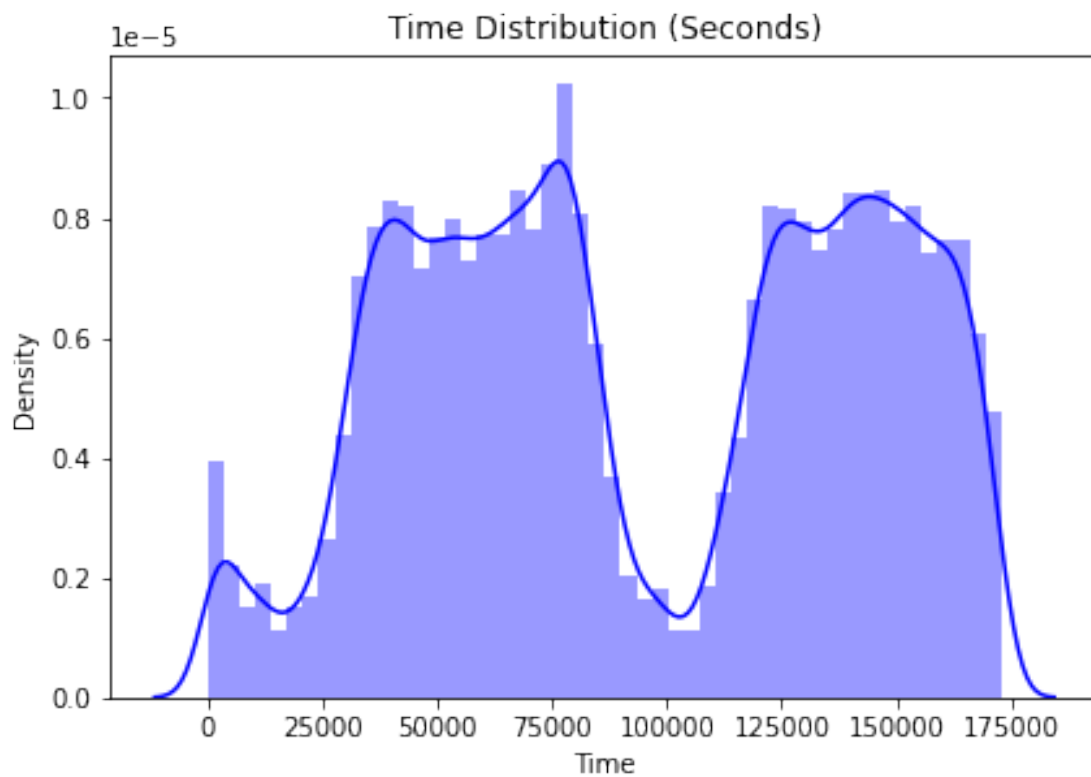
Time      0
V1         0
V2         0
V3         0
V4         0
V5         0
V6         0
V7         0
V8         0
V9         0
V10        0
V11        0
V12        0
V13        0
V14        0

```

```
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
dtype: int64
```

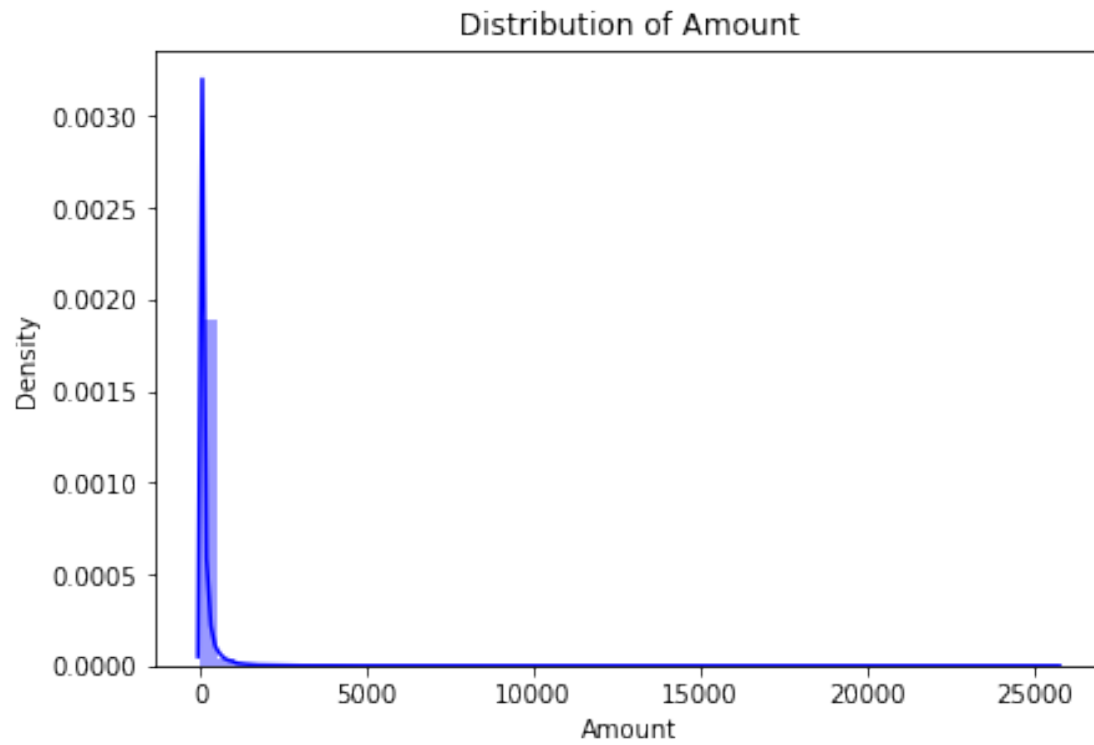
```
plt.figure(figsize=(15,10))
plt.subplot(2, 2, 1)
plt.title('Time Distribution (Seconds)')
sns.distplot(df['Time'], color='blue')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fce0fa0fcd0>
```

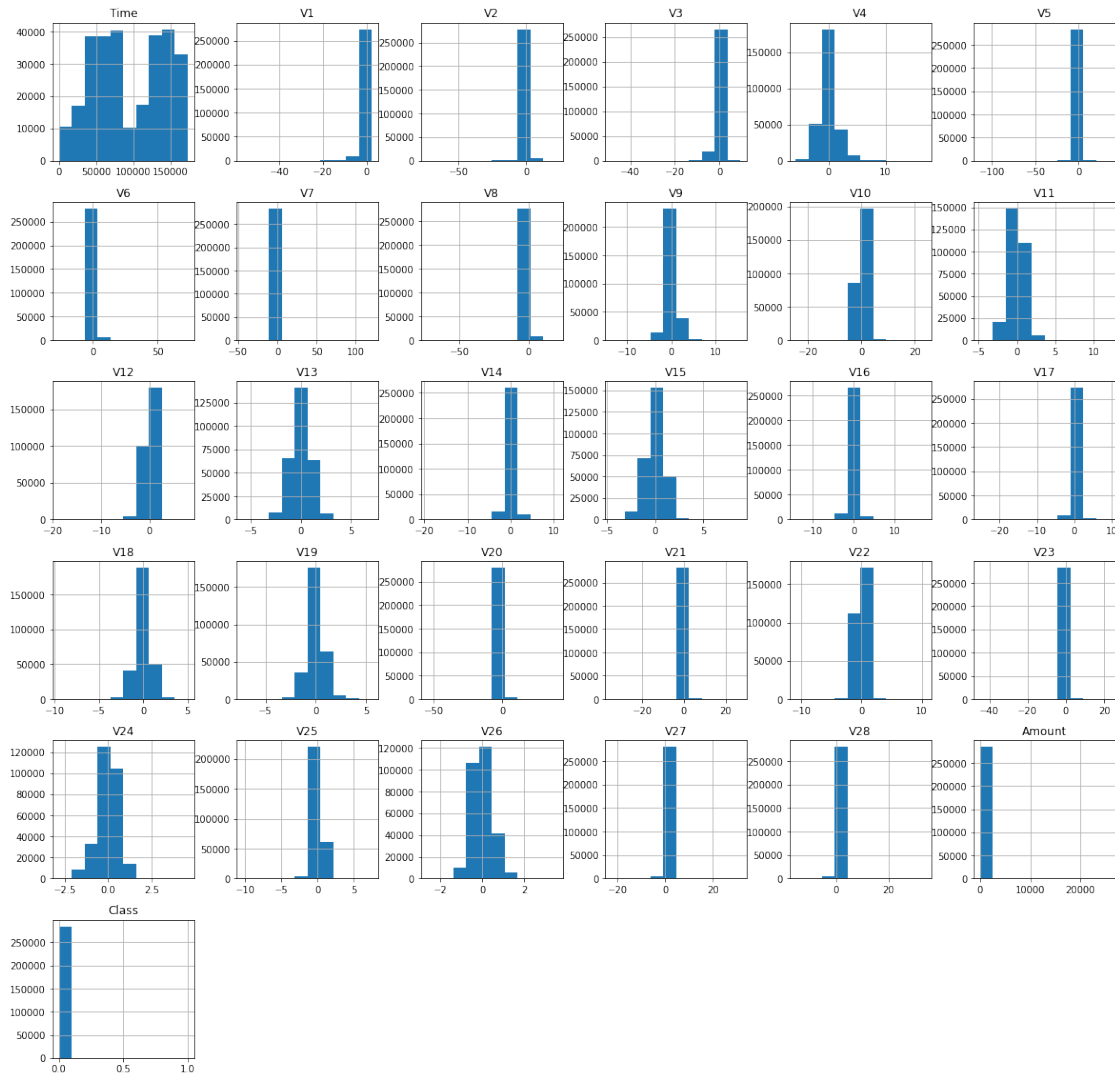


```
plt.figure(figsize=(15,10))
plt.subplot(2, 2, 2)
```

```
plt.title('Distribution of Amount')
sns.distplot(df['Amount'],color='blue')
<matplotlib.axes._subplots.AxesSubplot at 0x7fce0662e310>
```

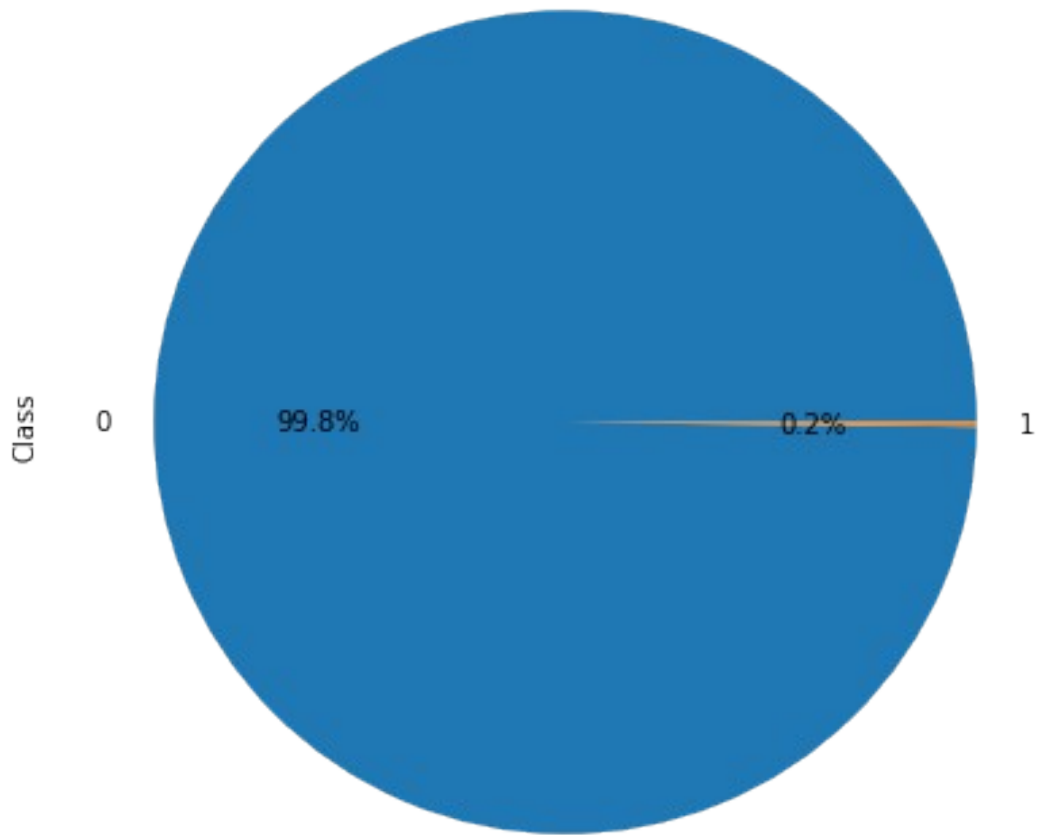


```
df.hist(figsize=(20, 20));
```



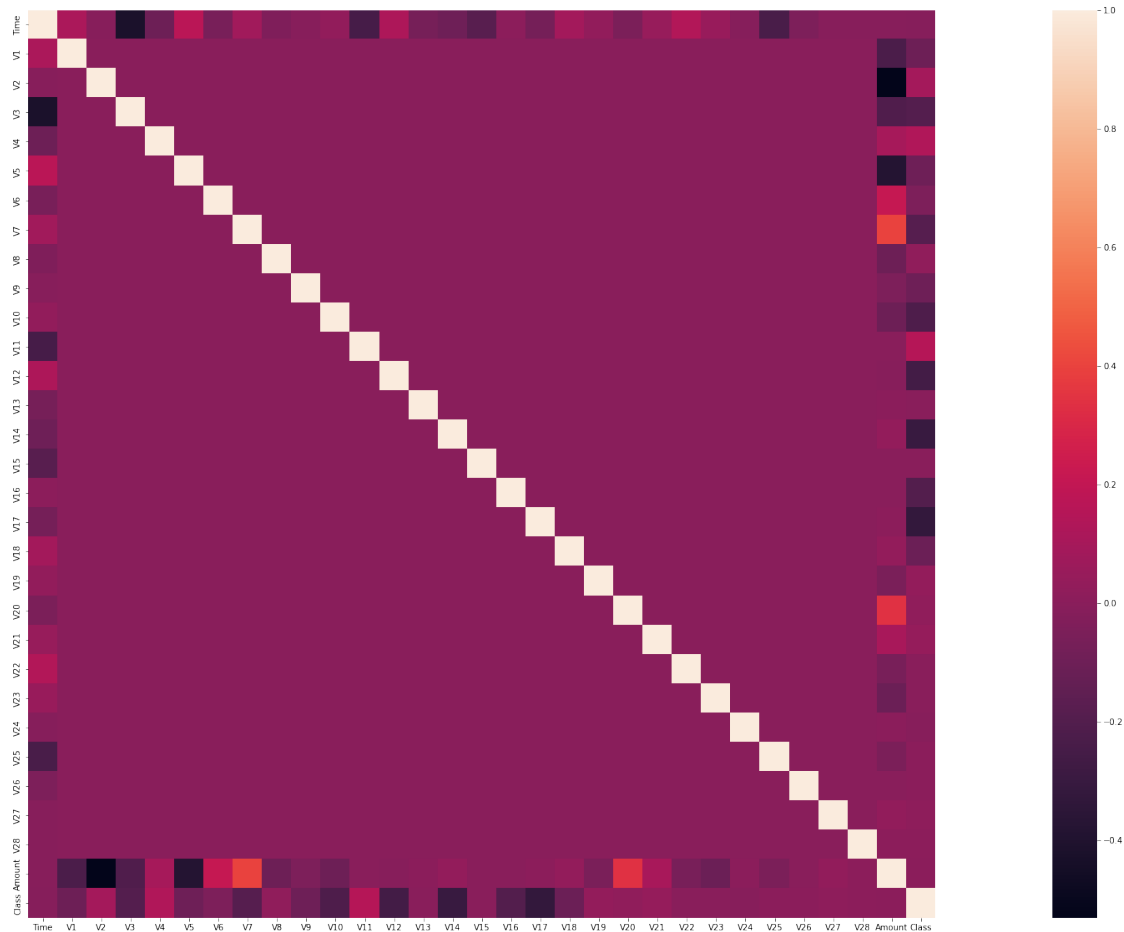
```
print(df["Class"].value_counts())
print("-----")
plt.figure(figsize=(7,7))
df["Class"].value_counts().plot.pie(autopct="%.1f%%")
plt.show()
```

```
0    284315
1      492
Name: Class, dtype: int64
-----
```



```
ax = plt.subplots(figsize=(50, 20))
sns.heatmap(df.corr(), square=True,)
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fce05095390>



```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

y = df["Class"]
X = df.drop("Class",axis=1)
X_train,X_test,y_train,y_test =
train_test_split(X,y,test_size=0.3,random_state=1,stratify=y)

ss = StandardScaler()
X_train_ss = ss.fit_transform(X_train)
X_test_ss = ss.transform(X_test)
```

BaseLine Model

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

model = Sequential()
model.add(Dense(16,activation="relu", input_dim=30))
model.add(Dense(16,activation="relu"))
model.add(Dense(16,activation="relu"))
model.add(Dense(1, activation="sigmoid"))
```

```

model.compile(optimizer="adam", loss="binary_crossentropy")
model.fit(X_train_ss,y_train,epochs=5, batch_size=8)

Epoch 1/5
24921/24921 [=====] - 37s 1ms/step - loss:
0.0071
Epoch 2/5
24921/24921 [=====] - 37s 1ms/step - loss:
0.0038
Epoch 3/5
24921/24921 [=====] - 37s 1ms/step - loss:
0.0032
Epoch 4/5
24921/24921 [=====] - 37s 1ms/step - loss:
0.0031
Epoch 5/5
24921/24921 [=====] - 37s 1ms/step - loss:
0.0030

```

```
<keras.callbacks.History at 0x7fcdb73c4bd0>
```

```
# testing
```

```
y_pred= model.predict(X_test_ss)
```

```
y_pred = y_pred.argmax(axis=1)
```

```
from sklearn.metrics import classification_report
```

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

	precision	recall	f1-score	support
0	1.00	1.00	1.00	85295
1	0.00	0.00	0.00	148
accuracy			1.00	85443
macro avg	0.50	0.50	0.50	85443
weighted avg	1.00	1.00	1.00	85443

Using RandomOverSampler to Balanced the training data

```
from imblearn.over_sampling import RandomOverSampler
```

```
ros = RandomOverSampler(random_state=1)
```

```
X_sample1, y_sample1 = ros.fit_sample(X_train_ss,y_train)
```

```
pd.Series(y_sample1).value_counts()
```



```

1      199020
0      199020
dtype: int64

model1 = Sequential()
model1.add(Dense(16,activation="relu", input_dim=30))
model1.add(Dense(16,activation="relu"))
model1.add(Dense(16,activation="relu"))
model1.add(Dense(1, activation="sigmoid"))

model.compile(optimizer="adam", loss="binary_crossentropy")

model.fit(X_sample1, y_sample1,epochs=10, batch_size=8)

Epoch 1/10
49755/49755 [=====] - 71s 1ms/step - loss:
0.0197
Epoch 2/10
49755/49755 [=====] - 71s 1ms/step - loss:
0.0057
Epoch 3/10
49755/49755 [=====] - 72s 1ms/step - loss:
0.0048
Epoch 4/10
49755/49755 [=====] - 71s 1ms/step - loss:
0.0039
Epoch 5/10
49755/49755 [=====] - 71s 1ms/step - loss:
0.0036
Epoch 6/10
49755/49755 [=====] - 73s 1ms/step - loss:
0.0036
Epoch 7/10
49755/49755 [=====] - 75s 2ms/step - loss:
0.0034
Epoch 8/10
49755/49755 [=====] - 73s 1ms/step - loss:
0.0031
Epoch 9/10
49755/49755 [=====] - 73s 1ms/step - loss:
0.0028
Epoch 10/10
49755/49755 [=====] - 73s 1ms/step - loss:
0.0027

<keras.callbacks.History at 0x7fcdb743ab90>

# testing
y_pred1= model.predict(X_test_ss)

y_pred = np.where(y_pred1>= 0.5,1,0)

```

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

	precision	recall	f1-score	support
0	1.00	1.00	1.00	85295
1	0.64	0.89	0.74	148
accuracy			1.00	85443
macro avg	0.82	0.94	0.87	85443
weighted avg	1.00	1.00	1.00	85443