

October 10, 2025

Credit Card Fraud Detection

Goal:

Predict the probability of an online credit card transaction being fraudulent, based on different properties of the transactions.

1 1. Setup Environment

```
[1]: # Data Manipulation
import numpy as np
import pandas as pd

# Data Visualization
import seaborn as sns
import matplotlib.pyplot as plt
import matplotlib.lines as mlines

# Time
import time
import datetime

# Machine Learning
from sklearn.preprocessing import LabelEncoder, minmax_scale
from sklearn.ensemble import RandomForestClassifier
from sklearn.decomposition import PCA
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.metrics import (confusion_matrix , classification_report,
    accuracy_score, roc_auc_score,
    plot_roc_curve, precision_recall_curve,
    plot_precision_recall_curve)
from sklearn.calibration import calibration_curve
from sklearn.calibration import CalibratedClassifierCV

from xgboost import XGBClassifier
from lightgbm import LGBMClassifier

from imblearn.over_sampling import RandomOverSampler
```

```

from   scipy.stats import chi2_contingency, f_oneway

import gc
import warnings
from   tqdm import tqdm

# Set Options
pd.set_option('display.max_rows', 100)
pd.set_option('display.max_columns', 500)
%matplotlib inline
warnings.filterwarnings("ignore")

```

2 2. Data Overview

[2]:

```

%%time
df_id    = pd.read_csv(r'C:\Users\admin\Desktop\PYTHON\_
                        ↪LEARNING\Data_Projects_01\MLp Projects\CREDIT CARD FRAUD\_
                        ↪DETECTION\Datasets\train_identity.csv')
df_tran = pd.read_csv(r'C:\Users\admin\Desktop\PYTHON\_
                        ↪LEARNING\Data_Projects_01\MLp Projects\CREDIT CARD FRAUD\_
                        ↪DETECTION\Datasets\train_transaction.csv')

```

Wall time: 26.5 s

[3]: df_id.head()

	TransactionID	id_01	id_02	id_03	id_04	id_05	id_06	id_07	id_08	\	
0	2987004	0.0	70787.0	NaN	NaN	NaN	NaN	NaN	NaN		
1	2987008	-5.0	98945.0	NaN	NaN	0.0	-5.0	NaN	NaN		
2	2987010	-5.0	191631.0	0.0	0.0	0.0	0.0	NaN	NaN		
3	2987011	-5.0	221832.0	NaN	NaN	0.0	-6.0	NaN	NaN		
4	2987016	0.0	7460.0	0.0	0.0	1.0	0.0	NaN	NaN		
	id_09	id_10	id_11	id_12	id_13	id_14	id_15	id_16	id_17	id_18	\
0	NaN	NaN	100.0	NotFound	NaN	-480.0	New	NotFound	166.0	NaN	
1	NaN	NaN	100.0	NotFound	49.0	-300.0	New	NotFound	166.0	NaN	
2	0.0	0.0	100.0	NotFound	52.0	NaN	Found	Found	121.0	NaN	
3	NaN	NaN	100.0	NotFound	52.0	NaN	New	NotFound	225.0	NaN	
4	0.0	0.0	100.0	NotFound	NaN	-300.0	Found	Found	166.0	15.0	
	id_19	id_20	id_21	id_22	id_23	id_24	id_25	id_26	id_27	id_28	\
0	542.0	144.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	New	
1	621.0	500.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	New	
2	410.0	142.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Found	
3	176.0	507.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	New	

```

4 529.0 575.0    NaN    NaN    NaN    NaN    NaN    NaN    NaN Found
          id_29      id_30      id_31    id_32      id_33 \
0  NotFound   Android 7.0  samsung browser 6.2  32.0  2220x1080
1  NotFound     iOS 11.1.2  mobile safari 11.0  32.0  1334x750
2   Found        NaN      chrome 62.0    NaN      NaN
3  NotFound        NaN      chrome 62.0    NaN      NaN
4   Found  Mac OS X 10_11_6      chrome 62.0  24.0  1280x800

          id_34    id_35    id_36    id_37    id_38 DeviceType \
0  match_status:2      T      F      T      T    mobile
1  match_status:1      T      F      F      T    mobile
2           NaN      F      F      T      T  desktop
3           NaN      F      F      T      T  desktop
4  match_status:2      T      F      T      T  desktop

          DeviceInfo
0  SAMSUNG SM-G892A Build/NRD90M
1           iOS Device
2           Windows
3           NaN
4           MacOS

```

2.0.1 Identity Data Description

Variables in this table are identity information – network connection information (IP, ISP, Proxy, etc) and digital signature (UA/browser/os/version, etc) associated with transactions. They're collected by Vesta's fraud protection system and digital security partners. (The field names are masked and pairwise dictionary will not be provided for privacy protection and contract agreement)

Categorical Features: - DeviceType - DeviceInfo - id_12 - id_38

[4] : df_tran.head()

```

[4]: TransactionID  isFraud  TransactionDT  TransactionAmt  ProductCD  card1 \
0      2987000      0          86400          68.5          W  13926
1      2987001      0          86401          29.0          W  2755
2      2987002      0          86469          59.0          W  4663
3      2987003      0          86499          50.0          W  18132
4      2987004      0          86506          50.0          H  4497

  card2  card3      card4  card5  card6  addr1  addr2  dist1  dist2 \
0  NaN  150.0  discover  142.0  credit  315.0  87.0  19.0  NaN
1  404.0  150.0  mastercard  102.0  credit  325.0  87.0  NaN  NaN
2  490.0  150.0       visa  166.0  debit  330.0  87.0  287.0  NaN
3  567.0  150.0  mastercard  117.0  debit  476.0  87.0  NaN  NaN
4  514.0  150.0  mastercard  102.0  credit  420.0  87.0  NaN  NaN

```

	P_emaildomain	R_emaildomain	C1	C2	C3	C4	C5	C6	C7	C8	C9	\				
0		NaN	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0				
1		gmail.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0				
2		outlook.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0				
3		yahoo.com	NaN	2.0	5.0	0.0	0.0	0.0	4.0	0.0	0.0	1.0				
4		gmail.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	1.0	0.0				
	C10	C11	C12	C13	C14	D1	D2	D3	D4	D5	D6	D7	D8	D9	\	
0	0.0	2.0	0.0	1.0	1.0	14.0	NaN	13.0	NaN	NaN	NaN	NaN	NaN	NaN		
1	0.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	NaN	NaN		
2	0.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	NaN	NaN		
3	0.0	1.0	0.0	25.0	1.0	112.0	112.0	0.0	94.0	0.0	NaN	NaN	NaN	NaN		
4	1.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
	D10	D11	D12	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	\	
0	13.0	13.0	NaN	NaN	NaN	0.0	T	T	T	M2	F	T	NaN	NaN		
1	0.0	NaN	NaN	NaN	NaN	0.0	NaN	NaN	MO	T	T	NaN	NaN			
2	0.0	315.0	NaN	NaN	NaN	315.0	T	T	T	MO	F	F	F	F		
3	84.0	NaN	NaN	NaN	NaN	111.0	NaN	NaN	NaN	MO	T	F	NaN	NaN		
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
	M9	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	\
0	NaN	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.0	0.0	1.0	
2	F	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V15	V16	V17	V18	V19	V20	V21	V22	V23	V24	V25	V26	V27	V28	V29	\
0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V30	V31	V32	V33	V34	V35	V36	V37	V38	V39	V40	V41	V42	V43	V44	\
0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
2	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
3	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V45	V46	V47	V48	V49	V50	V51	V52	V53	V54	V55	V56	V57	V58	V59	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
1	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	
2	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
3	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	

4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
	V60	V61	V62	V63	V64	V65	V66	V67	V68	V69	V70	V71	V72	V73	V74	\	
0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V75	V76	V77	V78	V79	V80	V81	V82	V83	V84	V85	V86	V87	V88	V89	\	
0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	
1	0.0	0.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	
2	1.0	1.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	
3	1.0	1.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V90	V91	V92	V93	V94	V95	V96	V97	V98	V99	V100	V101	V102	\			
0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0		
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
3	0.0	0.0	0.0	0.0	0.0	1.0	48.0	28.0	0.0	10.0	4.0	1.0	38.0				
4	NaN	NaN	NaN	NaN	NaN	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V103	V104	V105	V106	V107	V108	V109	V110	V111	V112	V113	V114	\				
0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
1	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
2	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
3	24.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
4	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
	V115	V116	V117	V118	V119	V120	V121	V122	V123	V124	V125	V126	\				
0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0		
1	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0		
2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0		
3	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	50.0		
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0		
	V127	V128	V129	V130	V131	V132		V133	V134	V135	V136	V137	\				
0	117.0	0.0	0.0	0.0	0.0	0.0		117.0	0.0	0.0	0.0	0.0					
1	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0					
2	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0					
3	1758.0	925.0	0.0	354.0	135.0	50.0		1404.0	790.0	0.0	0.0	0.0					
4	0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0					
	V138	V139	V140	V141	V142	V143	V144	V145	V146	V147	V148	V149	\				
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN					
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN					

2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	0.0	0.0	0.0	0.0	0.0	6.0	18.0	140.0	0.0	0.0	0.0	0.0	0.0
	V150	V151	V152	V153	V154	V155	V156	V157	V158		V159		\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN		
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN		
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN		
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN		
4	1803.0	49.0	64.0	0.0	0.0	0.0	0.0	0.0	0.0	15557.990234			
	V160	V161	V162	V163	V164	V165	V166	V167	V168	V169		\	
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	169690.796875	0.0	0.0	0.0	515.0	5155.0	2840.0	0.0	0.0	0.0	0.0	0.0	
	V170	V171	V172	V173	V174	V175	V176	V177	V178	V179	V180	V181	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	1.0	1.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	
	V182	V183	V184	V185	V186	V187	V188	V189	V190	V191	V192	V193	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
	V194	V195	V196	V197	V198	V199	V200	V201	V202	V203	V204	V205	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
	V206	V207	V208	V209	V210	V211	V212	V213	V214	V215	V216	V217	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V218	V219	V220	V221	V222	V223	V224	V225	V226	V227	V228	V229	\

0	NaN											
1	NaN											
2	NaN											
3	NaN											
4	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
	V230	V231	V232	V233	V234	V235	V236	V237	V238	V239	V240	V241
0	NaN											
1	NaN											
2	NaN											
3	NaN											
4	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
	V242	V243	V244	V245	V246	V247	V248	V249	V250	V251	V252	V253
0	NaN											
1	NaN											
2	NaN											
3	NaN											
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
	V254	V255	V256	V257	V258	V259	V260	V261	V262	V263	V264	V265
0	NaN											
1	NaN											
2	NaN											
3	NaN											
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0
	V266	V267	V268	V269	V270	V271	V272	V273	V274	V275	V276	V277
0	NaN											
1	NaN											
2	NaN											
3	NaN											
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	V278	V279	V280	V281	V282	V283	V284	V285	V286	V287	V288	V289
0	NaN	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
1	NaN	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
2	NaN	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
3	NaN	1.0	28.0	0.0	0.0	0.0	0.0	10.0	0.0	4.0	0.0	0.0
4	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
	V290	V291	V292	V293	V294	V295	V296	V297	V298	V299	V300	V301
0	1.0	1.0	1.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
1	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	1.0	1.0	1.0	1.0	38.0	24.0	0.0	0.0	0.0	0.0	0.0	0.0
4	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

	V302	V303	V304	V305	V306	V307	V308	V309	V310	V311	V312	\	
0	0.0	0.0	0.0	1.0	0.0	117.0	0.0	0.0	0.0	0.0	0.0		
1	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
2	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
3	0.0	0.0	0.0	1.0	50.0	1758.0	925.0	0.0	354.0	0.0	135.0		
4	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
	V313	V314	V315	V316	V317	V318	V319	V320	V321	V322	V323	V324	\
0	0.0	0.0	0.0	0.0	117.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN	
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN	
3	0.0	0.0	0.0	50.0	1404.0	790.0	0.0	0.0	0.0	NaN	NaN	NaN	
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V325	V326	V327	V328	V329	V330	V331	V332	V333	V334	V335	V336	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V337	V338	V339										
0	NaN	NaN	NaN										
1	NaN	NaN	NaN										
2	NaN	NaN	NaN										
3	NaN	NaN	NaN										
4	0.0	0.0	0.0										

2.0.2 Transaction Data Description

- **TransactionDT**: timedelta from a given reference datetime (not an actual timestamp)
- **TransactionAMT**: transaction payment amount in USD
- **ProductCD**: product code, the product for each transaction
- **card1 - card6**: payment card information, such as card type, card category, issue bank, country, etc.
- **addr**: address
- **dist**: distance
- ****P_ and (R____) emaildomain****: purchaser and recipient email domain
- **C1-C14**: counting, such as how many addresses are found to be associated with the payment card, etc. The actual meaning is masked.
- **D1-D15**: timedelta, such as days between previous transaction, etc.
- **M1-M9**: match, such as names on card and address, etc.
- **Vxxx**: Vesta engineered rich features, including ranking, counting, and other entity relations.

3 3. Optimize Memory Used by Data

Memory occupied by the dataframe (in mb)

```
[5]: df_id.memory_usage(deep=True).sum() / 1024**2
```

```
[5]: 157.63398933410645
```

```
[6]: df_tran.memory_usage(deep=True).sum() / 1024**2
```

```
[6]: 2100.701406478882
```

```
[7]: print('int64 min: ', np.iinfo(np.int64).min)
print('int64 max: ', np.iinfo(np.int64).max)
```

```
int64 min: -9223372036854775808
```

```
int64 max: 9223372036854775807
```

```
[8]: print('int8 min: ', np.iinfo(np.int8).min)
print('int8 max: ', np.iinfo(np.int8).max)
```

```
int8 min: -128
```

```
int8 max: 127
```

```
[9]: # Reduce memory usage
def reduce_mem_usage(df, verbose=True):
    numerics = ['int16', 'int32', 'int64', 'float16', 'float32', 'float64']
    start_mem = df.memory_usage(deep=True).sum() / 1024**2
    for col in df.columns:
        col_type = df[col].dtypes
        if col_type in numerics:
            c_min = df[col].min()
            c_max = df[col].max()
            if str(col_type)[-3] == 'int':
                if c_min > np.iinfo(np.int8).min and c_max < np.iinfo(np.int8).
                    ↵max:
                    df[col] = df[col].astype(np.int8)
                elif c_min > np.iinfo(np.int16).min and c_max < np.iinfo(np.
                    ↵int16).max:
                    df[col] = df[col].astype(np.int16)
                elif c_min > np.iinfo(np.int32).min and c_max < np.iinfo(np.
                    ↵int32).max:
                    df[col] = df[col].astype(np.int32)
                elif c_min > np.iinfo(np.int64).min and c_max < np.iinfo(np.
                    ↵int64).max:
                    df[col] = df[col].astype(np.int64)
                else:
                    if c_min > np.finfo(np.float16).min and c_max < np.finfo(np.
                        ↵float16).max:
                        df[col] = df[col].astype(np.float16)
```

```

        elif c_min > np.finfo(np.float32).min and c_max < np.finfo(np.
˓→float32).max:
            df[col] = df[col].astype(np.float32)
        else:
            df[col] = df[col].astype(np.float64)
    end_mem = df.memory_usage(deep=True).sum() / 1024**2
    if verbose: print('Mem. usage decreased to {:.2f} Mb {:.1f}% reduction'.
˓→format(end_mem, 100 * (start_mem - end_mem) / start_mem))
    return df

```

[10]: # Reduce the memory size of the dataframe

```

df_id = reduce_mem_usage(df_id)
df_tran = reduce_mem_usage(df_tran)

```

Mem. usage decreased to 138.38 Mb (12.2% reduction)

Mem. usage decreased to 867.89 Mb (58.7% reduction)

4 4. Basic Data Stats

Shape of dataframe

[11]: df_id.shape

[11]: (144233, 41)

[12]: df_tran.shape

[12]: (590540, 394)

Check how many transactions has ID info

[13]: df_tran.TransactionID.isin(df_id.TransactionID).sum()

[13]: 144233

Summary of dataframe

[14]: from pandas_summary import DataFrameSummary
df_id_summary = DataFrameSummary(df_id)
df_id_summary.summary()

	TransactionID	id_01	id_02	id_03	id_04	\
count	144233.0	144233.0	140872.0	66324.0	66324.0	
mean	3236329.311288	NaN	174716.59375	0.0	-0.0	
std	178849.571186	0.0	159651.8125	0.0	0.0	
min	2987004.0	-100.0	1.0	-13.0	-28.0	
25%	3077142.0	-10.0	67992.0	0.0	0.0	
50%	3198818.0	-5.0	125800.5	0.0	0.0	

75%	3392923.0	-5.0	228749.0	0.0	0.0		\
max	3577534.0	0.0	999595.0	10.0	0.0		
counts	144233	144233	140872	66324	66324		
uniques	144233	77	115655	24	15		
missing	0	0	3361	77909	77909		
missing_perc	0%	0%	2.33%	54.02%	54.02%		
types	numeric	numeric	numeric	numeric	numeric		
count	136865.0	136865.0	5155.0	5155.0	74926.0	74926.0	
mean	NaN	NaN	inf	-inf	0.0	-0.0	
std	0.0	0.0	11.382812	26.078125	0.0	0.0	
min	-72.0	-100.0	-46.0	-100.0	-36.0	-100.0	
25%	0.0	-6.0	5.0	-48.0	0.0	0.0	
50%	0.0	0.0	14.0	-34.0	0.0	0.0	
75%	1.0	0.0	22.0	-23.0	0.0	0.0	
max	52.0	0.0	61.0	0.0	25.0	0.0	
counts	136865	136865	5155	5155	74926	74926	
uniques	93	101	84	94	46	62	
missing	7368	7368	139078	139078	69307	69307	
missing_perc	5.11%	5.11%	96.43%	96.43%	48.05%	48.05%	
types	numeric	numeric	numeric	numeric	numeric	numeric	
count	140978.0	NaN	127320.0	80044.0	NaN	NaN	
mean	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	NaN	0.0	NaN	NaN	NaN	
min	90.0	NaN	10.0	-660.0	NaN	NaN	
25%	100.0	NaN	49.0	-360.0	NaN	NaN	
50%	100.0	NaN	52.0	-300.0	NaN	NaN	
75%	100.0	NaN	52.0	-300.0	NaN	NaN	
max	100.0	NaN	64.0	720.0	NaN	NaN	
counts	140978	144233	127320	80044	140985	129340	
uniques	146	2	54	25	3	2	
missing	3255	0	16913	64189	3248	14893	
missing_perc	2.26%	0%	11.73%	44.50%	2.25%	10.33%	
types	numeric	bool	numeric	numeric	categorical	bool	
count	139369.0	45113.0	139318.0	139261.0	5159.0	5169.0	
mean	NaN	inf	NaN	NaN	inf	inf	
std	0.0	1.561523	NaN	NaN	inf	6.898438	
min	100.0	10.0	100.0	100.0	100.0	10.0	
25%	166.0	13.0	266.0	256.0	252.0	14.0	
50%	166.0	15.0	341.0	472.0	252.0	14.0	
75%	225.0	15.0	427.0	533.0	486.5	14.0	
max	229.0	29.0	671.0	661.0	854.0	44.0	

counts	139369	45113	139318	139261	5159	5169
uniques	104	18	522	394	490	25
missing	4864	99120	4915	4972	139074	139064
missing_perc	3.37%	68.72%	3.41%	3.45%	96.42%	96.42%
types	numeric	numeric	numeric	numeric	numeric	numeric

count		id_23	id_24	id_25	id_26	id_27	id_28	\
mean		NaN	4747.0	5132.0	5163.0	NaN	NaN	
std		NaN	2.371094	97.4375	32.09375	NaN	NaN	
min		NaN	11.0	100.0	100.0	NaN	NaN	
25%		NaN	11.0	321.0	119.0	NaN	NaN	
50%		NaN	11.0	321.0	149.0	NaN	NaN	
75%		NaN	15.0	371.0	169.0	NaN	NaN	
max		NaN	26.0	548.0	216.0	NaN	NaN	
counts	5169	4747	5132	5163	5169	140978		
uniques	3	12	341	95	2	2		
missing	139064	139486	139101	139070	139064	3255		
missing_perc	96.42%	96.71%	96.44%	96.42%	96.42%	2.26%		
types	categorical	numeric	numeric	numeric	bool	bool		

count		id_29	id_30	id_31	id_32	id_33	\
mean		NaN	NaN	NaN	77586.0	NaN	
std		NaN	NaN	NaN	0.0	NaN	
min		NaN	NaN	NaN	0.0	NaN	
25%		NaN	NaN	NaN	24.0	NaN	
50%		NaN	NaN	NaN	24.0	NaN	
75%		NaN	NaN	NaN	32.0	NaN	
max		NaN	NaN	NaN	32.0	NaN	
counts	140978	77565	140282	77586	73289		
uniques	2	75	130	4	260		
missing	3255	66668	3951	66647	70944		
missing_perc	2.26%	46.22%	2.74%	46.21%	49.19%		
types	bool	categorical	categorical	numeric	categorical		

count		id_34	id_35	id_36	id_37	id_38	DeviceType	\
mean		NaN	NaN	NaN	NaN	NaN	NaN	
std		NaN	NaN	NaN	NaN	NaN	NaN	
min		NaN	NaN	NaN	NaN	NaN	NaN	
25%		NaN	NaN	NaN	NaN	NaN	NaN	
50%		NaN	NaN	NaN	NaN	NaN	NaN	
75%		NaN	NaN	NaN	NaN	NaN	NaN	
max		NaN	NaN	NaN	NaN	NaN	NaN	
counts	77805	140985	140985	140985	140985	140810		
uniques	4	2	2	2	2	2		

missing	66428	3248	3248	3248	3248	3423
missing_perc	46.06%	2.25%	2.25%	2.25%	2.25%	2.37%
types	categorical	bool	bool	bool	bool	bool
DeviceInfo						
count		NaN				
mean		NaN				
std		NaN				
min		NaN				
25%		NaN				
50%		NaN				
75%		NaN				
max		NaN				
counts		118666				
uniques		1786				
missing		25567				
missing_perc		17.73%				
types	categorical					

[15]: df_tran.describe()

	TransactionID	isFraud	TransactionDT	TransactionAmt	\			
count	5.905400e+05	590540.000000	5.905400e+05	590540.000000				
mean	3.282270e+06	0.034990	7.372311e+06		NaN			
std	1.704744e+05	0.183755	4.617224e+06		NaN			
min	2.987000e+06	0.000000	8.640000e+04		0.250977			
25%	3.134635e+06	0.000000	3.027058e+06		43.312500			
50%	3.282270e+06	0.000000	7.306528e+06		68.750000			
75%	3.429904e+06	0.000000	1.124662e+07		125.000000			
max	3.577539e+06	1.000000	1.581113e+07		31936.000000			
	card1	card2	card3	card5	addr1	addr2	\	
count	590540.000000	581607.0	588975.0	586281.0	524834.0	524834.0		
mean	9898.734658	NaN	NaN	NaN	NaN	NaN		
std	4901.170153	NaN	0.0	0.0	NaN	0.0		
min	1000.000000	100.0	100.0	100.0	100.0	10.0		
25%	6019.000000	214.0	150.0	166.0	204.0	87.0		
50%	9678.000000	361.0	150.0	226.0	299.0	87.0		
75%	14184.000000	512.0	150.0	226.0	330.0	87.0		
max	18396.000000	600.0	231.0	237.0	540.0	102.0		
	dist1	dist2	C1	C2	C3	C4	C5	\
count	238269.0	37627.0	590540.0	590540.0	590540.0	590540.0	590540.0	
mean	NaN	inf	NaN	NaN	0.0	NaN	NaN	
std	NaN	inf	NaN	NaN	0.0	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	3.0	7.0	1.0	1.0	0.0	0.0	0.0	

50%	8.0	37.0	1.0	1.0	0.0	0.0	0.0	\
75%	24.0	206.0	3.0	3.0	0.0	0.0	1.0	
max	10288.0	11624.0	4684.0	5692.0	26.0	2252.0	349.0	
	C6	C7	C8	C9	C10	C11	C12	\
count	590540.0	590540.0	590540.0	590540.0	590540.0	590540.0	590540.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	0.0	NaN	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	0.0	0.0	0.0	0.0	1.0	0.0	
50%	1.0	0.0	0.0	1.0	0.0	1.0	0.0	
75%	2.0	0.0	0.0	2.0	0.0	2.0	0.0	
max	2252.0	2256.0	3332.0	210.0	3256.0	3188.0	3188.0	
	C13	C14	D1	D2	D3	D4	D5	\
count	590540.0	590540.0	589271.0	309743.0	327662.0	421618.0	280699.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	-122.0	0.0	
25%	1.0	1.0	0.0	26.0	1.0	0.0	1.0	
50%	3.0	1.0	3.0	97.0	8.0	26.0	10.0	
75%	12.0	2.0	122.0	276.0	27.0	253.0	32.0	
max	2918.0	1429.0	640.0	640.0	819.0	869.0	819.0	
	D6	D7	D8	D9	D10	D11	D12	\
count	73187.0	38917.0	74926.000000	74926.000000	514518.0	311253.0	311253.0	
mean	NaN	inf	NaN	0.000000	NaN	NaN	NaN	
std	NaN	inf	NaN	0.000000	NaN	NaN	NaN	
min	-83.0	0.0	0.000000	0.000000	0.0	-53.0		
25%	0.0	0.0	0.958496	0.208374	0.0	0.0		
50%	0.0	0.0	37.875000	0.666504	15.0	43.0		
75%	40.0	17.0	188.000000	0.833496	197.0	274.0		
max	873.0	843.0	1708.000000	0.958496	876.0	670.0		
	D12	D13	D14	D15	V1	V2	V3	\
count	64717.0	61952.0	62187.0	501427.0	311253.0	311253.0	311253.0	
mean	inf	inf	inf	NaN	NaN	NaN	NaN	
std	inf	inf	inf	NaN	0.0	0.0	0.0	
min	-83.0	0.0	-193.0	-83.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
50%	0.0	0.0	0.0	52.0	1.0	1.0	1.0	
75%	13.0	0.0	2.0	314.0	1.0	1.0	1.0	
max	648.0	847.0	878.0	879.0	1.0	8.0	9.0	
	V4	V5	V6	V7	V8	V9	V10	\
count	311253.0	311253.0	311253.0	311253.0	311253.0	311253.0	311253.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	0.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	0.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	6.0	6.0	9.0	9.0	8.0	8.0	4.0	
	V11	V12	V13	V14	V15	V16	V17	\
count	311253.0	514467.0	514467.0	514467.0	514467.0	514467.0	514467.0	
mean	NaN	NaN	NaN	NaN	0.0	0.0	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	1.0	0.0	0.0	0.0	
50%	0.0	1.0	1.0	1.0	0.0	0.0	0.0	
75%	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
max	5.0	3.0	6.0	1.0	7.0	15.0	15.0	
	V18	V19	V20	V21	V22	V23	V24	\
count	514467.0	514467.0	514467.0	514467.0	514467.0	514467.0	514467.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
50%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
75%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
max	15.0	7.0	15.0	5.0	8.0	13.0	13.0	
	V25	V26	V27	V28	V29	V30	V31	\
count	514467.0	514467.0	514467.0	514467.0	514467.0	514467.0	514467.0	
mean	NaN	NaN	0.0	0.0	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
50%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
75%	1.0	1.0	0.0	0.0	1.0	1.0	0.0	
max	7.0	13.0	4.0	4.0	5.0	9.0	7.0	
	V32	V33	V34	V35	V36	V37	V38	\
count	514467.0	514467.0	514467.0	421571.0	421571.0	421571.0	421571.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
50%	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
75%	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
max	15.0	7.0	13.0	3.0	5.0	54.0	54.0	

	V39	V40	V41	V42	V43	V44	V45	\
count	421571.0	421571.0	421571.0	421571.0	421571.0	421571.0	421571.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	1.0	0.0	0.0	1.0	1.0	
50%	0.0	0.0	1.0	0.0	0.0	1.0	1.0	
75%	0.0	0.0	1.0	0.0	0.0	1.0	1.0	
max	15.0	24.0	1.0	8.0	8.0	48.0	48.0	
	V46	V47	V48	V49	V50	V51	V52	\
count	421571.0	421571.0	421571.0	421571.0	421571.0	421571.0	421571.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
50%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
75%	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
max	6.0	12.0	5.0	5.0	5.0	6.0	12.0	
	V53	V54	V55	V56	V57	V58	V59	\
count	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	1.0	1.0	0.0	0.0	0.0	
50%	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
75%	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
max	5.0	6.0	17.0	51.0	6.0	10.0	16.0	
	V60	V61	V62	V63	V64	V65	V66	\
count	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
50%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
75%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
max	16.0	6.0	10.0	7.0	7.0	1.0	7.0	
	V67	V68	V69	V70	V71	V72	V73	\
count	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	513444.0	
mean	NaN	0.0	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
50%	1.0	0.0	0.0	0.0	0.0	0.0	0.0	

75%	1.0	0.0	1.0	1.0	0.0	0.0	0.0	
max	8.0	2.0	5.0	6.0	6.0	10.0	7.0	
	V74	V75	V76	V77	V78	V79	V80	\
count	513444.0	501376.0	501376.0	501376.0	501376.0	501376.0	501376.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	1.0	1.0	0.0	0.0	
50%	0.0	1.0	1.0	1.0	1.0	0.0	0.0	
75%	0.0	1.0	1.0	1.0	1.0	0.0	0.0	
max	8.0	4.0	6.0	30.0	31.0	7.0	19.0	
	V81	V82	V83	V84	V85	V86	V87	\
count	501376.0	501376.0	501376.0	501376.0	501376.0	501376.0	501376.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
50%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
75%	0.0	1.0	1.0	0.0	0.0	1.0	1.0	
max	19.0	7.0	7.0	7.0	7.0	30.0	30.0	
	V88	V89	V90	V91	V92	V93	V94	\
count	501376.0	501376.0	501376.0	501376.0	501376.0	501376.0	501376.0	
mean	NaN	0.0	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
50%	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
75%	1.0	0.0	1.0	1.0	0.0	0.0	0.0	
max	1.0	2.0	5.0	6.0	7.0	7.0	2.0	
	V95	V96	V97	V98	V99	V100	V101	\
count	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	
mean	NaN	NaN	NaN	0.0	NaN	NaN	NaN	
std	NaN	NaN	NaN	0.0	0.0	0.0	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.0	1.0	0.0	0.0	1.0	0.0	0.0	
max	880.0	1410.0	976.0	12.0	88.0	28.0	869.0	
	V102	V103	V104	V105	V106	V107	V108	\
count	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	
mean	NaN	NaN	0.0	NaN	NaN	NaN	NaN	
std	NaN	NaN	0.0	0.0	0.0	0.0	0.0	

min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
25%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
50%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
75%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
max	1285.0	928.0	15.0	99.0	55.0	1.0	7.0	
	V109	V110	V111	V112	V113	V114	V115	\
count	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	7.0	7.0	9.0	9.0	9.0	6.0	6.0	
	V116	V117	V118	V119	V120	V121	V122	\
count	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	590226.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	6.0	3.0	3.0	3.0	3.0	3.0	3.0	
	V123	V124	V125	V126	V127			\
count	590226.0	590226.0	590226.0	590226.000000	590226.000000			
mean	NaN	NaN	NaN	129.979431	336.611603			
std	0.0	0.0	0.0	2346.951660	4238.666992			
min	0.0	0.0	0.0	0.000000	0.000000			
25%	1.0	1.0	1.0	0.000000	0.000000			
50%	1.0	1.0	1.0	0.000000	0.000000			
75%	1.0	1.0	1.0	0.000000	107.949997			
max	13.0	13.0	13.0	160000.000000	160000.000000			
	V128	V129	V130	V131	V132			\
count	590226.000000	590226.0	590226.0	590226.0	590226.000000			
mean	204.094025	NaN	NaN	NaN	103.513191			
std	3010.258789	NaN	NaN	NaN	2266.106201			
min	0.000000	0.0	0.0	0.0	0.000000			
25%	0.000000	0.0	0.0	0.0	0.000000			
50%	0.000000	0.0	0.0	0.0	0.000000			
75%	0.000000	0.0	59.0	0.0	0.000000			
max	160000.000000	55136.0	55136.0	55136.0	93736.000000			
	V133	V134	V135	V136				\

count	590226.00000	590226.000000	590226.000000	590226.000000				
mean	204.88916	145.972336	17.250130	38.821198				
std	3796.31665	2772.986816	293.847565	451.808411				
min	0.00000	0.000000	0.000000	0.000000				
25%	0.00000	0.000000	0.000000	0.000000				
50%	0.00000	0.000000	0.000000	0.000000				
75%	0.00000	0.000000	0.000000	0.000000				
max	133915.00000	98476.000000	90750.000000	90750.000000				
	V137	V138	V139	V140	V141	V142	V143	\
count	590226.000000	81945.0	81945.0	81945.0	81945.0	81945.0	81951.0	
mean	26.365089	0.0	NaN	NaN	0.0	0.0	NaN	
std	348.332703	0.0	0.0	0.0	0.0	0.0	NaN	
min	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.000000	0.0	1.0	1.0	0.0	0.0	0.0	
75%	0.000000	0.0	1.0	1.0	0.0	0.0	0.0	
max	90750.000000	22.0	33.0	33.0	5.0	9.0	869.0	
	V144	V145	V146	V147	V148	V149	V150	V151 \
count	81951.0	81951.0	81945.0	81945.0	81945.0	81945.0	81951.0	81951.0
mean	NaN	NaN	0.0	0.0	0.0	0.0	NaN	NaN
std	0.0	NaN	0.0	0.0	0.0	0.0	NaN	0.0
min	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
25%	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
50%	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0
75%	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0
max	62.0	297.0	24.0	26.0	20.0	20.0	3388.0	57.0
	V152	V153	V154	V155	V156	V157	V158	V159 \
count	81951.0	81945.0	81945.0	81945.0	81945.0	81945.0	81945.0	81951.0
mean	NaN	0.0	0.0	0.0	0.0	NaN	NaN	NaN
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN
min	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
25%	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0
max	69.0	18.0	18.0	24.0	24.0	24.0	24.0	55136.0
	V160	V161	V162	V163	V164	V165		\
count	81951.000000	81945.0	81945.0	81945.0	81951.000000	81951.000000		
mean	47453.191406	NaN	NaN	NaN	877.888794	2239.911865		
std	142076.062500	NaN	NaN	NaN	6049.166016	8223.258789		
min	0.000000	0.0	0.0	0.0	0.000000	0.000000		
25%	0.000000	0.0	0.0	0.0	0.000000	0.000000		
50%	0.000000	0.0	0.0	0.0	0.000000	0.000000		
75%	0.000000	0.0	0.0	0.0	0.000000	0.000000		

max	641511.437500	3300.0	3300.0	3300.0	93736.000000	98476.000000		\
count	81951.000000	139631.0	139631.0	139819.0	139819.0	139819.0	139819.0	
mean	359.469391	NaN	NaN	0.0	NaN	NaN	NaN	
std	1244.463257	NaN	NaN	0.0	0.0	0.0	0.0	
min	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.000000	0.0	0.0	0.0	1.0	1.0	1.0	
50%	0.000000	0.0	0.0	0.0	1.0	1.0	1.0	
75%	0.000000	0.0	1.0	0.0	1.0	1.0	1.0	
max	104060.000000	872.0	964.0	19.0	48.0	61.0		
count	139631.0	139631.0	139819.0	139819.0	139631.0	139631.0	139631.0	\
mean	0.0	0.0	0.0	0.0	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	1.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	1.0	0.0	0.0	
75%	0.0	0.0	0.0	0.0	1.0	0.0	0.0	
max	31.0	7.0	8.0	14.0	48.0	861.0	1235.0	
count	139631.0	139819.0	139631.0	139631.0	139631.0	139819.0	139819.0	\
mean	NaN	NaN	0.0	NaN	NaN	0.0	0.0	
std	NaN	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
max	920.0	83.0	24.0	83.0	41.0	16.0	31.0	
count	139631.0	139631.0	139819.0	139819.0	139631.0	139631.0	139631.0	\
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	38.0	218.0	30.0	30.0	42.0	21.0	44.0	
count	139631.0	139819.0	139819.0	139631.0	139819.0	139819.0	139631.0	\
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	

25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0
max	37.0	7.0	16.0	38.0	14.0	21.0	45.0

	V200	V201	V202	V203	V204	\
count	139819.0	139819.0	139631.000000	139631.000000	139631.000000	
mean	NaN	NaN	444.147125	1078.327515	686.956909	
std	0.0	0.0	4683.828125	9105.608398	6048.980957	
min	0.0	0.0	0.000000	0.000000	0.000000	
25%	1.0	1.0	0.000000	0.000000	0.000000	
50%	1.0	1.0	0.000000	0.000000	0.000000	
75%	1.0	1.0	0.000000	30.924400	20.000000	
max	45.0	55.0	104060.000000	139777.000000	104060.000000	

	V205	V206	V207	V208	V209	V210	\
count	139631.0	139631.0	139631.0	139819.0	139819.0	139819.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.0	0.0	0.0	0.0	0.0	0.0	
max	55136.0	55136.0	55136.0	3300.0	8048.0	3300.0	

	V211	V212	V213	V214	\
count	139631.000000	139631.000000	139631.000000	139631.000000	
mean	385.137024	765.988464	536.302734	38.43755	
std	4541.837891	7496.120605	5471.664551	571.83429	
min	0.000000	0.000000	0.000000	0.00000	
25%	0.000000	0.000000	0.000000	0.00000	
50%	0.000000	0.000000	0.000000	0.00000	
75%	0.000000	0.000000	0.000000	0.00000	
max	92888.000000	129006.000000	97628.000000	104060.000000	

	V215	V216	V217	V218	V219	V220	\
count	139631.000000	139631.000000	130430.0	130430.0	130430.0	141416.0	
mean	133.208221	71.107147	NaN	NaN	NaN	0.0	
std	1040.453735	680.267639	NaN	NaN	NaN	0.0	
min	0.000000	0.000000	0.0	0.0	0.0	0.0	
25%	0.000000	0.000000	0.0	0.0	0.0	0.0	
50%	0.000000	0.000000	0.0	0.0	0.0	0.0	
75%	0.000000	0.000000	0.0	1.0	1.0	0.0	
max	104060.000000	104060.000000	303.0	400.0	378.0	25.0	

	V221	V222	V223	V224	V225	V226	V227	\
count	141416.0	141416.0	130430.0	130430.0	130430.0	130430.0	141416.0	

mean	NaN	NaN	0.0	0.0	0.0	0.0	0.0	
std	NaN	NaN	0.0	0.0	0.0	0.0	0.0	NaN
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
50%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
75%	1.0	1.0	0.0	0.0	0.0	0.0	0.0	
max	384.0	384.0	16.0	144.0	51.0	242.0	360.0	
	V228	V229	V230	V231	V232	V233	V234	\
count	130430.0	130430.0	130430.0	130430.0	130430.0	130430.0	141416.0	
mean	NaN							
std	0.0	0.0	0.0	NaN	NaN	NaN	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
50%	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
75%	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
max	54.0	176.0	65.0	293.0	337.0	332.0	121.0	
	V235	V236	V237	V238	V239	V240	V241	\
count	130430.0	130430.0	130430.0	141416.0	141416.0	130430.0	130430.0	
mean	0.0	0.0	0.0	0.0	0.0	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
50%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
75%	0.0	0.0	0.0	0.0	0.0	1.0	1.0	
max	23.0	45.0	39.0	23.0	23.0	7.0	5.0	
	V242	V243	V244	V245	V246	V247	V248	\
count	130430.0	130430.0	130430.0	141416.0	130430.0	130430.0	130430.0	
mean	NaN							
std	0.0	0.0	0.0	NaN	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	20.0	57.0	22.0	262.0	45.0	18.0	36.0	
	V249	V250	V251	V252	V253	V254	V255	\
count	130430.0	141416.0	141416.0	130430.0	130430.0	130430.0	141416.0	
mean	NaN							
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	22.0	18.0	18.0	24.0	163.0	60.0	87.0	

	V256	V257	V258	V259	V260	V261	V262	\
count	141416.0	130430.0	130430.0	141416.0	130430.0	130430.0	130430.0	
mean	NaN							
std	0.0	0.0	0.0	NaN	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
50%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
75%	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
max	87.0	48.0	66.0	285.0	8.0	49.0	20.0	

	V263	V264	V265	V266	V267	\
count	130430.000000	130430.000000	130430.000000	130430.0	130430.0	
mean	117.390648	201.657608	153.520554	NaN	NaN	
std	1294.851562	2284.827637	1605.512329	NaN	NaN	
min	0.000000	0.000000	0.000000	0.0	0.0	
25%	0.000000	0.000000	0.000000	0.0	0.0	
50%	0.000000	0.000000	0.000000	0.0	0.0	
75%	0.000000	33.593498	20.897526	0.0	0.0	
max	153600.000000	153600.000000	153600.000000	55136.0	55136.0	

	V268	V269	V270	V271	V272	V273	\
count	130430.0	130430.0	141416.0	141416.0	141416.0	130430.0	
mean	NaN	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	NaN	NaN	
min	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.0	0.0	0.0	0.0	0.0	0.0	
max	55136.0	55136.0	4000.0	4000.0	4000.0	51200.0	

	V274	V275	V276	V277	V278	\
count	130430.000000	130430.0	130430.000000	130430.000000	130430.000000	
mean	107.151642	NaN	31.797276	51.956646	42.328224	
std	1258.734131	NaN	615.659729	732.145386	660.611816	
min	0.000000	0.0	0.000000	0.000000	0.000000	
25%	0.000000	0.0	0.000000	0.000000	0.000000	
50%	0.000000	0.0	0.000000	0.000000	0.000000	
75%	0.000000	0.0	0.000000	0.000000	0.000000	
max	66000.000000	51200.0	104060.000000	104060.000000	104060.000000	

	V279	V280	V281	V282	V283	V284	V285	\
count	590528.0	590528.0	589271.0	589271.0	589271.0	590528.0	590528.0	
mean	NaN	NaN	0.0	NaN	NaN	0.0	NaN	
std	NaN	NaN	0.0	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	

50%	0.0	0.0	0.0	1.0	1.0	0.0	0.0	\
75%	0.0	1.0	0.0	1.0	1.0	0.0	1.0	
max	880.0	975.0	22.0	32.0	68.0	12.0	95.0	
	V286	V287	V288	V289	V290	V291	V292	\
count	590528.0	590528.0	589271.0	589271.0	590528.0	590528.0	590528.0	
mean	0.0	NaN	NaN	NaN	NaN	NaN	NaN	
std	0.0	0.0	0.0	0.0	0.0	NaN	NaN	
min	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
25%	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
50%	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
75%	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
max	8.0	31.0	10.0	12.0	67.0	1055.0	323.0	
	V293	V294	V295	V296	V297	V298	V299	\
count	590528.0	590528.0	590528.0	589271.0	590528.0	590528.0	590528.0	
mean	NaN	NaN	NaN	NaN	0.0	NaN	NaN	
std	NaN	NaN	NaN	0.0	0.0	0.0	0.0	
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
max	869.0	1286.0	928.0	93.0	12.0	93.0	49.0	
	V300	V301	V302	V303	V304	V305	V309	\
count	589271.0	589271.0	590528.0	590528.0	590528.0	590528.0	590528.0	
mean	0.0	0.0	NaN	NaN	NaN	NaN		
std	0.0	0.0	0.0	0.0	0.0	0.0		
min	0.0	0.0	0.0	0.0	0.0	1.0		
25%	0.0	0.0	0.0	0.0	0.0	1.0		
50%	0.0	0.0	0.0	0.0	0.0	1.0		
75%	0.0	0.0	0.0	0.0	0.0	1.0		
max	11.0	13.0	16.0	20.0	16.0	2.0		
	V306	V307	V308	V309	V310	V311	V312	\
count	590528.000000	590528.000000	590528.000000	590528.000000	590528.0	590528.0000	590528.0000	
mean	139.748718	408.682404	230.413162	NaN	NaN	NaN	NaN	
std	2348.849609	4391.993164	3021.924316	NaN	NaN	NaN	NaN	
min	0.000000	0.000000	0.000000	0.0	0.0	0.0000	0.0000	
25%	0.000000	0.000000	0.000000	0.0	0.0	0.0000	0.0000	
50%	0.000000	0.000000	0.000000	0.0	0.0	0.0000	0.0000	
75%	0.000000	151.380676	35.970001	0.0	0.0	107.9375		
max	108800.000000	145765.000000	108800.000000	55136.0	55136.0	55136.0000		
	V313	V314	V315	V316	V317	V318	V319	\
count	590528.0	590528.0	589271.0	589271.0	589271.0	590528.000000	590528.000000	
mean	NaN	NaN	NaN	NaN	NaN	109.818542		

std	NaN	NaN	NaN	NaN	NaN	2270.033203
min	0.0	0.0	0.0	0.0	0.0	0.000000
25%	0.0	0.0	0.0	0.0	0.0	0.000000
50%	0.0	0.0	0.0	0.0	0.0	0.000000
75%	0.0	0.0	0.0	0.0	0.0	0.000000
max	55136.0	55136.0	4816.0	7520.0	4816.0	93736.000000

	V317	V318	V319	V320	\
count	590528.000000	590528.000000	590528.000000	590528.000000	
mean	247.606750	162.153412	18.372477	42.073135	
std	3980.042725	2793.343506	332.304840	473.499298	
min	0.000000	0.000000	0.000000	0.000000	
25%	0.000000	0.000000	0.000000	0.000000	
50%	0.000000	0.000000	0.000000	0.000000	
75%	0.000000	0.000000	0.000000	0.000000	
max	134021.000000	98476.000000	104060.000000	104060.000000	

	V321	V322	V323	V324	V325	V326	V327	\
count	590528.000000	82351.0	82351.0	82351.0	82351.0	82351.0	82351.0	
mean	28.326582	NaN	NaN	NaN	0.0	NaN	0.0	
std	382.053162	NaN	NaN	NaN	0.0	0.0	0.0	
min	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
25%	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
50%	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	
75%	0.000000	0.0	1.0	0.0	0.0	0.0	0.0	
max	104060.000000	880.0	1411.0	976.0	12.0	44.0	18.0	

	V328	V329	V330	V331	V332	V333	\
count	82351.0	82351.0	82351.0	82351.000000	82351.000000	82351.000000	
mean	0.0	NaN	0.0	721.741882	1375.783691	1014.622803	
std	0.0	0.0	0.0	6217.223633	11169.276367	7955.735352	
min	0.0	0.0	0.0	0.000000	0.000000	0.000000	
25%	0.0	0.0	0.0	0.000000	0.000000	0.000000	
50%	0.0	0.0	0.0	0.000000	0.000000	0.000000	
75%	0.0	0.0	0.0	0.000000	25.000000	0.000000	
max	15.0	99.0	55.0	160000.000000	160000.000000	160000.000000	

	V334	V335	V336	V337	V338	V339	
count	82351.0	82351.0	82351.0	82351.000000	82351.000000	82351.000000	
mean	NaN	NaN	NaN	55.352417	151.160538	100.700874	
std	NaN	NaN	NaN	668.486816	1095.034302	814.946716	
min	0.0	0.0	0.0	0.000000	0.000000	0.000000	
25%	0.0	0.0	0.0	0.000000	0.000000	0.000000	
50%	0.0	0.0	0.0	0.000000	0.000000	0.000000	
75%	0.0	0.0	0.0	0.000000	0.000000	0.000000	
max	55136.0	55136.0	55136.0	104060.000000	104060.000000	104060.000000	

Check class imbalance

```
[16]: df_tran.loc[:, 'isFraud'].value_counts()
```

```
[16]: 0    569877  
1    20663  
Name: isFraud, dtype: int64
```

```
[17]: df_tran.loc[:, 'isFraud'].value_counts(normalize=True)*100
```

```
[17]: 0    96.500999  
1    3.499001  
Name: isFraud, dtype: float64
```

Lot of interesting things can be observed here:

- Rows in identity dataset are less than transaction dataset, that means only a subset of transactions in transactions dataset has identity data
- Both datasets have the common and unique key as TransactionID, both can be joined at this unique key
- id_24, id_25, dist2, D7 and many more columns have 90%+ missing values, which means that these columns are probably useless so need to drop it for now
- Columns from V1 to V339 in transaction dataset are numeric whereas columns from id_01 to id_39 are of mixed datatype
- TransactionDT column is a timedelta from a given reference datetime (not an actual timestamp). But reference datetime is not known, so need to assume it and convert it to date format
- Target class is imbalanced. So no need to drop the columns where one category contains the majority of rows

5 5. Data Preprocessing for EDA

Merge the datasets

```
[18]: df = df_tran.merge(df_id, how='left', on='TransactionID')
```

```
del df_tran, df_id  
gc.collect()
```

```
[18]: 0
```

```
[19]: df.shape
```

```
[19]: (590540, 434)
```

Add missing flag

```
[20]: for col in df.columns:
    df[col+"_missing_flag"] = df[col].isnull()

df.head()
```

	TransactionID	isFraud	TransactionDT	TransactionAmt	ProductCD	card1	\
0	2987000	0	86400	68.5	W	13926	
1	2987001	0	86401	29.0	W	2755	
2	2987002	0	86469	59.0	W	4663	
3	2987003	0	86499	50.0	W	18132	
4	2987004	0	86506	50.0	H	4497	

	card2	card3	card4	card5	card6	addr1	addr2	dist1	dist2	\
0	NaN	150.0	discover	142.0	credit	315.0	87.0	19.0	NaN	
1	404.0	150.0	mastercard	102.0	credit	325.0	87.0	NaN	NaN	
2	490.0	150.0	visa	166.0	debit	330.0	87.0	287.0	NaN	
3	567.0	150.0	mastercard	117.0	debit	476.0	87.0	NaN	NaN	
4	514.0	150.0	mastercard	102.0	credit	420.0	87.0	NaN	NaN	

	P_emaildomain	R_emaildomain	C1	C2	C3	C4	C5	C6	C7	C8	C9	\
0	NaN	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	
1	gmail.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	
2	outlook.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	
3	yahoo.com	NaN	2.0	5.0	0.0	0.0	0.0	4.0	0.0	0.0	1.0	
4	gmail.com	NaN	1.0	1.0	0.0	0.0	0.0	1.0	0.0	1.0	0.0	

	C10	C11	C12	C13	C14	D1	D2	D3	D4	D5	D6	D7	D8	D9	\
0	0.0	2.0	0.0	1.0	1.0	14.0	NaN	13.0	NaN	NaN	NaN	NaN	NaN	NaN	
1	0.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	NaN	NaN	
2	0.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	NaN	NaN	
3	0.0	1.0	0.0	25.0	1.0	112.0	112.0	0.0	94.0	0.0	NaN	NaN	NaN	NaN	
4	1.0	1.0	0.0	1.0	1.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

	D10	D11	D12	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	\
0	13.0	13.0	NaN	NaN	NaN	0.0	T	T	T	M2	F	T	NaN	NaN	
1	0.0	NaN	NaN	NaN	NaN	0.0	NaN	NaN	NaN	M0	T	T	NaN	NaN	
2	0.0	315.0	NaN	NaN	NaN	315.0	T	T	T	M0	F	F	F	F	
3	84.0	NaN	NaN	NaN	NaN	111.0	NaN	NaN	NaN	MO	T	F	NaN	NaN	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

	M9	V1	V2	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	\
0	NaN	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0		
1	NaN	0.0	0.0	1.0												
2	F	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0		
3	NaN	1.0	1.0	1.0												
4	NaN															

	V15	V16	V17	V18	V19	V20	V21	V22	V23	V24	V25	V26	V27	V28	V29	\
0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
4	NaN	NaN														
	V30	V31	V32	V33	V34	V35	V36	V37	V38	V39	V40	V41	V42	V43	V44	\
0	0.0	0.0	0.0	0.0	0.0	NaN	NaN									
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
2	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
3	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	
4	NaN	NaN														
	V45	V46	V47	V48	V49	V50	V51	V52	V53	V54	V55	V56	V57	V58	V59	\
0	NaN	1.0	1.0	1.0	1.0	0.0	0.0	0.0								
1	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	
2	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
3	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	
4	NaN	NaN														
	V60	V61	V62	V63	V64	V65	V66	V67	V68	V69	V70	V71	V72	V73	V74	\
0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN														
	V75	V76	V77	V78	V79	V80	V81	V82	V83	V84	V85	V86	V87	V88	V89	\
0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	0.0	
1	0.0	0.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	
2	1.0	1.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	
3	1.0	1.0	1.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	
4	NaN	NaN														
	V90	V91	V92	V93	V94	V95	V96	V97	V98	V99	V100	V101	V102	\		
0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0		
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		
3	0.0	0.0	0.0	0.0	0.0	1.0	48.0	28.0	0.0	10.0	4.0	1.0	1.0	38.0		
4	NaN	NaN	NaN	NaN	NaN	NaN	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V103	V104	V105	V106	V107	V108	V109	V110	V111	V112	V113	V114	\			
0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
1	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
2	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		
3	24.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0		

4	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
0	V115	V116	V117	V118	V119	V120	V121	V122	V123	V124	V125	V126	V127	\
1	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0
2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0
3	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	50.0
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0
0	V127	V128	V129	V130	V131	V132	V133	V134	V135	V136	V137	V138	V139	\
1	117.0	0.0	0.0	0.0	0.0	0.0	117.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	1758.0	925.0	0.0	354.0	135.0	50.0	1404.0	790.0	0.0	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0	V138	V139	V140	V141	V142	V143	V144	V145	V146	V147	V148	V149	V150	\
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	0.0	0.0	0.0	0.0	0.0	6.0	18.0	140.0	0.0	0.0	0.0	0.0	0.0	0.0
0	V150	V151	V152	V153	V154	V155	V156	V157	V158	V159	V160	V161	V162	\
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	1803.0	49.0	64.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	15560.0	V163	V164	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	169690.796875	0.0	0.0	0.0	515.0	5155.0	2840.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0	V170	V171	V172	V173	V174	V175	V176	V177	V178	V179	V180	V181	V182	\
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	1.0	1.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0	V182	V183	V184	V185	V186	V187	V188	V189	V190	V191	V192	V193	V182	\
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN

2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
	V194	V195	...	V130_missing_flag	V131_missing_flag	V132_missing_flag	\						
0	NaN	NaN	...		False		False		False		False		
1	NaN	NaN	...		False		False		False		False		
2	NaN	NaN	...		False		False		False		False		
3	NaN	NaN	...		False		False		False		False		
4	1.0	1.0	...		False		False		False		False		
	V133_missing_flag	V134_missing_flag	V135_missing_flag	V136_missing_flag	\								
0		False		False		False		False		False			
1		False		False		False		False		False			
2		False		False		False		False		False			
3		False		False		False		False		False			
4		False		False		False		False		False			
	V137_missing_flag	V138_missing_flag	V139_missing_flag	V140_missing_flag	\								
0		False		True		True		True		True			
1		False		True		True		True		True			
2		False		True		True		True		True			
3		False		True		True		True		True			
4		False		False		False		False		False			
	V141_missing_flag	V142_missing_flag	V143_missing_flag	V144_missing_flag	\								
0		True		True		True		True		True			
1		True		True		True		True		True			
2		True		True		True		True		True			
3		True		True		True		True		True			
4		False		False		False		False		False			
	V145_missing_flag	V146_missing_flag	V147_missing_flag	V148_missing_flag	\								
0		True		True		True		True		True			
1		True		True		True		True		True			
2		True		True		True		True		True			
3		True		True		True		True		True			
4		False		False		False		False		False			
	V149_missing_flag	V150_missing_flag	V151_missing_flag	V152_missing_flag	\								
0		True		True		True		True		True			
1		True		True		True		True		True			
2		True		True		True		True		True			
3		True		True		True		True		True			
4		False		False		False		False		False			
	V153_missing_flag	V154_missing_flag	V155_missing_flag	V156_missing_flag	\								

0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V157_missing_flag	V158_missing_flag	V159_missing_flag	V160_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V161_missing_flag	V162_missing_flag	V163_missing_flag	V164_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V165_missing_flag	V166_missing_flag	V167_missing_flag	V168_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V169_missing_flag	V170_missing_flag	V171_missing_flag	V172_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V173_missing_flag	V174_missing_flag	V175_missing_flag	V176_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V177_missing_flag	V178_missing_flag	V179_missing_flag	V180_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V181_missing_flag	V182_missing_flag	V183_missing_flag	V184_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V185_missing_flag	V186_missing_flag	V187_missing_flag	V188_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V189_missing_flag	V190_missing_flag	V191_missing_flag	V192_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V193_missing_flag	V194_missing_flag	V195_missing_flag	V196_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V197_missing_flag	V198_missing_flag	V199_missing_flag	V200_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V201_missing_flag	V202_missing_flag	V203_missing_flag	V204_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V205_missing_flag	V206_missing_flag	V207_missing_flag	V208_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True

3	True	True	True	True	True
4	False	False	False	False	False
	V209_missing_flag	V210_missing_flag	V211_missing_flag	V212_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V213_missing_flag	V214_missing_flag	V215_missing_flag	V216_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V217_missing_flag	V218_missing_flag	V219_missing_flag	V220_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V221_missing_flag	V222_missing_flag	V223_missing_flag	V224_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V225_missing_flag	V226_missing_flag	V227_missing_flag	V228_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V229_missing_flag	V230_missing_flag	V231_missing_flag	V232_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V233_missing_flag	V234_missing_flag	V235_missing_flag	V236_missing_flag	\
0	True	True	True	True	True

1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V237_missing_flag	V238_missing_flag	V239_missing_flag	V240_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V241_missing_flag	V242_missing_flag	V243_missing_flag	V244_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V245_missing_flag	V246_missing_flag	V247_missing_flag	V248_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V249_missing_flag	V250_missing_flag	V251_missing_flag	V252_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V253_missing_flag	V254_missing_flag	V255_missing_flag	V256_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V257_missing_flag	V258_missing_flag	V259_missing_flag	V260_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V261_missing_flag	V262_missing_flag	V263_missing_flag	V264_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V265_missing_flag	V266_missing_flag	V267_missing_flag	V268_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V269_missing_flag	V270_missing_flag	V271_missing_flag	V272_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V273_missing_flag	V274_missing_flag	V275_missing_flag	V276_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V277_missing_flag	V278_missing_flag	V279_missing_flag	V280_missing_flag	\
0	True	True	False	False	False
1	True	True	False	False	False
2	True	True	False	False	False
3	True	True	False	False	False
4	False	False	False	False	False

	V281_missing_flag	V282_missing_flag	V283_missing_flag	V284_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False

	V285_missing_flag	V286_missing_flag	V287_missing_flag	V288_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False

4	False	False	False	False	
	V289_missing_flag	V290_missing_flag	V291_missing_flag	V292_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V293_missing_flag	V294_missing_flag	V295_missing_flag	V296_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V297_missing_flag	V298_missing_flag	V299_missing_flag	V300_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V301_missing_flag	V302_missing_flag	V303_missing_flag	V304_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V305_missing_flag	V306_missing_flag	V307_missing_flag	V308_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V309_missing_flag	V310_missing_flag	V311_missing_flag	V312_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V313_missing_flag	V314_missing_flag	V315_missing_flag	V316_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False

2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V317_missing_flag	V318_missing_flag	V319_missing_flag	V320_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V321_missing_flag	V322_missing_flag	V323_missing_flag	V324_missing_flag	\
0	False	True	True	True	True
1	False	True	True	True	True
2	False	True	True	True	True
3	False	True	True	True	True
4	False	False	False	False	False
	V325_missing_flag	V326_missing_flag	V327_missing_flag	V328_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V329_missing_flag	V330_missing_flag	V331_missing_flag	V332_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V333_missing_flag	V334_missing_flag	V335_missing_flag	V336_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V337_missing_flag	V338_missing_flag	V339_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	id_01_missing_flag	id_02_missing_flag	id_03_missing_flag	\	

0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
	id_04_missing_flag	id_05_missing_flag	id_06_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_07_missing_flag	id_08_missing_flag	id_09_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_10_missing_flag	id_11_missing_flag	id_12_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_13_missing_flag	id_14_missing_flag	id_15_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_16_missing_flag	id_17_missing_flag	id_18_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
	id_19_missing_flag	id_20_missing_flag	id_21_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	

```

    id_22_missing_flag  id_23_missing_flag  id_24_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          True           True           True

    id_25_missing_flag  id_26_missing_flag  id_27_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          True           True           True

    id_28_missing_flag  id_29_missing_flag  id_30_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          False          False          False

    id_31_missing_flag  id_32_missing_flag  id_33_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          False          False          False

    id_34_missing_flag  id_35_missing_flag  id_36_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          False          False          False

    id_37_missing_flag  id_38_missing_flag  DeviceType_missing_flag  \
0          True           True           True
1          True           True           True
2          True           True           True
3          True           True           True
4          False          False          False

    DeviceInfo_missing_flag
0          True
1          True
2          True

```

```
3          True
4          False
```

```
[5 rows x 868 columns]
```

Clean Data Let's drop the columns which may not be useful for our analysis

Create a missing value flag column for the columns we are dropping which have more than 90% missing values, there might be some specific pattern associated with missing values and transaction being fraud

```
[21]: drop_cols = []

for col in df.columns:
    missing_share = df[col].isnull().sum()/df.shape[0]
    if missing_share > 0.9:
        drop_cols.append(col)
        print(col)
    # df[col + "_missing_flag"] = df[col].isnull()

good_cols = [col for col in df.columns if col not in drop_cols]
```

```
dist2
D7
id_07
id_08
id_18
id_21
id_22
id_23
id_24
id_25
id_26
id_27
```

```
[22]: drop_cols = []
for col in good_cols:
    unique_value = df[col].nunique()
    if unique_value == 1:
        drop_cols.append(col)
        print(col)
good_cols = [col for col in good_cols if col not in drop_cols]
```

```
TransactionID_missing_flag
isFraud_missing_flag
TransactionDT_missing_flag
TransactionAmt_missing_flag
ProductCD_missing_flag
```

```
card1_missing_flag  
C1_missing_flag  
C2_missing_flag  
C3_missing_flag  
C4_missing_flag  
C5_missing_flag  
C6_missing_flag  
C7_missing_flag  
C8_missing_flag  
C9_missing_flag  
C10_missing_flag  
C11_missing_flag  
C12_missing_flag  
C13_missing_flag  
C14_missing_flag
```

```
[23]: # Filter the data for relevant columns only  
df = df[good_cols]
```

```
[24]: df.shape
```

```
[24]: (590540, 836)
```

Create date features

```
[25]: START_DATE      = '2017-12-01'  
startdate       = datetime.datetime.strptime(START_DATE, "%Y-%m-%d")  
df["Date"]       = df['TransactionDT'].apply(lambda x: (startdate + datetime.  
           timedelta(seconds=x)))  
  
df['_Weekdays']   = df['Date'].dt.dayofweek  
df['_Hours']      = df['Date'].dt.hour  
df['_Days']       = df['Date'].dt.day
```

```
[26]: df = reduce_mem_usage(df)
```

```
Mem. usage decreased to 1453.88 Mb (0.8% reduction)
```

6 Exploratory Data Analysis

6.0.1 Check distribution of target variable

```
[27]: df['isFraud'].value_counts()
```

```
[27]: 0    569877  
1    20663  
Name: isFraud, dtype: int64
```

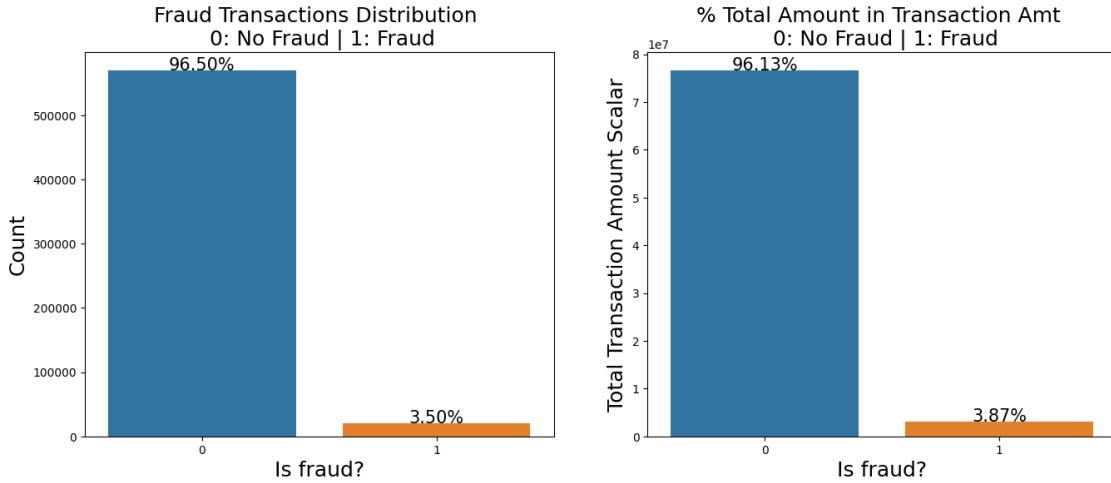
```
[28]: df['TransactionAmt'] = df['TransactionAmt'].astype(float)
total = len(df)
total_amt = df.groupby(['isFraud'])['TransactionAmt'].sum().sum()
plt.figure(figsize=(16,6))

plt.subplot(121)
g = sns.countplot(x='isFraud', data=df )
g.set_title("Fraud Transactions Distribution \n 0: No Fraud | 1: Fraud",□
    ↪fontsize=18)
g.set_xlabel("Is fraud?", fontsize=18)
g.set_ylabel('Count', fontsize=18)
for p in g.patches:
    height = p.get_height()
    g.text(p.get_x()+p.get_width()/2.,
           height + 3,
           '{:1.2f}%'.format(height/total*100),
           ha="center", fontsize=15)

perc_amt = (df.groupby(['isFraud'])['TransactionAmt'].sum())
perc_amt = perc_amt.reset_index()

plt.subplot(122)
g1 = sns.barplot(x='isFraud', y='TransactionAmt', dodge=True, data=perc_amt)
g1.set_title("% Total Amount in Transaction Amt \n 0: No Fraud | 1: Fraud",□
    ↪fontsize=18)
g1.set_xlabel("Is fraud?", fontsize=18)
g1.set_ylabel('Total Transaction Amount Scalar', fontsize=18)
for p in g1.patches:
    height = p.get_height()
    g1.text(p.get_x()+p.get_width()/2.,
           height + 3,
           '{:1.2f}%'.format(height/total_amt * 100),
           ha="center", fontsize=15)

plt.show()
```



```
[29]: df.groupby('isFraud')['TransactionAmt'].mean()
```

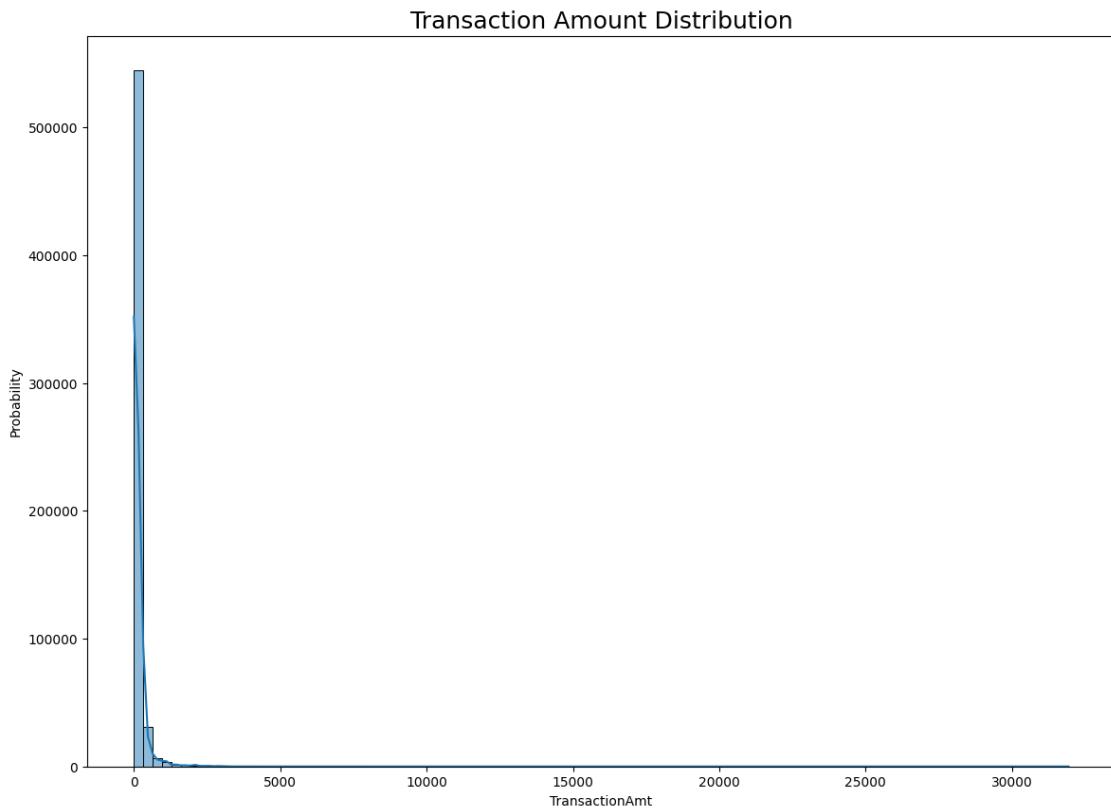
```
[29]: isFraud
0    134.511857
1    149.244353
Name: TransactionAmt, dtype: float64
```

- The target variable is **imbalanced**. 3.5% transactions are Fraud
- Around same % of transaction amounts are fraud

Let's explore the Transaction amount further

6.0.2 Check distribution of Transaction Amount

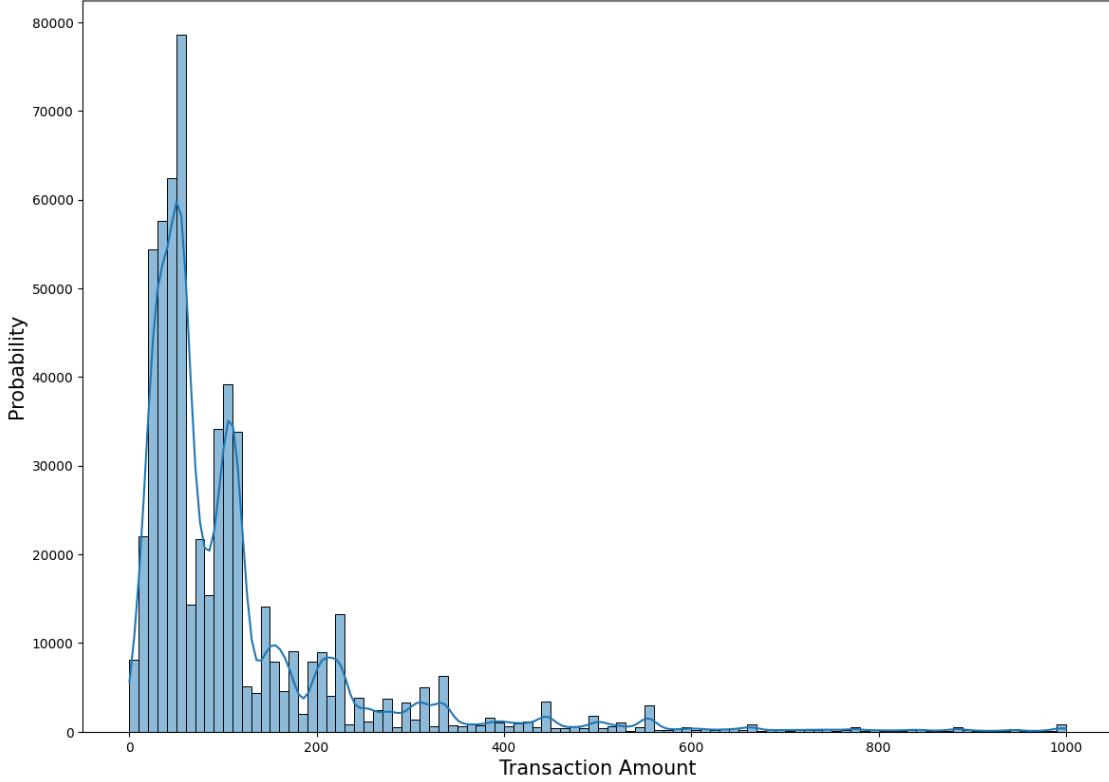
```
[30]: plt.figure(figsize=(14,10))
sns.histplot(df['TransactionAmt'], bins=100, kde=True)
plt.title("Transaction Amount Distribution", fontsize=18)
plt.ylabel("Probability")
plt.show()
```



```
[31]: # Distribution plot of Transaction Amount less than 1000
plt.figure(figsize=(14,10))
plt.suptitle('Transaction Values Distribution', fontsize=22)
sns.histplot(df[df['TransactionAmt'] <=
                1000]['TransactionAmt'], bins=100, kde=True)
plt.title("Transaction Amount Distribuition <= 1000", fontsize=18)
plt.xlabel("Transaction Amount", fontsize=15)
plt.ylabel("Probability", fontsize=15)
plt.show()
```

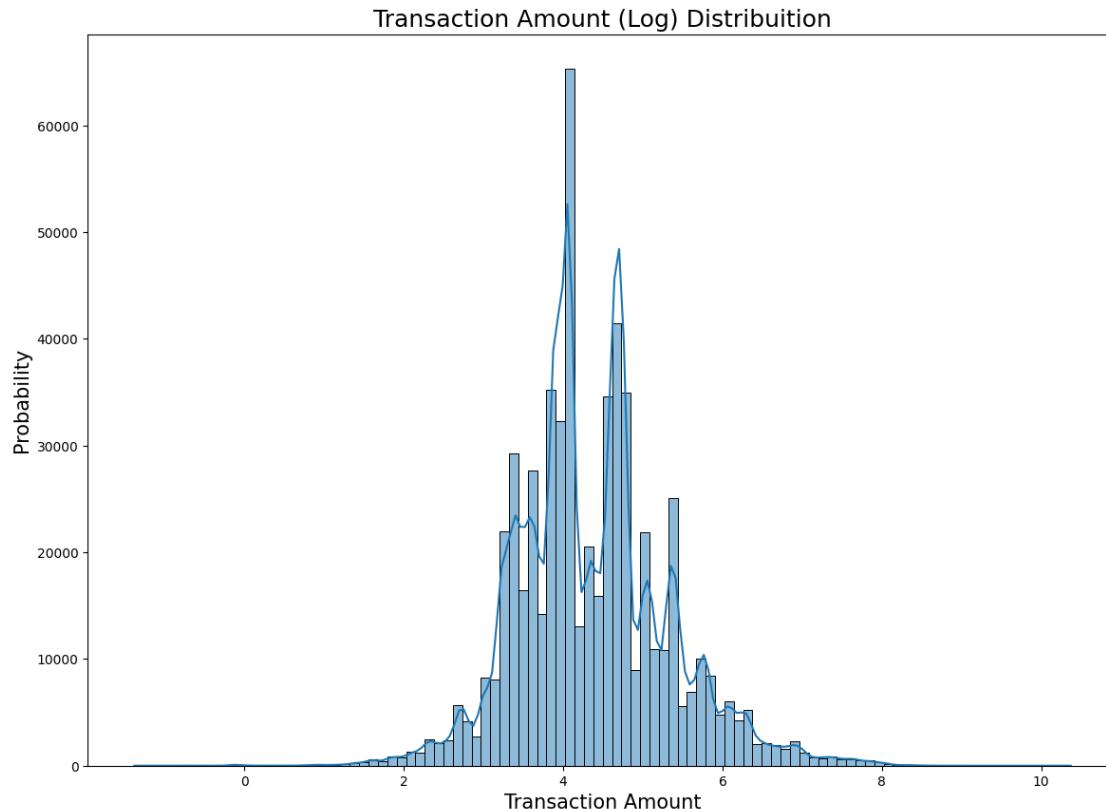
Transaction Values Distribution

Transaction Amount Distribuition <= 1000



```
[32]: # Distribution plot of Transaction Amount less than 1000
plt.figure(figsize=(14,10))
plt.suptitle('Transaction Values Distribution', fontsize=22)
sns.histplot(np.log(df['TransactionAmt']), kde=True, bins=100)
plt.title("Transaction Amount (Log) Distribuition", fontsize=18)
plt.xlabel("Transaction Amount", fontsize=15)
plt.ylabel("Probability", fontsize=15)
plt.show()
```

Transaction Values Distribution



- Transaction Amount is right skewed.
- Log of transaction amount is almost normally distributed, so use log of transaction amount while building the model

6.0.3 Product Features

- Distribution of ProductCD
- Distribution of Frauds by Product

```
[33]: def plot_cat_feat_dist(df, col):  
    tmp = pd.crosstab(df[col], df['isFraud'], normalize='index') * 100  
    tmp = tmp.reset_index()  
    tmp.rename(columns={0:'NoFraud', 1:'Fraud'}, inplace=True)  
  
    plt.figure(figsize=(16,12))  
    plt.suptitle(f'{col} Distributions', fontsize=22)  
  
    plt.subplot(221)  
    g = sns.countplot(x=col, data=df, order=tmp[col].values)
```

```

g.set_title(f"{col} Distribution", fontsize=16)
g.set_xlabel(f"{col} Name", fontsize=17)
g.set_ylabel("Count", fontsize=17)
for p in g.patches:
    height = p.get_height()
    g.text(p.get_x() + p.get_width() / 2.,
           height + 3,
           '{:1.2f}%'.format(height / total * 100),
           ha="center", fontsize=14)

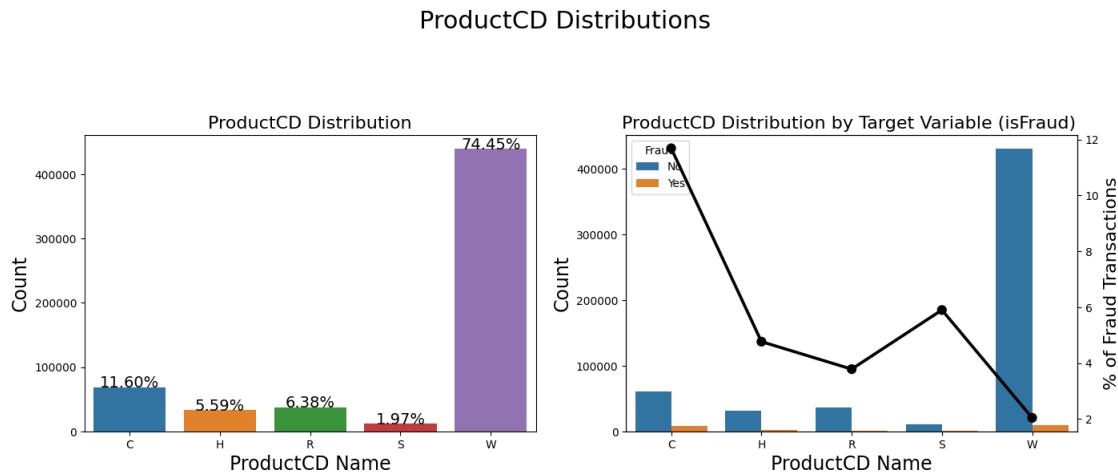
plt.subplot(222)
g1 = sns.countplot(x=col, hue='isFraud', data=df, order=tmp[col].values)
plt.legend(title='Fraud', loc='best', labels=['No', 'Yes'])
gt = g1.twinx()
gt = sns.pointplot(x=col, y='Fraud', data=tmp, color='black', order=tmp[col].values, legend=False)
gt.set_ylabel("% of Fraud Transactions", fontsize=16)

g1.set_title(f"{col} Distribution by Target Variable (isFraud) ", fontsize=16)
g1.set_xlabel(f"{col} Name", fontsize=17)
g1.set_ylabel("Count", fontsize=17)

plt.subplots_adjust(hspace = 0.4, top = 0.85)
plt.show()

```

[34]: plot_cat_feat_dist(df, "ProductCD")



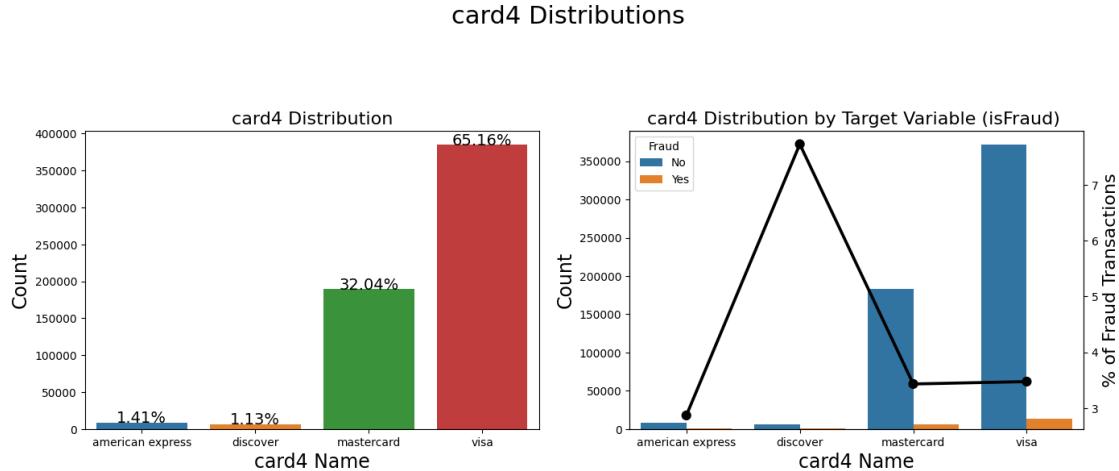
[35]: df.groupby('ProductCD')['isFraud'].mean()

```
[35]: ProductCD
C      0.116873
H      0.047662
R      0.037826
S      0.058996
W      0.020399
Name: isFraud, dtype: float64
```

- 75% of the transactions are for Product Category W
- 11.6% of the transactions are for Product Category C
- Fraud Transaction rate is maximum for Product Category C and minimum for Product Category W

6.0.4 Card Features

```
[36]: # Card 4
plot_cat_feat_dist(df, "card4")
```



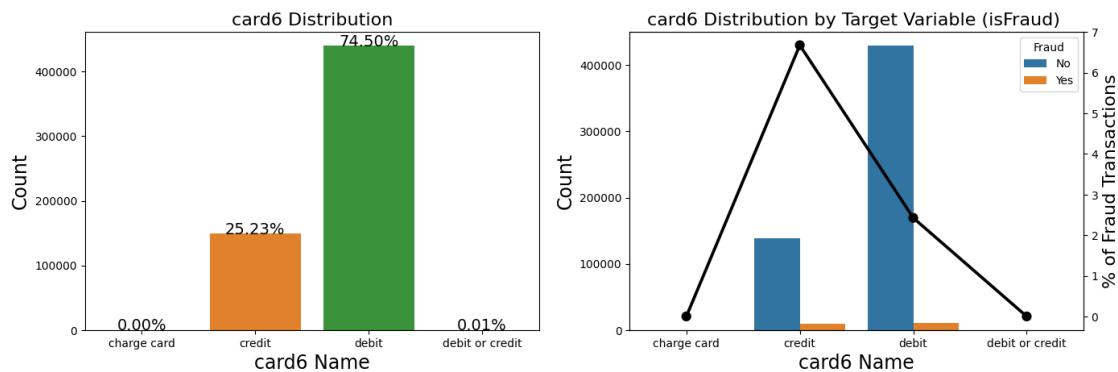
```
[37]: df.groupby('card4')['isFraud'].mean()
```

```
[37]: card4
american express      0.028698
discover              0.077282
mastercard            0.034331
visa                  0.034756
Name: isFraud, dtype: float64
```

- 97% of transactions are from Mastercard(32%) and Visa(65%)
- Fraud transaction rate is highest for discover cards(~8%) against ~3.5% of Mastercard and Visa and 2.87% in American Express

```
[38]: # Card 6
plot_cat_feat_dist(df, "card6")
```

card6 Distributions



```
[39]: df.groupby('card6')['isFraud'].mean()
```

```
[39]: card6
charge card      0.000000
credit          0.066785
debit           0.024263
debit or credit 0.000000
Name: isFraud, dtype: float64
```

- Almost all the transactions are from Credit and Debit cards.
- Debit card transactions are almost 3 times as compared to credit card transactions.
- Fraud transaction rate is high for Credit cards as compared to Debit cards.

6.0.5 P_emaildomain

```
[40]: df.loc[df['P_emaildomain'].isin(['gmail.com', 'gmail']), 'P_emaildomain'] = ↴'Google'

df.loc[df['P_emaildomain'].isin(['yahoo.com', 'yahoo.com.mx', 'yahoo.co.uk',
                                 'yahoo.co.jp', 'yahoo.de', 'yahoo.fr',
                                 'yahoo.es']), 'P_emaildomain'] = ↴'Yahoo Mail'

df.loc[df['P_emaildomain'].isin(['hotmail.com', 'outlook.com', 'msn.com', 'live.
                                 com.mx',
                                 'hotmail.es', 'hotmail.co.uk', 'hotmail.
                                 de',
                                 'outlook.es', 'live.com', 'live.fr',
                                 'live.de']), 'P_emaildomain'] = ↴'Hotmail / Outlook / MSN / Live
```

```

        'hotmail.fr']), 'P_emaildomain'] = □
↳ 'Microsoft'
df.loc[df.P_emaildomain.isin(df.P_emaildomain\
                               .value_counts()[df.P_emaildomain.
                               .value_counts() <= 500 ]\

                               .index), 'P_emaildomain'] = "Others"
df.P_emaildomain.fillna("NoInf", inplace=True)

```

```

[41]: def plot_cat_with_amt(df, col, lim=2000):
    tmp = pd.crosstab(df[col], df['isFraud'], normalize='index') * 100
    tmp = tmp.reset_index()
    tmp.rename(columns={0:'NoFraud', 1:'Fraud'}, inplace=True)

    plt.figure(figsize=(16,14))
    plt.suptitle(f'{col} Distributions ', fontsize=24)

    plt.subplot(211)
    g = sns.countplot( x=col, data=df, order=list(tmp[col].values))
    gt = g.twinx()
    gt = sns.pointplot(x=col, y='Fraud', data=tmp, order=list(tmp[col].values),
                        color='black', legend=False, )
    gt.set_ylim(0,tmp['Fraud'].max()*1.1)
    gt.set_ylabel("%Fraud Transactions", fontsize=16)
    g.set_title(f"Share of {col} categories and % of Fraud Transactions",□
    fontsize=18)
    g.set_xlabel(f"{col} Category Names", fontsize=16)
    g.set_ylabel("Count", fontsize=17)
    g.set_xticklabels(g.get_xticklabels(), rotation=45)
    sizes = []
    for p in g.patches:
        height = p.get_height()
        sizes.append(height)
        g.text(p.get_x()+p.get_width()/2.,
               height + 3,
               '{:1.2f}%'.format(height/total*100),
               ha="center", fontsize=12)

    g.set_ylim(0,max(sizes)*1.15)

#####
perc_amt = (df.groupby(['isFraud',col])['TransactionAmt'].sum() \
            / df.groupby([col])['TransactionAmt'].sum() * 100). \
            unstack('isFraud')
perc_amt = perc_amt.reset_index()
perc_amt.rename(columns={0:'NoFraud', 1:'Fraud'}, inplace=True)
amt = df.groupby([col])['TransactionAmt'].sum().reset_index()
perc_amt = perc_amt.fillna(0)

```

```

plt.subplot(212)
g1 = sns.barplot(x=col, y='TransactionAmt',
                  data=amt,
                  order=list(tmp[col].values))
g1t = g1.twinx()
g1t = sns.pointplot(x=col, y='Fraud', data=perc_amt,
                     order=list(tmp[col].values),
                     color='black', legend=False, )
g1t.set_ylim(0,perc_amt['Fraud'].max()*1.1)
g1t.set_ylabel("%Fraud Total Amount", fontsize=16)
g.set_xticklabels(g.get_xticklabels(), rotation=45)
g1.set_title(f"Transactions amount by {col} categories and % of Fraud")
g1.set_xlabel(f"{col} Category Names", fontsize=16)
g1.set_ylabel("Transaction Total Amount(U$)", fontsize=16)
g1.set_xticklabels(g.get_xticklabels(), rotation=45)

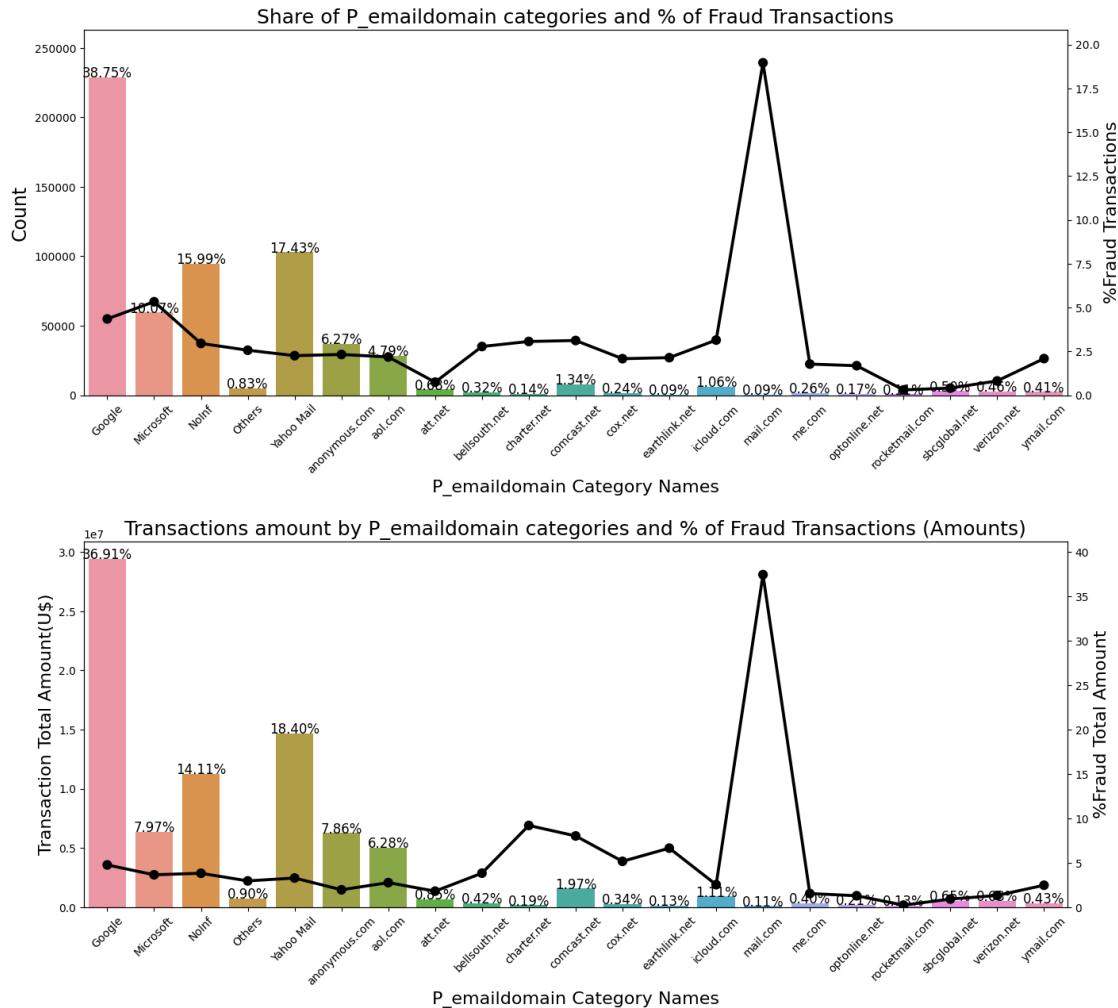
for p in g1.patches:
    height = p.get_height()
    g1.text(p.get_x()+p.get_width()/2.,
            height + 3,
            '{:1.2f}%'.format(height/total_amt*100),
            ha="center", fontsize=12)

plt.subplots_adjust(hspace=.4, top = 0.9)
plt.show()

```

[42]: plot_cat_with_amt(df, 'P_emaildomain')

P_emaildomain Distributions



```
[43]: df.groupby('P_emaildomain')['isFraud'].mean()
```

```
[43]: P_emaildomain
```

Google	0.043496
Microsoft	0.053298
NoInf	0.029538
Others	0.025646
Yahoo Mail	0.022544
anonymous.com	0.023217
aol.com	0.021811
att.net	0.007439
bellsouth.net	0.027763
charter.net	0.030637
comcast.net	0.031187

```

cox.net          0.020818
earthlink.net   0.021401
icloud.com      0.031434
mail.com         0.189624
me.com           0.017740
optonline.net    0.016815
rocketmail.com   0.003012
sbcglobal.net    0.004040
verizon.net      0.008133
ymail.com        0.020868
Name: isFraud, dtype: float64

```

- Majority of transactions are with P_emaildomain as Google, Microsoft and Yahoo Mail
- There isn't any information about P_emaildomain of around 16% transactions in terms of count and 14.11% in terms of amount
- Fraud transaction rate for Microsoft is high as compared to Google and Yahoo mail
- Fraud transaction rate (amount) for Google is high as compared to Microsoft and Yahoo mail

6.0.6 R-Email Domain

```
[44]: df.loc[df['R_emaildomain'].isin(['gmail.com', 'gmail']), 'R_emaildomain'] = "Google"

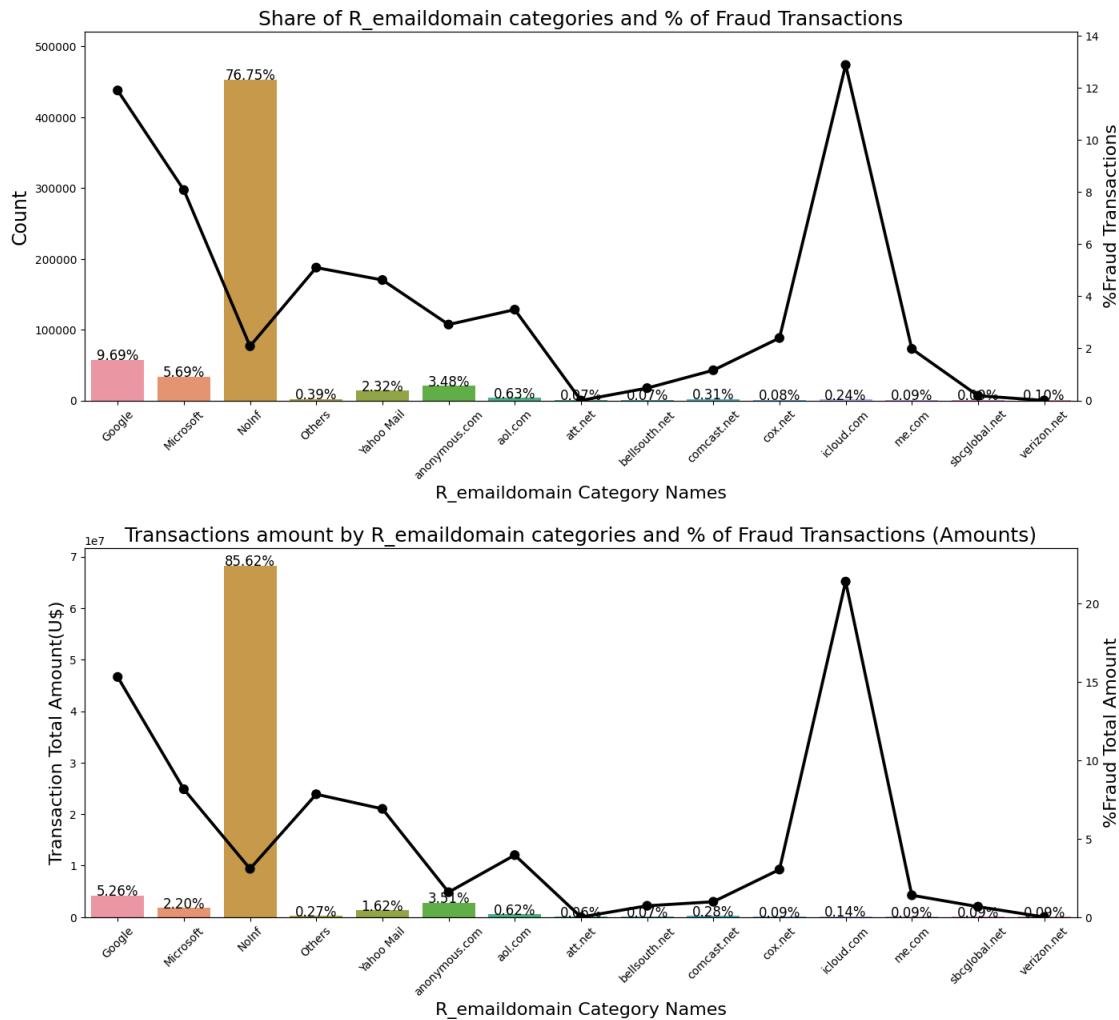
df.loc[df['R_emaildomain'].isin(['yahoo.com', 'yahoo.com.mx', 'yahoo.co.uk',
                                 'yahoo.co.jp', 'yahoo.de', 'yahoo.
                                 ↪fr',
                                 ↪'Yahoo Mail']),
       'R_emaildomain'] = "Yahoo Mail"

df.loc[df['R_emaildomain'].isin(['hotmail.com', 'outlook.com', 'msn.com', 'live.
                                 ↪com.mx',
                                 'hotmail.es', 'hotmail.co.uk',
                                 ↪'hotmail.de',
                                 'outlook.es', 'live.com', 'live.
                                 ↪fr',
                                 ↪'Microsoft']),
       'R_emaildomain'] = "Microsoft"

df.loc[df.R_emaildomain.isin(df.R_emaildomain\
                             .value_counts()[df.R_emaildomain.
                             ↪value_counts() <= 300 ])\ \
                             .index), 'R_emaildomain'] = "Others"
df.R_emaildomain.fillna("NoInf", inplace=True)

[45]: plot_cat_with_amt(df, 'R_emaildomain')
```

R_emaildomain Distributions



```
[46]: df.groupby('R_emaildomain')['isFraud'].mean()
```

```
[46]: R_emaildomain
Google          0.118986
Microsoft       0.080764
NoInf           0.020819
Others          0.050989
Yahoo Mail      0.046235
anonymous.com   0.029130
aol.com          0.034855
att.net          0.000000
bellsouth.net    0.004739
comcast.net     0.011589
cox.net          0.023965
```

```

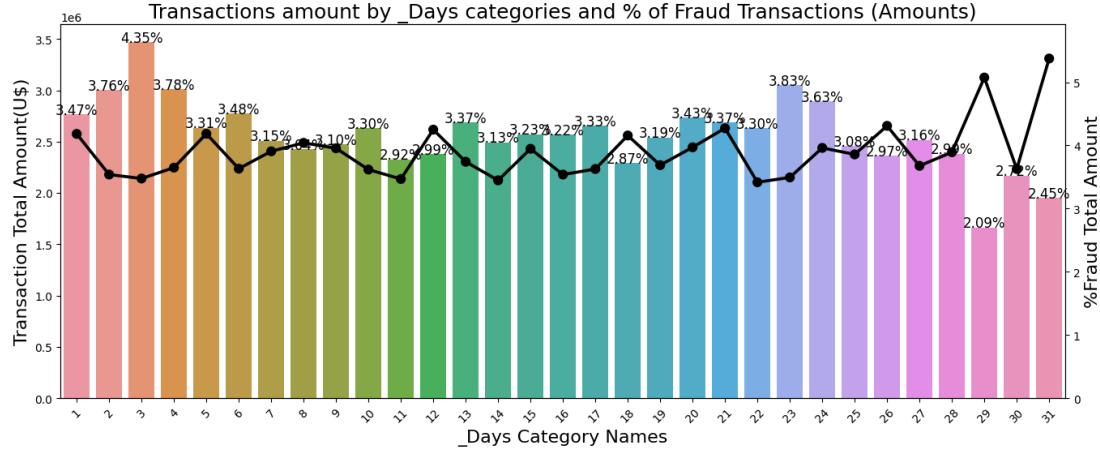
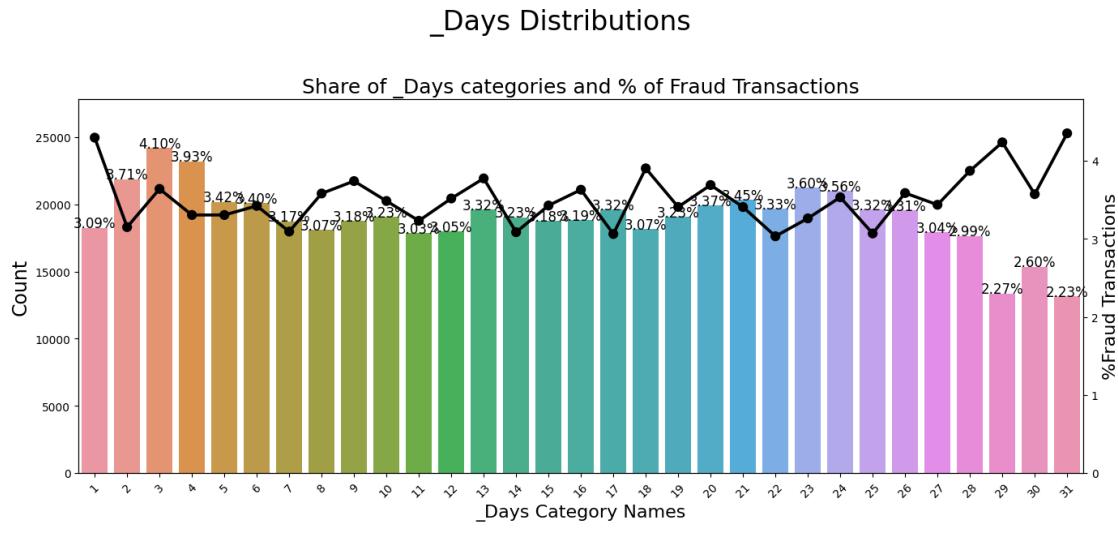
icloud.com      0.128755
me.com         0.019784
sbcglobal.net   0.001812
verizon.net     0.000000
Name: isFraud, dtype: float64

```

- There isn't any information about R_emaildomain for Majority of transactions (76.75% count, 85.62% amount)
- Fraud transaction rate for Google is high as compared to Yahoo, anaonymous.com and Microsoft

6.0.7 Days of the Month

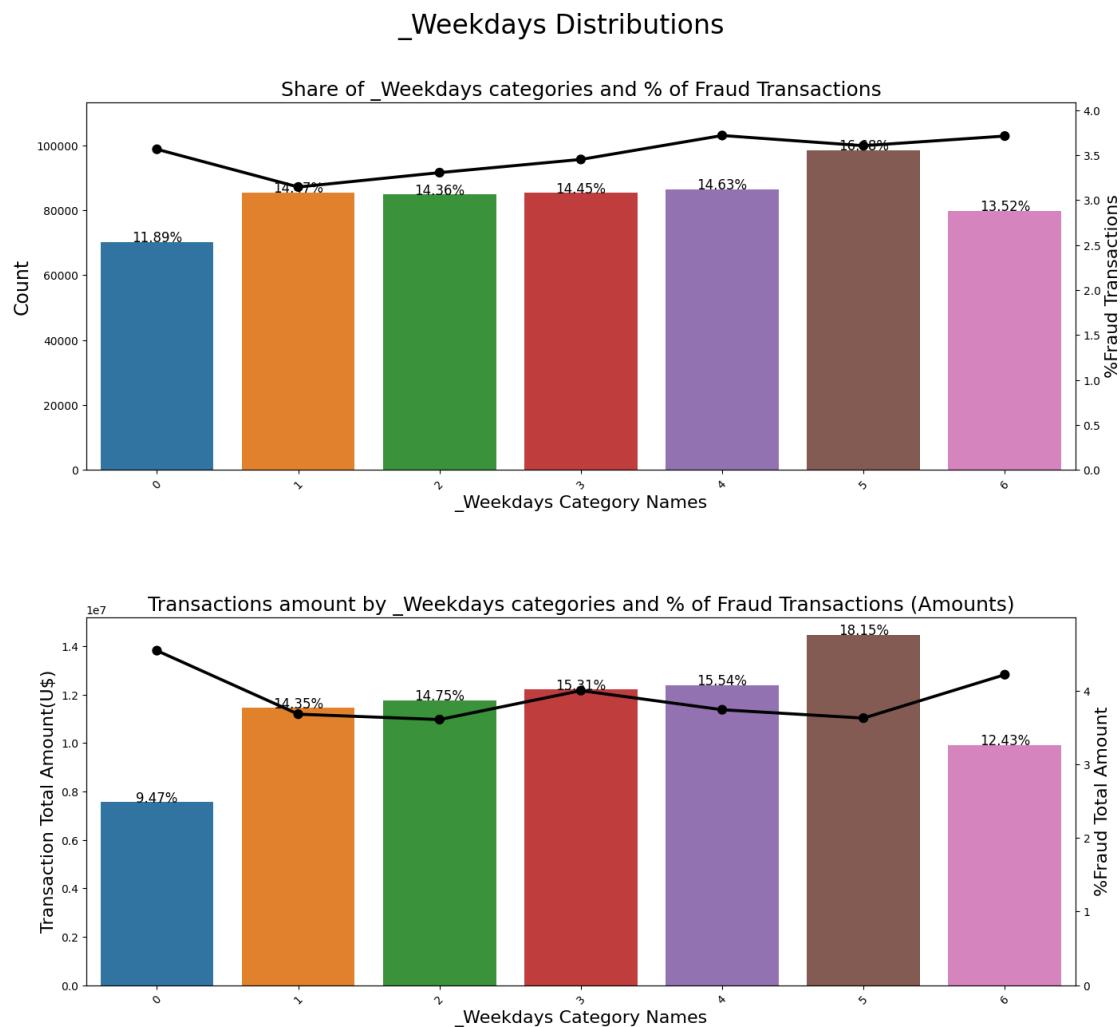
```
[47]: plot_cat_with_amt(df, '_Days')
```



- The perc of fraud transactions is highest towards the beginning and the end of the month. Might be accelerated at the time of receiving pay-checks.
- Incidentally, fraud transaction rate is high on the days when number of transactions are less
- Day 29,30 and 31 are having less transactions, looks like people are cautious with spending in those times.

6.0.8 Days of the week

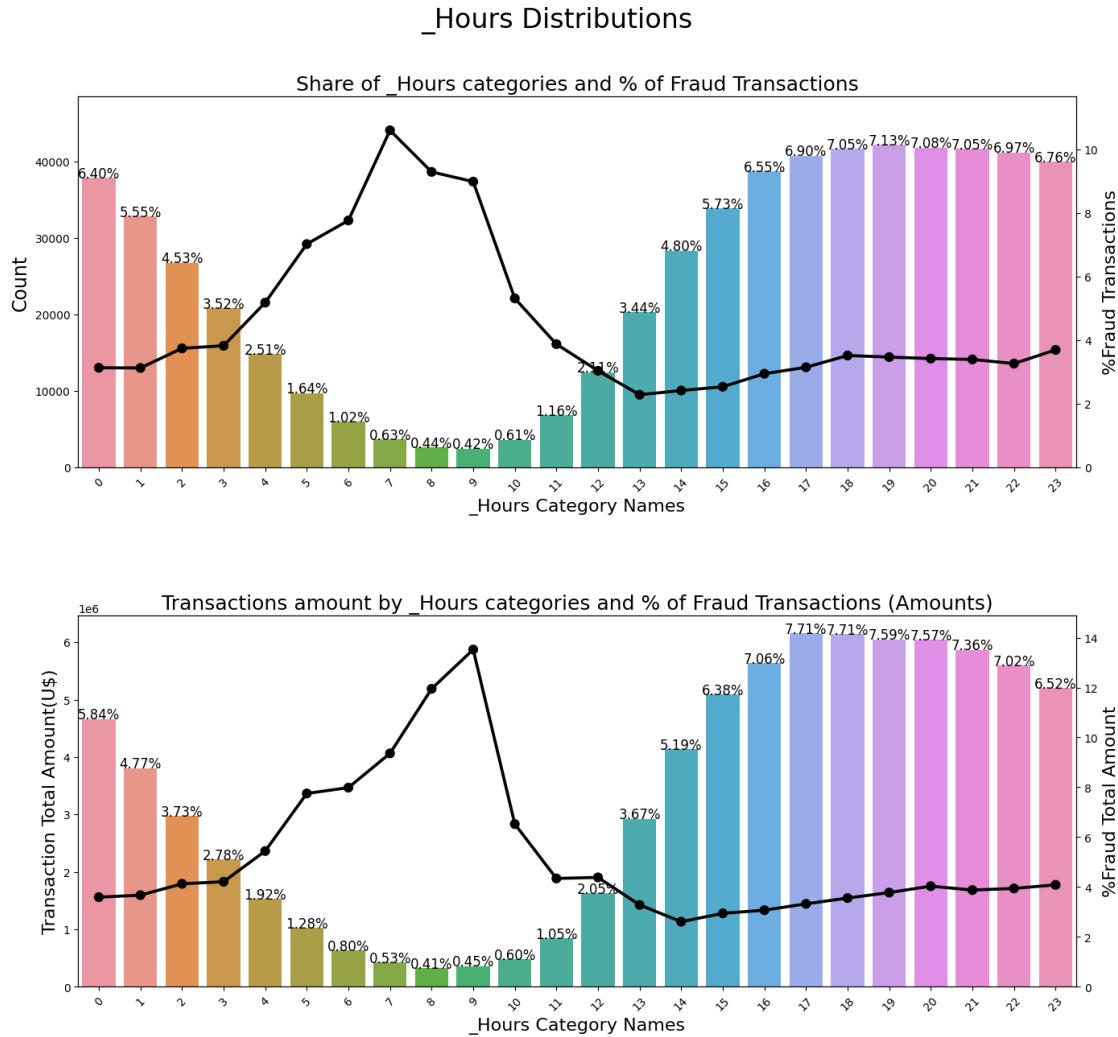
[48]: `plot_cat_with_amt(df, '_Weekdays')`



- Surprisingly fraud transaction rate is high on the days when number of transactions and transaction amounts are less. Day 0 and 6
- Day 0 and 6 have less transactions, these might be weekend days

6.0.9 Hour of the Day

```
[49]: plot_cat_with_amt(df, '_Hours')
```

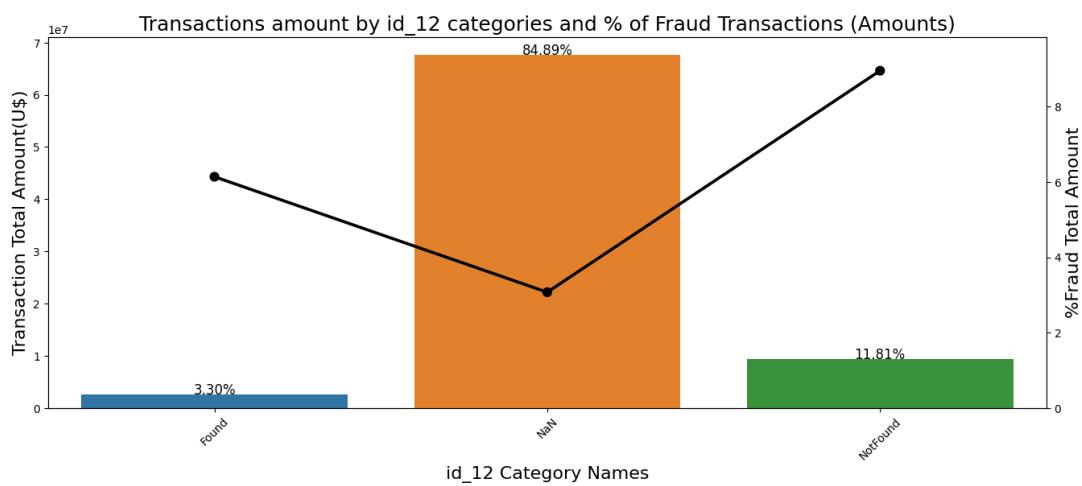
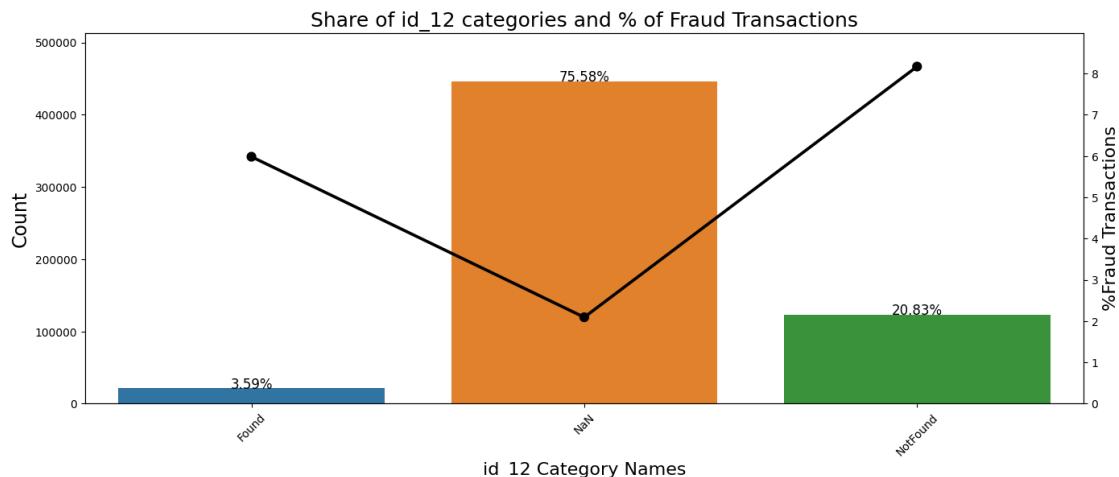


- Transactions start decreasing mid night but the fraud rate starts increasing
- Transactions from 3 AM to 12 PM needs to monitored very closely

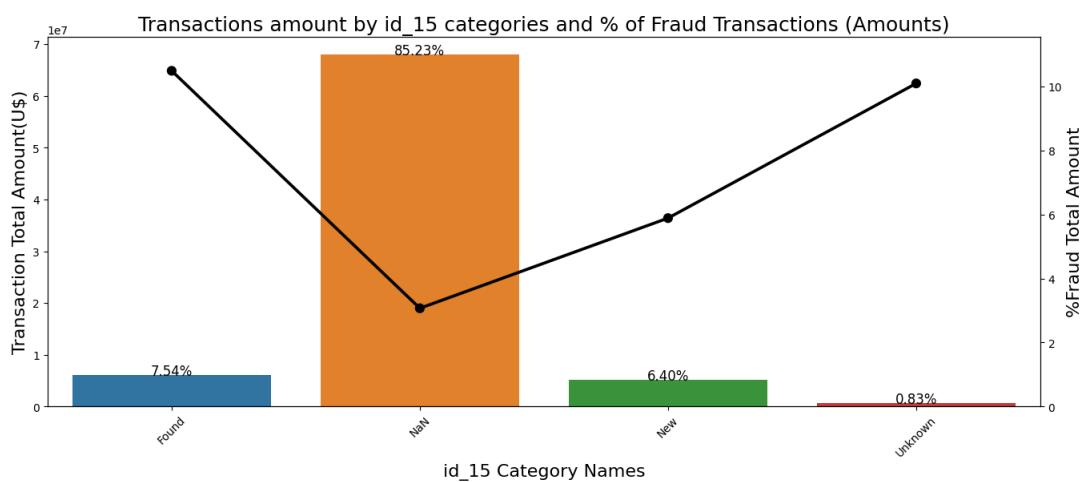
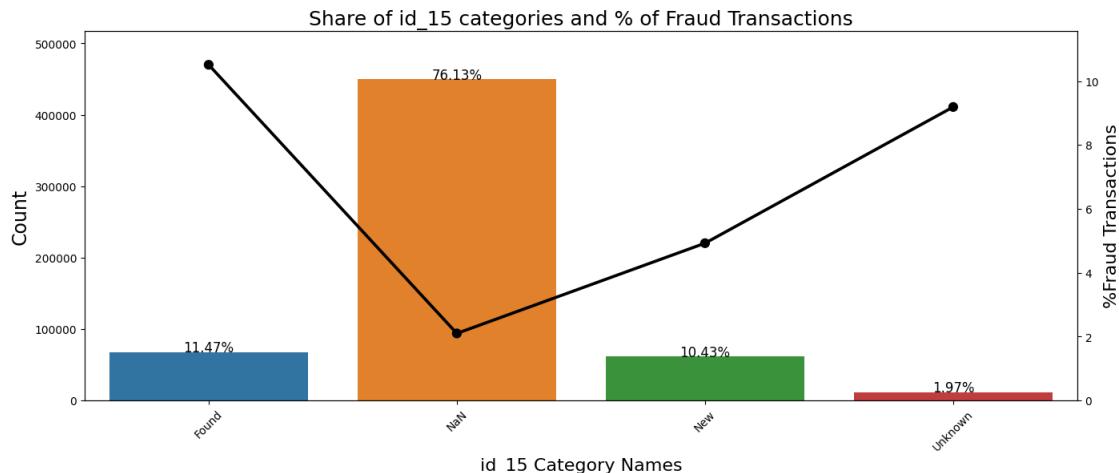
6.0.10 Columns from identity data

```
[50]: for col in ['id_12', 'id_15', 'id_16', 'id_28', 'id_29']:
    df[col] = df[col].fillna('NaN')
    plot_cat_with_amt(df, col)
```

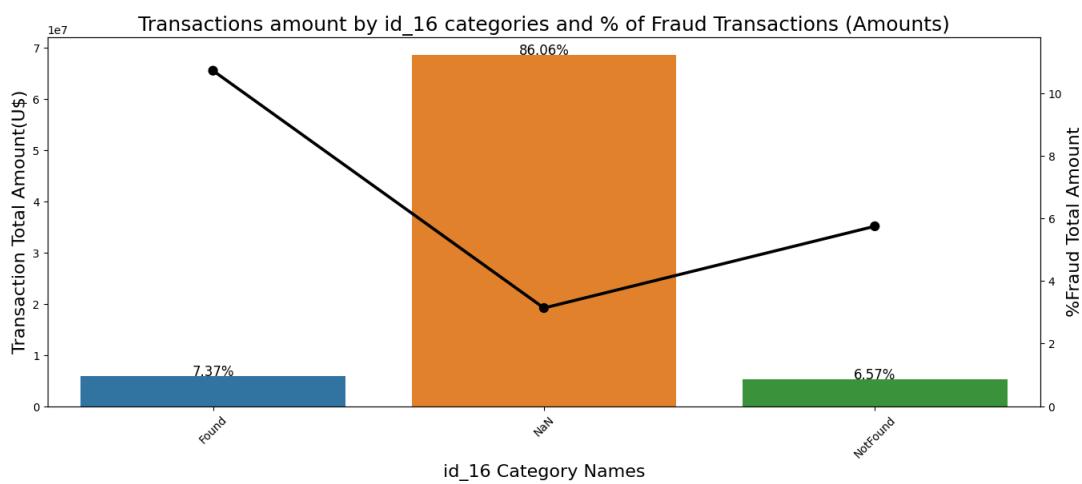
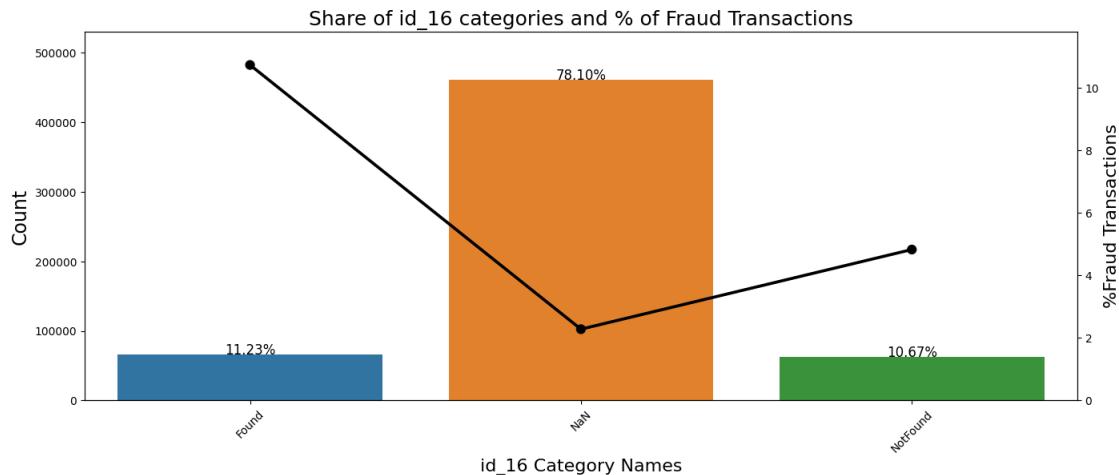
id_12 Distributions



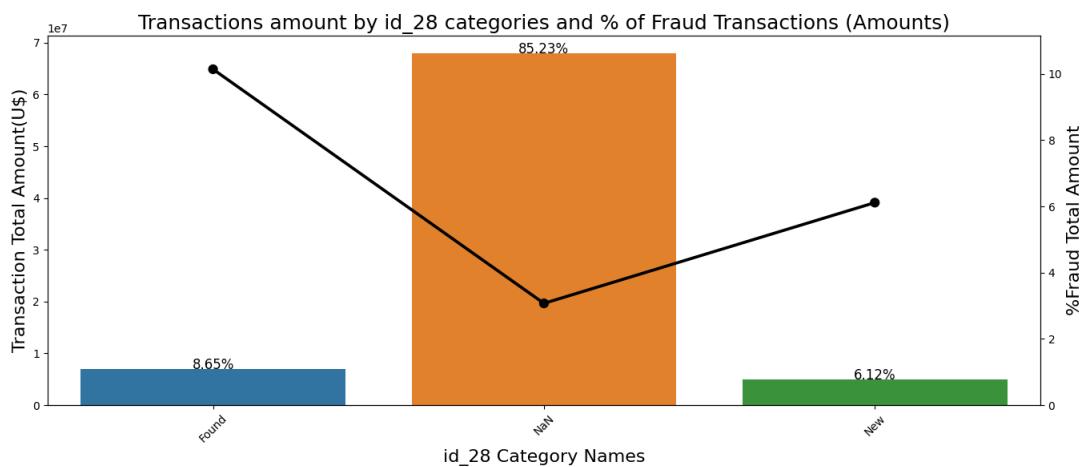
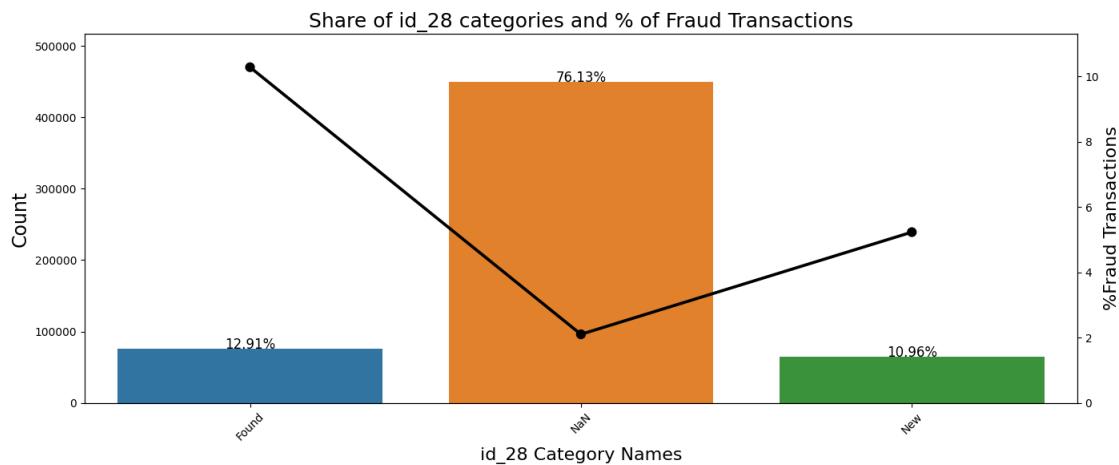
id_15 Distributions



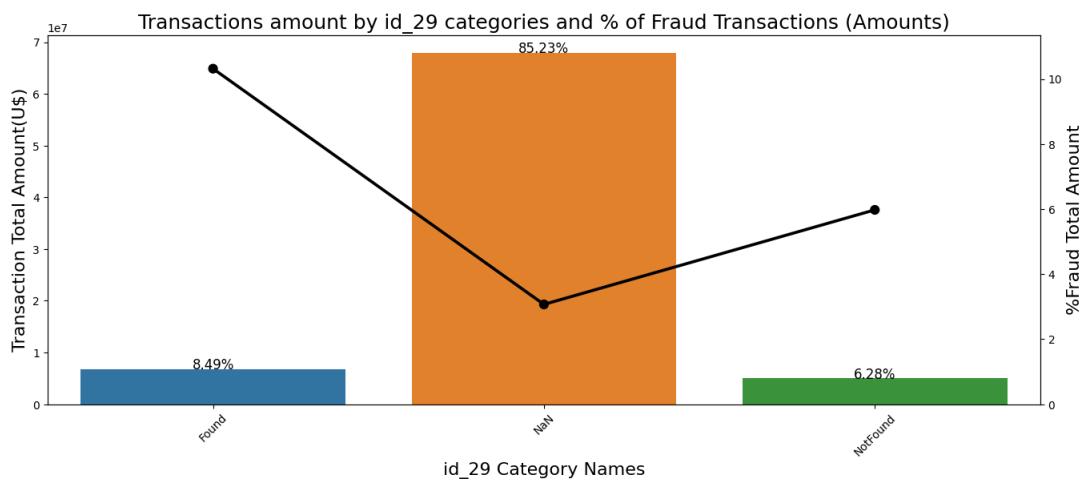
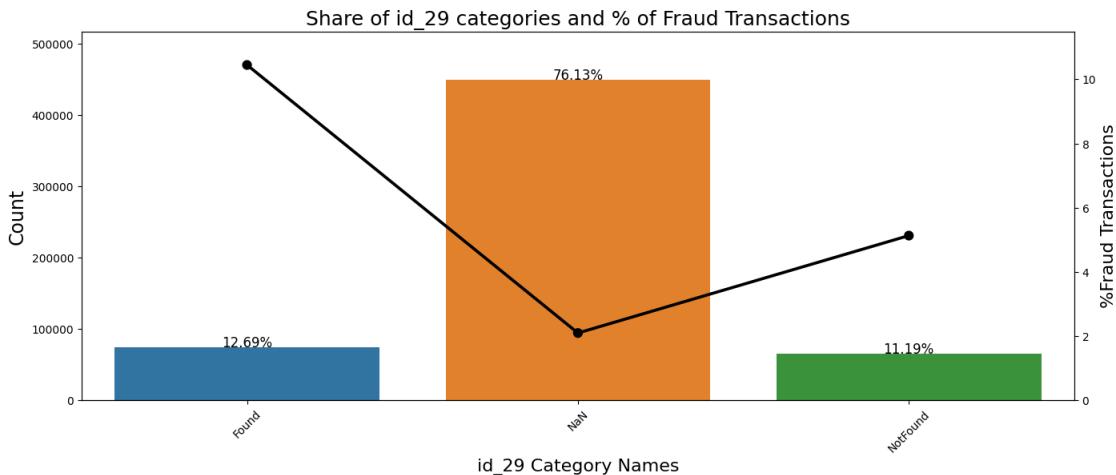
id_16 Distributions



id_28 Distributions



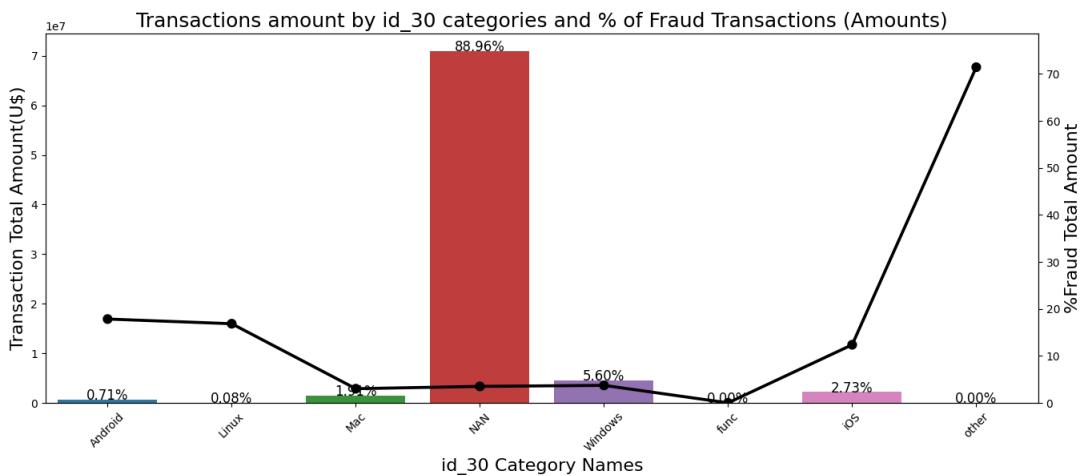
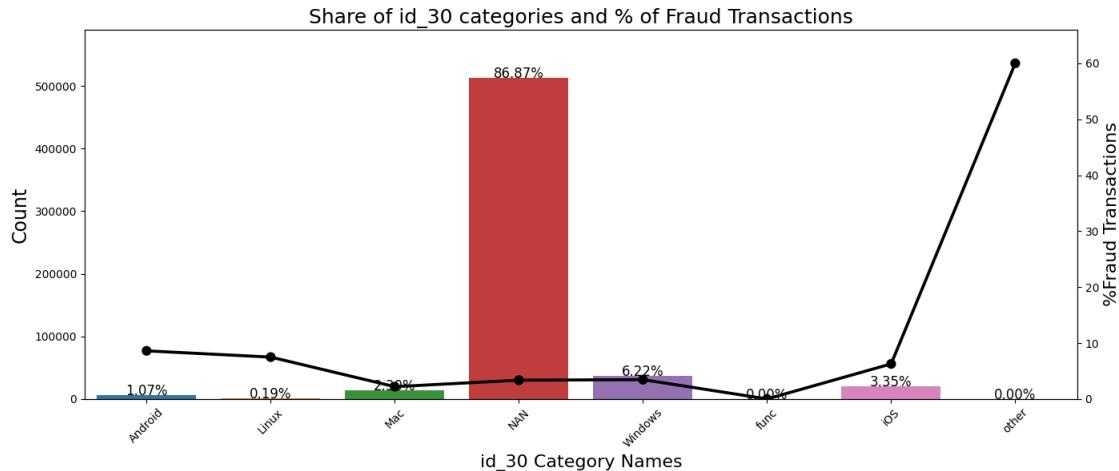
id_29 Distributions



```
[51]: df.loc[df['id_30'].str.contains('Windows', na=False), 'id_30'] = 'Windows'
df.loc[df['id_30'].str.contains('iOS', na=False), 'id_30'] = 'iOS'
df.loc[df['id_30'].str.contains('Mac OS', na=False), 'id_30'] = 'Mac'
df.loc[df['id_30'].str.contains('Android', na=False), 'id_30'] = 'Android'
df['id_30'].fillna("NAN", inplace=True)

plot_cat_with_amt(df, "id_30")
```

id_30 Distributions



```
[52]: cat_columns = df.select_dtypes(include=['object']).columns
len(cat_columns)
```

[52]: 29

```
[53]: binary_columns = [col for col in df.columns if df[col].nunique() == 2]
len(binary_columns)
```

[53]: 435

```
[54]: num_columns = [col for col in df.columns if (col not in cat_columns) & (col not in binary_columns)]
len(num_columns)
```

[54]: 389

[55]: cat_columns = cat_columns.to_list() + binary_columns

7 7. Feature Engineering

[56]: df.head()

```
TransactionID  isFraud  TransactionDT  TransactionAmt  ProductCD  card1  \
0            2987000      0            86400          68.5        W  13926
1            2987001      0            86401          29.0        W   2755
2            2987002      0            86469          59.0        W   4663
3            2987003      0            86499          50.0        W  18132
4            2987004      0            86506          50.0        H   4497

card2  card3  card4  card5  card6  addr1  addr2  dist1  P_emaildomain  \
0    NaN  150.0  discover  142.0  credit  315.0   87.0   19.0        NoInf
1  404.0  150.0  mastercard  102.0  credit  325.0   87.0     NaN        Google
2  490.0  150.0       visa  166.0  debit   330.0   87.0  287.0        Microsoft
3  567.0  150.0  mastercard  117.0  debit   476.0   87.0     NaN        Yahoo Mail
4  514.0  150.0  mastercard  102.0  credit  420.0   87.0     NaN        Google

R_emaildomain  C1  C2  C3  C4  C5  C6  C7  C8  C9  C10  C11  C12  \
0      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  2.0  0.0
1      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  0.0  0.0  1.0  0.0
2      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  1.0  0.0
3      NoInf  2.0  5.0  0.0  0.0  0.0  4.0  0.0  0.0  1.0  0.0  1.0  0.0
4      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  1.0  0.0  1.0  1.0  0.0

C13  C14  D1  D2  D3  D4  D5  D6  D8  D9  D10  D11  D12  \
0    1.0  1.0  14.0  NaN  13.0  NaN  NaN  NaN  NaN  13.0  13.0  NaN
1    1.0  1.0   0.0  NaN  NaN  0.0  NaN  NaN  NaN  0.0  NaN  NaN
2    1.0  1.0   0.0  NaN  NaN  0.0  NaN  NaN  NaN  0.0  315.0  NaN
3   25.0  1.0  112.0  112.0  0.0  94.0  0.0  NaN  NaN  84.0  NaN  NaN
4    1.0  1.0   0.0  NaN  NaN  NaN  NaN  NaN  NaN  NaN  NaN  NaN

D13  D14  D15  M1  M2  M3  M4  M5  M6  M7  M8  M9  V1  V2  \
0  NaN  NaN  0.0  T  T  T  M2  F  T  NaN  NaN  NaN  1.0  1.0
1  NaN  NaN  0.0  NaN  NaN  NaN  MO  T  T  NaN  NaN  NaN  NaN  NaN
2  NaN  NaN  315.0  T  T  T  MO  F  F  F  F  F  1.0  1.0
3  NaN  NaN  111.0  NaN  NaN  NaN  MO  T  F  NaN  NaN  NaN  NaN  NaN
4  NaN  NaN

V3  V4  V5  V6  V7  V8  V9  V10  V11  V12  V13  V14  V15  V16  V17  \
0  1.0  1.0  1.0  1.0  1.0  1.0  1.0  0.0  0.0  1.0  1.0  1.0  0.0  0.0
1  NaN  NaN  NaN  NaN  NaN  NaN  NaN  NaN  NaN  0.0  0.0  1.0  0.0  0.0
```

2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
3	NaN	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0									
4	NaN	NaN	NaN	NaN	NaN													
	V18	V19	V20	V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	\		
0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN													
	V33	V34	V35	V36	V37	V38	V39	V40	V41	V42	V43	V44	V45	V46	V47	\		
0	0.0	0.0	NaN	NaN	NaN	NaN	NaN											
1	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	
2	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	
3	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN													
	V48	V49	V50	V51	V52	V53	V54	V55	V56	V57	V58	V59	V60	V61	V62	\		
0	NaN	NaN	NaN	NaN	NaN	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
2	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
3	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN													
	V63	V64	V65	V66	V67	V68	V69	V70	V71	V72	V73	V74	V75	V76	V77	\		
0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
1	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	
2	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
3	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN													
	V78	V79	V80	V81	V82	V83	V84	V85	V86	V87	V88	V89	V90	V91	V92	\		
0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN													
	V93	V94	V95	V96	V97	V98	V99	V100	V101	V102	V103	V104	V105	\				
0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	0.0	1.0	48.0	28.0	0.0	10.0	4.0	1.0	38.0	24.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V106	V107	V108	V109	V110	V111	V112	V113	V114	V115	V116	V117	\					

0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
1	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
2	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
3	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
4	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
	V118	V119	V120	V121	V122	V123	V124	V125	V126	V127	V128	V129	\
0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	117.0	0.0	0.0	
1	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
3	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	50.0	1758.0	925.0	0.0	
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	
	V130	V131	V132	V133	V134	V135	V136	V137	V138	V139	V140	V141	\
0	0.0	0.0	0.0	117.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN		
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN		
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN		
3	354.0	135.0	50.0	1404.0	790.0	0.0	0.0	0.0	NaN	NaN	NaN		
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	V141	V142	V143	V144	V145	V146	V147	V148	V149	V150	V151	V152	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	0.0	6.0	18.0	140.0	0.0	0.0	0.0	0.0	1803.0	49.0	64.0	
	V153	V154	V155	V156	V157	V158	V159		V160	V161	V162	V163	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN		
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN		
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN		
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN	NaN		
4	0.0	0.0	0.0	0.0	0.0	0.0	15560.0	169690.796875	0.0	0.0			
	V163	V164	V165	V166	V167	V168	V169	V170	V171	V172	V173	V174	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
4	0.0	515.0	5155.0	2840.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	
	V174	V175	V176	V177	V178	V179	V180	V181	V182	V183	V184	V185	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	

	V186	V187	V188	V189	V190	V191	V192	V193	V194	V195	V196	V197	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	
	...	V134_missing_flag	V135_missing_flag	V136_missing_flag									\
0	...	False	False	False									
1	...	False	False	False									
2	...	False	False	False									
3	...	False	False	False									
4	...	False	False	False									
	V137_missing_flag	V138_missing_flag	V139_missing_flag	V140_missing_flag									\
0	False	True	True	True									
1	False	True	True	True									
2	False	True	True	True									
3	False	True	True	True									
4	False	False	False	False									
	V141_missing_flag	V142_missing_flag	V143_missing_flag	V144_missing_flag									\
0	True	True	True	True									
1	True	True	True	True									
2	True	True	True	True									
3	True	True	True	True									
4	False	False	False	False									
	V145_missing_flag	V146_missing_flag	V147_missing_flag	V148_missing_flag									\
0	True	True	True	True									
1	True	True	True	True									
2	True	True	True	True									
3	True	True	True	True									
4	False	False	False	False									
	V149_missing_flag	V150_missing_flag	V151_missing_flag	V152_missing_flag									\
0	True	True	True	True									
1	True	True	True	True									
2	True	True	True	True									
3	True	True	True	True									
4	False	False	False	False									
	V153_missing_flag	V154_missing_flag	V155_missing_flag	V156_missing_flag									\
0	True	True	True	True									
1	True	True	True	True									
2	True	True	True	True									

3	True	True	True	True	True
4	False	False	False	False	False
	V157_missing_flag	V158_missing_flag	V159_missing_flag	V160_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V161_missing_flag	V162_missing_flag	V163_missing_flag	V164_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V165_missing_flag	V166_missing_flag	V167_missing_flag	V168_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V169_missing_flag	V170_missing_flag	V171_missing_flag	V172_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V173_missing_flag	V174_missing_flag	V175_missing_flag	V176_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V177_missing_flag	V178_missing_flag	V179_missing_flag	V180_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V181_missing_flag	V182_missing_flag	V183_missing_flag	V184_missing_flag	\
0	True	True	True	True	True

1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V185_missing_flag	V186_missing_flag	V187_missing_flag	V188_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V189_missing_flag	V190_missing_flag	V191_missing_flag	V192_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V193_missing_flag	V194_missing_flag	V195_missing_flag	V196_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V197_missing_flag	V198_missing_flag	V199_missing_flag	V200_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V201_missing_flag	V202_missing_flag	V203_missing_flag	V204_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V205_missing_flag	V206_missing_flag	V207_missing_flag	V208_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V209_missing_flag	V210_missing_flag	V211_missing_flag	V212_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V213_missing_flag	V214_missing_flag	V215_missing_flag	V216_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V217_missing_flag	V218_missing_flag	V219_missing_flag	V220_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V221_missing_flag	V222_missing_flag	V223_missing_flag	V224_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V225_missing_flag	V226_missing_flag	V227_missing_flag	V228_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V229_missing_flag	V230_missing_flag	V231_missing_flag	V232_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V233_missing_flag	V234_missing_flag	V235_missing_flag	V236_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True

4	False	False	False	False
0	V237_missing_flag True	V238_missing_flag True	V239_missing_flag True	V240_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V241_missing_flag True	V242_missing_flag True	V243_missing_flag True	V244_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V245_missing_flag True	V246_missing_flag True	V247_missing_flag True	V248_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V249_missing_flag True	V250_missing_flag True	V251_missing_flag True	V252_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V253_missing_flag True	V254_missing_flag True	V255_missing_flag True	V256_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V257_missing_flag True	V258_missing_flag True	V259_missing_flag True	V260_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V261_missing_flag True	V262_missing_flag True	V263_missing_flag True	V264_missing_flag True
1	True	True	True	True

2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V265_missing_flag	V266_missing_flag	V267_missing_flag	V268_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V269_missing_flag	V270_missing_flag	V271_missing_flag	V272_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V273_missing_flag	V274_missing_flag	V275_missing_flag	V276_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V277_missing_flag	V278_missing_flag	V279_missing_flag	V280_missing_flag	\
0	True	True	False	False	
1	True	True	False	False	
2	True	True	False	False	
3	True	True	False	False	
4	False	False	False	False	
	V281_missing_flag	V282_missing_flag	V283_missing_flag	V284_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V285_missing_flag	V286_missing_flag	V287_missing_flag	V288_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V289_missing_flag	V290_missing_flag	V291_missing_flag	V292_missing_flag	\

0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V293_missing_flag	V294_missing_flag	V295_missing_flag	V296_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V297_missing_flag	V298_missing_flag	V299_missing_flag	V300_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V301_missing_flag	V302_missing_flag	V303_missing_flag	V304_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V305_missing_flag	V306_missing_flag	V307_missing_flag	V308_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V309_missing_flag	V310_missing_flag	V311_missing_flag	V312_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V313_missing_flag	V314_missing_flag	V315_missing_flag	V316_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False

	V317_missing_flag	V318_missing_flag	V319_missing_flag	V320_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False

	V321_missing_flag	V322_missing_flag	V323_missing_flag	V324_missing_flag	\
0	False	True	True	True	True
1	False	True	True	True	True
2	False	True	True	True	True
3	False	True	True	True	True
4	False	False	False	False	False

	V325_missing_flag	V326_missing_flag	V327_missing_flag	V328_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V329_missing_flag	V330_missing_flag	V331_missing_flag	V332_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V333_missing_flag	V334_missing_flag	V335_missing_flag	V336_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V337_missing_flag	V338_missing_flag	V339_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	

	id_01_missing_flag	id_02_missing_flag	id_03_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	

3	True	True	True	
4	False	False	True	
0	id_04_missing_flag	id_05_missing_flag	id_06_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
0	id_07_missing_flag	id_08_missing_flag	id_09_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
0	id_10_missing_flag	id_11_missing_flag	id_12_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
0	id_13_missing_flag	id_14_missing_flag	id_15_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
0	id_16_missing_flag	id_17_missing_flag	id_18_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
0	id_19_missing_flag	id_20_missing_flag	id_21_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
0	id_22_missing_flag	id_23_missing_flag	id_24_missing_flag	\
	True	True	True	

	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	True	True	True		
	id_25_missing_flag	id_26_missing_flag	id_27_missing_flag	\	
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	True	True	True	True	
	id_28_missing_flag	id_29_missing_flag	id_30_missing_flag	\	
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	id_31_missing_flag	id_32_missing_flag	id_33_missing_flag	\	
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	id_34_missing_flag	id_35_missing_flag	id_36_missing_flag	\	
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	id_37_missing_flag	id_38_missing_flag	DeviceType_missing_flag	\	
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	DeviceInfo_missing_flag	Date	_Weekdays	_Hours	_Days
0	True	2017-12-02 00:00:00	5	0	2
1	True	2017-12-02 00:00:01	5	0	2
2	True	2017-12-02 00:01:09	5	0	2
3	True	2017-12-02 00:01:39	5	0	2
4	False	2017-12-02 00:01:46	5	0	2

[5 rows x 840 columns]

7.0.1 Domain Specific Features

```
[57]: df['Trans_min_mean'] = df['TransactionAmt'] - np.  
      ↪nanmean(df['TransactionAmt'], dtype="float64")  
df['Trans_min_std'] = df['Trans_min_mean'] / np.nanstd(df['TransactionAmt']).  
      ↪astype("float64"), dtype="float64")
```

```
[58]: df['TransactionAmt_to_mean_card1'] = df['TransactionAmt'] / df.  
      ↪groupby(['card1'])['TransactionAmt'].transform('mean')  
df['TransactionAmt_to_mean_card4'] = df['TransactionAmt'] / df.  
      ↪groupby(['card4'])['TransactionAmt'].transform('mean')  
df['TransactionAmt_to_std_card1'] = df['TransactionAmt'] / df.  
      ↪groupby(['card1'])['TransactionAmt'].transform('std')  
df['TransactionAmt_to_std_card4'] = df['TransactionAmt'] / df.  
      ↪groupby(['card4'])['TransactionAmt'].transform('std')
```

```
[59]: df['TransactionAmt'] = np.log(df['TransactionAmt'])
```

```
[60]: df.head()
```

```
[60]:   TransactionID  isFraud  TransactionDT  TransactionAmt  ProductCD  card1  \  
0        2987000       0          86400        4.226834      W  13926  
1        2987001       0          86401        3.367296      W  2755  
2        2987002       0          86469        4.077537      W  4663  
3        2987003       0          86499        3.912023      W  18132  
4        2987004       0          86506        3.912023      H  4497  
  
    card2  card3      card4  card5  card6  addr1  addr2  dist1 P_emaildomain  \  
0     NaN  150.0  discover  142.0  credit  315.0  87.0  19.0      NoInf  
1  404.0  150.0  mastercard  102.0  credit  325.0  87.0    NaN      Google  
2  490.0  150.0       visa  166.0  debit   330.0  87.0  287.0  Microsoft  
3  567.0  150.0  mastercard  117.0  debit   476.0  87.0    NaN  Yahoo Mail  
4  514.0  150.0  mastercard  102.0  credit  420.0  87.0    NaN      Google  
  
R_emaildomain  C1  C2  C3  C4  C5  C6  C7  C8  C9  C10  C11  C12  \  
0      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  2.0  0.0  
1      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  0.0  0.0  1.0  0.0  
2      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  1.0  0.0  
3      NoInf  2.0  5.0  0.0  0.0  0.0  4.0  0.0  0.0  1.0  0.0  1.0  0.0  
4      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  1.0  0.0  1.0  1.0  0.0  
  
C13  C14      D1      D2      D3      D4      D5      D6      D8      D9      D10      D11      D12  \  
0    1.0  1.0  14.0    NaN  13.0    NaN    NaN    NaN    NaN  13.0  13.0    NaN  
1    1.0  1.0    0.0    NaN    NaN    0.0    NaN    NaN    NaN    NaN    0.0    NaN    NaN
```

2	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	NaN	0.0	315.0	NaN			
3	25.0	1.0	112.0	112.0	0.0	94.0	0.0	NaN	NaN	NaN	84.0	NaN	NaN			
4	1.0	1.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN			
	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	M9	V1	V2	\	
0	NaN	NaN	0.0	T	T	T	M2	F	T	NaN	NaN	NaN	1.0	1.0		
1	NaN	NaN	0.0	NaN	NaN	NaN	M0	T	T	NaN	NaN	NaN	NaN	NaN		
2	NaN	NaN	315.0	T	T	T	M0	F	F	F	F	F	1.0	1.0		
3	NaN	NaN	111.0	NaN	NaN	NaN	M0	T	F	NaN	NaN	NaN	NaN	NaN		
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		
	V3	V4	V5	V6	V7	V8	V9	V10	V11	V12	V13	V14	V15	V16	V17	\
0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.0	0.0	1.0	0.0	0.0	0.0	
2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1.0	1.0	1.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V18	V19	V20	V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	\
0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
3	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V33	V34	V35	V36	V37	V38	V39	V40	V41	V42	V43	V44	V45	V46	V47	\
0	0.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	
2	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	
3	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V48	V49	V50	V51	V52	V53	V54	V55	V56	V57	V58	V59	V60	V61	V62	\
0	NaN	NaN	NaN	NaN	NaN	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
2	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
3	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V63	V64	V65	V66	V67	V68	V69	V70	V71	V72	V73	V74	V75	V76	V77	\
0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
1	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	
2	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
3	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	V78	V79	V80	V81	V82	V83	V84	V85	V86	V87	V88	V89	V90	V91	V92	\

0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0
1	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0
2	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0
3	1.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
	V93	V94	V95	V96	V97	V98	V99	V100	V101	V102	V103	V104	V105	\		
0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	0.0	0.0	1.0	48.0	28.0	0.0	10.0	4.0	1.0	38.0	24.0	0.0	0.0	0.0	0.0	0.0
4	NaN	NaN	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	V106	V107	V108	V109	V110	V111	V112	V113	V114	V115	V116	V117	\			
0	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
1	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
2	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
3	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
4	0.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
	V118	V119	V120	V121	V122	V123	V124	V125	V126	V127	V128	V129	\			
0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	117.0	0.0	0.0				
1	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0				
2	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0				
3	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	50.0	1758.0	925.0	0.0				
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0	0.0	0.0				
	V130	V131	V132	V133	V134	V135	V136	V137	V138	V139	V140	\				
0	0.0	0.0	0.0	117.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN					
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN					
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	NaN	NaN	NaN					
3	354.0	135.0	50.0	1404.0	790.0	0.0	0.0	0.0	NaN	NaN	NaN					
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0				
	V141	V142	V143	V144	V145	V146	V147	V148	V149	V150	V151	V152	\			
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN				
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN				
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN				
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN				
4	0.0	0.0	6.0	18.0	140.0	0.0	0.0	0.0	0.0	1803.0	49.0	64.0				
	V153	V154	V155	V156	V157	V158	V159			V160	V161	V162	\			
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN			NaN	NaN	NaN				
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN			NaN	NaN	NaN				
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN			NaN	NaN	NaN				
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN			NaN	NaN	NaN				
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	15560.0	169690.796875	0.0	0.0	0.0				

	V163	V164	V165	V166	V167	V168	V169	V170	V171	V172	V173	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
4	0.0	515.0	5155.0	2840.0	0.0	0.0	0.0	1.0	1.0	0.0	0.0	
	V174	V175	V176	V177	V178	V179	V180	V181	V182	V183	V184	V185 \
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	0.0	0.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	V186	V187	V188	V189	V190	V191	V192	V193	V194	V195	V196	V197 \
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
	... V140_missing_flag	V141_missing_flag	V142_missing_flag	V143_missing_flag	V144_missing_flag	V145_missing_flag	V146_missing_flag	V147_missing_flag	V148_missing_flag	V149_missing_flag	V150_missing_flag	\
0	...	True	True	...	True	True	True	...	True	True	True	
1	...	True	True	...	True	True	True	...	True	True	True	
2	...	True	True	...	True	True	True	...	True	True	True	
3	...	True	True	...	True	True	True	...	True	True	True	
4	...	False	False	...	False	False	False	...	False	False	False	
	V143_missing_flag	V144_missing_flag	V145_missing_flag	V146_missing_flag	V147_missing_flag	V148_missing_flag	V149_missing_flag	V150_missing_flag	V151_missing_flag	V152_missing_flag	V153_missing_flag	V154_missing_flag \
0	True	True	True	True	True	True	True	True	True	True	True	True
1	True	True	True	True	True	True	True	True	True	True	True	True
2	True	True	True	True	True	True	True	True	True	True	True	True
3	True	True	True	True	True	True	True	True	True	True	True	True
4	False	False	False	False	False	False	False	False	False	False	False	False
	V151_missing_flag	V152_missing_flag	V153_missing_flag	V154_missing_flag	V151_missing_flag	V152_missing_flag	V153_missing_flag	V154_missing_flag	V151_missing_flag	V152_missing_flag	V153_missing_flag	V154_missing_flag \
0	True	True	True	True	True	True	True	True	True	True	True	True
1	True	True	True	True	True	True	True	True	True	True	True	True
2	True	True	True	True	True	True	True	True	True	True	True	True

3	True	True	True	True	True
4	False	False	False	False	False
	V155_missing_flag	V156_missing_flag	V157_missing_flag	V158_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V159_missing_flag	V160_missing_flag	V161_missing_flag	V162_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V163_missing_flag	V164_missing_flag	V165_missing_flag	V166_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V167_missing_flag	V168_missing_flag	V169_missing_flag	V170_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V171_missing_flag	V172_missing_flag	V173_missing_flag	V174_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V175_missing_flag	V176_missing_flag	V177_missing_flag	V178_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V179_missing_flag	V180_missing_flag	V181_missing_flag	V182_missing_flag	\
0	True	True	True	True	True

1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V183_missing_flag	V184_missing_flag	V185_missing_flag	V186_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V187_missing_flag	V188_missing_flag	V189_missing_flag	V190_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V191_missing_flag	V192_missing_flag	V193_missing_flag	V194_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V195_missing_flag	V196_missing_flag	V197_missing_flag	V198_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V199_missing_flag	V200_missing_flag	V201_missing_flag	V202_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	V203_missing_flag	V204_missing_flag	V205_missing_flag	V206_missing_flag
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False

	V207_missing_flag	V208_missing_flag	V209_missing_flag	V210_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V211_missing_flag	V212_missing_flag	V213_missing_flag	V214_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V215_missing_flag	V216_missing_flag	V217_missing_flag	V218_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V219_missing_flag	V220_missing_flag	V221_missing_flag	V222_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V223_missing_flag	V224_missing_flag	V225_missing_flag	V226_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V227_missing_flag	V228_missing_flag	V229_missing_flag	V230_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V231_missing_flag	V232_missing_flag	V233_missing_flag	V234_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True

4	False	False	False	False
0	V235_missing_flag True	V236_missing_flag True	V237_missing_flag True	V238_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V239_missing_flag True	V240_missing_flag True	V241_missing_flag True	V242_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V243_missing_flag True	V244_missing_flag True	V245_missing_flag True	V246_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V247_missing_flag True	V248_missing_flag True	V249_missing_flag True	V250_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V251_missing_flag True	V252_missing_flag True	V253_missing_flag True	V254_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V255_missing_flag True	V256_missing_flag True	V257_missing_flag True	V258_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V259_missing_flag True	V260_missing_flag True	V261_missing_flag True	V262_missing_flag True
1	True	True	True	True

2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V263_missing_flag	V264_missing_flag	V265_missing_flag	V266_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V267_missing_flag	V268_missing_flag	V269_missing_flag	V270_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V271_missing_flag	V272_missing_flag	V273_missing_flag	V274_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V275_missing_flag	V276_missing_flag	V277_missing_flag	V278_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V279_missing_flag	V280_missing_flag	V281_missing_flag	V282_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V283_missing_flag	V284_missing_flag	V285_missing_flag	V286_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V287_missing_flag	V288_missing_flag	V289_missing_flag	V290_missing_flag	\

0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V291_missing_flag	V292_missing_flag	V293_missing_flag	V294_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V295_missing_flag	V296_missing_flag	V297_missing_flag	V298_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V299_missing_flag	V300_missing_flag	V301_missing_flag	V302_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V303_missing_flag	V304_missing_flag	V305_missing_flag	V306_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V307_missing_flag	V308_missing_flag	V309_missing_flag	V310_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
	V311_missing_flag	V312_missing_flag	V313_missing_flag	V314_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False

	V315_missing_flag	V316_missing_flag	V317_missing_flag	V318_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V319_missing_flag	V320_missing_flag	V321_missing_flag	V322_missing_flag	\
0	False	False	False	True	
1	False	False	False	True	
2	False	False	False	True	
3	False	False	False	True	
4	False	False	False	False	
	V323_missing_flag	V324_missing_flag	V325_missing_flag	V326_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	False
	V327_missing_flag	V328_missing_flag	V329_missing_flag	V330_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	False
	V331_missing_flag	V332_missing_flag	V333_missing_flag	V334_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	False
	V335_missing_flag	V336_missing_flag	V337_missing_flag	V338_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	False
	V339_missing_flag	id_01_missing_flag	id_02_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		

3	True	True	True	
4	False	False	False	
0	id_03_missing_flag	id_04_missing_flag	id_05_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
0	id_06_missing_flag	id_07_missing_flag	id_08_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
0	id_09_missing_flag	id_10_missing_flag	id_11_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	False	
0	id_12_missing_flag	id_13_missing_flag	id_14_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	True	False	
0	id_15_missing_flag	id_16_missing_flag	id_17_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	
0	id_18_missing_flag	id_19_missing_flag	id_20_missing_flag	\
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
0	id_21_missing_flag	id_22_missing_flag	id_23_missing_flag	\
	True	True	True	

	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_24_missing_flag	id_25_missing_flag	id_26_missing_flag	\
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	True	True	True	True
	id_27_missing_flag	id_28_missing_flag	id_29_missing_flag	\
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	True	False	False	False
	id_30_missing_flag	id_31_missing_flag	id_32_missing_flag	\
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	id_33_missing_flag	id_34_missing_flag	id_35_missing_flag	\
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	id_36_missing_flag	id_37_missing_flag	id_38_missing_flag	\
0	True	True	True	True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
	DeviceType_missing_flag	DeviceInfo_missing_flag	Date	\
0	True	True	2017-12-02 00:00:00	
1	True	True	2017-12-02 00:00:01	
2	True	True	2017-12-02 00:01:09	
3	True	True	2017-12-02 00:01:39	
4	False	False	2017-12-02 00:01:46	

```

_Weekdays _Hours _Days Trans_min_mean Trans_min_std \
0 5 0 2 -66.527347 -0.278174
1 5 0 2 -106.027347 -0.443337
2 5 0 2 -76.027347 -0.317897
3 5 0 2 -85.027347 -0.355529
4 5 0 2 -85.027347 -0.355529

TransactionAmt_to_mean_card1 TransactionAmt_to_mean_card4 \
0 0.194638 0.257761
1 0.123780 0.219053
2 0.608151 0.443070
3 0.405136 0.377678
4 0.515616 0.377678

TransactionAmt_to_std_card1 TransactionAmt_to_std_card4
0 0.184560 0.170241
1 0.063004 0.114214
2 0.589226 0.258550
3 0.259460 0.196921
4 0.882898 0.196921

[5 rows x 846 columns]

```

8 8. Dimensionality Reduction - PCA

```
[61]: def perform_PCA(df, cols, n_components, prefix='PCA_', rand_seed=4):
    pca = PCA(n_components=n_components, random_state=rand_seed)
    principalComponents = pca.fit_transform(df[cols])
    principalDf = pd.DataFrame(principalComponents)
    df.drop(cols, axis=1, inplace=True)

    principalDf.rename(columns=lambda x: str(prefix)+str(x), inplace=True)
    df = pd.concat([df, principalDf], axis=1)
    return df
```

```
[62]: # Columns starting from V1 to V339
filter_col = df.columns[53:392]
```

Impute missing values in the mas_v columns, later use minmax_scale function to scale the values in these columns

```
[63]: from sklearn.preprocessing import minmax_scale

# Fill na values and scale V columns
for col in filter_col:
    df[col] = df[col].fillna((df[col].min() - 2))
```

```

df[col] = (minmax_scale(df[col], feature_range=(0,1)))

# Perform PCA
df = perform_PCA(df, filter_col, prefix='PCA_V_', n_components=30)

[64]: df = reduce_mem_usage(df)

Mem. usage decreased to 1170.85 Mb (4.4% reduction)

[65]: df.head()

[65]:   TransactionID  isFraud  TransactionDT  TransactionAmt  ProductCD  card1 \
0          2987000      0           86400        4.226562       W  13926
1          2987001      0           86401        3.367188       W  2755
2          2987002      0           86469        4.078125       W  4663
3          2987003      0           86499        3.912109       W  18132
4          2987004      0           86506        3.912109       H  4497

    card2  card3      card4  card5  card6  addr1  addr2  dist1 P_emaildomain \
0    NaN  150.0  discover  142.0  credit  315.0   87.0   19.0      NoInf
1  404.0  150.0  mastercard  102.0  credit  325.0   87.0    NaN      Google
2  490.0  150.0      visa  166.0  debit   330.0   87.0  287.0      Microsoft
3  567.0  150.0  mastercard  117.0  debit   476.0   87.0    NaN      Yahoo Mail
4  514.0  150.0  mastercard  102.0  credit  420.0   87.0    NaN      Google

R_emaildomain  C1  C2  C3  C4  C5  C6  C7  C8  C9  C10  C11  C12 \
0      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  2.0  0.0
1      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  0.0  0.0  1.0  0.0
2      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  1.0  0.0
3      NoInf  2.0  5.0  0.0  0.0  0.0  4.0  0.0  0.0  1.0  0.0  1.0  0.0
4      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  1.0  0.0  1.0  1.0  0.0

C13  C14      D1      D2      D3      D4      D5      D6      D8      D9      D10      D11      D12 \
0    1.0  1.0   14.0    NaN   13.0    NaN    NaN    NaN    NaN  13.0   13.0    NaN
1    1.0  1.0    0.0    NaN    NaN    0.0    NaN    NaN    NaN    0.0    NaN    NaN
2    1.0  1.0    0.0    NaN    NaN    0.0    NaN    NaN    NaN    0.0  315.0    NaN
3   25.0  1.0  112.0  112.0    0.0   94.0    0.0    NaN    NaN    NaN  84.0    NaN    NaN
4    1.0  1.0    0.0    NaN    NaN    NaN    NaN    NaN    NaN    NaN    NaN    NaN    NaN

D13  D14      D15      M1      M2      M3      M4      M5      M6      M7      M8      M9      id_01 \
0    NaN    NaN    0.0      T      T      T     M2      F      T    NaN    NaN    NaN    NaN
1    NaN    NaN    0.0    NaN    NaN    NaN    MO      T      T    NaN    NaN    NaN    NaN
2    NaN    NaN  315.0      T      T      T     MO      F      F      F      F      F    NaN
3    NaN    NaN  111.0    NaN    NaN    NaN    MO      T      F    NaN    NaN    NaN    NaN
4    NaN    0.0

id_02  id_03  id_04  id_05  id_06  id_09  id_10  id_11      id_12  id_13 \

```

0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	70787.0	NaN	NaN	NaN	NaN	NaN	NaN	100.0	NotFound	NaN
	id_14	id_15	id_16	id_17	id_19	id_20	id_28	id_29	id_30	\
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NAN	
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NAN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NAN	
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NAN	
4	-480.0	New	NotFound	166.0	542.0	144.0	New	NotFound	Android	
	id_31	id_32	id_33		id_34	id_35	id_36	id_37	\	
0	NaN	NaN	NaN		NaN	NaN	NaN	NaN		
1	NaN	NaN	NaN		NaN	NaN	NaN	NaN		
2	NaN	NaN	NaN		NaN	NaN	NaN	NaN		
3	NaN	NaN	NaN		NaN	NaN	NaN	NaN		
4	samsung	browser	6.2	32.0	2220x1080	match_status:2	T	F	T	
	id_38	DeviceType		DeviceInfo	card2_missing_flag	\				
0	NaN	NaN		NaN	True					
1	NaN	NaN		NaN	False					
2	NaN	NaN		NaN	False					
3	NaN	NaN		NaN	False					
4	T	mobile	SAMSUNG SM-G892A Build/NRD90M		False					
	card3_missing_flag	card4_missing_flag	card5_missing_flag	\						
0	False	False	False							
1	False	False	False							
2	False	False	False							
3	False	False	False							
4	False	False	False							
	card6_missing_flag	addr1_missing_flag	addr2_missing_flag	\						
0	False	False	False							
1	False	False	False							
2	False	False	False							
3	False	False	False							
4	False	False	False							
	dist1_missing_flag	dist2_missing_flag	P_emaildomain_missing_flag	\						
0	False	True	True							
1	True	True	True							
2	False	True	True							
3	True	True	True							
4	True	True	True							

	R_emaildomain_missing_flag	D1_missing_flag	D2_missing_flag	\
0	True	False	True	
1	True	False	True	
2	True	False	True	
3	True	False	False	
4	True	False	True	
	D3_missing_flag	D4_missing_flag	D5_missing_flag	D6_missing_flag \
0	False	True	True	True
1	True	False	True	True
2	True	False	True	True
3	False	False	False	True
4	True	True	True	True
	D7_missing_flag	D8_missing_flag	D9_missing_flag	D10_missing_flag \
0	True	True	True	False
1	True	True	True	False
2	True	True	True	False
3	True	True	True	False
4	True	True	True	True
	D11_missing_flag	D12_missing_flag	D13_missing_flag	D14_missing_flag \
0	False	True	True	True
1	True	True	True	True
2	False	True	True	True
3	True	True	True	True
4	True	True	True	True
	D15_missing_flag	M1_missing_flag	M2_missing_flag	M3_missing_flag \
0	False	False	False	False
1	False	True	True	True
2	False	False	False	False
3	False	True	True	True
4	True	True	True	True
	M4_missing_flag	M5_missing_flag	M6_missing_flag	M7_missing_flag \
0	False	False	False	True
1	False	False	False	True
2	False	False	False	False
3	False	False	False	True
4	True	True	True	True
	M8_missing_flag	M9_missing_flag	V1_missing_flag	V2_missing_flag \
0	True	True	False	False
1	True	True	True	True
2	False	False	False	False

3	True	True	True	True	
4	True	True	True	True	
	V3_missing_flag	V4_missing_flag	V5_missing_flag	V6_missing_flag	\
0	False	False	False	False	False
1	True	True	True	True	True
2	False	False	False	False	False
3	True	True	True	True	True
4	True	True	True	True	True
	V7_missing_flag	V8_missing_flag	V9_missing_flag	V10_missing_flag	\
0	False	False	False	False	False
1	True	True	True	True	True
2	False	False	False	False	False
3	True	True	True	True	True
4	True	True	True	True	True
	V11_missing_flag	V12_missing_flag	V13_missing_flag	V14_missing_flag	\
0	False	False	False	False	False
1	True	False	False	False	False
2	False	False	False	False	False
3	True	False	False	False	False
4	True	True	True	True	True
	V15_missing_flag	V16_missing_flag	V17_missing_flag	V18_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V19_missing_flag	V20_missing_flag	V21_missing_flag	V22_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V23_missing_flag	V24_missing_flag	V25_missing_flag	V26_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V27_missing_flag	V28_missing_flag	V29_missing_flag	V30_missing_flag	\
0	False	False	False	False	False

1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V31_missing_flag	V32_missing_flag	V33_missing_flag	V34_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V35_missing_flag	V36_missing_flag	V37_missing_flag	V38_missing_flag
0	True	True	True	True
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V39_missing_flag	V40_missing_flag	V41_missing_flag	V42_missing_flag
0	True	True	True	True
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V43_missing_flag	V44_missing_flag	V45_missing_flag	V46_missing_flag
0	True	True	True	True
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V47_missing_flag	V48_missing_flag	V49_missing_flag	V50_missing_flag
0	True	True	True	True
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V51_missing_flag	V52_missing_flag	V53_missing_flag	V54_missing_flag
0	True	True	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True

	V55_missing_flag	V56_missing_flag	V57_missing_flag	V58_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V59_missing_flag	V60_missing_flag	V61_missing_flag	V62_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V63_missing_flag	V64_missing_flag	V65_missing_flag	V66_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V67_missing_flag	V68_missing_flag	V69_missing_flag	V70_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V71_missing_flag	V72_missing_flag	V73_missing_flag	V74_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V75_missing_flag	V76_missing_flag	V77_missing_flag	V78_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V79_missing_flag	V80_missing_flag	V81_missing_flag	V82_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	

	4	True	True	True	True
0	V83_missing_flag	V84_missing_flag	V85_missing_flag	V86_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
0	V87_missing_flag	V88_missing_flag	V89_missing_flag	V90_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
0	V91_missing_flag	V92_missing_flag	V93_missing_flag	V94_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
0	V95_missing_flag	V96_missing_flag	V97_missing_flag	V98_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
0	V99_missing_flag	V100_missing_flag	V101_missing_flag	V102_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
0	V103_missing_flag	V104_missing_flag	V105_missing_flag	V106_missing_flag	\
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
0	V107_missing_flag	V108_missing_flag	V109_missing_flag	V110_missing_flag	\
1	False	False	False	False	False

2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V111_missing_flag	V112_missing_flag	V113_missing_flag	V114_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V115_missing_flag	V116_missing_flag	V117_missing_flag	V118_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V119_missing_flag	V120_missing_flag	V121_missing_flag	V122_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V123_missing_flag	V124_missing_flag	V125_missing_flag	V126_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V127_missing_flag	V128_missing_flag	V129_missing_flag	V130_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	V131_missing_flag	V132_missing_flag	...	V170_missing_flag	\
0	False	False	...	True	
1	False	False	...	True	
2	False	False	...	True	
3	False	False	...	True	
4	False	False	...	False	
	V171_missing_flag	V172_missing_flag	V173_missing_flag	V174_missing_flag	\

0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V175_missing_flag	V176_missing_flag	V177_missing_flag	V178_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V179_missing_flag	V180_missing_flag	V181_missing_flag	V182_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V183_missing_flag	V184_missing_flag	V185_missing_flag	V186_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V187_missing_flag	V188_missing_flag	V189_missing_flag	V190_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V191_missing_flag	V192_missing_flag	V193_missing_flag	V194_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V195_missing_flag	V196_missing_flag	V197_missing_flag	V198_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V199_missing_flag	V200_missing_flag	V201_missing_flag	V202_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V203_missing_flag	V204_missing_flag	V205_missing_flag	V206_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V207_missing_flag	V208_missing_flag	V209_missing_flag	V210_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V211_missing_flag	V212_missing_flag	V213_missing_flag	V214_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V215_missing_flag	V216_missing_flag	V217_missing_flag	V218_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V219_missing_flag	V220_missing_flag	V221_missing_flag	V222_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V223_missing_flag	V224_missing_flag	V225_missing_flag	V226_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True

3	True	True	True	True	True
4	False	False	False	False	False
	V227_missing_flag	V228_missing_flag	V229_missing_flag	V230_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V231_missing_flag	V232_missing_flag	V233_missing_flag	V234_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V235_missing_flag	V236_missing_flag	V237_missing_flag	V238_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V239_missing_flag	V240_missing_flag	V241_missing_flag	V242_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V243_missing_flag	V244_missing_flag	V245_missing_flag	V246_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V247_missing_flag	V248_missing_flag	V249_missing_flag	V250_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V251_missing_flag	V252_missing_flag	V253_missing_flag	V254_missing_flag	\
0	True	True	True	True	True

1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V255_missing_flag	V256_missing_flag	V257_missing_flag	V258_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V259_missing_flag	V260_missing_flag	V261_missing_flag	V262_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V263_missing_flag	V264_missing_flag	V265_missing_flag	V266_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V267_missing_flag	V268_missing_flag	V269_missing_flag	V270_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V271_missing_flag	V272_missing_flag	V273_missing_flag	V274_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V275_missing_flag	V276_missing_flag	V277_missing_flag	V278_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V279_missing_flag	V280_missing_flag	V281_missing_flag	V282_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V283_missing_flag	V284_missing_flag	V285_missing_flag	V286_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V287_missing_flag	V288_missing_flag	V289_missing_flag	V290_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V291_missing_flag	V292_missing_flag	V293_missing_flag	V294_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V295_missing_flag	V296_missing_flag	V297_missing_flag	V298_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V299_missing_flag	V300_missing_flag	V301_missing_flag	V302_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V303_missing_flag	V304_missing_flag	V305_missing_flag	V306_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	

4	False	False	False	False
0	V307_missing_flag	V308_missing_flag	V309_missing_flag	V310_missing_flag \
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
0	V311_missing_flag	V312_missing_flag	V313_missing_flag	V314_missing_flag \
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
0	V315_missing_flag	V316_missing_flag	V317_missing_flag	V318_missing_flag \
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
0	V319_missing_flag	V320_missing_flag	V321_missing_flag	V322_missing_flag \
1	False	False	False	True
2	False	False	False	True
3	False	False	False	True
4	False	False	False	False
0	V323_missing_flag	V324_missing_flag	V325_missing_flag	V326_missing_flag \
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V327_missing_flag	V328_missing_flag	V329_missing_flag	V330_missing_flag \
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V331_missing_flag	V332_missing_flag	V333_missing_flag	V334_missing_flag \
1	True	True	True	True

2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V335_missing_flag	V336_missing_flag	V337_missing_flag	V338_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V339_missing_flag	id_01_missing_flag	id_02_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	id_03_missing_flag	id_04_missing_flag	id_05_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	True	True	True		
	id_06_missing_flag	id_07_missing_flag	id_08_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	True	True	True		
	id_09_missing_flag	id_10_missing_flag	id_11_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	True	True	False		
	id_12_missing_flag	id_13_missing_flag	id_14_missing_flag	\	
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	True	False		
	id_15_missing_flag	id_16_missing_flag	id_17_missing_flag	\	

0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	
	id_18_missing_flag	id_19_missing_flag	id_20_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_21_missing_flag	id_22_missing_flag	id_23_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_24_missing_flag	id_25_missing_flag	id_26_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_27_missing_flag	id_28_missing_flag	id_29_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_30_missing_flag	id_31_missing_flag	id_32_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	
	id_33_missing_flag	id_34_missing_flag	id_35_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	

```

      id_36_missing_flag  id_37_missing_flag  id_38_missing_flag  \
0              True                  True                  True
1              True                  True                  True
2              True                  True                  True
3              True                  True                  True
4             False                 False                 False

      DeviceType_missing_flag DeviceInfo_missing_flag          Date  \
0                      True                     True 2017-12-02 00:00:00
1                      True                     True 2017-12-02 00:00:01
2                      True                     True 2017-12-02 00:01:09
3                      True                     True 2017-12-02 00:01:39
4                     False                    False 2017-12-02 00:01:46

      _Weekdays   _Hours   _Days  Trans_min_mean  Trans_min_std  \
0          5       0       2        -66.5      -0.278076
1          5       0       2       -106.0      -0.443359
2          5       0       2        -76.0      -0.317871
3          5       0       2        -85.0      -0.355469
4          5       0       2        -85.0      -0.355469

      TransactionAmt_to_mean_card1  TransactionAmt_to_mean_card4  \
0            0.194580                0.257812
1            0.123779                0.218994
2            0.607910                0.443115
3            0.405029                0.377686
4            0.515625                0.377686

      TransactionAmt_to_std_card1  TransactionAmt_to_std_card4  PCA_V_0  \
0            0.184560                0.170288 -0.157349
1            0.063004                0.114197 -0.086426
2            0.589226                0.258545 -0.800781
3            0.259460                0.196899 -0.237305
4            0.882898                0.196899  2.904297

      PCA_V_1    PCA_V_2    PCA_V_3    PCA_V_4    PCA_V_5    PCA_V_6    PCA_V_7  \
0  0.918945 -0.843750  0.307861 -0.089417  0.003073 -0.020004 -0.187622
1 -0.800293 -0.152222 -0.363281 -0.101929 -0.002268  0.032410 -0.068970
2  0.316895  0.273193 -0.026337  0.043182 -0.008049 -0.039246 -0.217041
3 -0.811523 -0.123718 -0.423584 -0.067444  0.025040  0.110596 -0.253906
4  0.380127  0.480469 -0.009628 -0.171631  1.170898 -0.178345  0.004597

      PCA_V_8    PCA_V_9    PCA_V_10   PCA_V_11   PCA_V_12   PCA_V_13   PCA_V_14  \
0  0.038269  0.002508 -0.010406  0.034119 -0.044281 -0.089722  0.045013
1  0.040070 -0.180054 -0.059326  0.002270  0.018936 -0.029465  0.016541
2  0.017593  0.033508 -0.000374 -0.015335  0.020706 -0.046051 -0.006680

```

```

3  0.004715  0.170288 -0.012352 -0.014534  0.005116  0.031982 -0.013611
4  0.043732 -0.001608  0.015762 -0.017914 -0.020660 -0.005581 -0.010246

    PCA_V_15  PCA_V_16  PCA_V_17  PCA_V_18  PCA_V_19  PCA_V_20  PCA_V_21  \
0  0.001526 -0.003292  0.018463 -0.018341  0.009972 -0.026993 -0.021591
1 -0.006153 -0.004082  0.010201 -0.001767 -0.022842  0.006413 -0.021011
2  0.004841  0.001108  0.007488 -0.007820 -0.006615 -0.008324  0.009743
3 -0.017181 -0.001876 -0.084717  0.050446  0.140747  0.059052 -0.020187
4 -0.003710 -0.011406 -0.008102  0.013367  0.017807 -0.013626  0.000090

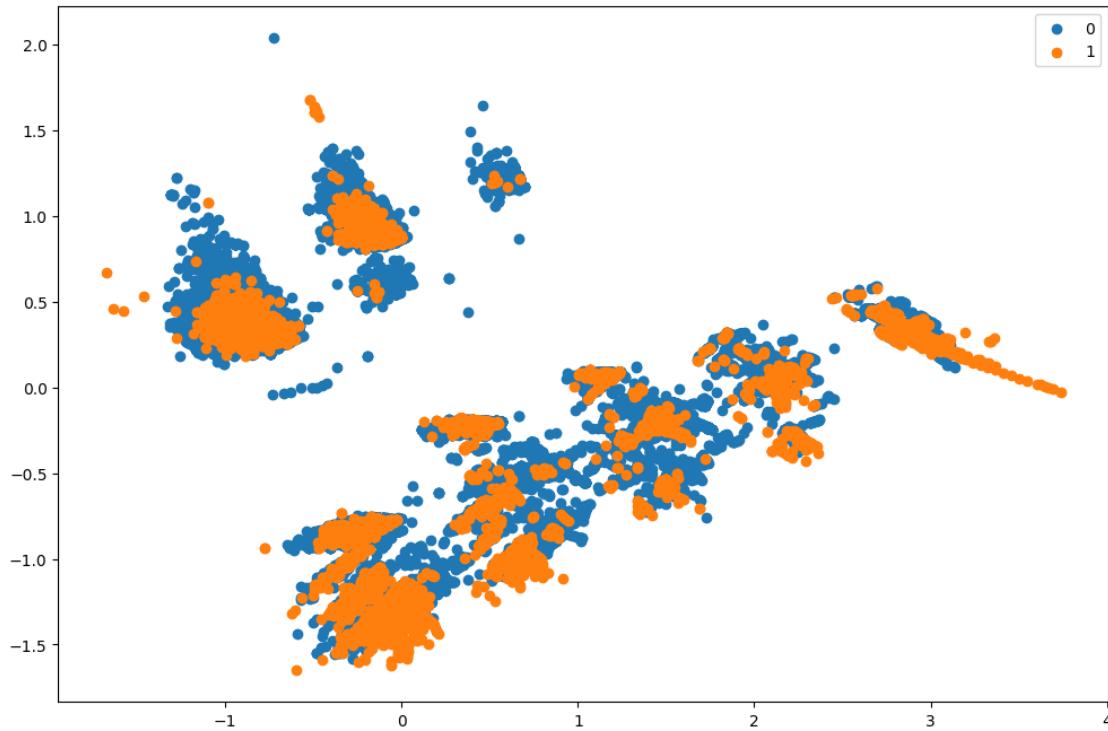
    PCA_V_22  PCA_V_23  PCA_V_24  PCA_V_25  PCA_V_26  PCA_V_27  PCA_V_28  \
0 -0.054565  0.025299  0.018677 -0.006100  0.004215 -0.042999  0.008232
1  0.054779 -0.042572 -0.026566  0.003536  0.001620  0.001561 -0.003544
2 -0.007683  0.003206  0.002855  0.001932  0.003839  0.002985 -0.019638
3  0.066467 -0.010803 -0.017715  0.025543  0.003864  0.003452  0.026138
4 -0.005260 -0.141479  0.203003  0.014557  0.012848  0.002218  0.014000

    PCA_V_29
0 -0.007919
1 -0.003748
2 -0.003443
3 -0.041901
4 -0.001691

[5 rows x 537 columns]

```

```
[66]: # Plot first 2 PCA features and colour by target variable
plt.figure(figsize=(12, 8));
groups = df.groupby("isFraud")
for name, group in groups:
    plt.scatter(group["PCA_V_0"], group["PCA_V_1"], label=name)
plt.legend()
plt.show()
```



9 9. Feature Encoding

```
[67]: cat_cols = df.select_dtypes(include=['object','category']).columns.tolist()
```

```
[68]: binary_cols = [col for col in df.columns if df[col].dropna().nunique() ==2 and col not in cat_cols]
```

```
[69]: num_cols = [col for col in df.select_dtypes(include=['int64','int32','int16','int8','float64','float32','float16']).columns if col not in binary_cols]
```

```
[70]: len(cat_cols)
```

```
[70]: 29
```

```
[71]: len(binary_cols)
```

```
[71]: 415
```

```
[72]: len(num_cols)
```

```
[72]: 92
```

```
[73]: cat_cols = cat_cols + binary_cols  
len(cat_cols)
```

```
[73]: 444
```

```
[ ]:
```

```
[74]: # Frequency encoding variables  
frequency_encoded_variables = []  
for col in cat_cols:  
    if df[col].nunique() > 30:  
        print(col, df[col].nunique())  
        frequency_encoded_variables.append(col)
```

```
id_31 130  
id_33 260  
DeviceInfo 1786
```

```
[75]: for variable in tqdm(frequency_encoded_variables):  
    # group by frequency  
    fq = df.groupby(variable).size()/len(df)  
    # mapping values to dataframe  
    df.loc[:, "{}".format(variable)] = df[variable].map(fq)  
    cat_columns.remove(variable)
```

```
100%|  
| 3/3 [00:00<00:00, 4.14it/s]
```

```
[76]: df.head()
```

```
[76]: TransactionID  isFraud  TransactionDT  TransactionAmt  ProductCD  card1  \  
0      2987000       0          86400        4.226562      W  13926  
1      2987001       0          86401        3.367188      W  2755  
2      2987002       0          86469        4.078125      W  4663  
3      2987003       0          86499        3.912109      W  18132  
4      2987004       0          86506        3.912109      H  4497  
  
    card2  card3      card4  card5  card6  addr1  addr2  dist1 P_emaildomain  \  
0   NaN  150.0  discover  142.0  credit  315.0   87.0   19.0      NoInf  
1  404.0  150.0  mastercard  102.0  credit  325.0   87.0     NaN      Google  
2  490.0  150.0      visa  166.0  debit   330.0   87.0  287.0      Microsoft  
3  567.0  150.0  mastercard  117.0  debit   476.0   87.0     NaN  Yahoo Mail  
4  514.0  150.0  mastercard  102.0  credit  420.0   87.0     NaN      Google  
  
R_emaildomain  C1  C2  C3  C4  C5  C6  C7  C8  C9  C10  C11  C12  \  
0      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  1.0  0.0  2.0  0.0  
1      NoInf  1.0  1.0  0.0  0.0  0.0  1.0  0.0  0.0  0.0  0.0  1.0  0.0
```

2	NoInf	1.0	1.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0
3	NoInf	2.0	5.0	0.0	0.0	0.0	4.0	0.0	0.0	1.0	0.0	1.0	0.0
4	NoInf	1.0	1.0	0.0	0.0	0.0	1.0	0.0	1.0	0.0	1.0	1.0	0.0
	C13	C14	D1	D2	D3	D4	D5	D6	D8	D9	D10	D11	D12
0	1.0	1.0	14.0	NaN	13.0	NaN	NaN	NaN	NaN	13.0	13.0	NaN	
1	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	0.0	NaN	NaN	
2	1.0	1.0	0.0	NaN	NaN	0.0	NaN	NaN	NaN	0.0	315.0	NaN	
3	25.0	1.0	112.0	112.0	0.0	94.0	0.0	NaN	NaN	84.0	NaN	NaN	
4	1.0	1.0	0.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	M9	id_01
0	NaN	NaN	0.0	T	T	T	M2	F	T	NaN	NaN	NaN	NaN
1	NaN	NaN	0.0	NaN	NaN	NaN	M0	T	T	NaN	NaN	NaN	NaN
2	NaN	NaN	315.0	T	T	T	M0	F	F	F	F	F	NaN
3	NaN	NaN	111.0	NaN	NaN	NaN	M0	T	F	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	0.0
	id_02	id_03	id_04	id_05	id_06	id_09	id_10	id_11		id_12	id_13		
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN		
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN		
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN		
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NaN		
4	70787.0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	100.0	NotFound	NaN		
	id_14	id_15	id_16	id_17	id_19	id_20	id_28		id_29	id_30			
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NAN			
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NAN			
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NAN			
3	NaN	NaN	NaN	NaN	NaN	NaN	NaN		NaN	NAN			
4	-480.0	New	NotFound	166.0	542.0	144.0	New	NotFound	Android				
	id_31	id_32	id_33		id_34	id_35	id_36	id_37	id_38				
0	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN				
1	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN				
2	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN				
3	NaN	NaN	NaN		NaN	NaN	NaN	NaN	NaN				
4	0.001797	32.0	0.000921	match_status:2		T	F	T	T				
	DeviceType	DeviceInfo	card2_missing_flag	card3_missing_flag									
0	NaN	NaN	True										
1	NaN	NaN	False										
2	NaN	NaN	False										
3	NaN	NaN	False										
4	mobile	0.000015	False										
	card4_missing_flag	card5_missing_flag	card6_missing_flag										

0	False	False	False		
1	False	False	False		
2	False	False	False		
3	False	False	False		
4	False	False	False		
	addr1_missing_flag	addr2_missing_flag	dist1_missing_flag	\	
0	False	False	False		
1	False	False	True		
2	False	False	False		
3	False	False	True		
4	False	False	True		
	dist2_missing_flag	P_emaildomain_missing_flag	R_emaildomain_missing_flag	\	
0	True	True	True		
1	True	False	True		
2	True	False	True		
3	True	False	True		
4	True	False	True		
	D1_missing_flag	D2_missing_flag	D3_missing_flag	D4_missing_flag	\
0	False	True	False	True	
1	False	True	True	False	
2	False	True	True	False	
3	False	False	False	False	
4	False	True	True	True	
	D5_missing_flag	D6_missing_flag	D7_missing_flag	D8_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	False	True	True	True	
4	True	True	True	True	
	D9_missing_flag	D10_missing_flag	D11_missing_flag	D12_missing_flag	\
0	True	False	False	True	
1	True	False	True	True	
2	True	False	False	True	
3	True	False	True	True	
4	True	True	True	True	
	D13_missing_flag	D14_missing_flag	D15_missing_flag	M1_missing_flag	\
0	True	True	False	False	
1	True	True	False	True	
2	True	True	False	False	
3	True	True	False	True	
4	True	True	True	True	

	M2_missing_flag	M3_missing_flag	M4_missing_flag	M5_missing_flag	\
0	False	False	False	False	
1	True	True	False	False	
2	False	False	False	False	
3	True	True	False	False	
4	True	True	True	True	

	M6_missing_flag	M7_missing_flag	M8_missing_flag	M9_missing_flag	\
0	False	True	True	True	
1	False	True	True	True	
2	False	False	False	False	
3	False	True	True	True	
4	True	True	True	True	

	V1_missing_flag	V2_missing_flag	V3_missing_flag	V4_missing_flag	\
0	False	False	False	False	
1	True	True	True	True	
2	False	False	False	False	
3	True	True	True	True	
4	True	True	True	True	

	V5_missing_flag	V6_missing_flag	V7_missing_flag	V8_missing_flag	\
0	False	False	False	False	
1	True	True	True	True	
2	False	False	False	False	
3	True	True	True	True	
4	True	True	True	True	

	V9_missing_flag	V10_missing_flag	V11_missing_flag	V12_missing_flag	\
0	False	False	False	False	
1	True	True	True	False	
2	False	False	False	False	
3	True	True	True	False	
4	True	True	True	True	

	V13_missing_flag	V14_missing_flag	V15_missing_flag	V16_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	

	V17_missing_flag	V18_missing_flag	V19_missing_flag	V20_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	

3	False	False	False	False	
4	True	True	True	True	
	V21_missing_flag	V22_missing_flag	V23_missing_flag	V24_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V25_missing_flag	V26_missing_flag	V27_missing_flag	V28_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V29_missing_flag	V30_missing_flag	V31_missing_flag	V32_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V33_missing_flag	V34_missing_flag	V35_missing_flag	V36_missing_flag	\
0	False	False	True	True	True
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V37_missing_flag	V38_missing_flag	V39_missing_flag	V40_missing_flag	\
0	True	True	True	True	True
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V41_missing_flag	V42_missing_flag	V43_missing_flag	V44_missing_flag	\
0	True	True	True	True	True
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	True	True	True	True	True
	V45_missing_flag	V46_missing_flag	V47_missing_flag	V48_missing_flag	\
0	True	True	True	True	True

1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V49_missing_flag	V50_missing_flag	V51_missing_flag	V52_missing_flag
0	True	True	True	True
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V53_missing_flag	V54_missing_flag	V55_missing_flag	V56_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V57_missing_flag	V58_missing_flag	V59_missing_flag	V60_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V61_missing_flag	V62_missing_flag	V63_missing_flag	V64_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V65_missing_flag	V66_missing_flag	V67_missing_flag	V68_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True
	V69_missing_flag	V70_missing_flag	V71_missing_flag	V72_missing_flag
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	True	True	True	True

	V73_missing_flag	V74_missing_flag	V75_missing_flag	V76_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V77_missing_flag	V78_missing_flag	V79_missing_flag	V80_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V81_missing_flag	V82_missing_flag	V83_missing_flag	V84_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V85_missing_flag	V86_missing_flag	V87_missing_flag	V88_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V89_missing_flag	V90_missing_flag	V91_missing_flag	V92_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	True	True	
	V93_missing_flag	V94_missing_flag	V95_missing_flag	V96_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	True	True	False	False	
	V97_missing_flag	V98_missing_flag	V99_missing_flag	V100_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	

4	False	False	False	False	
0	V101_missing_flag	V102_missing_flag	V103_missing_flag	V104_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V105_missing_flag	V106_missing_flag	V107_missing_flag	V108_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V109_missing_flag	V110_missing_flag	V111_missing_flag	V112_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V113_missing_flag	V114_missing_flag	V115_missing_flag	V116_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V117_missing_flag	V118_missing_flag	V119_missing_flag	V120_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V121_missing_flag	V122_missing_flag	V123_missing_flag	V124_missing_flag	\
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
0	V125_missing_flag	V126_missing_flag	V127_missing_flag	V128_missing_flag	\
1	False	False	False	False	

2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V129_missing_flag	V130_missing_flag	V131_missing_flag	V132_missing_flag	\
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
	... V170_missing_flag	V171_missing_flag	V172_missing_flag	\	
0	... True	True	True	True	
1	... True	True	True	True	
2	... True	True	True	True	
3	... True	True	True	True	
4	... False	False	False	False	
	V173_missing_flag	V174_missing_flag	V175_missing_flag	V176_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V177_missing_flag	V178_missing_flag	V179_missing_flag	V180_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V181_missing_flag	V182_missing_flag	V183_missing_flag	V184_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V185_missing_flag	V186_missing_flag	V187_missing_flag	V188_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V189_missing_flag	V190_missing_flag	V191_missing_flag	V192_missing_flag	\

0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V193_missing_flag	V194_missing_flag	V195_missing_flag	V196_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V197_missing_flag	V198_missing_flag	V199_missing_flag	V200_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V201_missing_flag	V202_missing_flag	V203_missing_flag	V204_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V205_missing_flag	V206_missing_flag	V207_missing_flag	V208_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V209_missing_flag	V210_missing_flag	V211_missing_flag	V212_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V213_missing_flag	V214_missing_flag	V215_missing_flag	V216_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False

	V217_missing_flag	V218_missing_flag	V219_missing_flag	V220_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V221_missing_flag	V222_missing_flag	V223_missing_flag	V224_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V225_missing_flag	V226_missing_flag	V227_missing_flag	V228_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V229_missing_flag	V230_missing_flag	V231_missing_flag	V232_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V233_missing_flag	V234_missing_flag	V235_missing_flag	V236_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V237_missing_flag	V238_missing_flag	V239_missing_flag	V240_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V241_missing_flag	V242_missing_flag	V243_missing_flag	V244_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True

3	True	True	True	True	True
4	False	False	False	False	False
	V245_missing_flag	V246_missing_flag	V247_missing_flag	V248_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V249_missing_flag	V250_missing_flag	V251_missing_flag	V252_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V253_missing_flag	V254_missing_flag	V255_missing_flag	V256_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V257_missing_flag	V258_missing_flag	V259_missing_flag	V260_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V261_missing_flag	V262_missing_flag	V263_missing_flag	V264_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V265_missing_flag	V266_missing_flag	V267_missing_flag	V268_missing_flag	\
0	True	True	True	True	True
1	True	True	True	True	True
2	True	True	True	True	True
3	True	True	True	True	True
4	False	False	False	False	False
	V269_missing_flag	V270_missing_flag	V271_missing_flag	V272_missing_flag	\
0	True	True	True	True	True

1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V273_missing_flag	V274_missing_flag	V275_missing_flag	V276_missing_flag	\
0	True	True	True	True	
1	True	True	True	True	
2	True	True	True	True	
3	True	True	True	True	
4	False	False	False	False	
	V277_missing_flag	V278_missing_flag	V279_missing_flag	V280_missing_flag	\
0	True	True	False	False	
1	True	True	False	False	
2	True	True	False	False	
3	True	True	False	False	
4	False	False	False	False	
	V281_missing_flag	V282_missing_flag	V283_missing_flag	V284_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V285_missing_flag	V286_missing_flag	V287_missing_flag	V288_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V289_missing_flag	V290_missing_flag	V291_missing_flag	V292_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V293_missing_flag	V294_missing_flag	V295_missing_flag	V296_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	

	V297_missing_flag	V298_missing_flag	V299_missing_flag	V300_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V301_missing_flag	V302_missing_flag	V303_missing_flag	V304_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V305_missing_flag	V306_missing_flag	V307_missing_flag	V308_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V309_missing_flag	V310_missing_flag	V311_missing_flag	V312_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V313_missing_flag	V314_missing_flag	V315_missing_flag	V316_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V317_missing_flag	V318_missing_flag	V319_missing_flag	V320_missing_flag	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
	V321_missing_flag	V322_missing_flag	V323_missing_flag	V324_missing_flag	\
0	False	True	True	True	
1	False	True	True	True	
2	False	True	True	True	
3	False	True	True	True	

4	False	False	False	False
0	V325_missing_flag True	V326_missing_flag True	V327_missing_flag True	V328_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V329_missing_flag True	V330_missing_flag True	V331_missing_flag True	V332_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V333_missing_flag True	V334_missing_flag True	V335_missing_flag True	V336_missing_flag True
1	True	True	True	True
2	True	True	True	True
3	True	True	True	True
4	False	False	False	False
0	V337_missing_flag True	V338_missing_flag True	V339_missing_flag True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	False	
0	id_01_missing_flag True	id_02_missing_flag True	id_03_missing_flag True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
0	id_04_missing_flag True	id_05_missing_flag True	id_06_missing_flag True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
0	id_07_missing_flag True	id_08_missing_flag True	id_09_missing_flag True	
1	True	True	True	

2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_10_missing_flag	id_11_missing_flag	id_12_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_13_missing_flag	id_14_missing_flag	id_15_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	False	False	
	id_16_missing_flag	id_17_missing_flag	id_18_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
	id_19_missing_flag	id_20_missing_flag	id_21_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	False	False	True	
	id_22_missing_flag	id_23_missing_flag	id_24_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_25_missing_flag	id_26_missing_flag	id_27_missing_flag	\
0	True	True	True	
1	True	True	True	
2	True	True	True	
3	True	True	True	
4	True	True	True	
	id_28_missing_flag	id_29_missing_flag	id_30_missing_flag	\

0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	id_31_missing_flag	id_32_missing_flag	id_33_missing_flag		
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	id_34_missing_flag	id_35_missing_flag	id_36_missing_flag		
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	id_37_missing_flag	id_38_missing_flag	DeviceType_missing_flag		
0	True	True	True		
1	True	True	True		
2	True	True	True		
3	True	True	True		
4	False	False	False		
	DeviceInfo_missing_flag	Date	_Weekdays	_Hours	_Days
0	True	2017-12-02 00:00:00	5	0	2
1	True	2017-12-02 00:00:01	5	0	2
2	True	2017-12-02 00:01:09	5	0	2
3	True	2017-12-02 00:01:39	5	0	2
4	False	2017-12-02 00:01:46	5	0	2
	Trans_min_mean	Trans_min_std	TransactionAmt_to_mean_card1		
0	-66.5	-0.278076	0.194580		
1	-106.0	-0.443359	0.123779		
2	-76.0	-0.317871	0.607910		
3	-85.0	-0.355469	0.405029		
4	-85.0	-0.355469	0.515625		
	TransactionAmt_to_mean_card4	TransactionAmt_to_std_card1			
0	0.257812	0.184560			
1	0.218994	0.063004			
2	0.443115	0.589226			
3	0.377686	0.259460			
4	0.377686	0.882898			

```

    TransactionAmt_to_std_card4  PCA_V_0  PCA_V_1  PCA_V_2  PCA_V_3  \
0          0.170288 -0.157349  0.918945 -0.843750  0.307861
1          0.114197 -0.086426 -0.800293 -0.152222 -0.363281
2          0.258545 -0.800781  0.316895  0.273193 -0.026337
3          0.196899 -0.237305 -0.811523 -0.123718 -0.423584
4          0.196899  2.904297  0.380127  0.480469 -0.009628

    PCA_V_4  PCA_V_5  PCA_V_6  PCA_V_7  PCA_V_8  PCA_V_9  PCA_V_10  \
0 -0.089417  0.003073 -0.020004 -0.187622  0.038269  0.002508 -0.010406
1 -0.101929 -0.002268  0.032410 -0.068970  0.040070 -0.180054 -0.059326
2  0.043182 -0.008049 -0.039246 -0.217041  0.017593  0.033508 -0.000374
3 -0.067444  0.025040  0.110596 -0.253906  0.004715  0.170288 -0.012352
4 -0.171631  1.170898 -0.178345  0.004597  0.043732 -0.001608  0.015762

    PCA_V_11  PCA_V_12  PCA_V_13  PCA_V_14  PCA_V_15  PCA_V_16  PCA_V_17  \
0  0.034119 -0.044281 -0.089722  0.045013  0.001526 -0.003292  0.018463
1  0.002270  0.018936 -0.029465  0.016541 -0.006153 -0.004082  0.010201
2 -0.015335  0.020706 -0.046051 -0.006680  0.004841  0.001108  0.007488
3 -0.014534  0.005116  0.031982 -0.013611 -0.017181 -0.001876 -0.084717
4 -0.017914 -0.020660 -0.005581 -0.010246 -0.003710 -0.011406 -0.008102

    PCA_V_18  PCA_V_19  PCA_V_20  PCA_V_21  PCA_V_22  PCA_V_23  PCA_V_24  \
0 -0.018341  0.009972 -0.026993 -0.021591 -0.054565  0.025299  0.018677
1 -0.001767 -0.022842  0.006413 -0.021011  0.054779 -0.042572 -0.026566
2 -0.007820 -0.006615 -0.008324  0.009743 -0.007683  0.003206  0.002855
3  0.050446  0.140747  0.059052 -0.020187  0.066467 -0.010803 -0.017715
4  0.013367  0.017807 -0.013626  0.000090 -0.005260 -0.141479  0.203003

    PCA_V_25  PCA_V_26  PCA_V_27  PCA_V_28  PCA_V_29
0 -0.006100  0.004215 -0.042999  0.008232 -0.007919
1  0.003536  0.001620  0.001561 -0.003544 -0.003748
2  0.001932  0.003839  0.002985 -0.019638 -0.003443
3  0.025543  0.003864  0.003452  0.026138 -0.041901
4  0.014557  0.012848  0.002218  0.014000 -0.001691

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[5 rows x 537 columns]

```
[77]: # Label encode the variables
for col in cat_cols:
    lbl = LabelEncoder()
    lbl.fit(list(df[col].values))
    df[col] = lbl.transform(list(df[col].values))
```

```
[78]: df = reduce_mem_usage(df)
```

Mem. usage decreased to 384.45 Mb (81.5% reduction)

10 10. Data Preprocessing for Model Building

```
[79]: df.loc[:, 'isFraud'].value_counts()
```

```
[79]: 0    569877  
1    20663  
Name: isFraud, dtype: int64
```

```
[80]: df = df.drop(['TransactionID', 'TransactionDT', 'Date'], axis=1)
```

```
[81]: # Split the y variable series and x variables dataset  
X = df.drop(['isFraud'], axis=1)  
y = df.isFraud.astype(bool)  
  
# Delete train df  
del df  
  
# Collect garbage  
gc.collect()
```

```
[81]: 0
```

```
[82]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3,  
random_state = 0, stratify=y)
```

```
[83]: X_train.head()
```

```
[83]: TransactionAmt  ProductCD  card1  card2  card3  card4  card5  card6  \\\n448539      4.679688      4   6598  111.0  150.0     2  195.0     2  
321311      4.355469      4  12839  321.0  150.0     4  226.0     2  
497320      3.892578      4  14649  548.0  150.0     4  226.0     2  
350951      4.058594      4   6489  295.0  150.0     4  226.0     2  
98132       5.296875      2   5714  170.0  150.0     4  195.0     1  
  
      addr1  addr2  dist1  P_emaildomain  R_emaildomain      C1      C2      C3  \\\n448539  264.0   87.0    6.0           0          2    2.0    4.0    0.0  
321311  264.0   87.0    0.0           0          2    1.0    1.0    0.0  
497320  441.0   87.0   86.0           4          2    2.0    4.0    0.0  
350951  184.0   87.0    NaN           0          2  154.0  148.0    0.0  
98132   498.0   87.0    NaN           5          6    1.0    1.0    0.0  
  
      C4      C5      C6      C7      C8      C9      C10      C11      C12      C13      C14  \\\n448539  0.0    1.0    1.0    0.0    0.0    1.0    0.0    2.0    0.0    7.0    2.0  
321311  0.0    2.0    1.0    0.0    0.0    1.0    0.0    1.0    0.0   12.0    1.0  
497320  0.0    0.0    2.0    0.0    0.0    3.0    0.0    4.0    0.0   45.0    2.0  
350951  0.0  108.0  101.0    0.0    0.0   92.0    0.0  107.0    0.0  504.0  129.0  
98132   1.0    0.0    1.0    0.0    1.0    0.0    1.0    1.0    0.0    1.0    1.0
```

	D1	D2	D3	D4	D5	D6	D8	D9	D10	D11	\										
448539	64.0	64.0	18.0	63.0	10.0	NaN	NaN	NaN	63.0	35.0											
321311	114.0	114.0	14.0	115.0	14.0	NaN	NaN	NaN	115.0	45.0											
497320	85.0	85.0	3.0	329.0	3.0	NaN	NaN	NaN	329.0	85.0											
350951	577.0	577.0	54.0	54.0	54.0	NaN	NaN	NaN	577.0	NaN											
98132	0.0	NaN	NaN	NaN	NaN	NaN	7.917969	0.916504	NaN	NaN											
	D12	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	M9	id_01	\						
448539	NaN	NaN	NaN	63.0	1	1	1	3	2	0	0	1	1	NaN							
321311	NaN	NaN	NaN	115.0	1	1	1	3	2	1	1	1	1	NaN							
497320	NaN	NaN	NaN	85.0	1	1	1	0	0	0	0	0	0	NaN							
350951	NaN	NaN	NaN	110.0	2	2	2	3	2	0	2	2	2	NaN							
98132	NaN	NaN	NaN	NaN	2	2	2	3	2	2	2	2	2	-5.0							
	id_02	id_03	id_04	id_05	id_06	id_09	id_10	id_11	id_12	id_13	id_14	id_15	id_16	id_17	id_19	id_20	id_28	id_29	id_30	\	
448539	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1		
321311	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1		
497320	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1		
350951	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1		
98132	105681.0	0.0	0.0	2.0	0.0	0.0	0.0	0.0	0.0	100.0									2		
	id_13	id_14	id_15	id_16	id_17	id_19	id_20	id_28	id_29	id_30	id_31	id_32	id_33	id_34	id_35	id_36	id_37	id_38	DeviceType	\	
448539	NaN	NaN	1	1	NaN	NaN	NaN	1	1	3	103	NaN	74	4	2	2	2	2	2		
321311	NaN	NaN	1	1	NaN	NaN	NaN	1	1	3	103	NaN	74	4	2	2	2	2	2		
497320	NaN	NaN	1	1	NaN	NaN	NaN	1	1	3	103	NaN	74	4	2	2	2	2	2		
350951	NaN	NaN	1	1	NaN	NaN	NaN	1	1	3	103	NaN	74	4	2	2	2	2	2		
98132	52.0	-360.0	0	0	166.0	621.0	272.0	0	0	4	84	24.0	72	3	1	0	1	1	0		
	DeviceInfo	card2_missing_flag	card3_missing_flag	card4_missing_flag	card5_missing_flag	card6_missing_flag															
448539	141	0	0	0	0	0															
321311	141	0	0	0	0	0															
497320	141	0	0	0	0	0															
350951	141	0	0	0	0	0															
98132	140	0	0	0	0	0															

350951	0	0	0	
98132	0	0	0	
	addr1_missing_flag	addr2_missing_flag	dist1_missing_flag	
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	1	
98132	0	0	1	
	dist2_missing_flag	P_emaildomain_missing_flag		
448539	1	0		
321311	1	0		
497320	1	0		
350951	1	0		
98132	1	0		
	R_emaildomain_missing_flag	D1_missing_flag	D2_missing_flag	
448539	1	0	0	
321311	1	0	0	
497320	1	0	0	
350951	1	0	0	
98132	0	0	1	
	D3_missing_flag	D4_missing_flag	D5_missing_flag	D6_missing_flag
448539	0	0	0	1
321311	0	0	0	1
497320	0	0	0	1
350951	0	0	0	1
98132	1	1	1	1
	D7_missing_flag	D8_missing_flag	D9_missing_flag	D10_missing_flag
448539	1	1	1	0
321311	1	1	1	0
497320	1	1	1	0
350951	1	1	1	0
98132	1	0	0	1
	D11_missing_flag	D12_missing_flag	D13_missing_flag	
448539	0	1	1	
321311	0	1	1	
497320	0	1	1	
350951	1	1	1	
98132	1	1	1	
	D14_missing_flag	D15_missing_flag	M1_missing_flag	M2_missing_flag
448539	1	0	0	0

321311	1	0	0	0
497320	1	0	0	0
350951	1	0	1	1
98132	1	1	1	1
	M3_missing_flag	M4_missing_flag	M5_missing_flag	M6_missing_flag
448539	0	1	1	0
321311	0	1	1	0
497320	0	0	0	0
350951	1	1	1	0
98132	1	1	1	1
	M7_missing_flag	M8_missing_flag	M9_missing_flag	V1_missing_flag
448539	0	0	0	0
321311	0	0	0	0
497320	0	0	0	0
350951	1	1	1	1
98132	1	1	1	1
	V2_missing_flag	V3_missing_flag	V4_missing_flag	V5_missing_flag
448539	0	0	0	0
321311	0	0	0	0
497320	0	0	0	0
350951	1	1	1	1
98132	1	1	1	1
	V6_missing_flag	V7_missing_flag	V8_missing_flag	V9_missing_flag
448539	0	0	0	0
321311	0	0	0	0
497320	0	0	0	0
350951	1	1	1	1
98132	1	1	1	1
	V10_missing_flag	V11_missing_flag	V12_missing_flag	
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	1	1	0	
98132	1	1	1	
	V13_missing_flag	V14_missing_flag	V15_missing_flag	
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	

	V16_missing_flag	V17_missing_flag	V18_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V19_missing_flag	V20_missing_flag	V21_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V22_missing_flag	V23_missing_flag	V24_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V25_missing_flag	V26_missing_flag	V27_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V28_missing_flag	V29_missing_flag	V30_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V31_missing_flag	V32_missing_flag	V33_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V34_missing_flag	V35_missing_flag	V36_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	

98132	1	1	1
	V37_missing_flag	V38_missing_flag	V39_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V40_missing_flag	V41_missing_flag	V42_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V43_missing_flag	V44_missing_flag	V45_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V46_missing_flag	V47_missing_flag	V48_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V49_missing_flag	V50_missing_flag	V51_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V52_missing_flag	V53_missing_flag	V54_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V55_missing_flag	V56_missing_flag	V57_missing_flag \
448539	0	0	0
321311	0	0	0

497320	0	0	0
350951	0	0	0
98132	1	1	1
	V58_missing_flag	V59_missing_flag	V60_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V61_missing_flag	V62_missing_flag	V63_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V64_missing_flag	V65_missing_flag	V66_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V67_missing_flag	V68_missing_flag	V69_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V70_missing_flag	V71_missing_flag	V72_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V73_missing_flag	V74_missing_flag	V75_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V76_missing_flag	V77_missing_flag	V78_missing_flag \

448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V79_missing_flag	V80_missing_flag	V81_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V82_missing_flag	V83_missing_flag	V84_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V85_missing_flag	V86_missing_flag	V87_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V88_missing_flag	V89_missing_flag	V90_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V91_missing_flag	V92_missing_flag	V93_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
	V94_missing_flag	V95_missing_flag	V96_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	0	0

	V97_missing_flag	V98_missing_flag	V99_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V100_missing_flag	V101_missing_flag	V102_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V103_missing_flag	V104_missing_flag	V105_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V106_missing_flag	V107_missing_flag	V108_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V109_missing_flag	V110_missing_flag	V111_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V112_missing_flag	V113_missing_flag	V114_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V115_missing_flag	V116_missing_flag	V117_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	

350951	0	0	0
98132	0	0	0
	V118_missing_flag	V119_missing_flag	V120_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V121_missing_flag	V122_missing_flag	V123_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V124_missing_flag	V125_missing_flag	V126_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V127_missing_flag	V128_missing_flag	V129_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V130_missing_flag	V131_missing_flag	V132_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V133_missing_flag	V134_missing_flag	V135_missing_flag
448539	0	0	0 ...
321311	0	0	0 ...
497320	0	0	0 ...
350951	0	0	0 ...
98132	0	0	0 ...
	V169_missing_flag	V170_missing_flag	V171_missing_flag
448539	1	1	1

321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V172_missing_flag	V173_missing_flag	V174_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V175_missing_flag	V176_missing_flag	V177_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V178_missing_flag	V179_missing_flag	V180_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V181_missing_flag	V182_missing_flag	V183_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V184_missing_flag	V185_missing_flag	V186_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V187_missing_flag	V188_missing_flag	V189_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	V190_missing_flag	V191_missing_flag	V192_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V193_missing_flag	V194_missing_flag	V195_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V196_missing_flag	V197_missing_flag	V198_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V199_missing_flag	V200_missing_flag	V201_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V202_missing_flag	V203_missing_flag	V204_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V205_missing_flag	V206_missing_flag	V207_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V208_missing_flag	V209_missing_flag	V210_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	

98132	0	0	0	
	V211_missing_flag	V212_missing_flag	V213_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V214_missing_flag	V215_missing_flag	V216_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V217_missing_flag	V218_missing_flag	V219_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V220_missing_flag	V221_missing_flag	V222_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V223_missing_flag	V224_missing_flag	V225_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V226_missing_flag	V227_missing_flag	V228_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V229_missing_flag	V230_missing_flag	V231_missing_flag	\
448539	1	1	1	
321311	1	1	1	

497320	1	1	1
350951	1	1	1
98132	0	0	0
	V232_missing_flag	V233_missing_flag	V234_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V235_missing_flag	V236_missing_flag	V237_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V238_missing_flag	V239_missing_flag	V240_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V241_missing_flag	V242_missing_flag	V243_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V244_missing_flag	V245_missing_flag	V246_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V247_missing_flag	V248_missing_flag	V249_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V250_missing_flag	V251_missing_flag	V252_missing_flag

448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V253_missing_flag	V254_missing_flag	V255_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V256_missing_flag	V257_missing_flag	V258_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V259_missing_flag	V260_missing_flag	V261_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V262_missing_flag	V263_missing_flag	V264_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V265_missing_flag	V266_missing_flag	V267_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V268_missing_flag	V269_missing_flag	V270_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	V271_missing_flag	V272_missing_flag	V273_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V274_missing_flag	V275_missing_flag	V276_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V277_missing_flag	V278_missing_flag	V279_missing_flag	\
448539	1	1	0	
321311	1	1	0	
497320	1	1	0	
350951	1	1	0	
98132	0	0	0	

	V280_missing_flag	V281_missing_flag	V282_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V283_missing_flag	V284_missing_flag	V285_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V286_missing_flag	V287_missing_flag	V288_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	

	V289_missing_flag	V290_missing_flag	V291_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	

350951	0	0	0
98132	0	0	0
	V292_missing_flag	V293_missing_flag	V294_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V295_missing_flag	V296_missing_flag	V297_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V298_missing_flag	V299_missing_flag	V300_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V301_missing_flag	V302_missing_flag	V303_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V304_missing_flag	V305_missing_flag	V306_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V307_missing_flag	V308_missing_flag	V309_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V310_missing_flag	V311_missing_flag	V312_missing_flag
448539	0	0	0

321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V313_missing_flag	V314_missing_flag	V315_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V316_missing_flag	V317_missing_flag	V318_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V319_missing_flag	V320_missing_flag	V321_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V322_missing_flag	V323_missing_flag	V324_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V325_missing_flag	V326_missing_flag	V327_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V328_missing_flag	V329_missing_flag	V330_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	V331_missing_flag	V332_missing_flag	V333_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V334_missing_flag	V335_missing_flag	V336_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V337_missing_flag	V338_missing_flag	V339_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	id_01_missing_flag	id_02_missing_flag	id_03_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	id_04_missing_flag	id_05_missing_flag	id_06_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	id_07_missing_flag	id_08_missing_flag	id_09_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	1	1	0	
	id_10_missing_flag	id_11_missing_flag	id_12_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	

98132	0	0	0
	id_13_missing_flag	id_14_missing_flag	id_15_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	id_16_missing_flag	id_17_missing_flag	id_18_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	1
	id_19_missing_flag	id_20_missing_flag	id_21_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	1
	id_22_missing_flag	id_23_missing_flag	id_24_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	1	1	1
	id_25_missing_flag	id_26_missing_flag	id_27_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	1	1	1
	id_28_missing_flag	id_29_missing_flag	id_30_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	id_31_missing_flag	id_32_missing_flag	id_33_missing_flag
448539	1	1	1
321311	1	1	1

497320	1	1	1					
350951	1	1	1					
98132	0	0	0					
	id_34_missing_flag	id_35_missing_flag	id_36_missing_flag	\				
448539	1	1	1					
321311	1	1	1					
497320	1	1	1					
350951	1	1	1					
98132	0	0	0					
	id_37_missing_flag	id_38_missing_flag	DeviceType_missing_flag	\				
448539	1	1	1					
321311	1	1	1					
497320	1	1	1					
350951	1	1	1					
98132	0	0	0					
	DeviceInfo_missing_flag	_Weekdays	_Hours	_Days	Trans_min_mean	\		
448539	1	3	3	12	-27.09375			
321311	1	5	14	3	-57.09375			
497320	1	0	19	30	-86.00000			
350951	1	6	1	11	-77.06250			
98132	0	5	22	23	65.00000			
	Trans_min_std	TransactionAmt_to_mean_card1	\					
448539	-0.113281	0.461914						
321311	-0.238770	0.645508						
497320	-0.359619	0.470703						
350951	-0.322266	0.555176						
98132	0.271729	1.193359						
	TransactionAmt_to_mean_card4	TransactionAmt_to_std_card1	\					
448539	0.815430	0.313331						
321311	0.585449	0.401900						
497320	0.367920	0.326992						
350951	0.435059	0.570935						
98132	1.501953	0.718662						
	TransactionAmt_to_std_card4	PCA_V_0	PCA_V_1	PCA_V_2	PCA_V_3	\		
448539	0.425049	-0.782227	0.368896	0.263428	0.007843			
321311	0.341553	-0.782227	0.368896	0.263428	0.007904			
497320	0.214722	-0.801758	0.317871	0.270020	-0.027130			
350951	0.253906	-0.304443	-0.825684	-0.107727	-0.468994			
98132	0.876465	2.806641	0.345703	0.377197	-0.017654			
	PCA_V_4	PCA_V_5	PCA_V_6	PCA_V_7	PCA_V_8	PCA_V_9	PCA_V_10	\

```

448539  0.013977 -0.007053  0.002455  0.334717 -0.010490 -0.220459 -0.042755
321311  0.013786 -0.007095  0.002304  0.334961 -0.010521 -0.220825 -0.042938
497320  0.058411 -0.003780 -0.014931 -0.225342  0.021576  0.033905 -0.024246
350951  -0.162598  0.029373  0.138550  0.065613 -0.067749  0.260010  0.052063
98132   -0.042450 -0.244263  0.023148  0.011810  0.000937  0.023010 -0.009560

    PCA_V_11  PCA_V_12  PCA_V_13  PCA_V_14  PCA_V_15  PCA_V_16  PCA_V_17 \
448539  0.021561 -0.005577 -0.020416  0.005253 -0.000285  0.004505 -0.006889
321311  0.020325 -0.003403 -0.022324  0.005749  0.000531  0.004292 -0.009888
497320  0.010124  0.020370  0.017105 -0.014145 -0.013794 -0.005302 -0.077637
350951  -0.091125  0.068176  0.014664  0.010857 -0.011421 -0.020767  0.009926
98132   -0.022293 -0.021683 -0.013680 -0.005581 -0.009674  0.001465 -0.008263

    PCA_V_18  PCA_V_19  PCA_V_20  PCA_V_21  PCA_V_22  PCA_V_23  PCA_V_24 \
448539  0.008362 -0.011055 -0.006924 -0.033722  0.011093 -0.011307 -0.007355
321311  0.009216 -0.005642 -0.005966 -0.034119  0.012009 -0.009644 -0.006516
497320  0.056580  0.115479  0.013458 -0.012100  0.003590  0.023392  0.003922
350951  0.008850  0.005054 -0.041290  0.041382 -0.051819  0.042450  0.026703
98132   -0.008881  0.005920 -0.004478 -0.000713 -0.003057  0.008438 -0.017868

    PCA_V_25  PCA_V_26  PCA_V_27  PCA_V_28  PCA_V_29
448539  -0.004627  0.002234 -0.005222  0.004154 -0.001096
321311  -0.002327  0.000332 -0.006203  0.002287 -0.005352
497320  0.032867  0.010475  0.007317  0.008362 -0.037079
350951  -0.046417 -0.029648  0.022217  0.004562 -0.008606
98132   0.000072  0.013344 -0.005264 -0.000665  0.007454

[5 rows x 533 columns]

```

11 11. Model Building

12 XGBoost Classifier

[]:

```
[84]: X_train = X_train.replace([np.inf, -np.inf], np.nan)
X_test = X_test.replace([np.inf, -np.inf], np.nan)

median_values = X_train.median()
X_train = X_train.fillna(median_values)
X_test = X_test.fillna(median_values)
```

```
[85]: %%time
xgb = XGBClassifier(nthread=-1, random_state=0)
xgb.fit(X_train, y_train)
xgb
```

```
[13:34:41] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release_1.4.0/src/learner.cc:1095: Starting in XGBoost 1.3.0, the default evaluation metric used with the objective 'binary:logistic' was changed from 'error' to 'logloss'. Explicitly set eval_metric if you'd like to restore the old behavior.
```

```
Wall time: 7min 4s
```

```
[85]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
                   colsample_bynode=1, colsample_bytree=1, gamma=0, gpu_id=-1,
                   importance_type='gain', interaction_constraints='',
                   learning_rate=0.300000012, max_delta_step=0, max_depth=6,
                   min_child_weight=1, missing=nan, monotone_constraints='()',
                   n_estimators=100, n_jobs=8, nthread=-1, num_parallel_tree=1,
                   random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
                   subsample=1, tree_method='exact', validate_parameters=1,
                   verbosity=None)
```

```
[86]: y_pred_xgb = xgb.predict(X_test)
y_prob_pred_xgb = xgb.predict_proba(X_test)
y_prob_pred_xgb = [x[1] for x in y_prob_pred_xgb]
print("Y predicted : ",y_pred_xgb)
print("Y probability predicted : ",y_prob_pred_xgb[:5])
```

```
Y predicted :  [False False False ... False False False]
Y probability predicted :  [0.0032196122, 0.0040678284, 0.007803707,
 0.023133403, 0.0024363862]
```

13 12. Evaluation Metrics

- Accuracy Score
- Confusion Matrix
- Classification Report
- AUC Score
- Concordance Index
- ROC curve
- PR curve

```
[87]: from bisect import bisect_left, bisect_right

def concordance(actuals, preds):
    ones_preds = [p for a,p in zip(actuals, preds) if a == 1]
    zeros_preds = [p for a,p in zip(actuals, preds) if a == 0]
    n_ones = len([x for x in actuals if x == 1])
    n_total_pairs = float(n_ones) * float(len(actuals) - n_ones)
    # print("Total Pairs: ", n_total_pairs)

    zeros_sorted = sorted(zeros_preds)
```

```

conc = 0; disc = 0; ties = 0;
for i, one_pred in enumerate(ones_preds):
    cur_conc = bisect_left(zeros_sorted, one_pred)
    cur_ties = bisect_right(zeros_sorted, one_pred) - cur_conc
    conc += cur_conc
    ties += cur_ties
    disc += float(len(zeros_sorted)) - cur_ties - cur_conc

concordance = conc/n_total_pairs
discordance = disc/n_total_pairs
ties_perc = ties/n_total_pairs
return concordance

```

```

[88]: def compute_evaluation_metric(model, x_test, y_actual, y_predicted, y_predicted_prob):
    print("\nAccuracy Score : ", accuracy_score(y_actual, y_predicted))
    print("\nAUC Score : ", roc_auc_score(y_actual, y_predicted_prob))
    print("\nConfusion Matrix : \n", confusion_matrix(y_actual, y_predicted))
    print("\nClassification Report : \n", classification_report(y_actual, y_predicted))

    # Concordance index if function exists
    try:
        print("\nConcordance Index : ", concordance(y_actual, y_predicted_prob))
    except:
        print("\nConcordance Index : Function 'concordance' not defined")

    # ROC curve
    print("\nROC curve : \n")
    plot_roc_curve(model, x_test, y_actual)
    plt.show()

    # PR curve
    print("\nPR curve : \n")
    plot_precision_recall_curve(model, x_test, y_actual)
    plt.show()

    # --- Compute TPR, FPR, TNR, FNR ---
    tn, fp, fn, tp = confusion_matrix(y_actual, y_predicted).ravel()

    tpr = tp / (tp + fn)  # True Positive Rate / Recall
    fpr = fp / (fp + tn)  # False Positive Rate
    tnr = tn / (tn + fp)  # True Negative Rate / Specificity
    fnr = fn / (fn + tp)  # False Negative Rate

    print("\nAdditional Metrics:")
    print(f"TPR (Recall) : {tpr:.4f}")

```

```
print(f"FPR : {fpr:.4f}")
print(f"TNR (Specificity) : {tnr:.4f}")
print(f"FNR : {fnr:.4f})
```

[89]: concordance(y_test.values, y_prob_pred_xgb)

[89]: 0.9363753273299149

[90]: compute_evaluation_metric(xgb, X_test, y_test, y_pred_xgb, y_prob_pred_xgb)

Accuracy Score : 0.9801086011672933

AUC Score : 0.9363753419553209

Confusion Matrix :

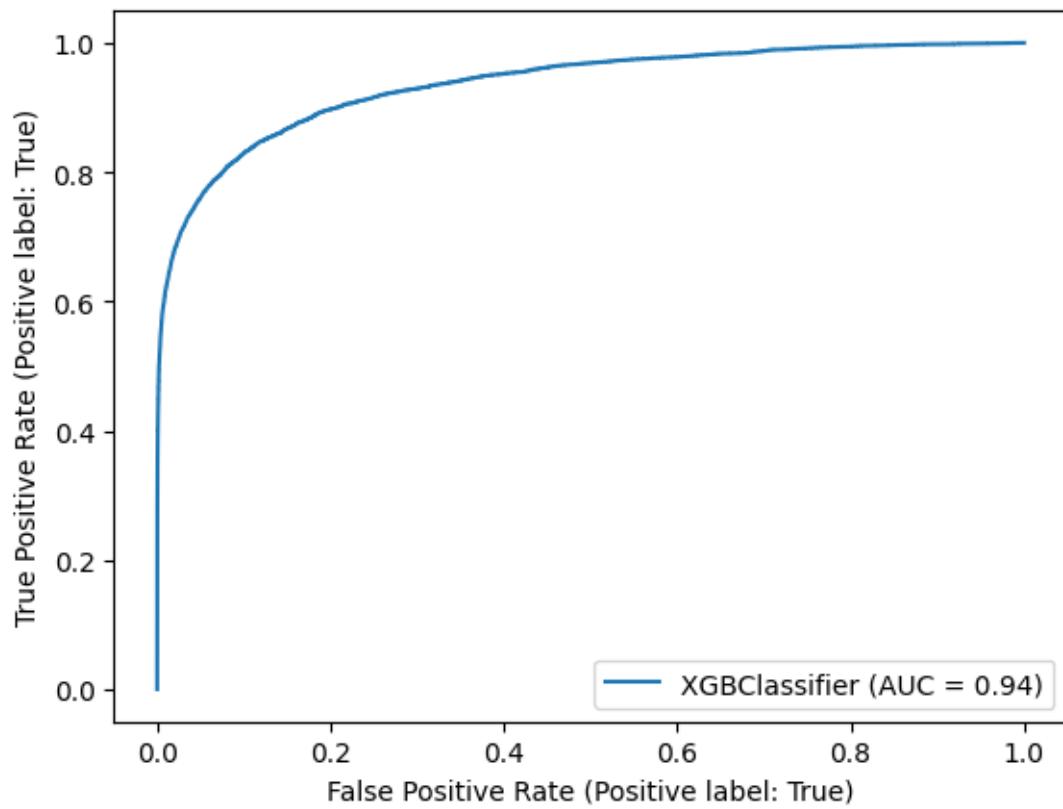
```
[[170646    317]
 [ 3207   2992]]
```

Classification Report :

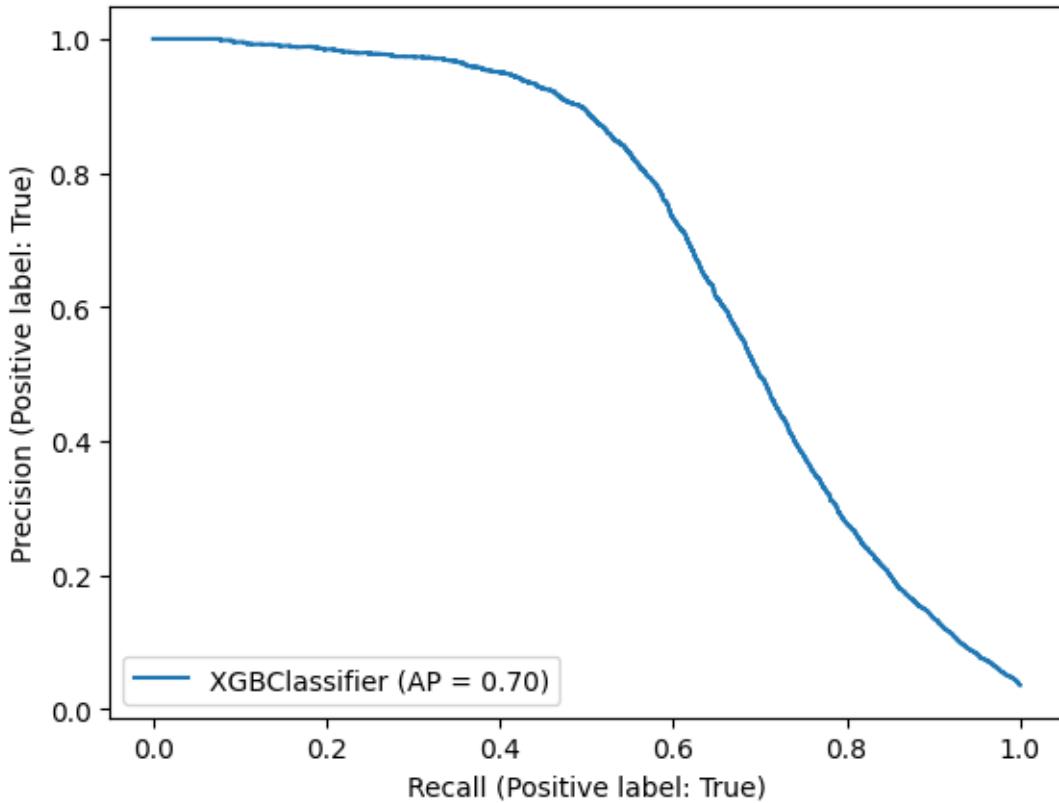
	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.90	0.48	0.63	6199
accuracy			0.98	177162
macro avg	0.94	0.74	0.81	177162
weighted avg	0.98	0.98	0.98	177162

Concordance Index : 0.9363753273299149

ROC curve :



PR curve :



Additional Metrics:

TPR (Recall) : 0.4827
 FPR : 0.0019
 TNR (Specificity) : 0.9981
 FNR : 0.5173

14 13. Capture Rates and Calibration Curve

Divide the data in 10 equal bins as per predicted probability scores. Then, compute the percentage of the total target class 1 captured in every bin.

Ideally the proportion should be decreasing as we go down ever bin. Let's check it out

```
[91]: # Create Validation set
validation_df = {'y_test' : y_test, 'y_pred' : y_pred_xgb, 'y_pred_prob' : y_prob_pred_xgb}
validation_df = pd.DataFrame(data = validation_df)

# Add binning column to the dataframe
validation_df['bin_y_pred_prob'] = pd.qcut(validation_df['y_pred_prob'], q=10)
```

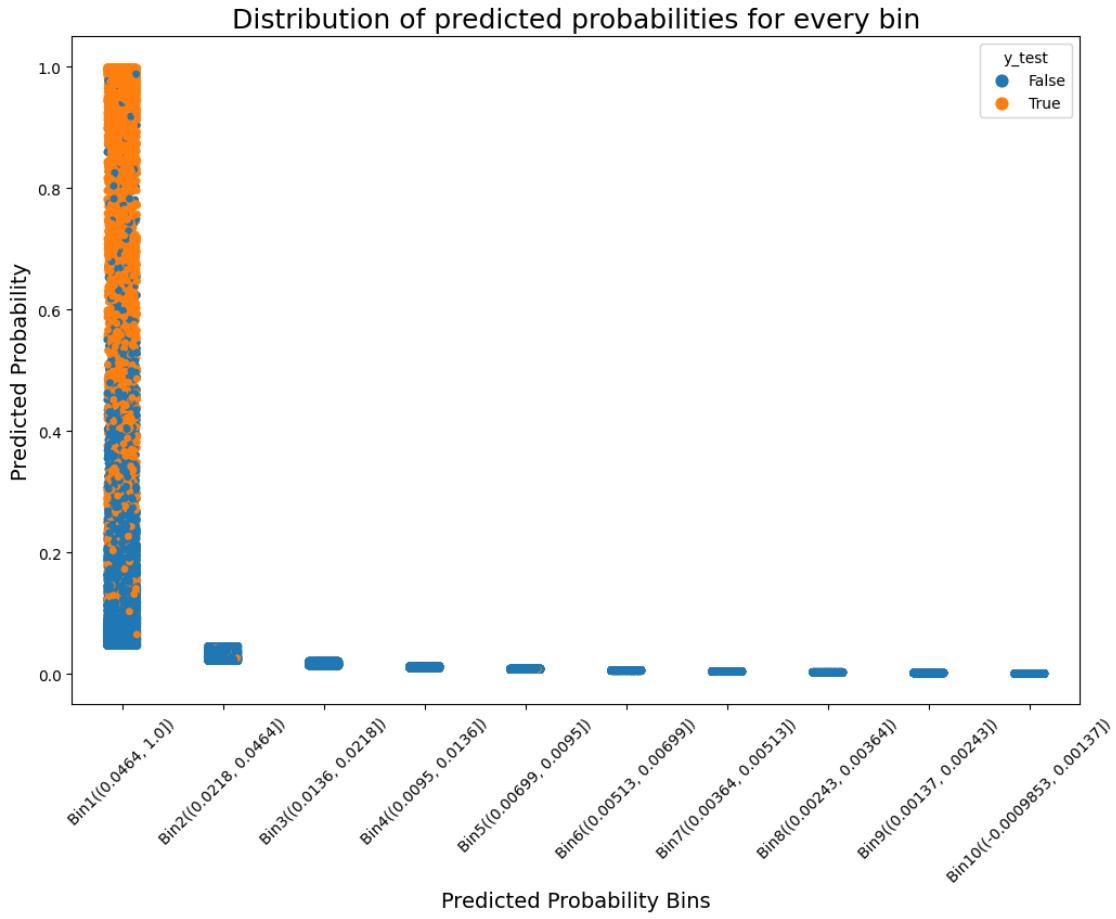
```
validation_df.head()
```

```
[91]:      y_test  y_pred  y_pred_prob    bin_y_pred_prob
274635    False   False    0.003220  (0.00243, 0.00364]
115501    False   False    0.004068  (0.00364, 0.00513]
387765    False   False    0.007804  (0.00699, 0.0095]
123591    False   False    0.023133  (0.0218, 0.0464]
462833    False   False    0.002436  (0.00243, 0.00364]
```

```
[92]: x_label = []
for i in range(len(validation_df['bin_y_pred_prob'].cat.categories[::-1].astype('str'))):
    x_label.append("Bin" + str(i + 1) + "(" + validation_df['bin_y_pred_prob'].cat.categories[::-1].astype('str')[i] + ")")
```

Capture Rates Plot

```
[93]: # Plot Distribution of predicted probabilities for every bin
plt.figure(figsize=(12, 8));
sns.stripplot(validation_df.bin_y_pred_prob, validation_df.y_pred_prob, jitter=True,
             hue = validation_df.y_test, order = validation_df['bin_y_pred_prob'].cat.categories[::-1])
plt.title("Distribution of predicted probabilities for every bin", fontsize=18)
plt.xlabel("Predicted Probability Bins", fontsize=14);
plt.ylabel("Predicted Probability", fontsize=14);
plt.xticks(np.arange(10), x_label, rotation=45);
plt.show()
```



Gains Table

```
[94]: # Aggregate the data
gains_df = validation_df.groupby(["bin_y_pred_prob", "y_test"]).
    agg({'y_test': ['count']})
gains_df.columns = gains_df.columns.map(''.join)
gains_df['prob_bin'] = gains_df.index.get_level_values(0)
gains_df['y_test'] = gains_df.index.get_level_values(1)
gains_df.reset_index(drop = True, inplace = True)
gains_df

# Get infection rate and percentage infections
gains_table = gains_df.pivot(index='prob_bin', columns='y_test', u
    values='y_testcount')
gains_table['prob_bin'] = gains_table.index
gains_table = gains_table.iloc[::-1]
gains_table['prob_bin'] = x_label
gains_table.reset_index(drop = True, inplace = True)
gains_table = gains_table[['prob_bin', 0, 1]]
```

```

gains_table.columns = ['prob_bin', "not_fraud", "fraud"]
gains_table['perc_fraud'] = gains_table['fraud']/gains_table['fraud'].sum()
gains_table['perc_not_fraud'] = gains_table['not_fraud']/
    ↪gains_table['not_fraud'].sum()
gains_table['cum_perc_fraud'] = 100*(gains_table.fraud.cumsum() / gains_table.
    ↪fraud.sum())
gains_table['cum_perc_not_fraud'] = 100*(gains_table.not_fraud.cumsum() / ↪
    ↪gains_table.not_fraud.sum())
gains_table

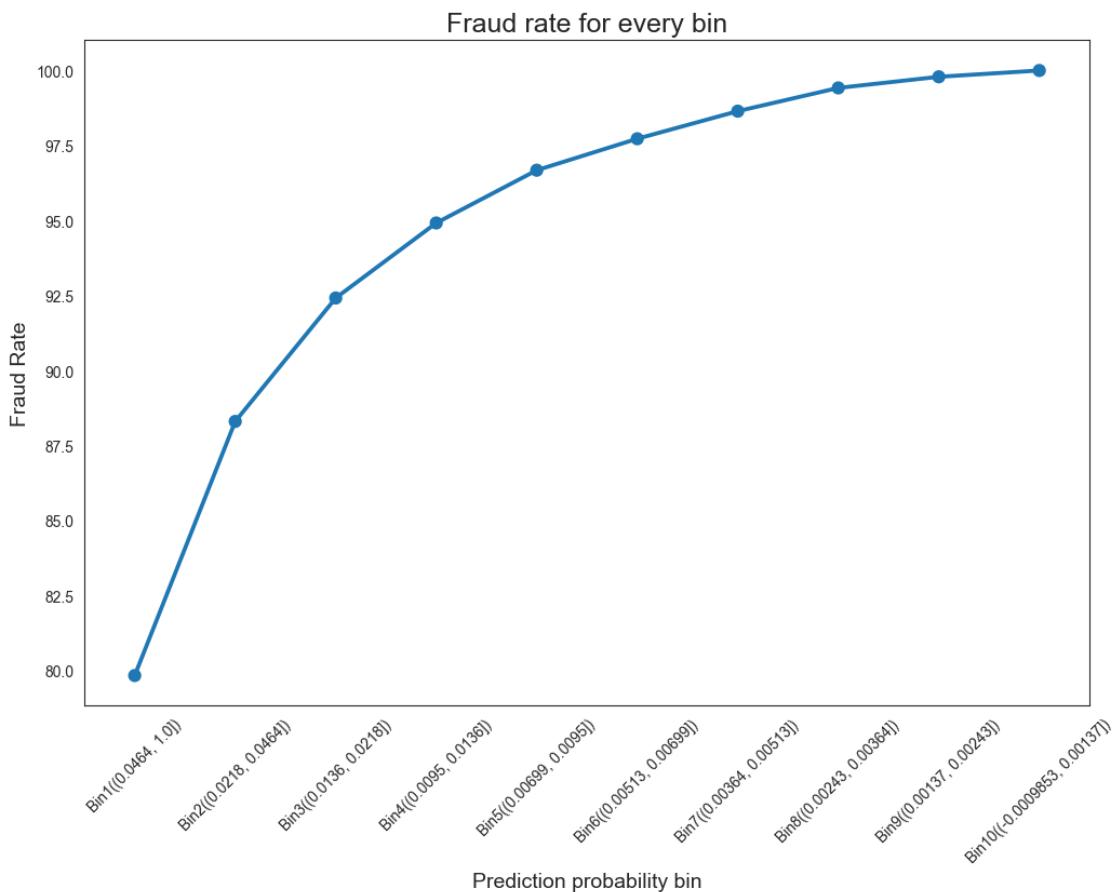
```

Plot

```

plt.figure(figsize=(12, 8));
sns.set_style("white")
sns.pointplot(x = "prob_bin", y = "cum_perc_fraud", data = gains_table, legend_=False, order=gains_table.prob_bin)
plt.xticks(rotation=45);
plt.ylabel("Fraud Rate", fontsize=14)
plt.xlabel("Prediction probability bin", fontsize=14)
plt.title("Fraud rate for every bin", fontsize=18)
plt.show()

```



Ideally the slope should be high initially and should decrease as we move further to the right. This is not really a good model.

```
[95]: def captures(y_test, y_pred, y_pred_prob):
    # Create Validation set
    validation_df = {'y_test' : y_test, 'y_pred' : y_pred, 'y_pred_prob' : y_pred_prob}
    validation_df = pd.DataFrame(data = validation_df)

    # Add binning column to the dataframe
    try:
        validation_df['bin_y_pred_prob'] = pd.qcut(validation_df['y_pred_prob'], q=10)
    except:
        validation_df['bin_y_pred_prob'] = pd.qcut(validation_df['y_pred_prob'], q=10, duplicates='drop')

    # Change x label and column names
    x_label = []
    for i in range(len(validation_df['bin_y_pred_prob'].cat.categories[::-1].astype('str'))):
        x_label.append("Bin" + str(i + 1) + "(" + validation_df['bin_y_pred_prob'].cat.categories[::-1].astype('str')[i] + ")")

    # Plot Distribution of predicted probabilities for every bin
    plt.figure(figsize=(12, 8));
    sns.stripplot(validation_df.bin_y_pred_prob, validation_df.y_pred_prob, jitter = 0.15, hue = validation_df.y_test, order = validation_df['bin_y_pred_prob'].cat.categories[::-1])
    plt.title("Distribution of predicted probabilities for every bin", fontsize=18)
    plt.xlabel("Predicted Probability Bins", fontsize=14);
    plt.ylabel("Predicted Probability", fontsize=14);
    try:
        plt.xticks(np.arange(10), x_label, rotation=45);
    except:
        pass
    plt.show()

    # Aggregate the data
    gains_df = validation_df.groupby(["bin_y_pred_prob", "y_test"]).
    agg({'y_test': ['count']})
    gains_df.columns = gains_df.columns.map(''.join)
    gains_df['prob_bin'] = gains_df.index.get_level_values(0)
```

```

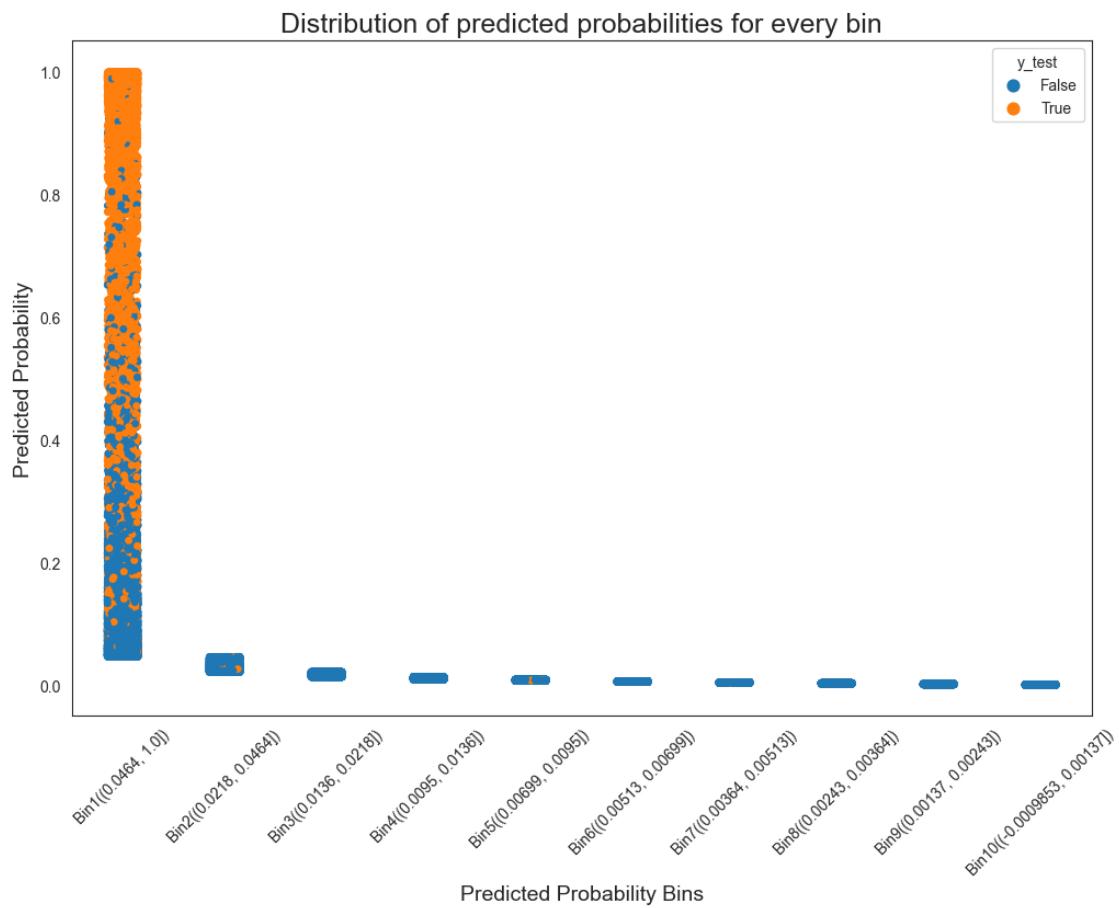
gains_df['y_test'] = gains_df.index.get_level_values(1)
gains_df.reset_index(drop = True, inplace = True)
gains_df

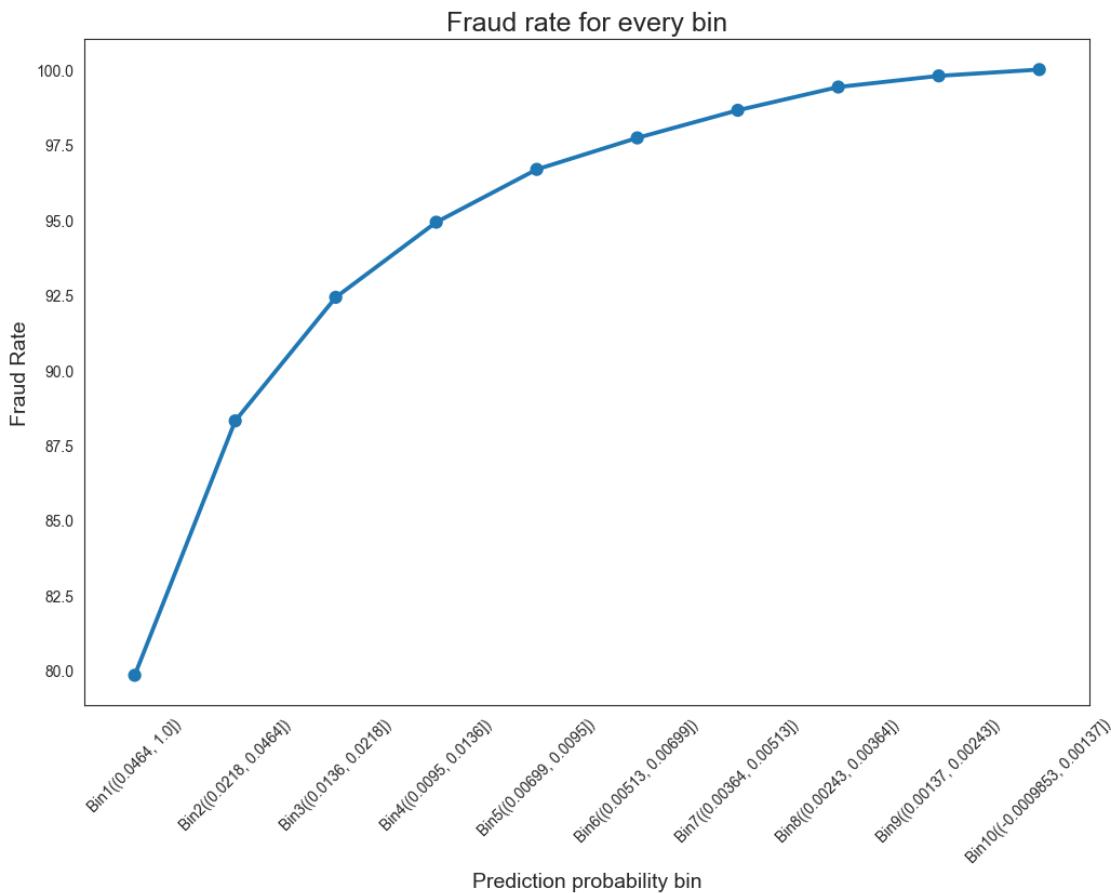
# Get infection rate and percentage infections
gains_table = gains_df.pivot(index='prob_bin', columns='y_test', u
values='y_testcount')
gains_table['prob_bin'] = gains_table.index
gains_table = gains_table.iloc[::-1]
gains_table['prob_bin'] = x_label
gains_table.reset_index(drop = True, inplace = True)
gains_table = gains_table[['prob_bin', 0, 1]]
gains_table.columns = ['prob_bin', "not_fraud", "fraud"]
gains_table['perc_fraud'] = gains_table['fraud']/gains_table['fraud'].sum()
gains_table['perc_not_fraud'] = gains_table['not_fraud']/
gains_table['not_fraud'].sum()
gains_table['cum_perc_fraud'] = 100*(gains_table.fraud.cumsum() / u
gains_table.fraud.sum())
gains_table['cum_perc_not_fraud'] = 100*(gains_table.not_fraud.cumsum() / u
gains_table.not_fraud.sum())
gains_table

# Plot
plt.figure(figsize=(12, 8));
sns.set_style("white")
sns.pointplot(x = "prob_bin", y = "cum_perc_fraud", data = gains_table,u
legend = False, order=gains_table.prob_bin)
plt.xticks(rotation=45);
plt.ylabel("Fraud Rate", fontsize=14)
plt.xlabel("Prediction probability bin", fontsize=14)
plt.title("Fraud rate for every bin", fontsize=18)
plt.show()
return gains_table

```

[96]: captures(y_test, y_pred_xgb, y_prob_pred_xgb)





[96] :

	prob_bin	not_fraud	fraud	perc_fraud	perc_not_fraud	\
0	Bin1((0.0464, 1.0])	12768	4949	0.798355	0.074683	
1	Bin2((0.0218, 0.0464])	17191	525	0.084691	0.100554	
2	Bin3((0.0136, 0.0218])	17461	255	0.041136	0.102133	
3	Bin4((0.0095, 0.0136])	17561	155	0.025004	0.102718	
4	Bin5((0.00699, 0.0095])	17607	109	0.017583	0.102987	
5	Bin6((0.00513, 0.00699])	17651	65	0.010486	0.103245	
6	Bin7((0.00364, 0.00513])	17659	57	0.009195	0.103291	
7	Bin8((0.00243, 0.00364])	17668	48	0.007743	0.103344	
8	Bin9((0.00137, 0.00243])	17693	23	0.003710	0.103490	
9	Bin10((-0.0009853, 0.00137])	17704	13	0.002097	0.103555	
	cum_perc_fraud	cum_perc_not_fraud				
0	79.835457	7.468283				
1	88.304565	17.523675				
2	92.418132	27.736996				
3	94.918535	38.008809				
4	96.676883	48.307529				
5	97.725440	58.631985				

```

6      98.644943      68.961120
7      99.419261      79.295520
8      99.790289      89.644543
9     100.000000     100.000000

```

14.1 Calibration Curve

```
[97]: from sklearn.calibration import calibration_curve
import matplotlib.pyplot as plt
```

```
[98]: def draw_calibration_curve(y_test, y_prob, n_bins=10):
    plt.figure(figsize=(7, 7), dpi=120)
    ax1 = plt.subplot2grid((3, 1), (0, 0), rowspan=2)
    ax2 = plt.subplot2grid((3, 1), (2, 0))
    ax1.plot([0, 1], [0, 1], "k:", label="Perfectly calibrated")

    fraction_of_positives, mean_predicted_value = calibration_curve(y_test, y_prob, n_bins=10)

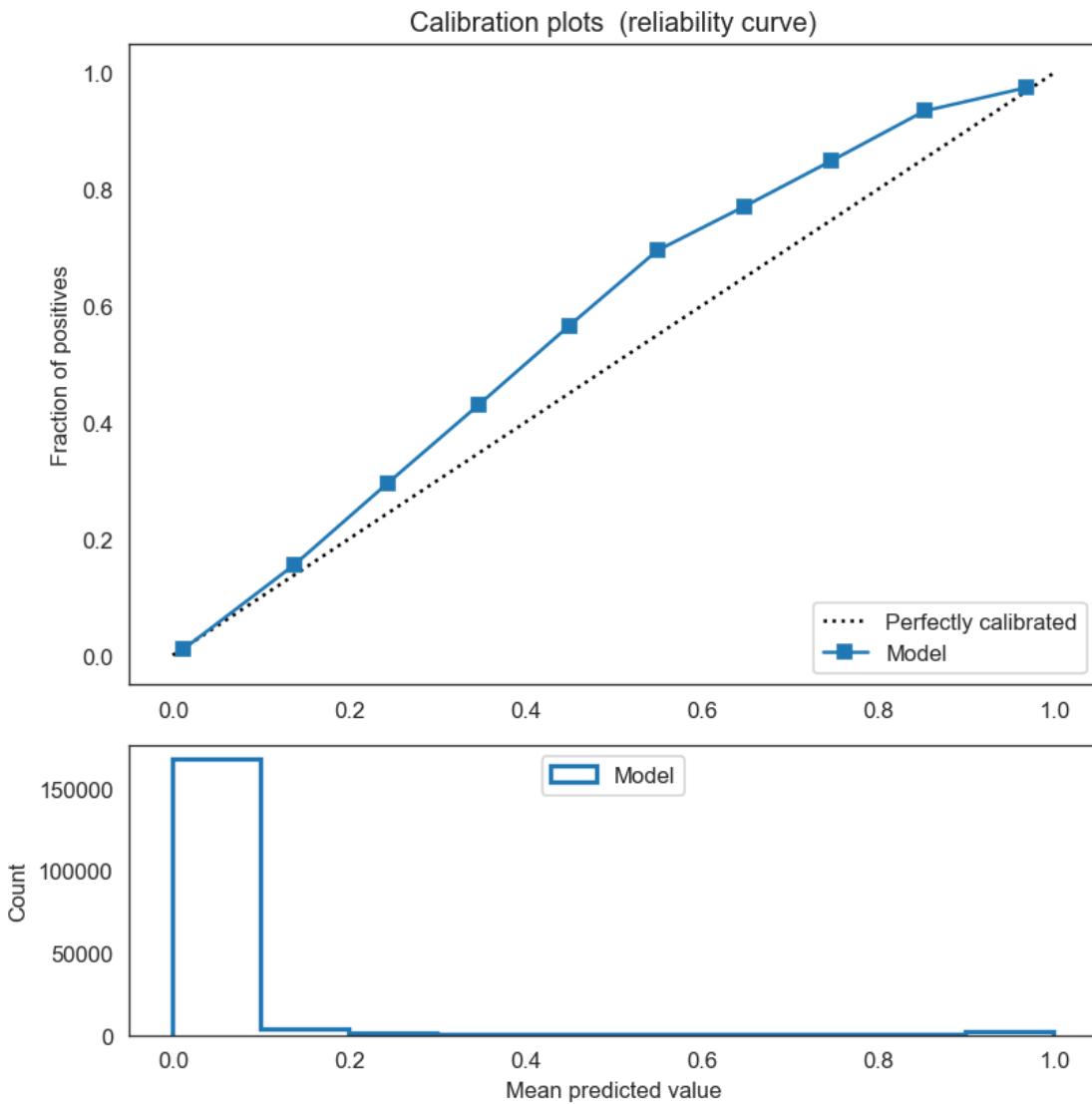
    ax1.plot(mean_predicted_value, fraction_of_positives, "s-", label="%s %s" % ("Model", ))
    ax2.hist(y_prob, range=(0, 1), bins=10, label="Model", histtype="step", lw=2)

    # Labels
    ax1.set_ylabel("Fraction of positives")
    ax1.set_ylim([-0.05, 1.05])
    ax1.legend(loc="lower right")
    ax1.set_title('Calibration plots (reliability curve)')

    ax2.set_xlabel("Mean predicted value")
    ax2.set_ylabel("Count")
    ax2.legend(loc="upper center", ncol=2)

    plt.tight_layout()
    plt.show()
```

```
[99]: draw_calibration_curve(y_test, y_prob_pred_xgb, n_bins=10)
```



15 14. Model Comparison and Evaluation

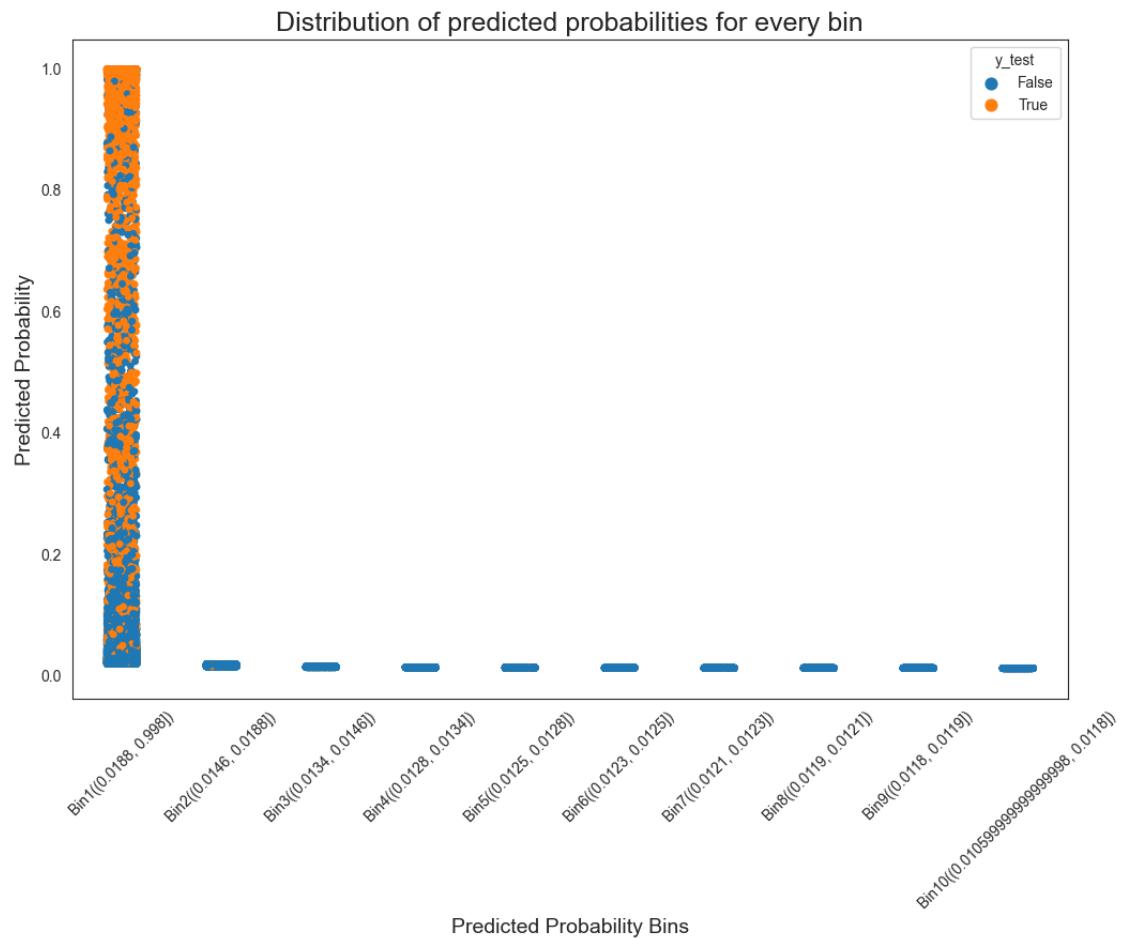
15.1 A) Logistic regression

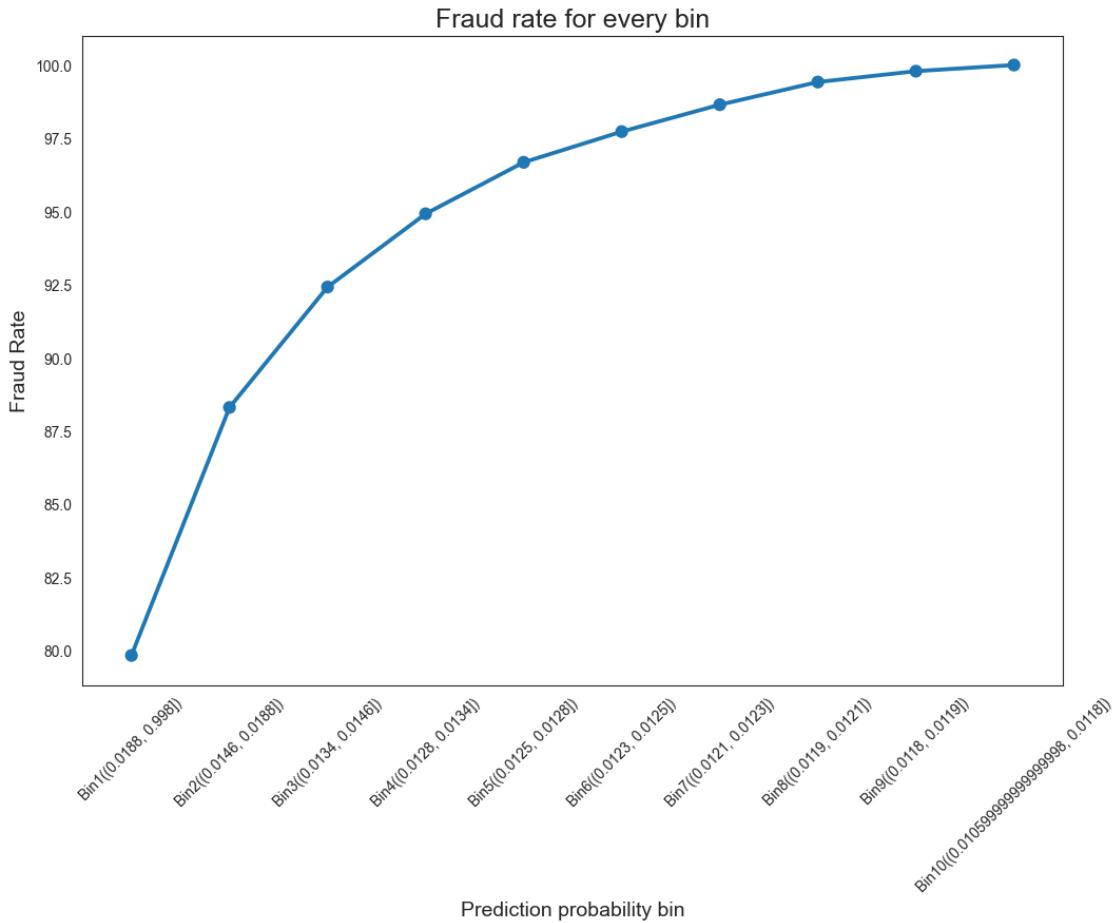
```
[100]: # Prediction
y_pred_xgb_test = xgb.predict(X_test)
y_prob_pred_xgb_test = xgb.predict_proba(X_test)[:, 1]
```

```
[101]: from sklearn.linear_model import LogisticRegression
X = np.array(y_prob_pred_xgb_test)
clf = LogisticRegression(random_state=0).fit(X.reshape(-1, 1), y_test)
```

```
[102]: y_prob_pred_calib = clf.predict_proba(X.reshape(-1, 1))[:, 1]
y_pred_calib       = clf.predict(X.reshape(-1, 1))
```

```
[103]: captures(y_test, y_pred_calib, y_prob_pred_calib)
```



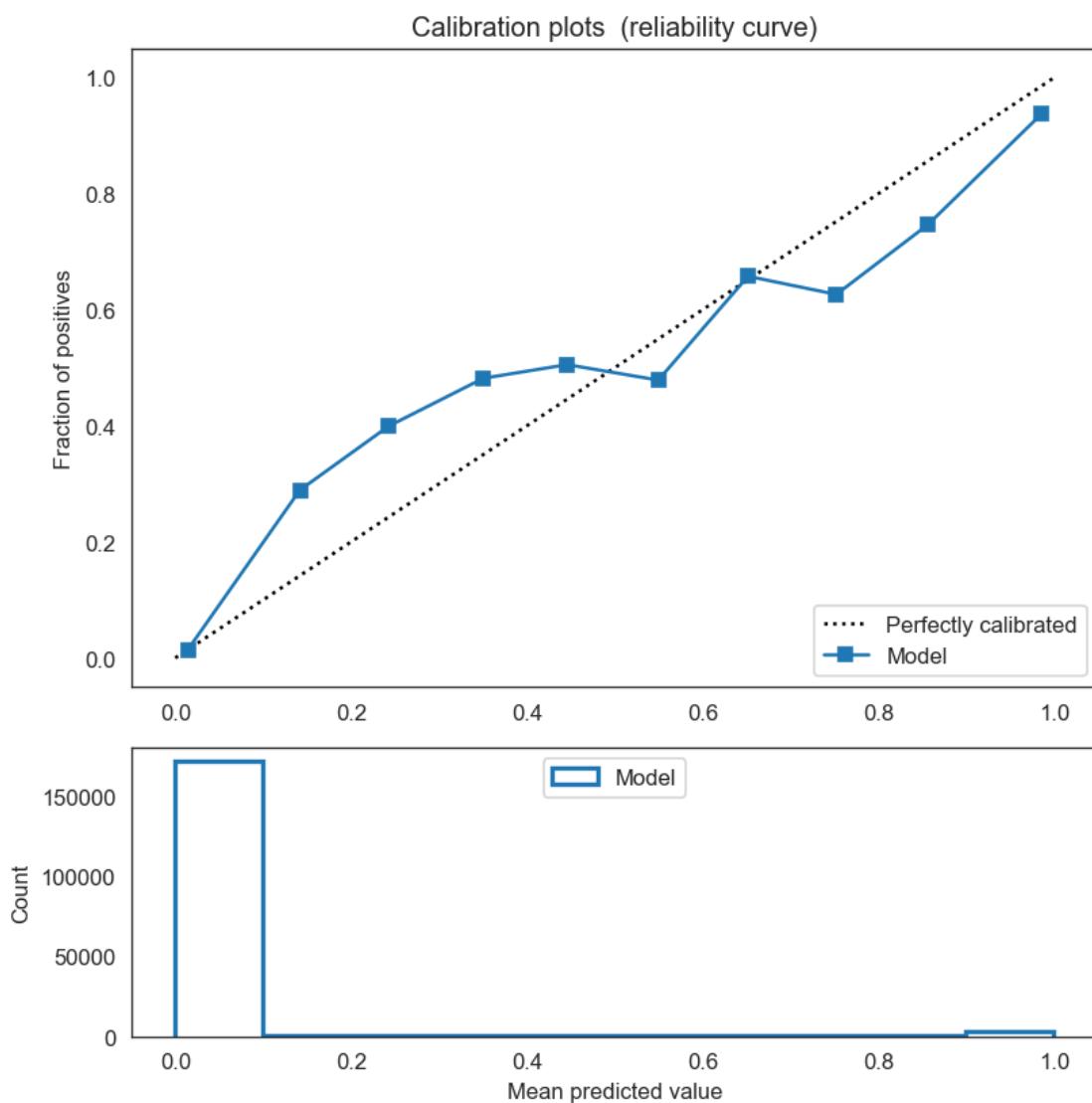


[103]:

	prob_bin	not_fraud	fraud	perc_fraud	\
0	Bin1((0.0188, 0.998])	12768	4949	0.798355	
1	Bin2((0.0146, 0.0188])	17191	525	0.084691	
2	Bin3((0.0134, 0.0146])	17461	255	0.041136	
3	Bin4((0.0128, 0.0134])	17561	155	0.025004	
4	Bin5((0.0125, 0.0128])	17607	109	0.017583	
5	Bin6((0.0123, 0.0125])	17651	65	0.010486	
6	Bin7((0.0121, 0.0123])	17659	57	0.009195	
7	Bin8((0.0119, 0.0121])	17668	48	0.007743	
8	Bin9((0.0118, 0.0119])	17693	23	0.003710	
9	Bin10((0.010599999999999998, 0.0118])	17704	13	0.002097	
	perc_not_fraud	cum_perc_fraud	cum_perc_not_fraud		
0	0.074683	79.835457	7.468283		
1	0.100554	88.304565	17.523675		
2	0.102133	92.418132	27.736996		
3	0.102718	94.918535	38.008809		
4	0.102987	96.676883	48.307529		

5	0.103245	97.725440	58.631985
6	0.103291	98.644943	68.961120
7	0.103344	99.419261	79.295520
8	0.103490	99.790289	89.644543
9	0.103555	100.000000	100.000000

```
[104]: draw_calibration_curve(y_test, y_prob_pred_calib, n_bins=10)
```



15.2 B) LightGBM

```
[107]: from lightgbm import LGBMClassifier
```

```
[108]: %%time  
lgbc = LGBMClassifier(random_state=0, n_jobs = -1)  
lgbc.fit(X_train,y_train)  
lgbc
```

Wall time: 29.5 s

```
[108]: LGBMClassifier(random_state=0)
```

```
[109]: y_pred_lgbc = lgbc.predict(X_test)  
y_prob_pred_lgbc = lgbc.predict_proba(X_test)  
y_prob_pred_lgbc = [x[1] for x in y_prob_pred_lgbc]  
print("Y predicted : ",y_pred_lgbc)  
print("Y probability predicted : ",y_prob_pred_lgbc[:5])
```

Y predicted : [False False False ... False False False]
Y probability predicted : [0.016160027443867835, 0.0080773782284121,
0.00825509576482752, 0.06708271917846045, 0.007287115458060899]

```
[110]: compute_evaluation_metric(lgbc, X_test, y_test, y_pred_lgbc, y_prob_pred_lgbc)
```

Accuracy Score : 0.9774613066007383

AUC Score : 0.9239690737882372

Confusion Matrix :

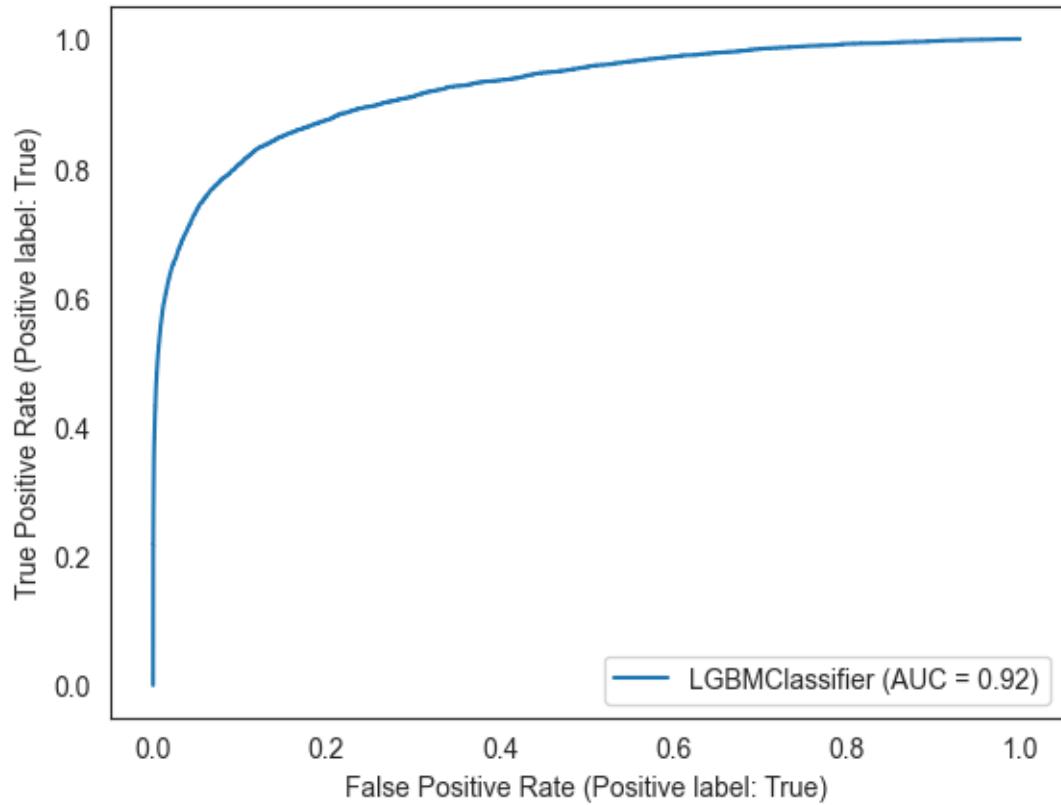
```
[[170585 378]  
 [ 3615 2584]]
```

Classification Report :

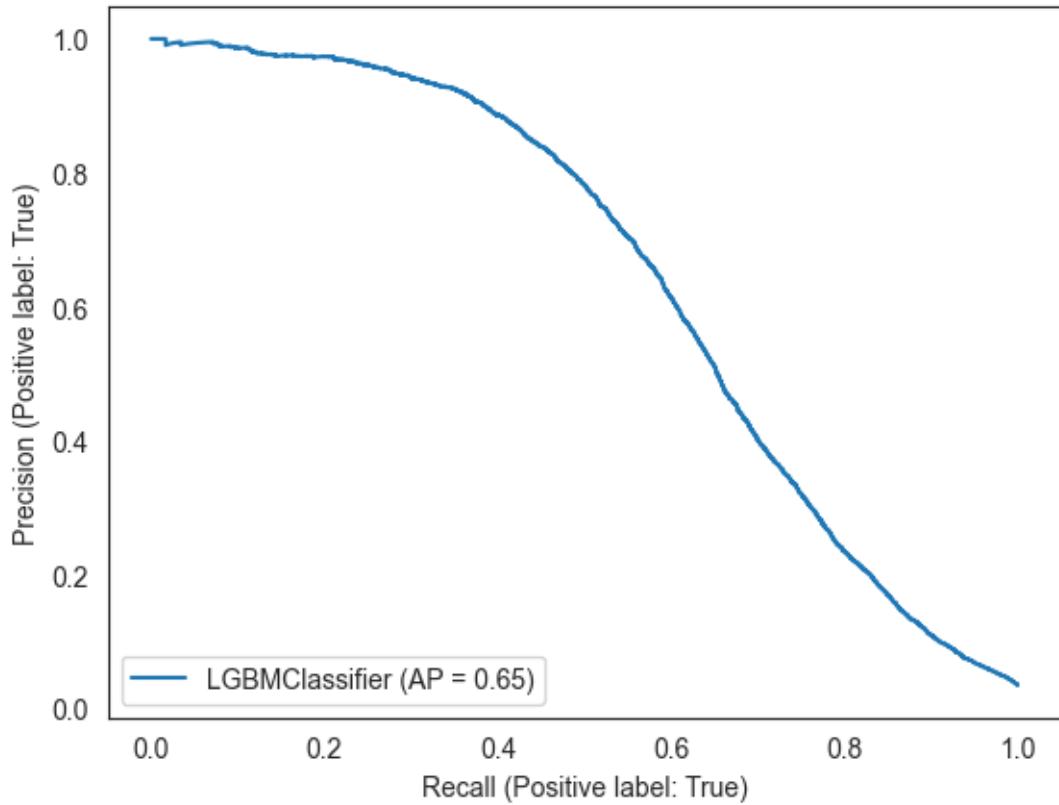
	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.87	0.42	0.56	6199
accuracy			0.98	177162
macro avg	0.93	0.71	0.78	177162
weighted avg	0.98	0.98	0.97	177162

Concordance Index : 0.9239690501988727

ROC curve :



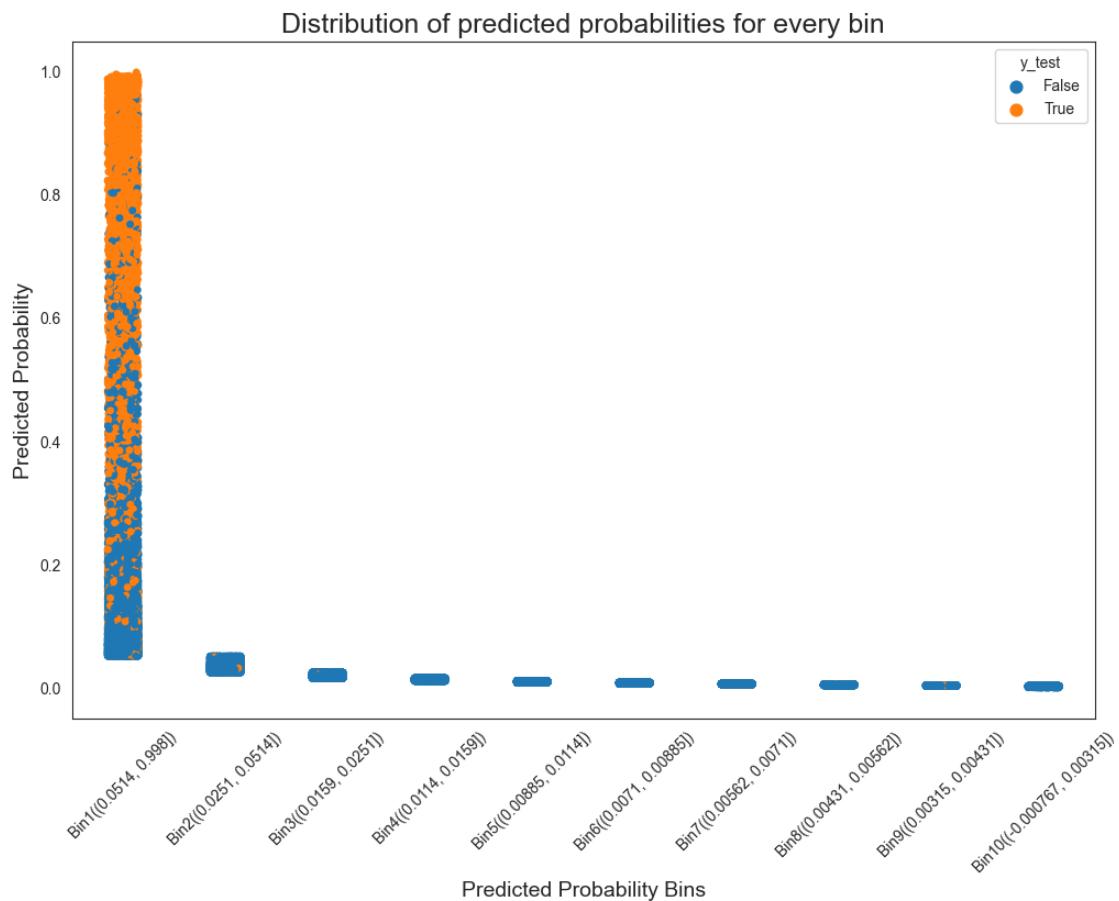
PR curve :

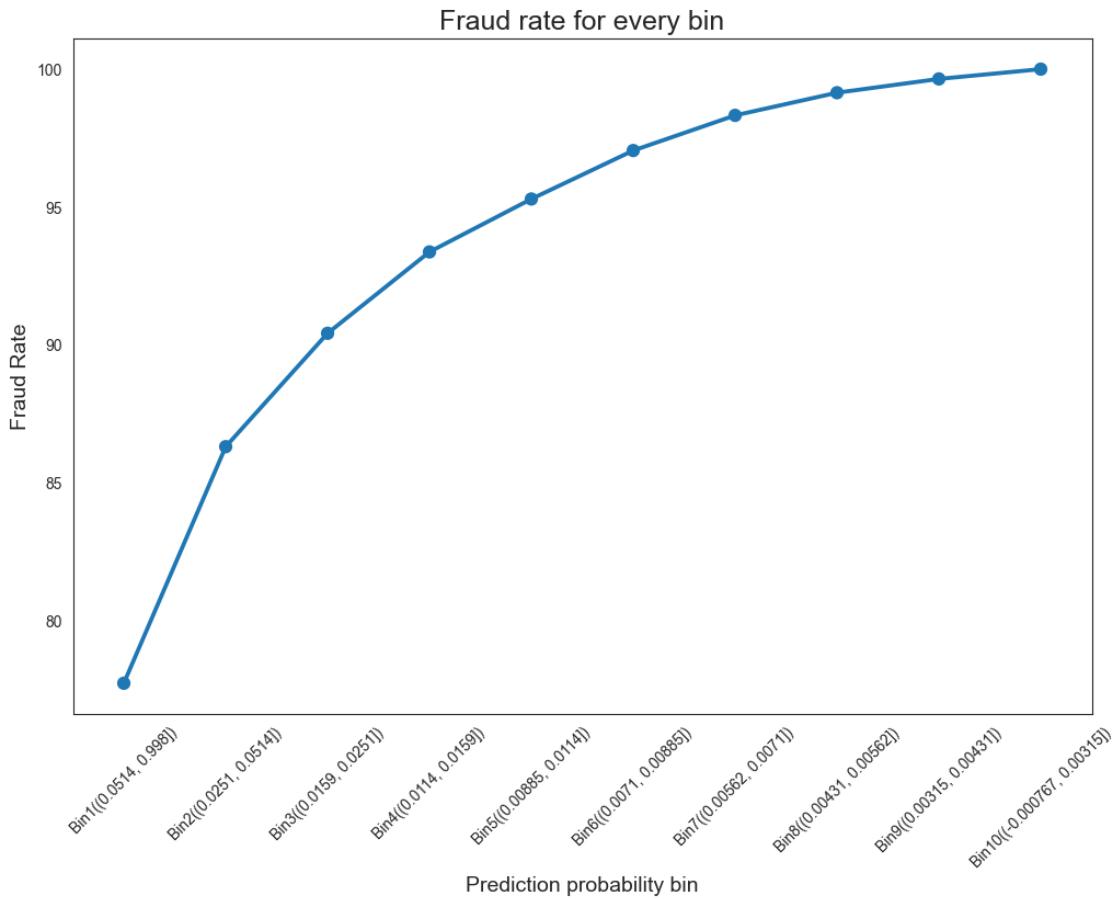


Additional Metrics:

TPR (Recall) : 0.4168
FPR : 0.0022
TNR (Specificity) : 0.9978
FNR : 0.5832

```
[111]: captures(y_test, y_pred_lgbc, y_prob_pred_lgbc)
```





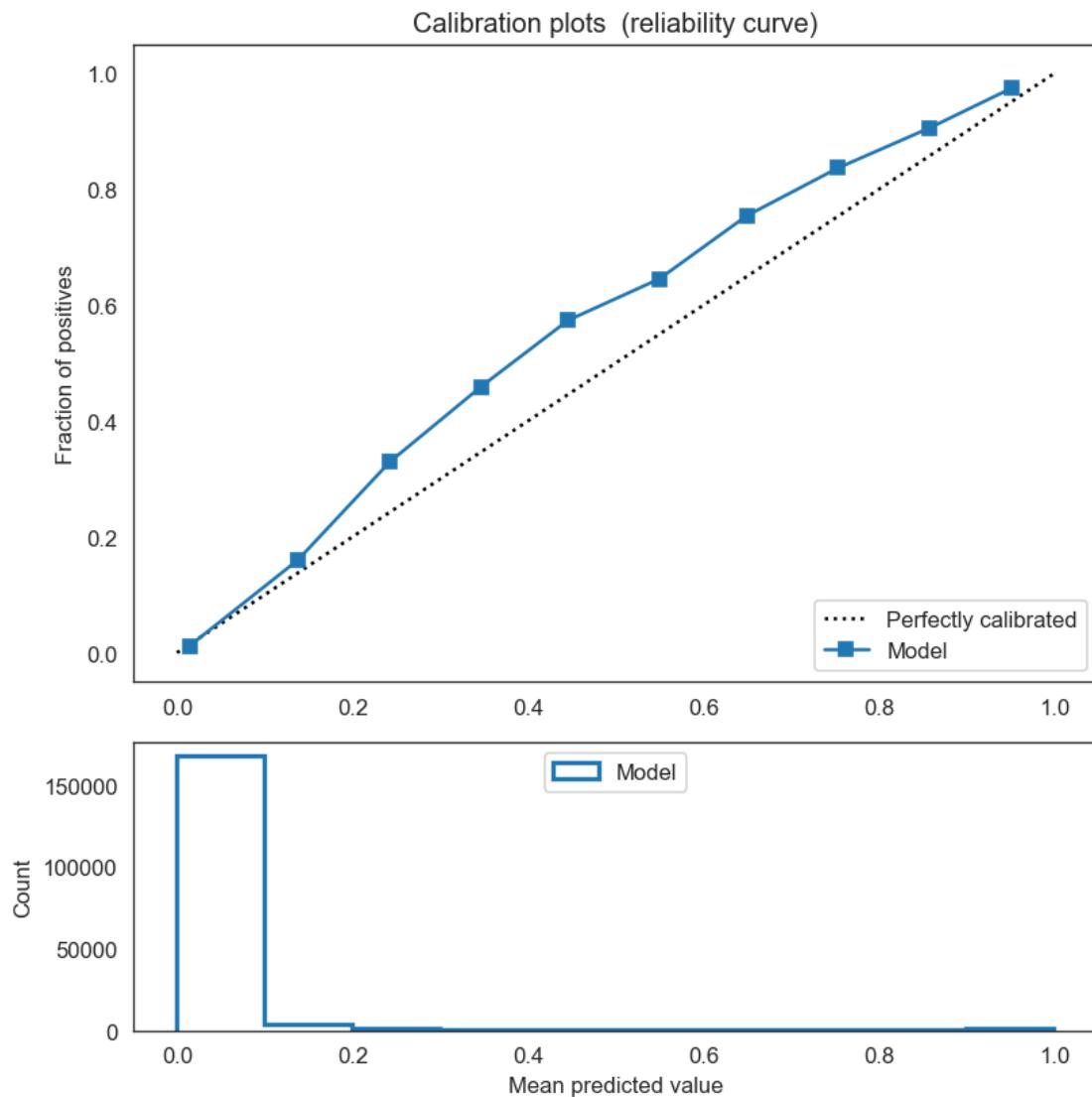
```
[111]:
```

	prob_bin	not_fraud	fraud	perc_fraud	perc_not_fraud	\
0	Bin1((0.0514, 0.998])	12899	4818	0.777222	0.075449	
1	Bin2((0.0251, 0.0514])	17184	532	0.085820	0.100513	
2	Bin3((0.0159, 0.0251])	17461	255	0.041136	0.102133	
3	Bin4((0.0114, 0.0159])	17533	183	0.029521	0.102554	
4	Bin5((0.00885, 0.0114])	17597	119	0.019197	0.102929	
5	Bin6((0.0071, 0.00885])	17607	109	0.017583	0.102987	
6	Bin7((0.00562, 0.0071])	17637	79	0.012744	0.103163	
7	Bin8((0.00431, 0.00562])	17665	51	0.008227	0.103326	
8	Bin9((0.00315, 0.00431])	17685	31	0.005001	0.103443	
9	Bin10((-0.000767, 0.00315])	17695	22	0.003549	0.103502	

	cum_perc_fraud	cum_perc_not_fraud
0	77.722213	7.544907
1	86.304243	17.596205
2	90.417809	27.809526
3	93.369898	38.064961
4	95.289563	48.357832
5	97.047911	58.656551

6	98.322310	68.972819
7	99.145023	79.305464
8	99.645104	89.649807
9	100.000000	100.000000

```
[112]: draw_calibration_curve(y_test, y_prob_pred_lgbc, n_bins=10)
```



- With LGBM, Accuracy score is 97.7%. It's almost similar to XGBoost model
- AUC score has improved to 92.6 from 88.5
- Recall and f-1 score have also improved, but it's still not up to the mark

15.3 C) Random Forest Classifier

```
[113]: from sklearn.ensemble import RandomForestClassifier, ExtraTreesClassifier, GradientBoostingClassifier
```

```
[114]: X_train.head()
```

	TransactionAmt	ProductCD	card1	card2	card3	card4	card5	card6	\							
448539	4.679688	4	6598	111.0	150.0	2	195.0	2								
321311	4.355469	4	12839	321.0	150.0	4	226.0	2								
497320	3.892578	4	14649	548.0	150.0	4	226.0	2								
350951	4.058594	4	6489	295.0	150.0	4	226.0	2								
98132	5.296875	2	5714	170.0	150.0	4	195.0	1								
	addr1	addr2	dist1	P_emaildomain	R_emaildomain	C1	C2	C3	\							
448539	264.0	87.0	6.0		0	2	2.0	4.0	0.0							
321311	264.0	87.0	0.0		0	2	1.0	1.0	0.0							
497320	441.0	87.0	86.0		4	2	2.0	4.0	0.0							
350951	184.0	87.0	8.0		0	2	154.0	148.0	0.0							
98132	498.0	87.0	8.0		5	6	1.0	1.0	0.0							
	C4	C5	C6	C7	C8	C9	C10	C11	C12	C13	C14	\				
448539	0.0	1.0	1.0	0.0	0.0	1.0	0.0	2.0	0.0	7.0	2.0					
321311	0.0	2.0	1.0	0.0	0.0	1.0	0.0	1.0	0.0	12.0	1.0					
497320	0.0	0.0	2.0	0.0	0.0	3.0	0.0	4.0	0.0	45.0	2.0					
350951	0.0	108.0	101.0	0.0	0.0	92.0	0.0	107.0	0.0	504.0	129.0					
98132	1.0	0.0	1.0	0.0	1.0	0.0	1.0	1.0	0.0	1.0	1.0					
	D1	D2	D3	D4	D5	D6		D8	D9	D10	\					
448539	64.0	64.0	18.0	63.0	10.0	0.0	37.843750	0.666504	63.0							
321311	114.0	114.0	14.0	115.0	14.0	0.0	37.843750	0.666504	115.0							
497320	85.0	85.0	3.0	329.0	3.0	0.0	37.843750	0.666504	329.0							
350951	577.0	577.0	54.0	54.0	54.0	0.0	37.843750	0.666504	577.0							
98132	0.0	97.0	8.0	26.0	10.0	0.0	7.917969	0.916504	15.0							
	D11	D12	D13	D14	D15	M1	M2	M3	M4	M5	M6	M7	M8	M9	id_01	\
448539	35.0	0.0	0.0	0.0	63.0	1	1	1	3	2	0	0	1	1	-5.0	
321311	45.0	0.0	0.0	0.0	115.0	1	1	1	3	2	1	1	1	1	-5.0	
497320	85.0	0.0	0.0	0.0	85.0	1	1	1	0	0	0	0	0	0	-5.0	
350951	43.0	0.0	0.0	0.0	110.0	2	2	2	3	2	0	2	2	2	-5.0	
98132	43.0	0.0	0.0	0.0	52.0	2	2	2	3	2	2	2	2	2	-5.0	
	id_02	id_03	id_04	id_05	id_06	id_09	id_10	id_11	id_12	\						
448539	125673.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100.0		1					
321311	125673.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100.0		1					
497320	125673.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100.0		1					
350951	125673.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100.0		1					

98132	105681.0	0.0	0.0	2.0	0.0	0.0	0.0	100.0	2		
	id_13	id_14	id_15	id_16	id_17	id_19	id_20	id_28	id_29	id_30	\
448539	52.0	-300.0	1	1	166.0	341.0	472.0	1	1	3	
321311	52.0	-300.0	1	1	166.0	341.0	472.0	1	1	3	
497320	52.0	-300.0	1	1	166.0	341.0	472.0	1	1	3	
350951	52.0	-300.0	1	1	166.0	341.0	472.0	1	1	3	
98132	52.0	-360.0	0	0	166.0	621.0	272.0	0	0	4	
	id_31	id_32	id_33	id_34	id_35	id_36	id_37	id_38	DeviceType	\	
448539	103	24.0	74	4	2	2	2	2	2		
321311	103	24.0	74	4	2	2	2	2	2		
497320	103	24.0	74	4	2	2	2	2	2		
350951	103	24.0	74	4	2	2	2	2	2		
98132	84	24.0	72	3	1	0	1	1	0		
	DeviceInfo	card2_missing_flag	card3_missing_flag	\							
448539	141	0	0								
321311	141	0	0								
497320	141	0	0								
350951	141	0	0								
98132	140	0	0								
	card4_missing_flag	card5_missing_flag	card6_missing_flag	\							
448539	0	0	0								
321311	0	0	0								
497320	0	0	0								
350951	0	0	0								
98132	0	0	0								
	addr1_missing_flag	addr2_missing_flag	dist1_missing_flag	\							
448539	0	0	0								
321311	0	0	0								
497320	0	0	0								
350951	0	0	0						1		
98132	0	0	0						1		
	dist2_missing_flag	P_emaildomain_missing_flag	\								
448539	1	0									
321311	1	0									
497320	1	0									
350951	1	0									
98132	1	0									
	R_emaildomain_missing_flag	D1_missing_flag	D2_missing_flag	\							
448539	1	0	0								
321311	1	0	0								

497320	1	0	0		
350951	1	0	0		
98132	0	0	1		
	D3_missing_flag	D4_missing_flag	D5_missing_flag	D6_missing_flag	\
448539	0	0	0	1	
321311	0	0	0	1	
497320	0	0	0	1	
350951	0	0	0	1	
98132	1	1	1	1	
	D7_missing_flag	D8_missing_flag	D9_missing_flag	D10_missing_flag	\
448539	1	1	1	0	
321311	1	1	1	0	
497320	1	1	1	0	
350951	1	1	1	0	
98132	1	0	0	1	
	D11_missing_flag	D12_missing_flag	D13_missing_flag	\	
448539	0	1	1		
321311	0	1	1		
497320	0	1	1		
350951	1	1	1		
98132	1	1	1		
	D14_missing_flag	D15_missing_flag	M1_missing_flag	M2_missing_flag	\
448539	1	0	0	0	
321311	1	0	0	0	
497320	1	0	0	0	
350951	1	0	1	1	
98132	1	1	1	1	
	M3_missing_flag	M4_missing_flag	M5_missing_flag	M6_missing_flag	\
448539	0	1	1	0	
321311	0	1	1	0	
497320	0	0	0	0	
350951	1	1	1	0	
98132	1	1	1	1	
	M7_missing_flag	M8_missing_flag	M9_missing_flag	V1_missing_flag	\
448539	0	0	0	0	
321311	0	0	0	0	
497320	0	0	0	0	
350951	1	1	1	1	
98132	1	1	1	1	
	V2_missing_flag	V3_missing_flag	V4_missing_flag	V5_missing_flag	\

448539	0	0	0	0
321311	0	0	0	0
497320	0	0	0	0
350951	1	1	1	1
98132	1	1	1	1
	V6_missing_flag	V7_missing_flag	V8_missing_flag	V9_missing_flag
448539	0	0	0	0
321311	0	0	0	0
497320	0	0	0	0
350951	1	1	1	1
98132	1	1	1	1
	V10_missing_flag	V11_missing_flag	V12_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	1	1	0	
98132	1	1	1	
	V13_missing_flag	V14_missing_flag	V15_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V16_missing_flag	V17_missing_flag	V18_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V19_missing_flag	V20_missing_flag	V21_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V22_missing_flag	V23_missing_flag	V24_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	

	V25_missing_flag	V26_missing_flag	V27_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V28_missing_flag	V29_missing_flag	V30_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V31_missing_flag	V32_missing_flag	V33_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V34_missing_flag	V35_missing_flag	V36_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V37_missing_flag	V38_missing_flag	V39_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V40_missing_flag	V41_missing_flag	V42_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V43_missing_flag	V44_missing_flag	V45_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	

350951	0	0	0	
98132	1	1	1	
	V46_missing_flag	V47_missing_flag	V48_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V49_missing_flag	V50_missing_flag	V51_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V52_missing_flag	V53_missing_flag	V54_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V55_missing_flag	V56_missing_flag	V57_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V58_missing_flag	V59_missing_flag	V60_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V61_missing_flag	V62_missing_flag	V63_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V64_missing_flag	V65_missing_flag	V66_missing_flag	\
448539	0	0	0	

321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	1	1	1

	V85_missing_flag	V86_missing_flag	V87_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V88_missing_flag	V89_missing_flag	V90_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V91_missing_flag	V92_missing_flag	V93_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	1	1	
	V94_missing_flag	V95_missing_flag	V96_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	1	0	0	
	V97_missing_flag	V98_missing_flag	V99_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V100_missing_flag	V101_missing_flag	V102_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V103_missing_flag	V104_missing_flag	V105_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	

98132	0	0	0	
	V106_missing_flag	V107_missing_flag	V108_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V109_missing_flag	V110_missing_flag	V111_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V112_missing_flag	V113_missing_flag	V114_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V115_missing_flag	V116_missing_flag	V117_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V118_missing_flag	V119_missing_flag	V120_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V121_missing_flag	V122_missing_flag	V123_missing_flag	\
448539	0	0	0	
321311	0	0	0	
497320	0	0	0	
350951	0	0	0	
98132	0	0	0	
	V124_missing_flag	V125_missing_flag	V126_missing_flag	\
448539	0	0	0	
321311	0	0	0	

497320	0	0	0
350951	0	0	0
98132	0	0	0
	V127_missing_flag	V128_missing_flag	V129_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V130_missing_flag	V131_missing_flag	V132_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V133_missing_flag	V134_missing_flag	V135_missing_flag
448539	0	0	0 ...
321311	0	0	0 ...
497320	0	0	0 ...
350951	0	0	0 ...
98132	0	0	0 ...
	V169_missing_flag	V170_missing_flag	V171_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V172_missing_flag	V173_missing_flag	V174_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V175_missing_flag	V176_missing_flag	V177_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V178_missing_flag	V179_missing_flag	V180_missing_flag

448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V181_missing_flag	V182_missing_flag	V183_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V184_missing_flag	V185_missing_flag	V186_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V187_missing_flag	V188_missing_flag	V189_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V190_missing_flag	V191_missing_flag	V192_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V193_missing_flag	V194_missing_flag	V195_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V196_missing_flag	V197_missing_flag	V198_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	V199_missing_flag	V200_missing_flag	V201_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V202_missing_flag	V203_missing_flag	V204_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V205_missing_flag	V206_missing_flag	V207_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V208_missing_flag	V209_missing_flag	V210_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V211_missing_flag	V212_missing_flag	V213_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V214_missing_flag	V215_missing_flag	V216_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	V217_missing_flag	V218_missing_flag	V219_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	

350951	1	1	1	
98132	0	0	0	
	V220_missing_flag	V221_missing_flag	V222_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V223_missing_flag	V224_missing_flag	V225_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V226_missing_flag	V227_missing_flag	V228_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V229_missing_flag	V230_missing_flag	V231_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V232_missing_flag	V233_missing_flag	V234_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V235_missing_flag	V236_missing_flag	V237_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V238_missing_flag	V239_missing_flag	V240_missing_flag	\
448539	1	1	1	

321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V241_missing_flag	V242_missing_flag	V243_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V244_missing_flag	V245_missing_flag	V246_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V247_missing_flag	V248_missing_flag	V249_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V250_missing_flag	V251_missing_flag	V252_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V253_missing_flag	V254_missing_flag	V255_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
V256_missing_flag	V257_missing_flag	V258_missing_flag	\
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	V259_missing_flag	V260_missing_flag	V261_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V262_missing_flag	V263_missing_flag	V264_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V265_missing_flag	V266_missing_flag	V267_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V268_missing_flag	V269_missing_flag	V270_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V271_missing_flag	V272_missing_flag	V273_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V274_missing_flag	V275_missing_flag	V276_missing_flag	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	
	V277_missing_flag	V278_missing_flag	V279_missing_flag	\
448539	1	1	0	
321311	1	1	0	
497320	1	1	0	
350951	1	1	0	

98132	0	0	0
	V280_missing_flag	V281_missing_flag	V282_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V283_missing_flag	V284_missing_flag	V285_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V286_missing_flag	V287_missing_flag	V288_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V289_missing_flag	V290_missing_flag	V291_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V292_missing_flag	V293_missing_flag	V294_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V295_missing_flag	V296_missing_flag	V297_missing_flag \
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V298_missing_flag	V299_missing_flag	V300_missing_flag \
448539	0	0	0
321311	0	0	0

497320	0	0	0
350951	0	0	0
98132	0	0	0
	V301_missing_flag	V302_missing_flag	V303_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V304_missing_flag	V305_missing_flag	V306_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V307_missing_flag	V308_missing_flag	V309_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V310_missing_flag	V311_missing_flag	V312_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V313_missing_flag	V314_missing_flag	V315_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V316_missing_flag	V317_missing_flag	V318_missing_flag
448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V319_missing_flag	V320_missing_flag	V321_missing_flag

448539	0	0	0
321311	0	0	0
497320	0	0	0
350951	0	0	0
98132	0	0	0
	V322_missing_flag	V323_missing_flag	V324_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V325_missing_flag	V326_missing_flag	V327_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V328_missing_flag	V329_missing_flag	V330_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V331_missing_flag	V332_missing_flag	V333_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V334_missing_flag	V335_missing_flag	V336_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0
	V337_missing_flag	V338_missing_flag	V339_missing_flag
448539	1	1	1
321311	1	1	1
497320	1	1	1
350951	1	1	1
98132	0	0	0

	<code>id_01_missing_flag</code>	<code>id_02_missing_flag</code>	<code>id_03_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	<code>id_04_missing_flag</code>	<code>id_05_missing_flag</code>	<code>id_06_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	<code>id_07_missing_flag</code>	<code>id_08_missing_flag</code>	<code>id_09_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	1	1	0	

	<code>id_10_missing_flag</code>	<code>id_11_missing_flag</code>	<code>id_12_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	<code>id_13_missing_flag</code>	<code>id_14_missing_flag</code>	<code>id_15_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	0	

	<code>id_16_missing_flag</code>	<code>id_17_missing_flag</code>	<code>id_18_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	
350951	1	1	1	
98132	0	0	1	

	<code>id_19_missing_flag</code>	<code>id_20_missing_flag</code>	<code>id_21_missing_flag</code>	\
448539	1	1	1	
321311	1	1	1	
497320	1	1	1	

	1	1	1		
350951	1	1	1		
98132	0	0	1		
			\		
id_22_missing_flag	id_23_missing_flag	id_24_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	1	1	1		
			\		
id_25_missing_flag	id_26_missing_flag	id_27_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	1	1	1		
			\		
id_28_missing_flag	id_29_missing_flag	id_30_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	0	0	0		
			\		
id_31_missing_flag	id_32_missing_flag	id_33_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	0	0	0		
			\		
id_34_missing_flag	id_35_missing_flag	id_36_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	0	0	0		
			\		
id_37_missing_flag	id_38_missing_flag	DeviceType_missing_flag	\		
448539	1	1	1		
321311	1	1	1		
497320	1	1	1		
350951	1	1	1		
98132	0	0	0		
			\		
DeviceInfo_missing_flag	_Weekdays	_Hours	_Days	Trans_min_mean	\
448539	1	3	3	12	-27.09375

321311		1	5	14	3	-57.09375		
497320		1	0	19	30	-86.00000		
350951		1	6	1	11	-77.06250		
98132		0	5	22	23	65.00000		
	Trans_min_std	TransactionAmt_to_mean_card1				\		
448539	-0.113281		0.461914					
321311	-0.238770		0.645508					
497320	-0.359619		0.470703					
350951	-0.322266		0.555176					
98132	0.271729		1.193359					
	TransactionAmt_to_mean_card4	TransactionAmt_to_std_card1				\		
448539		0.815430		0.313331				
321311		0.585449		0.401900				
497320		0.367920		0.326992				
350951		0.435059		0.570935				
98132		1.501953		0.718662				
	TransactionAmt_to_std_card4	PCA_V_0	PCA_V_1	PCA_V_2	PCA_V_3	\		
448539		0.425049	-0.782227	0.368896	0.263428	0.007843		
321311		0.341553	-0.782227	0.368896	0.263428	0.007904		
497320		0.214722	-0.801758	0.317871	0.270020	-0.027130		
350951		0.253906	-0.304443	-0.825684	-0.107727	-0.468994		
98132		0.876465	2.806641	0.345703	0.377197	-0.017654		
	PCA_V_4	PCA_V_5	PCA_V_6	PCA_V_7	PCA_V_8	PCA_V_9	PCA_V_10	\
448539	0.013977	-0.007053	0.002455	0.334717	-0.010490	-0.220459	-0.042755	
321311	0.013786	-0.007095	0.002304	0.334961	-0.010521	-0.220825	-0.042938	
497320	0.058411	-0.003780	-0.014931	-0.225342	0.021576	0.033905	-0.024246	
350951	-0.162598	0.029373	0.138550	0.065613	-0.067749	0.260010	0.052063	
98132	-0.042450	-0.244263	0.023148	0.011810	0.000937	0.023010	-0.009560	
	PCA_V_11	PCA_V_12	PCA_V_13	PCA_V_14	PCA_V_15	PCA_V_16	PCA_V_17	\
448539	0.021561	-0.005577	-0.020416	0.005253	-0.000285	0.004505	-0.006889	
321311	0.020325	-0.003403	-0.022324	0.005749	0.000531	0.004292	-0.009888	
497320	0.010124	0.020370	0.017105	-0.014145	-0.013794	-0.005302	-0.077637	
350951	-0.091125	0.068176	0.014664	0.010857	-0.011421	-0.020767	0.009926	
98132	-0.022293	-0.021683	-0.013680	-0.005581	-0.009674	0.001465	-0.008263	
	PCA_V_18	PCA_V_19	PCA_V_20	PCA_V_21	PCA_V_22	PCA_V_23	PCA_V_24	\
448539	0.008362	-0.011055	-0.006924	-0.033722	0.011093	-0.011307	-0.007355	
321311	0.009216	-0.005642	-0.005966	-0.034119	0.012009	-0.009644	-0.006516	
497320	0.056580	0.115479	0.013458	-0.012100	0.003590	0.023392	0.003922	
350951	0.008850	0.005054	-0.041290	0.041382	-0.051819	0.042450	0.026703	
98132	-0.008881	0.005920	-0.004478	-0.000713	-0.003057	0.008438	-0.017868	

```
PCA_V_25  PCA_V_26  PCA_V_27  PCA_V_28  PCA_V_29  
448539 -0.004627  0.002234 -0.005222  0.004154 -0.001096  
321311 -0.002327  0.000332 -0.006203  0.002287 -0.005352  
497320  0.032867  0.010475  0.007317  0.008362 -0.037079  
350951 -0.046417 -0.029648  0.022217  0.004562 -0.008606  
98132   0.000072  0.013344 -0.005264 -0.000665  0.007454
```

[5 rows x 533 columns]

```
[118]: X_train.isnull().sum()
```

```
TransactionAmt      0  
ProductCD          0  
card1               0  
card2               0  
card3               0  
..  
PCA_V_25            0  
PCA_V_26            0  
PCA_V_27            0  
PCA_V_28            0  
PCA_V_29            0  
Length: 533, dtype: int64
```

```
[120]: %%time  
rfc = RandomForestClassifier(random_state=0, n_jobs = -1)  
rfc.fit(X_train, y_train)  
rfc
```

Wall time: 2min 45s

```
[120]: RandomForestClassifier(n_jobs=-1, random_state=0)
```

```
[121]: y_pred_rfc = rfc.predict(X_test)  
y_prob_pred_rfc = rfc.predict_proba(X_test)[:, 1]  
  
print("Y predicted : ",y_pred_rfc)  
print("Y probability predicted : ",y_prob_pred_rfc[:5])
```

```
Y predicted : [False False False ... False False False]  
Y probability predicted : [0.02 0. 0. 0.06 0.02]
```

```
[123]: compute_evaluation_metric(rfc, X_test, y_test, y_pred_rfc, y_prob_pred_rfc)
```

Accuracy Score : 0.9792901412266739

AUC Score : 0.9295181538168427

Confusion Matrix :

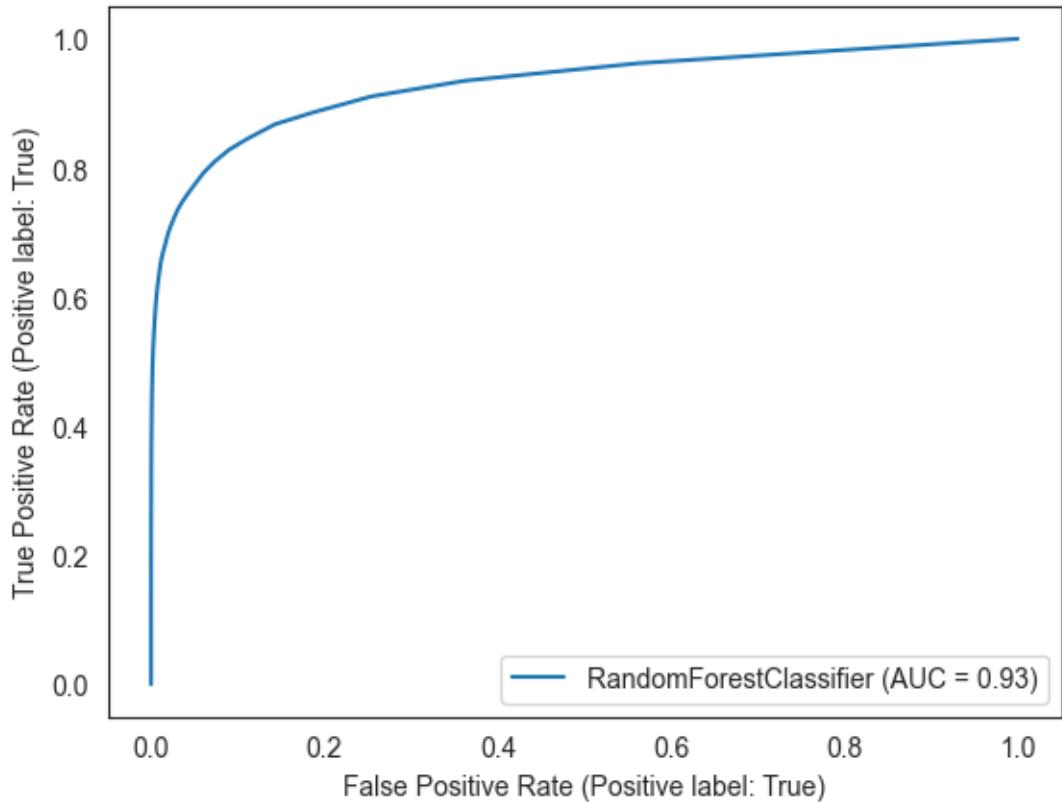
```
[[170811    152]
 [ 3517   2682]]
```

Classification Report :

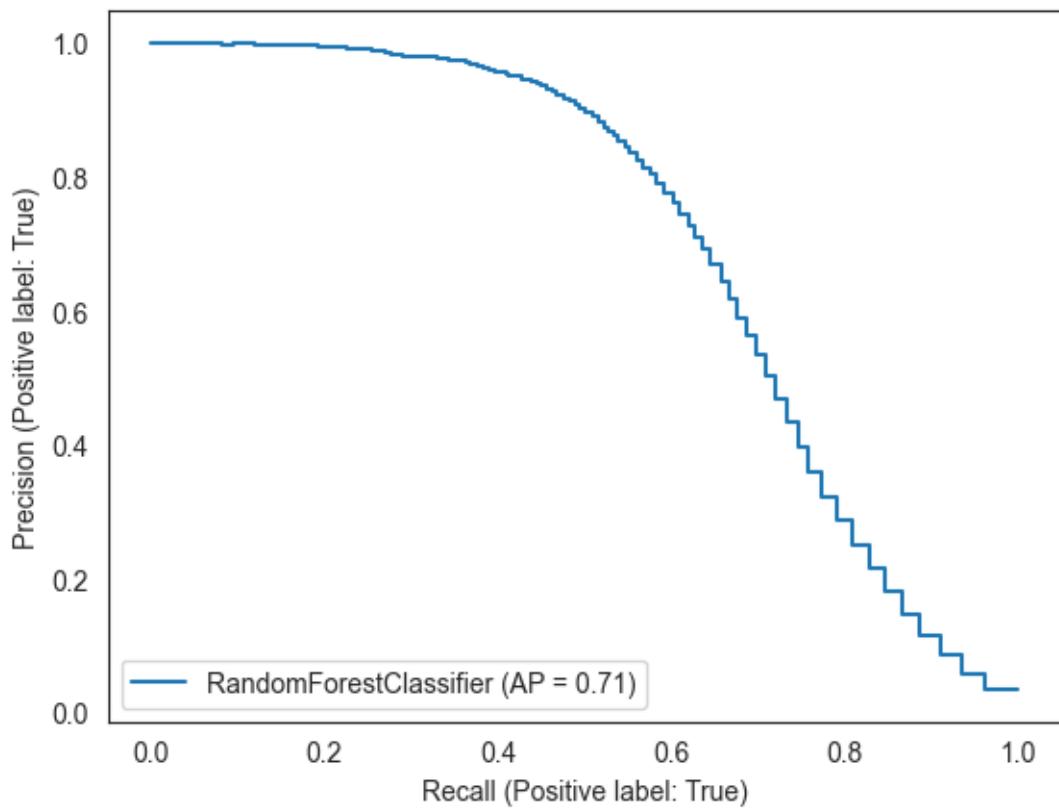
	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.95	0.43	0.59	6199
accuracy			0.98	177162
macro avg	0.96	0.72	0.79	177162
weighted avg	0.98	0.98	0.98	177162

Concordance Index : 0.914790096309497

ROC curve :



PR curve :



Additional Metrics:

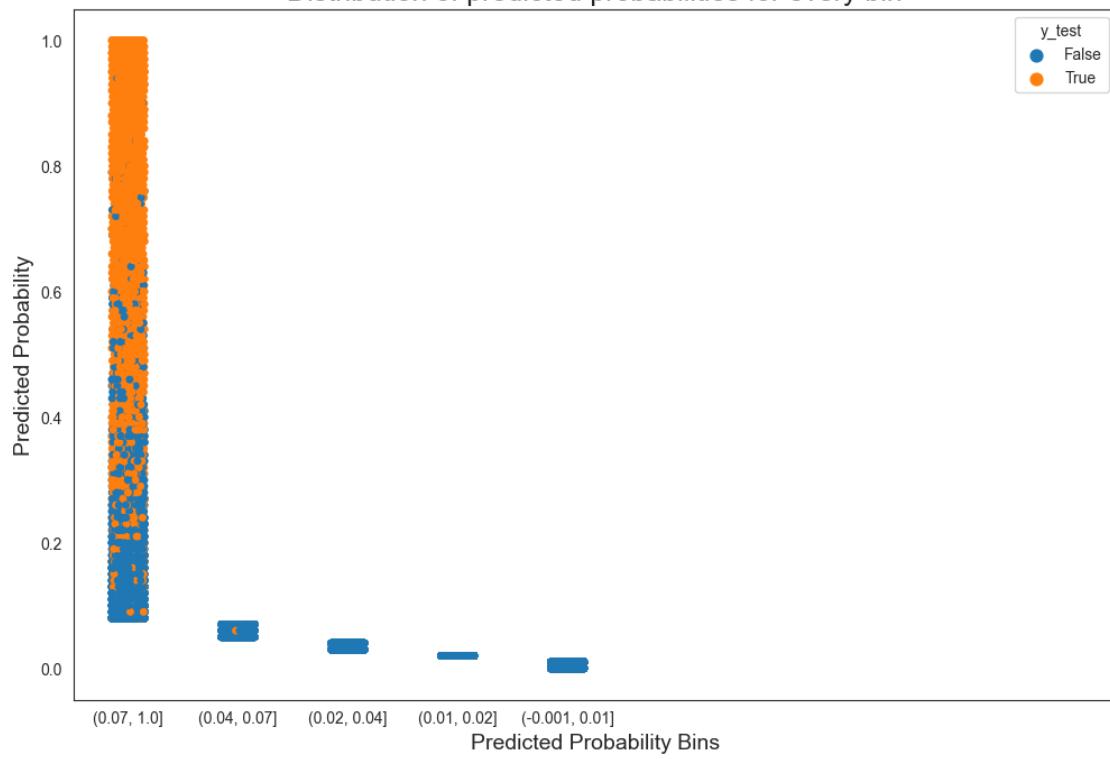
TPR (Recall) : 0.4327
FPR : 0.0009
TNR (Specificity) : 0.9991
FNR : 0.5673

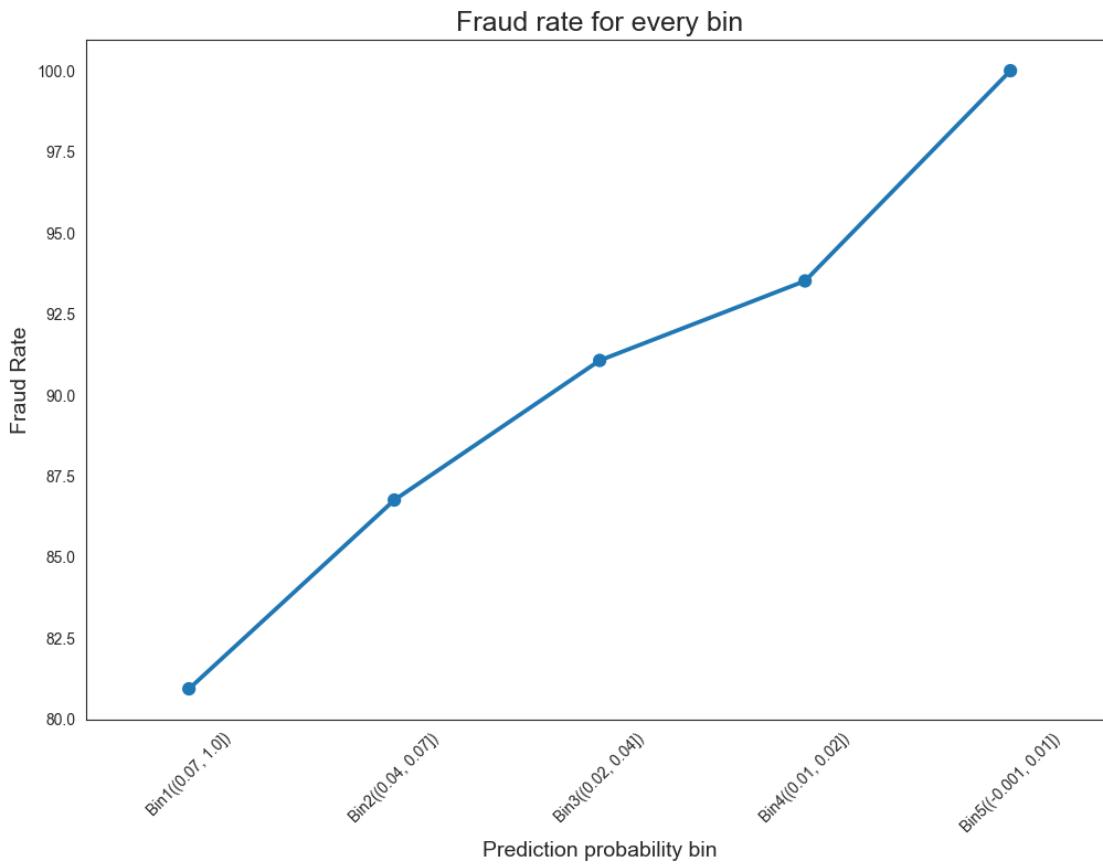
```
[124]: concordance(y_test.values, y_prob_pred_rfc)
```

```
[124]: 0.914790096309497
```

```
[125]: captures(y_test, y_pred_rfc, y_prob_pred_rfc)
```

Distribution of predicted probabilities for every bin

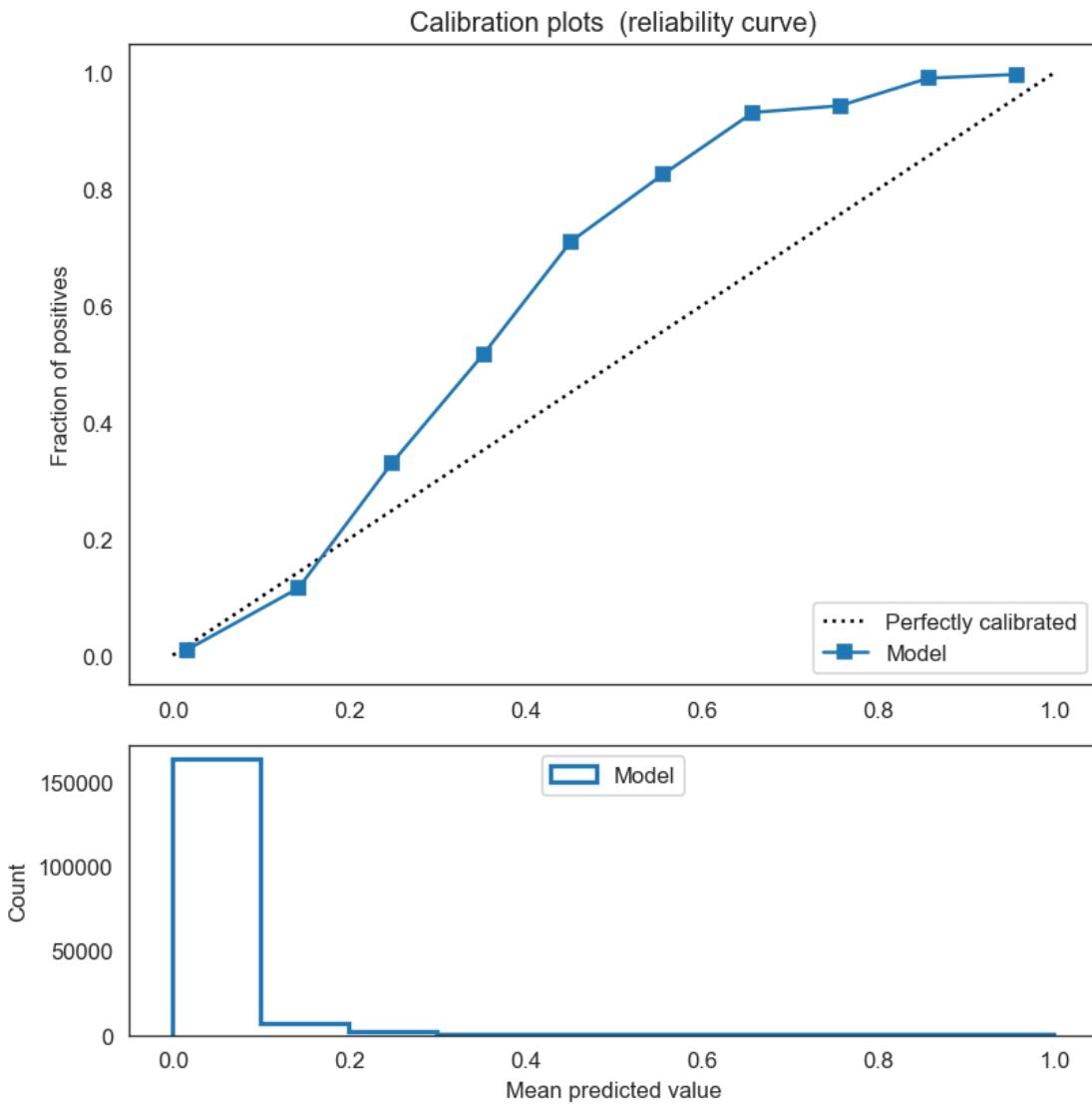




```
[125]:      prob_bin  not_fraud  fraud  perc_fraud  perc_not_fraud  \
0    Bin1((0.07, 1.0])      12482    5017    0.809324    0.073010
1    Bin2((0.04, 0.07])     11939     361    0.058235    0.069834
2    Bin3((0.02, 0.04])     18984     267    0.043071    0.111042
3    Bin4((0.01, 0.02])     18595     152    0.024520    0.108766
4    Bin5((-0.001, 0.01])   108963     402    0.064849    0.637348

      cum_perc_fraud  cum_perc_not_fraud
0            80.932408            7.300995
1            86.755928           14.284377
2            91.063075           25.388534
3            93.515083           36.265157
4           100.000000          100.000000
```

```
[126]: draw_calibration_curve(y_test, y_prob_pred_rfc, n_bins=10)
```



16 15. Handling Class Imbalance

16.0.1 Handle Class Imbalance with Random Oversampler

```
[130]: from imblearn.over_sampling import RandomOverSampler
ros = RandomOverSampler()
X_train_ros, y_train_ros = ros.fit_resample(X_train, y_train)
y_train_ros.value_counts()
```

```
[130]: False    398914
      True     398914
Name: isFraud, dtype: int64
```

```
[131]: %%time  
lgbc_ros = LGBMClassifier(random_state=0)  
lgbc_ros.fit(X_train_ros,y_train_ros)  
lgbc_ros
```

Wall time: 56.9 s

```
[131]: LGBMClassifier(random_state=0)
```

```
[133]: y_pred_lgbcros = lgbc_ros.predict(X_test)  
y_prob_pred_lgbcros = lgbc_ros.predict_proba(X_test)[:, 1]  
  
print("Y predicted : ",y_pred_lgbcros)  
print("Y probability predicted : ",y_prob_pred_lgbcros[:5])
```

Y predicted : [False False False ... False False False]
Y probability predicted : [0.23093612 0.12247041 0.13222139 0.56481875
0.10460942]

```
[135]: compute_evaluation_metric(lgbc_ros, X_test, y_test, y_pred_lgbcros,  
                                ↪y_prob_pred_lgbcros)
```

Accuracy Score : 0.886696921461713

AUC Score : 0.9271778128567145

Confusion Matrix :

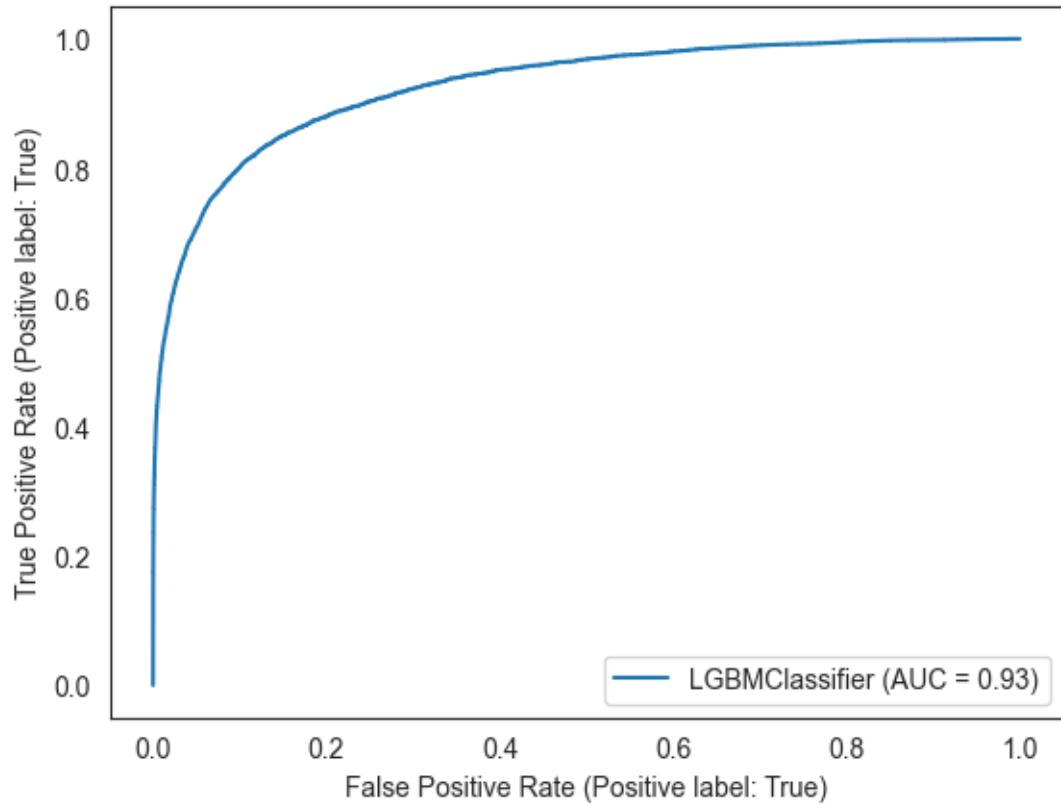
```
[[152037 18926]  
 [ 1147 5052]]
```

Classification Report :

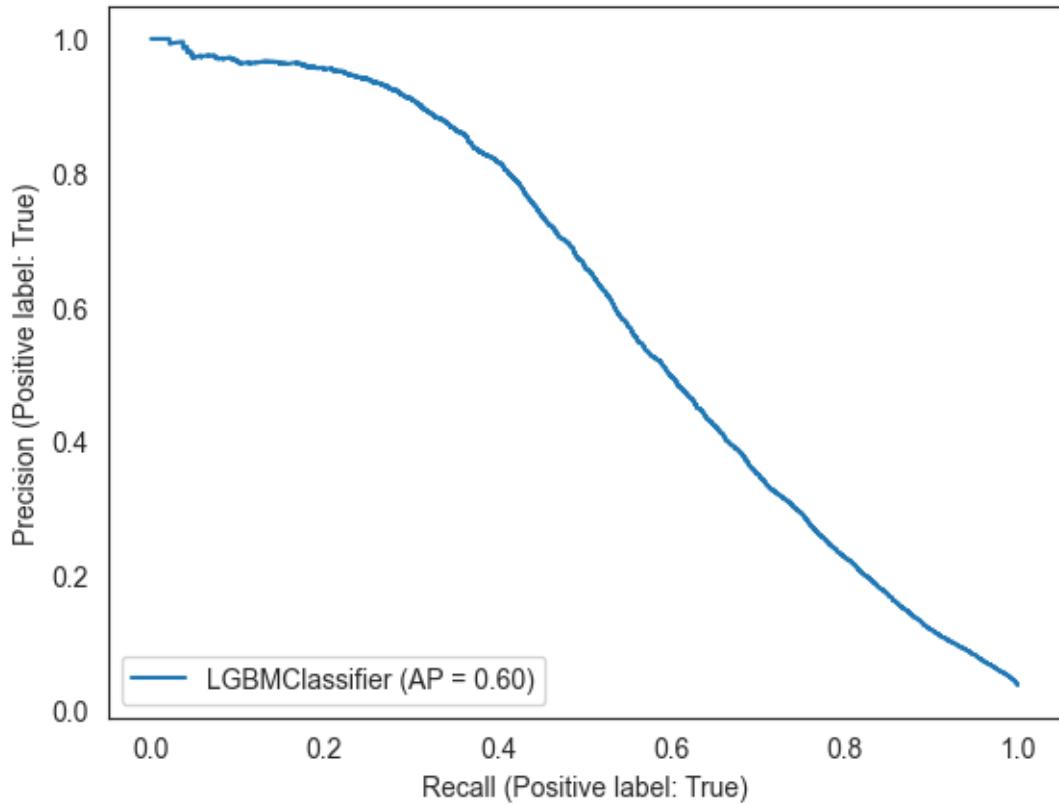
	precision	recall	f1-score	support
False	0.99	0.89	0.94	170963
True	0.21	0.81	0.33	6199
accuracy			0.89	177162
macro avg	0.60	0.85	0.64	177162
weighted avg	0.97	0.89	0.92	177162

Concordance Index : 0.9271778029491814

ROC curve :



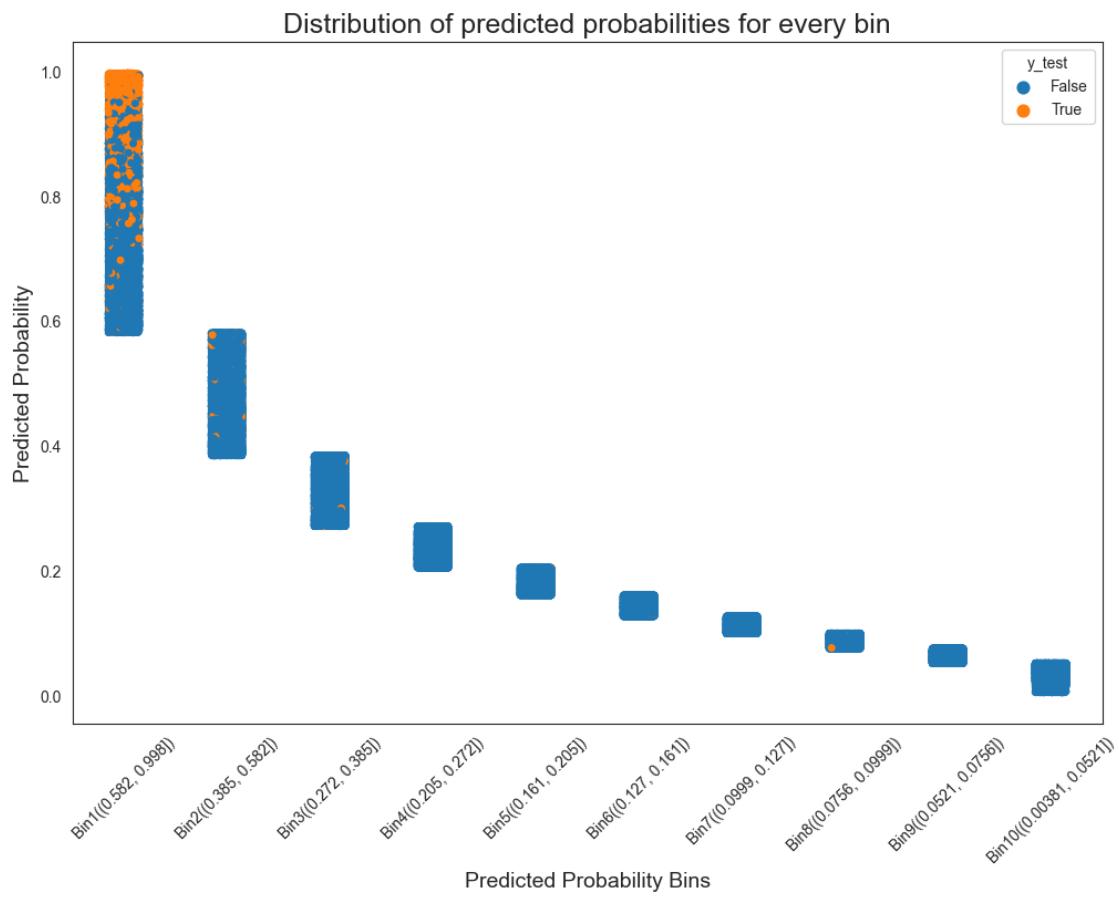
PR curve :

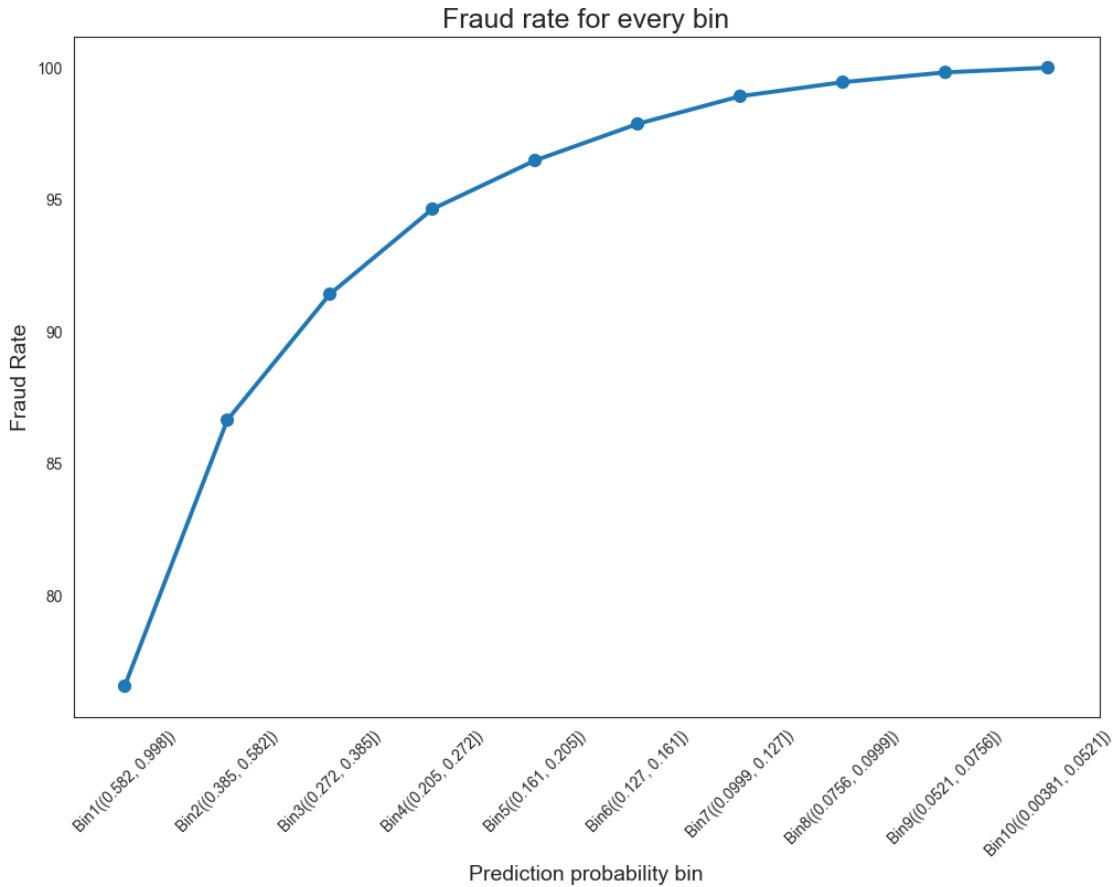


Additional Metrics:

TPR (Recall) : 0.8150
FPR : 0.1107
TNR (Specificity) : 0.8893
FNR : 0.1850

```
[136]: captures(y_test, y_pred_lgbcros, y_prob_pred_lgbcros)
```





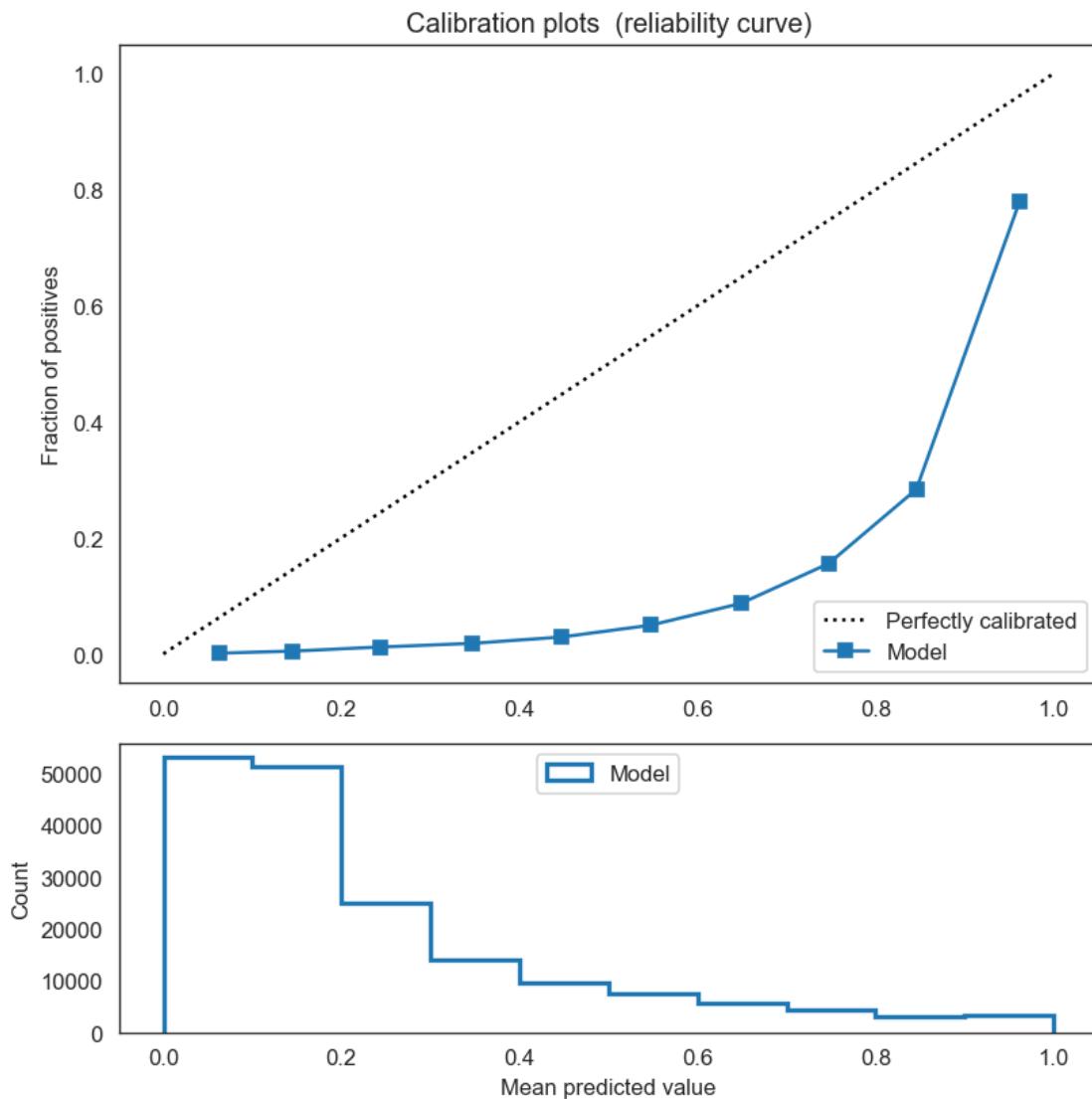
```
[136]:
```

	prob_bin	not_fraud	fraud	perc_fraud	perc_not_fraud	\
0	Bin1((0.582, 0.998])	12971	4746	0.765607	0.075870	
1	Bin2((0.385, 0.582])	17091	625	0.100823	0.099969	
2	Bin3((0.272, 0.385])	17420	296	0.047750	0.101893	
3	Bin4((0.205, 0.272])	17516	200	0.032263	0.102455	
4	Bin5((0.161, 0.205])	17602	114	0.018390	0.102958	
5	Bin6((0.127, 0.161])	17630	86	0.013873	0.103122	
6	Bin7((0.0999, 0.127])	17651	65	0.010486	0.103245	
7	Bin8((0.0756, 0.0999])	17683	33	0.005323	0.103432	
8	Bin9((0.0521, 0.0756])	17693	23	0.003710	0.103490	
9	Bin10((0.00381, 0.0521])	17706	11	0.001774	0.103566	

	cum_perc_fraud	cum_perc_not_fraud
0	76.560736	7.587022
1	86.643007	17.583922
2	91.417971	27.773261
3	94.644297	38.018753
4	96.483304	48.314548
5	97.870624	58.626720

6	98.919181	68.951177
7	99.451524	79.294350
8	99.822552	89.643373
9	100.000000	100.000000

```
[137]: draw_calibration_curve(y_test, y_prob_pred_lgbcros, n_bins=10)
```



- After balancing the class, accuracy score is 0.88 and AUC score is 92.5%
- Accuracy has decreased as compared to the previous model, but AUC has improved
- Additionally the recall has improved significantly at the cost of precision.

17 16. Cost Sensitive Learning with Class weights

```
[139]: %%time
lgbc_bal = LGBMClassifier(random_state=0, class_weight='balanced')
lgbc_bal.fit(X_train, y_train)
lgbc_bal
```

Wall time: 32 s

```
[139]: LGBMClassifier(class_weight='balanced', random_state=0)
```

```
[140]: y_pred_lgbcbal = lgbc_bal.predict(X_test)
y_prob_pred_lgbcbal = lgbc_bal.predict_proba(X_test)[:, 1]

print("Y predicted : ",y_pred_lgbcbal)
print("Y probability predicted : ",y_prob_pred_lgbcbal[:5])
```

Y predicted : [False False False ... False False False]
Y probability predicted : [0.22700362 0.15390235 0.13087413 0.52349433
0.14032882]

```
[141]: compute_evaluation_metric(lgbc_bal, X_test, y_test, y_pred_lgbcbal, y_prob_pred_lgbcbal)
```

Accuracy Score : 0.8862284237025999

AUC Score : 0.927124006459723

Confusion Matrix :

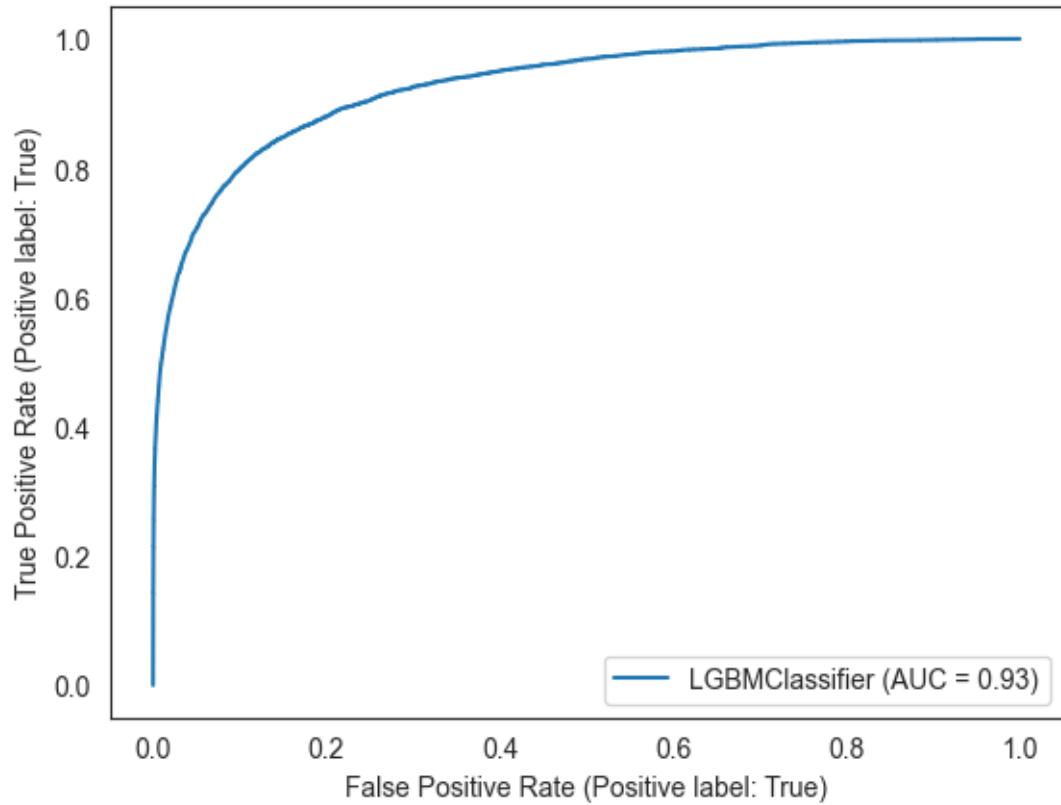
```
[[151969 18994]
 [ 1162  5037]]
```

Classification Report :

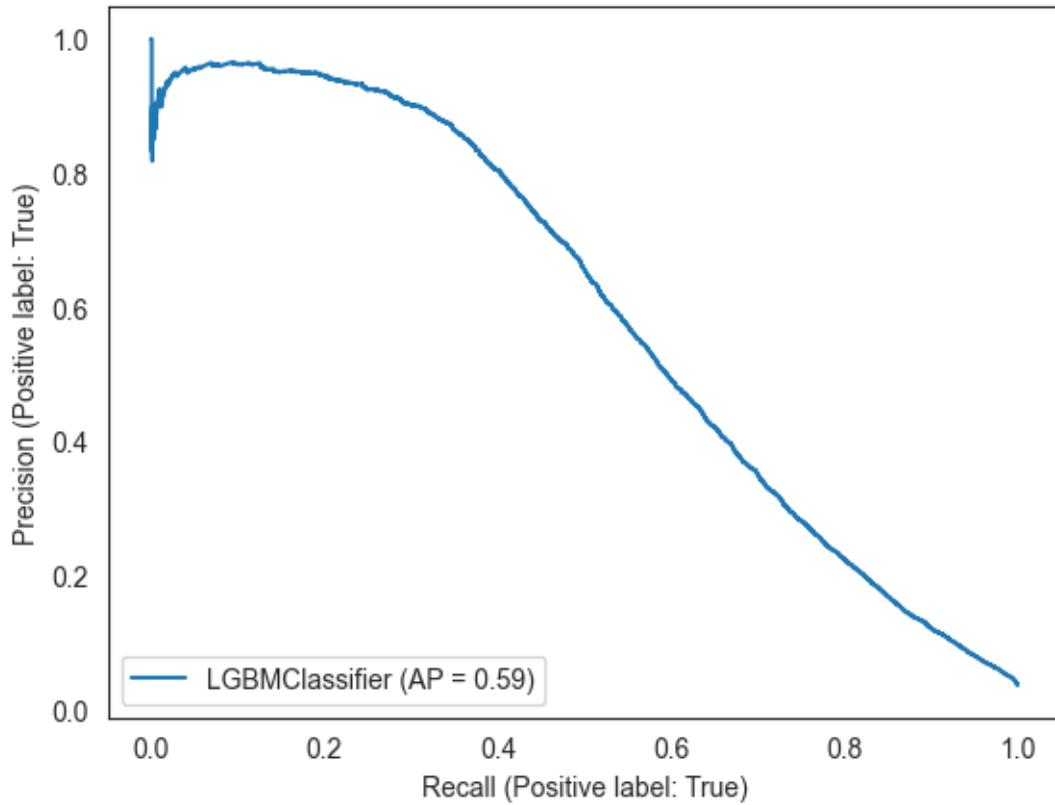
	precision	recall	f1-score	support
False	0.99	0.89	0.94	170963
True	0.21	0.81	0.33	6199
accuracy			0.89	177162
macro avg	0.60	0.85	0.64	177162
weighted avg	0.97	0.89	0.92	177162

Concordance Index : 0.9271239946650406

ROC curve :



PR curve :



Additional Metrics:

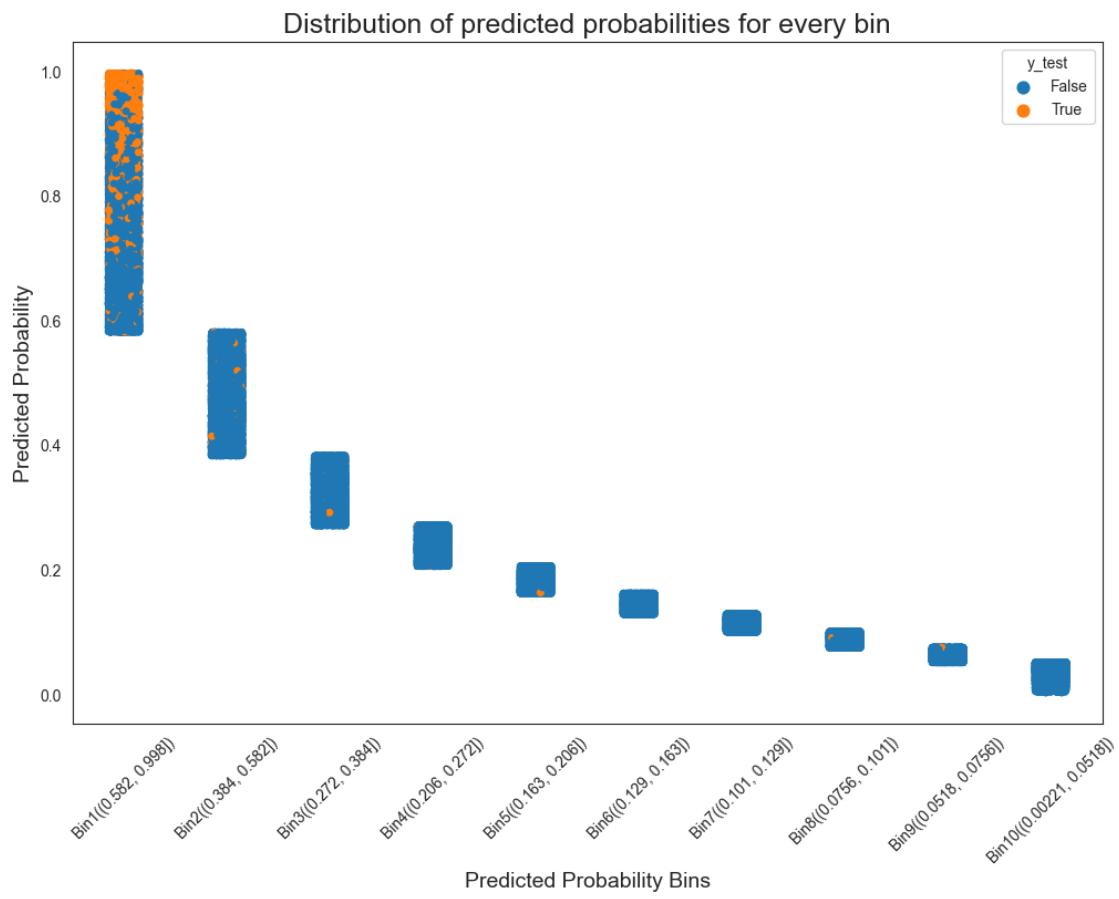
TPR (Recall) : 0.8126

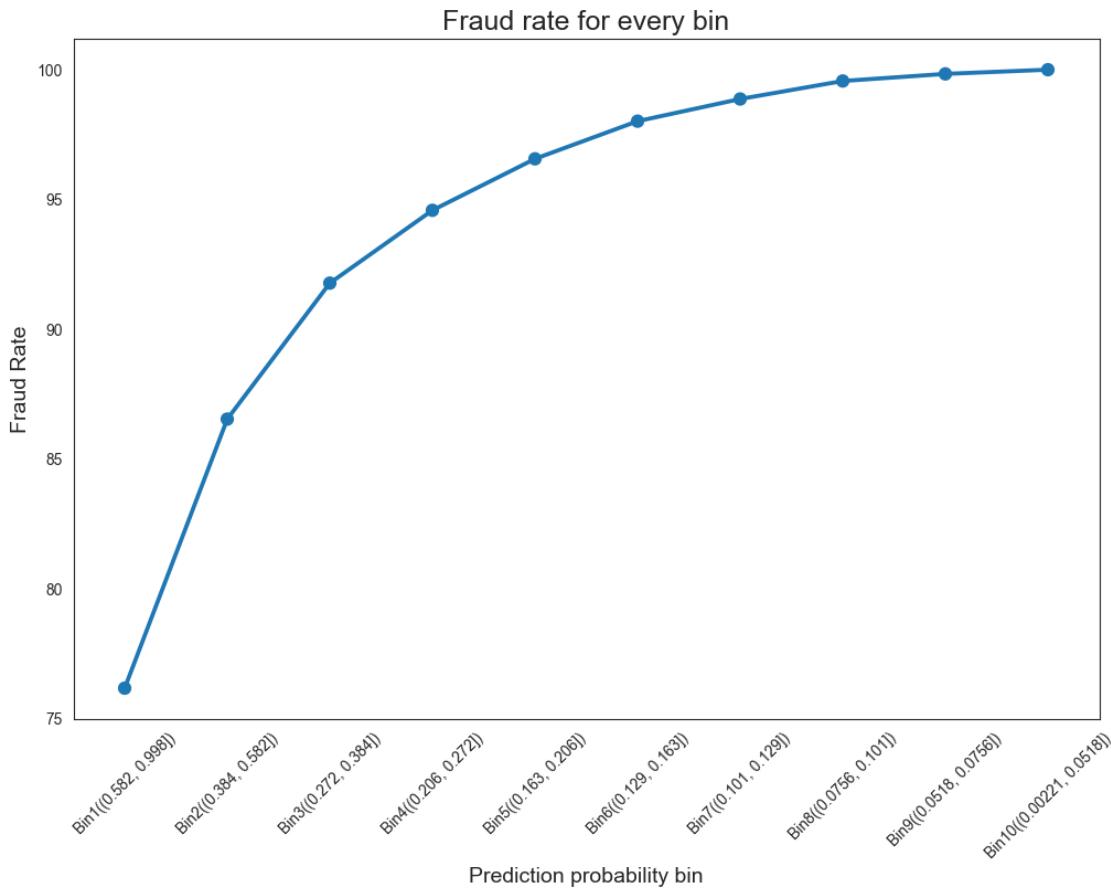
FPR : 0.1111

TNR (Specificity) : 0.8889

FNR : 0.1874

```
[142]: captures(y_test, y_pred_lgbcbal, y_prob_pred_lgbcbal)
```





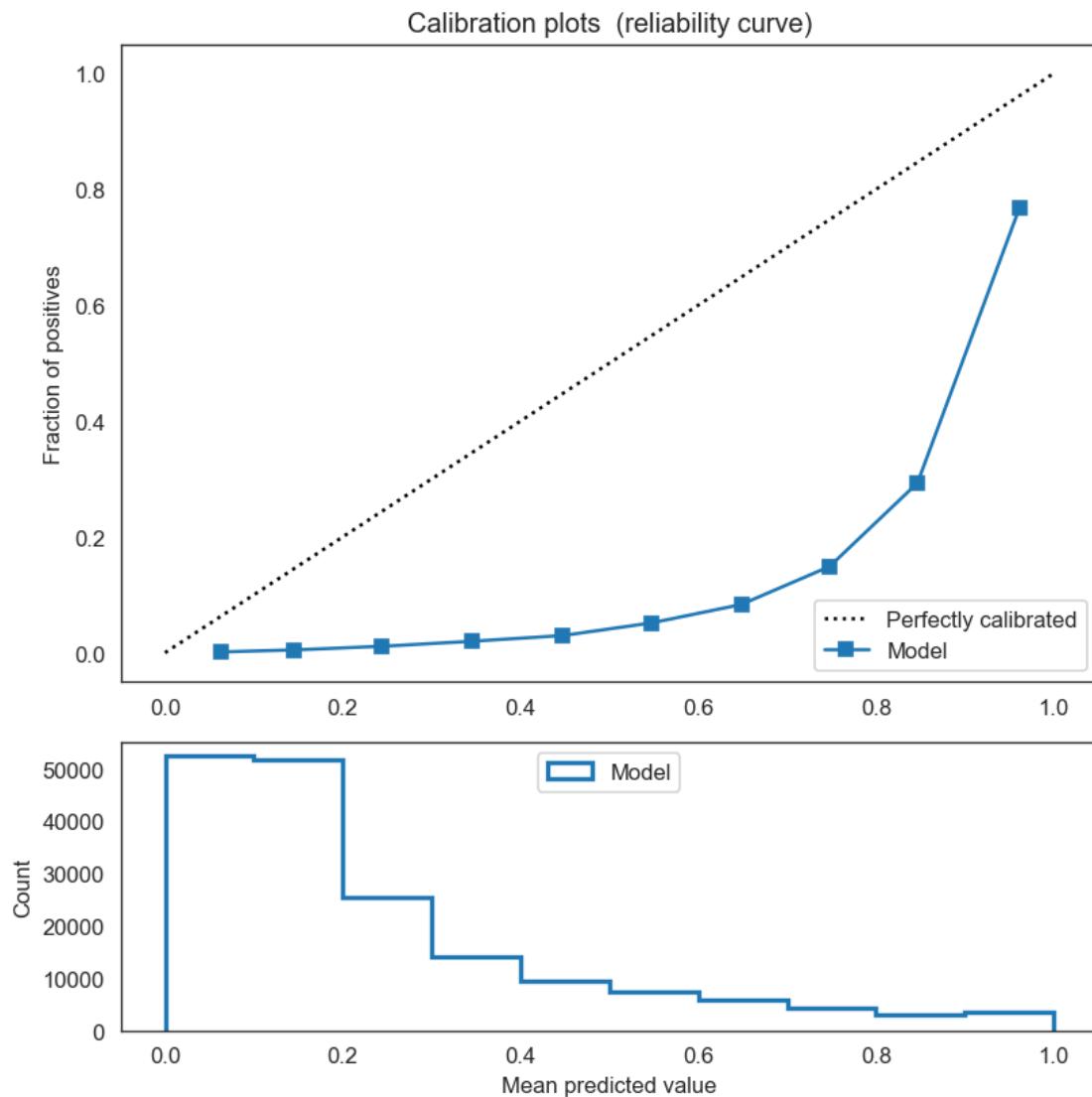
```
[142]:
```

	prob_bin	not_fraud	fraud	perc_fraud	perc_not_fraud	\
0	Bin1((0.582, 0.998])	12996	4721	0.761574	0.076016	
1	Bin2((0.384, 0.582])	17073	643	0.103726	0.099864	
2	Bin3((0.272, 0.384])	17391	325	0.052428	0.101724	
3	Bin4((0.206, 0.272])	17542	174	0.028069	0.102607	
4	Bin5((0.163, 0.206])	17593	123	0.019842	0.102905	
5	Bin6((0.129, 0.163])	17626	90	0.014518	0.103098	
6	Bin7((0.101, 0.129])	17663	53	0.008550	0.103315	
7	Bin8((0.0756, 0.101])	17673	43	0.006937	0.103373	
8	Bin9((0.0518, 0.0756])	17699	17	0.002742	0.103525	
9	Bin10((0.00221, 0.0518])	17707	10	0.001613	0.103572	

	cum_perc_fraud	cum_perc_not_fraud
0	76.157445	7.601645
1	86.530085	17.588016
2	91.772867	27.760393
3	94.579771	38.021092
4	96.563962	48.311623
5	98.015809	58.621456

6	98.870786	68.952931
7	99.564446	79.290256
8	99.838684	89.642788
9	100.000000	100.000000

```
[143]: draw_calibration_curve(y_test, y_prob_pred_lgbcbal, n_bins=10)
```



18 17. Model Calibration

```
[144]: from sklearn.calibration import CalibratedClassifierCV
```

```
[145]: lgbc_bal = LGBMClassifier(random_state=0)
calibrated_clf = CalibratedClassifierCV(base_estimator=lgbc_bal, cv=3, method='sigmoid')
calibrated_clf.fit(X_train, y_train)
```

```
[145]: CalibratedClassifierCV(base_estimator=LGBMClassifier(random_state=0), cv=3)
```

```
[146]: y_pred_calib = calibrated_clf.predict(X_test)
y_prob_pred_calib = calibrated_clf.predict_proba(X_test)[:, 1]
```

```
[147]: len(calibrated_clf.calibrated_classifiers_)
```

```
[147]: 3
```

```
[148]: print("Y predicted : ", y_pred_calib)
print("Y probability predicted : ", y_prob_pred_calib[:5])
```

```
Y predicted : [False False False ... False False False]
Y probability predicted : [0.01510527 0.013539 0.01360709 0.0235388
 0.01342325]
```

```
[150]: compute_evaluation_metric(calibrated_clf, X_test, y_test, y_pred_calib, y_prob_pred_calib)
```

Accuracy Score : 0.9781160745532338

AUC Score : 0.9292560410642979

Confusion Matrix :

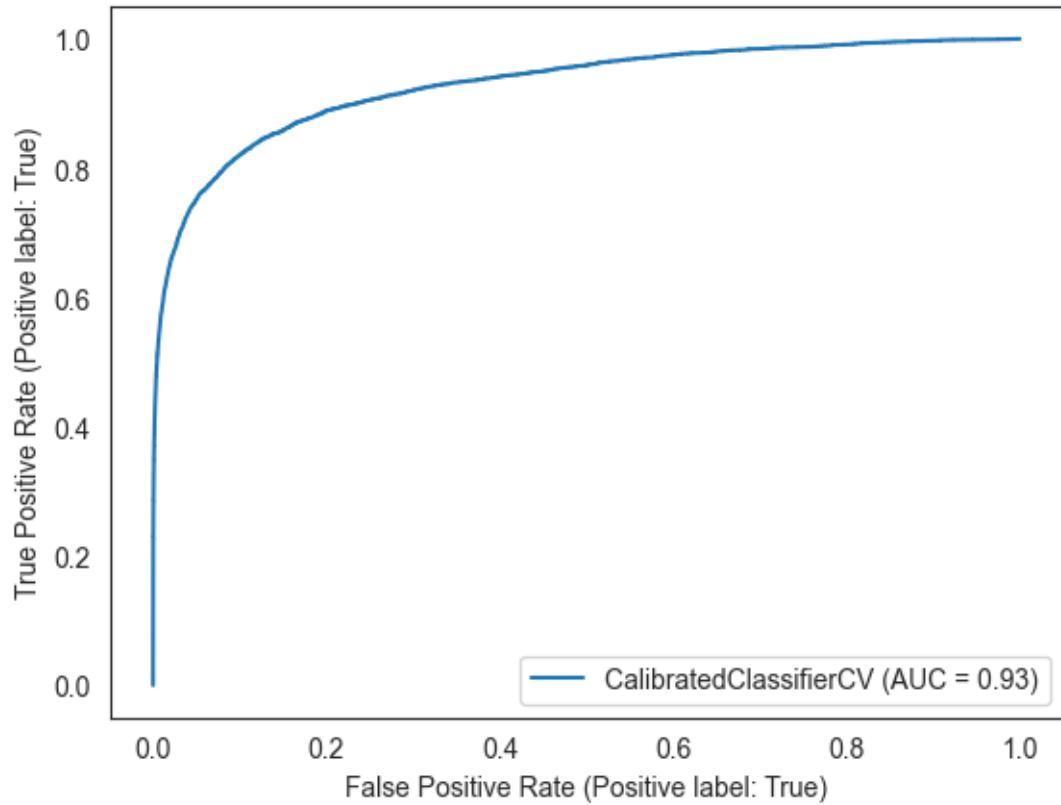
```
[[170502 461]
 [ 3416 2783]]
```

Classification Report :

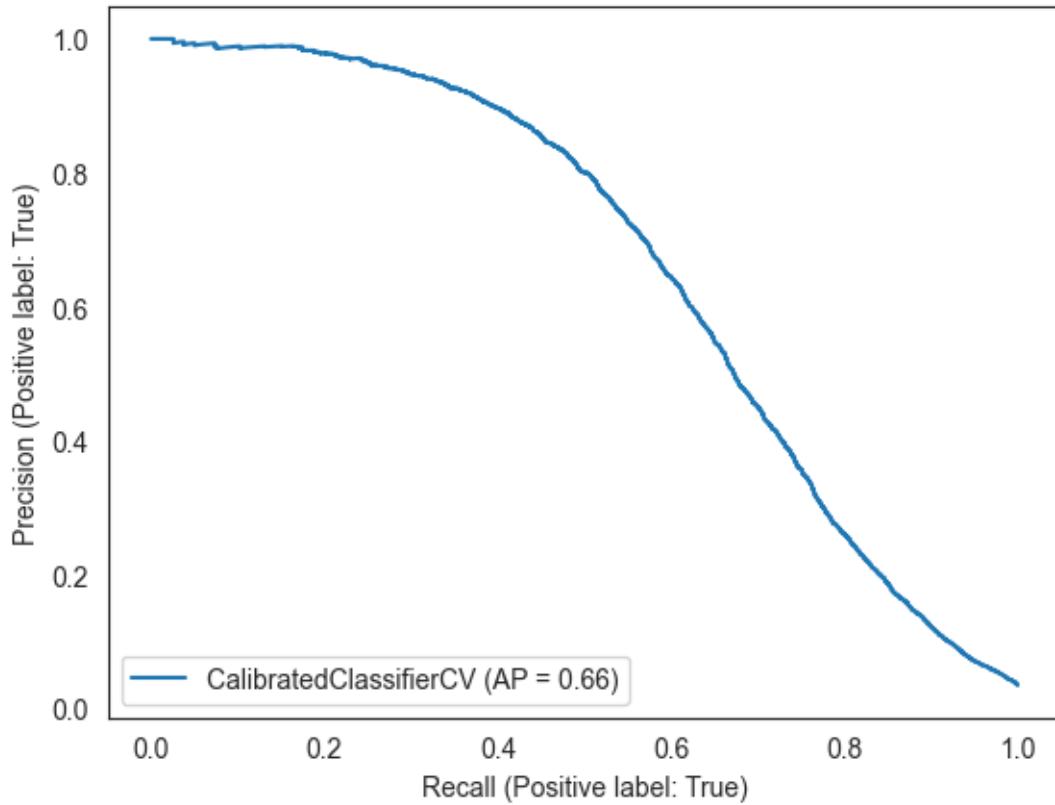
	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.86	0.45	0.59	6199
accuracy			0.98	177162
macro avg	0.92	0.72	0.79	177162
weighted avg	0.98	0.98	0.97	177162

Concordance Index : 0.9292560391771487

ROC curve :



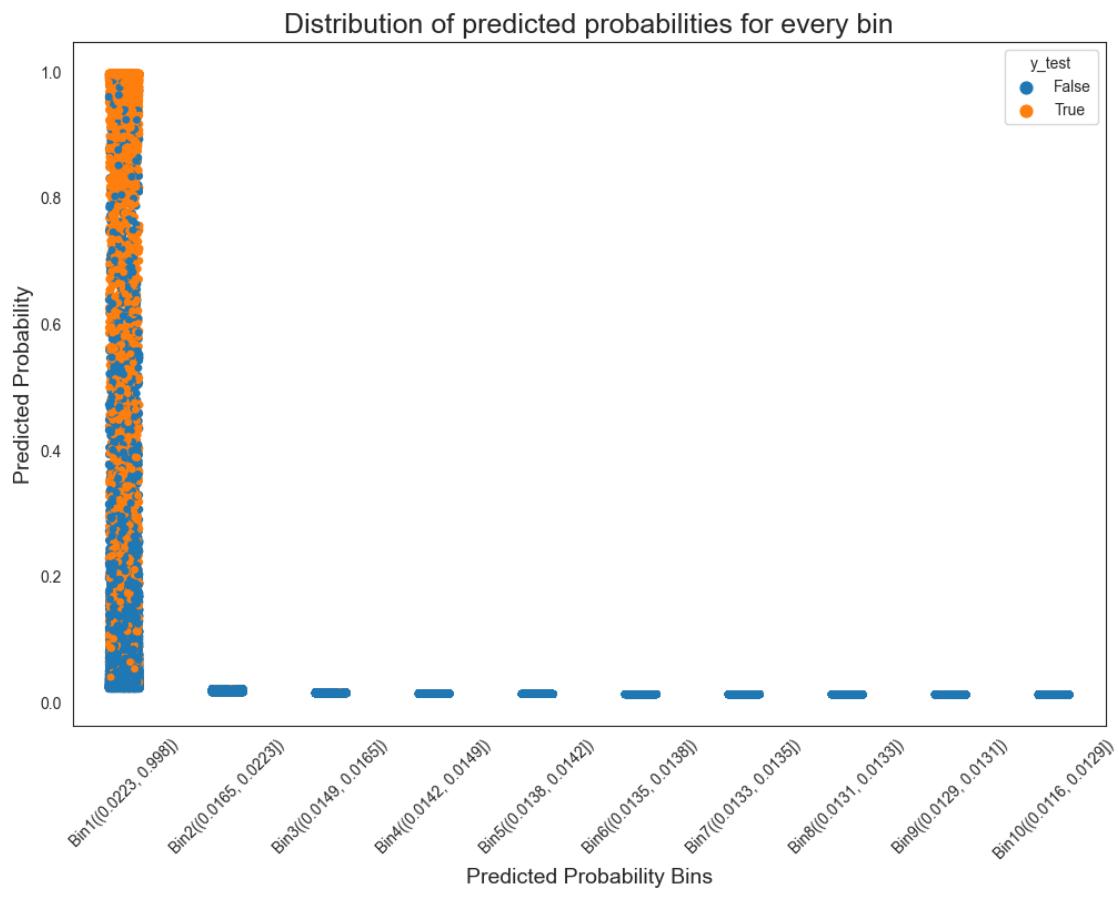
PR curve :

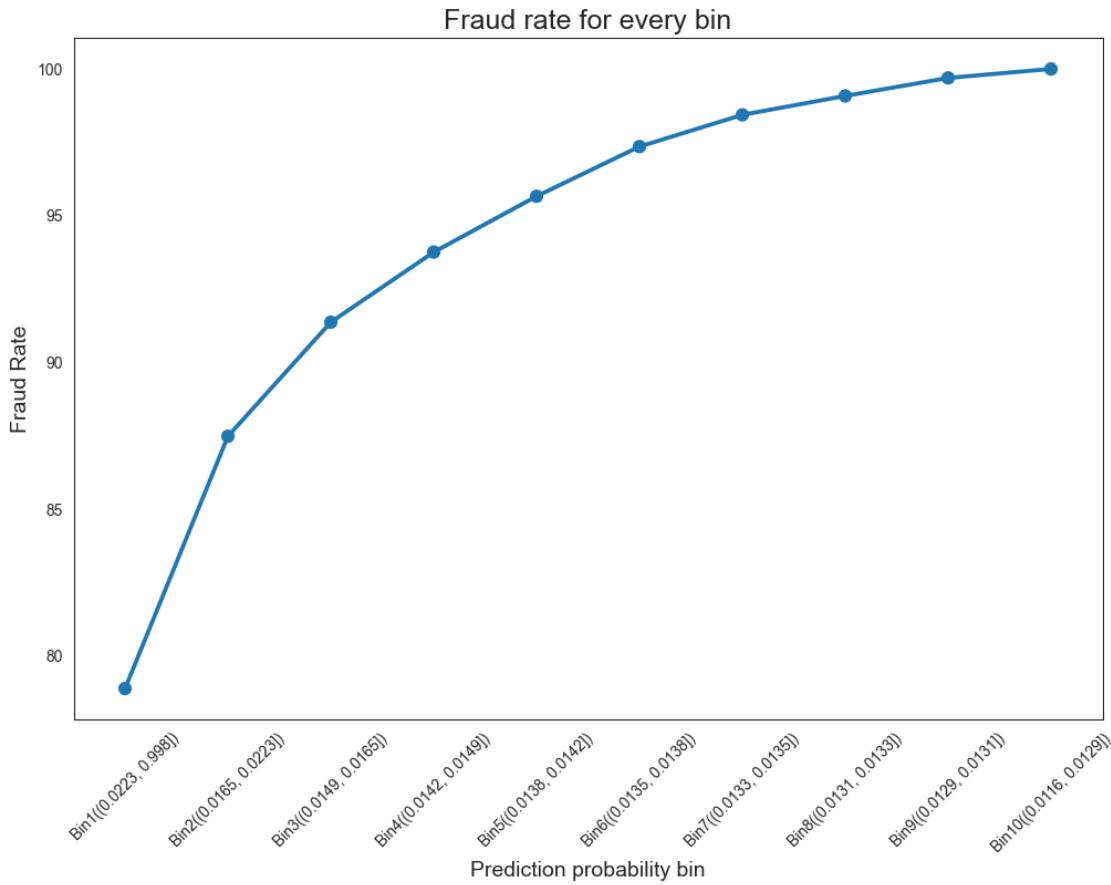


Additional Metrics:

TPR (Recall) : 0.4489
FPR : 0.0027
TNR (Specificity) : 0.9973
FNR : 0.5511

```
[151]: captures(y_test, y_pred_calib, y_prob_pred_calib)
```





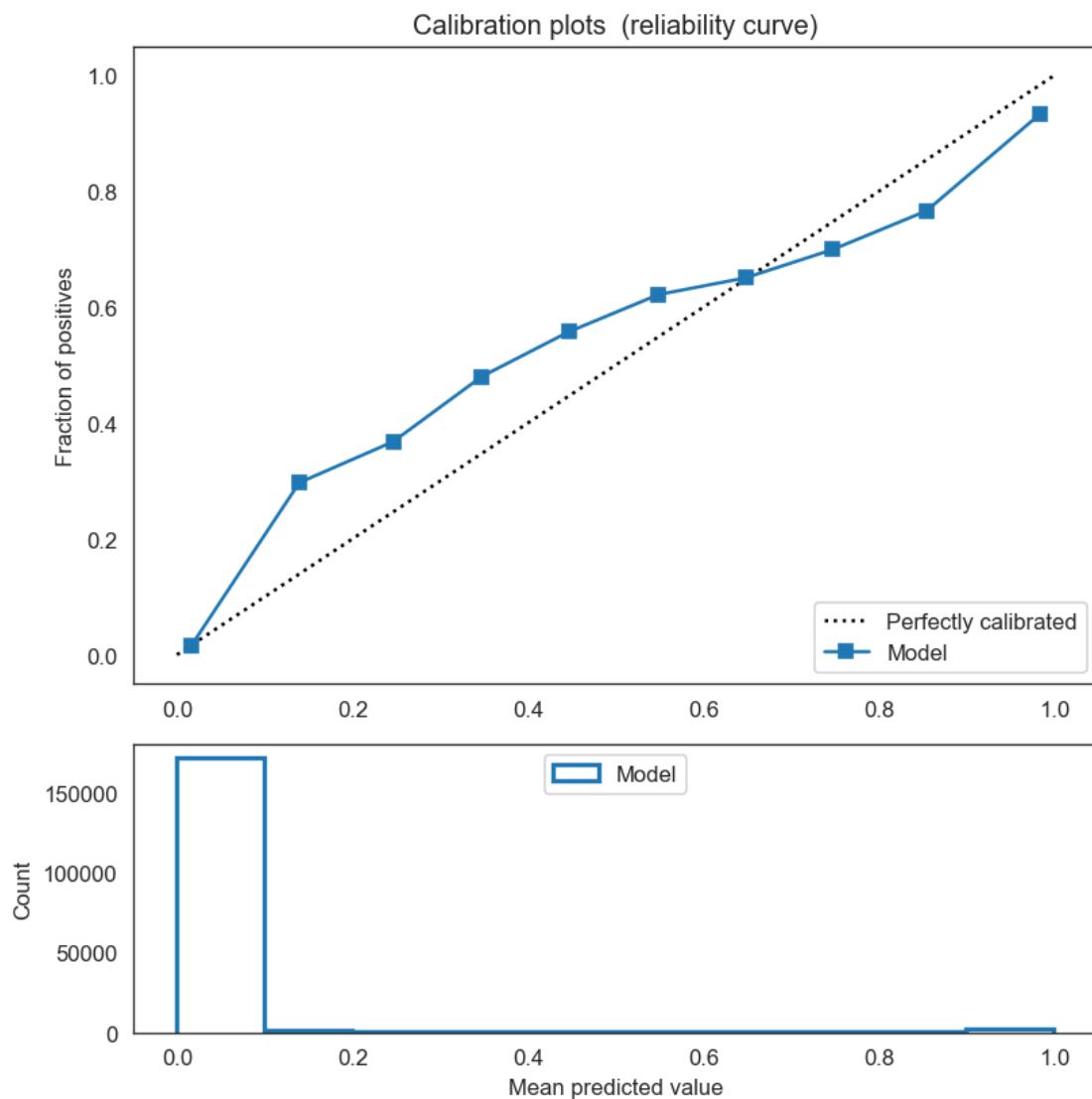
```
[151]:
```

	prob_bin	not_fraud	fraud	perc_fraud	perc_not_fraud	\
0	Bin1((0.0223, 0.998])	12827	4890	0.788837	0.075028	
1	Bin2((0.0165, 0.0223])	17183	533	0.085982	0.100507	
2	Bin3((0.0149, 0.0165])	17475	241	0.038877	0.102215	
3	Bin4((0.0142, 0.0149])	17568	148	0.023875	0.102759	
4	Bin5((0.0138, 0.0142])	17598	118	0.019035	0.102935	
5	Bin6((0.0135, 0.0138])	17611	105	0.016938	0.103011	
6	Bin7((0.0133, 0.0135])	17649	67	0.010808	0.103233	
7	Bin8((0.0131, 0.0133])	17676	40	0.006453	0.103391	
8	Bin9((0.0129, 0.0131])	17678	38	0.006130	0.103402	
9	Bin10((0.0116, 0.0129))	17698	19	0.003065	0.103519	

	cum_perc_fraud	cum_perc_not_fraud
0	78.883691	7.502793
1	87.481852	17.553506
2	91.369576	27.775016
3	93.757058	38.050923
4	95.660590	48.344379
5	97.354412	58.645438

6	98.435231	68.968724
7	99.080497	79.307803
8	99.693499	89.648053
9	100.000000	100.000000

```
[152]: draw_calibration_curve(y_test, y_prob_pred_calib, n_bins=10)
```



19 18. Model Tuning

```
[153]: %%time
```

```
lgbmclassifier = LGBMClassifier(  
    class_weight=None,  
    learning_rate=0.1,  
    max_depth=8,  
    n_estimators=100,  
    num_leaves=256,  
    reg_alpha=0.5,  
    random_state=0  
)  
  
lgbmclassifier.fit(X_train, y_train)
```

Wall time: 34.7 s

```
[153]: LGBMClassifier(max_depth=8, num_leaves=256, random_state=0, reg_alpha=0.5)
```

```
[154]: y_g_pred = lgbmclassifier.predict(X_test)  
y_prob_g_pred = lgbmclassifier.predict_proba(X_test)[:, 1]  
print("Y predicted : ", y_g_pred)  
print("Y probability predicted : ", y_prob_g_pred[:5])
```

```
Y predicted : [False False False ... False False False]  
Y probability predicted : [0.00549227 0.00615294 0.007173 0.09576845  
0.00734595]
```

```
[155]: compute_evaluation_metric(lgbmclassifier, X_test, y_test, y_g_pred,  
                                y_prob_g_pred)
```

Accuracy Score : 0.9796909043700116

AUC Score : 0.9439534102237099

Confusion Matrix :

```
[[170676 287]  
 [ 3311 2888]]
```

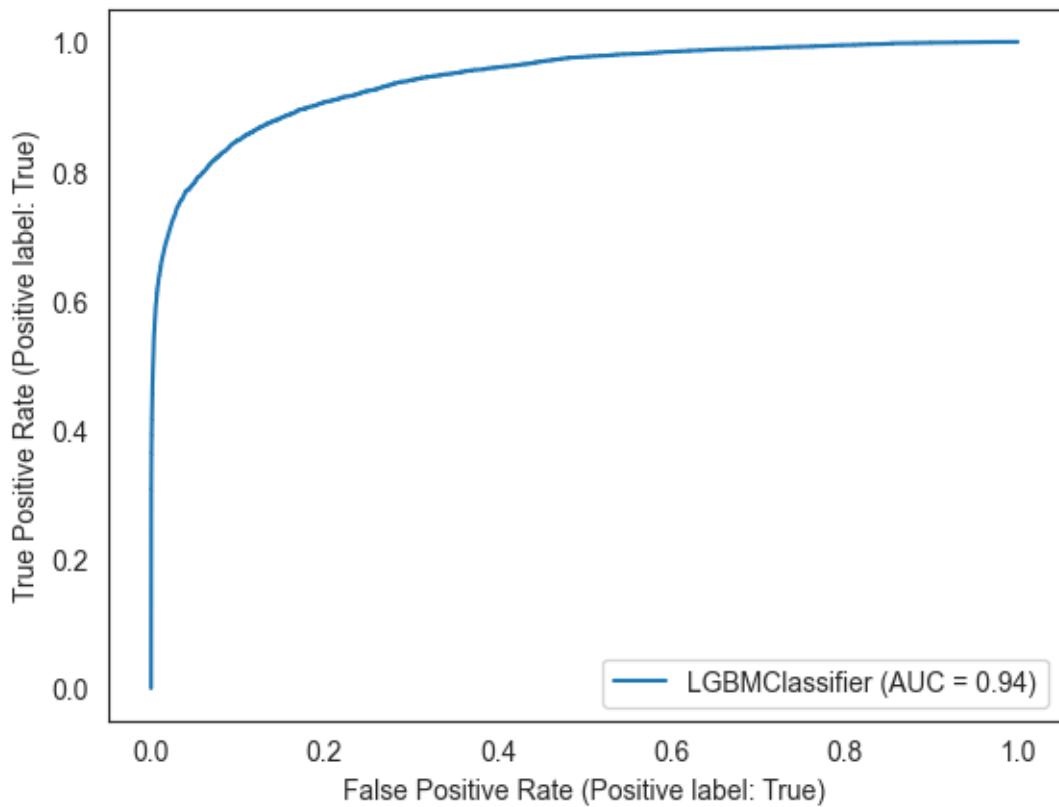
Classification Report :

	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.91	0.47	0.62	6199
accuracy			0.98	177162

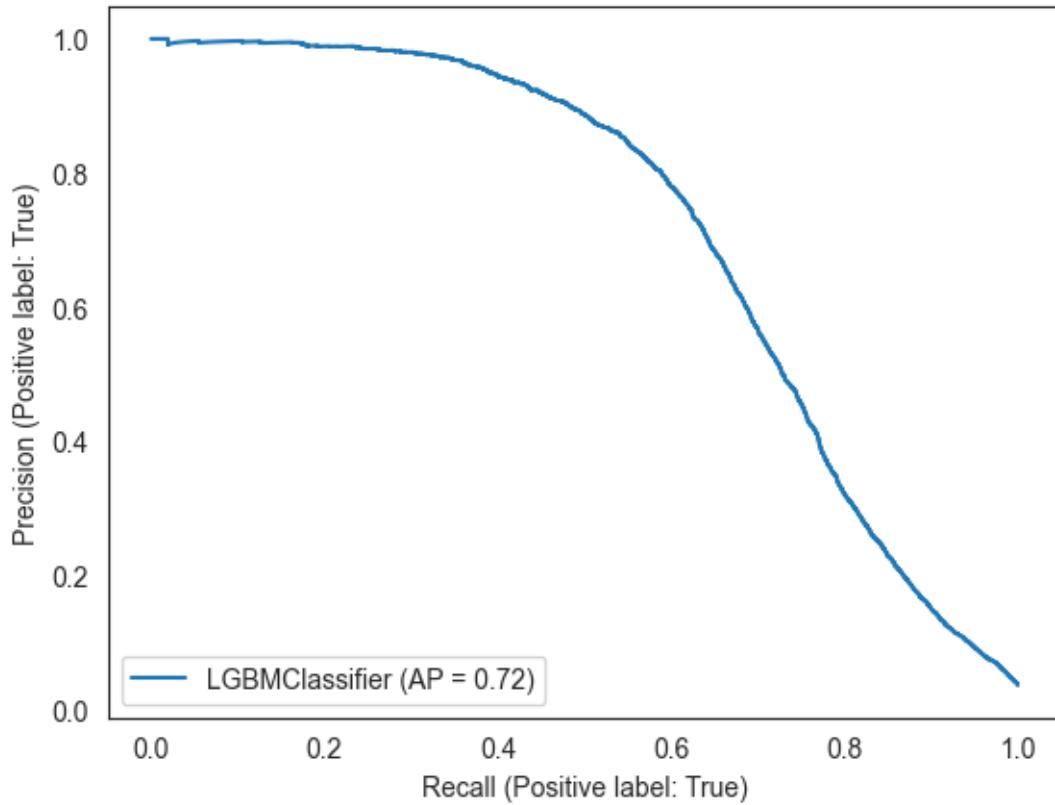
macro avg	0.95	0.73	0.80	177162
weighted avg	0.98	0.98	0.98	177162

Concordance Index : 0.9439534040904752

ROC curve :



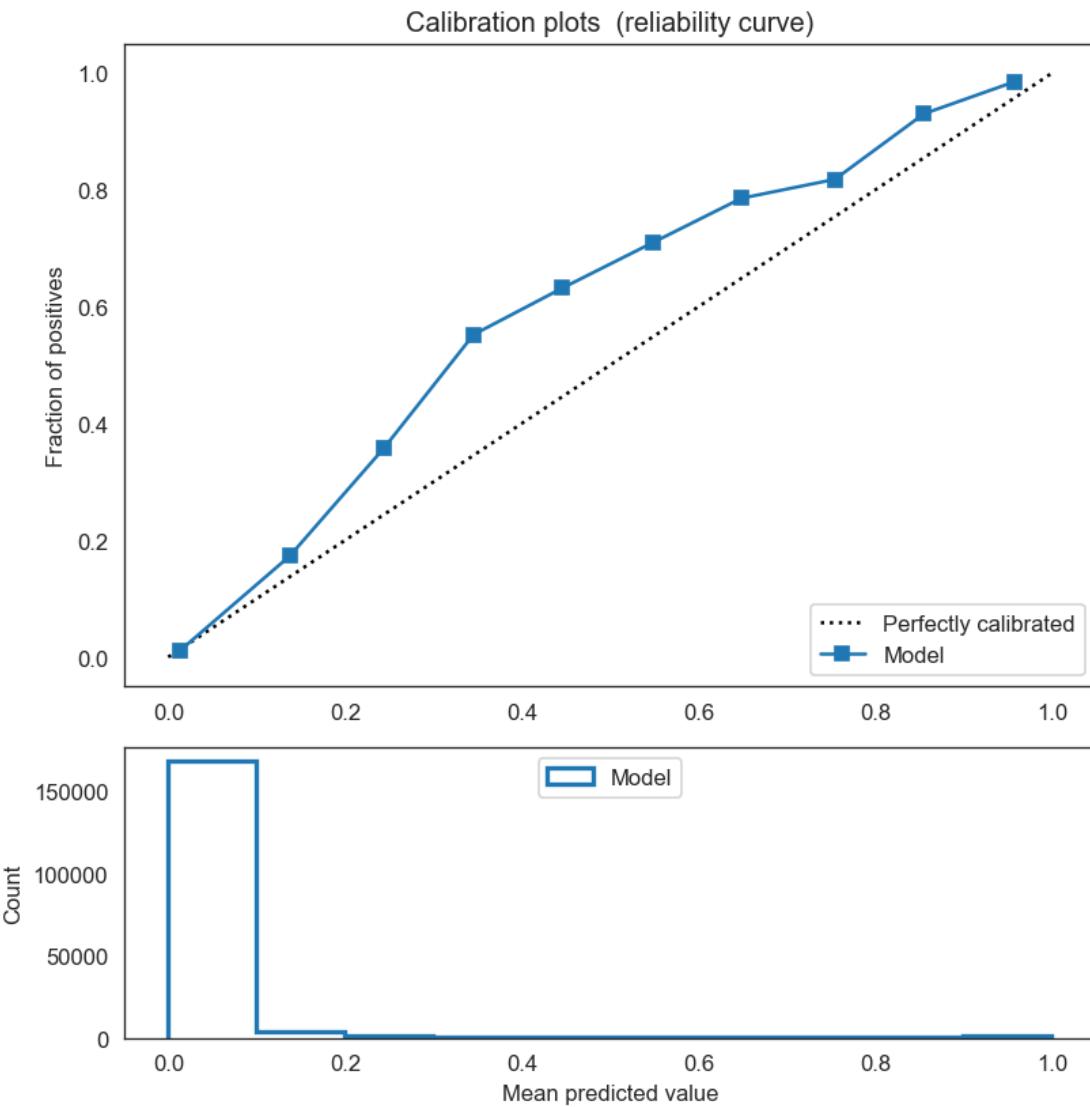
PR curve :



Additional Metrics:

TPR (Recall) : 0.4659
FPR : 0.0017
TNR (Specificity) : 0.9983
FNR : 0.5341

```
[156]: draw_calibration_curve(y_test, y_prob_g_pred, n_bins=10)
```



```
[157]: # Calibrate
calibrated_clf = CalibratedClassifierCV(base_estimator=lgbmclassifier, cv=3)
calibrated_clf.fit(X_train, y_train)
y_pred_calib = calibrated_clf.predict(X_test)
y_prob_pred_calib = calibrated_clf.predict_proba(X_test)[:, 1]
```

```
[158]: compute_evaluation_metric(calibrated_clf, X_test, y_test, y_pred_calib, y_prob_pred_calib)
```

Accuracy Score : 0.9804416296948556

AUC Score : 0.9456479526988175

Confusion Matrix :

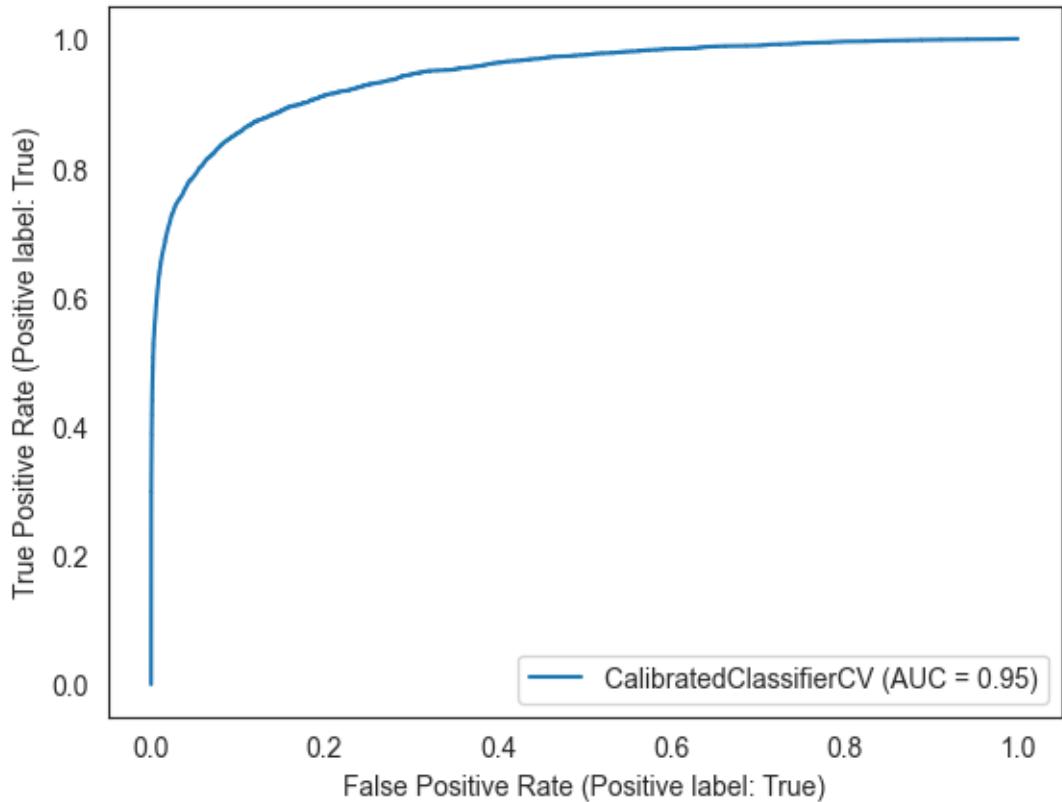
```
[[170572    391]
 [ 3074   3125]]
```

Classification Report :

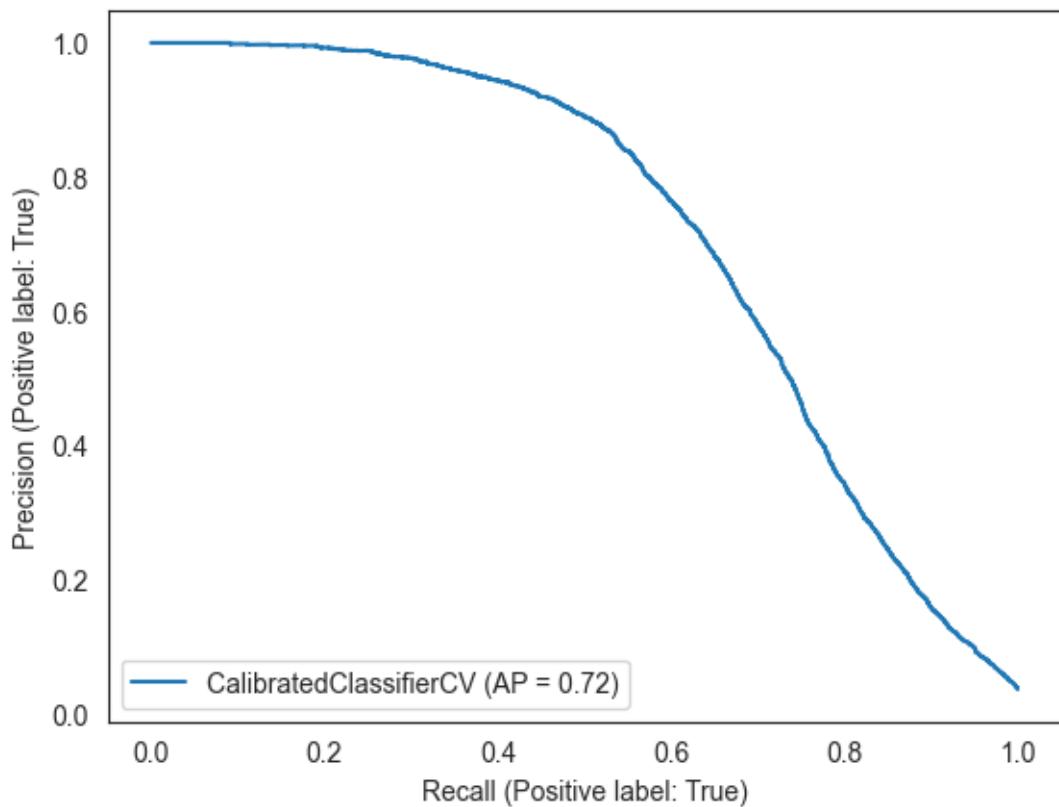
	precision	recall	f1-score	support
False	0.98	1.00	0.99	170963
True	0.89	0.50	0.64	6199
accuracy			0.98	177162
macro avg	0.94	0.75	0.82	177162
weighted avg	0.98	0.98	0.98	177162

Concordance Index : 0.9456479526988175

ROC curve :



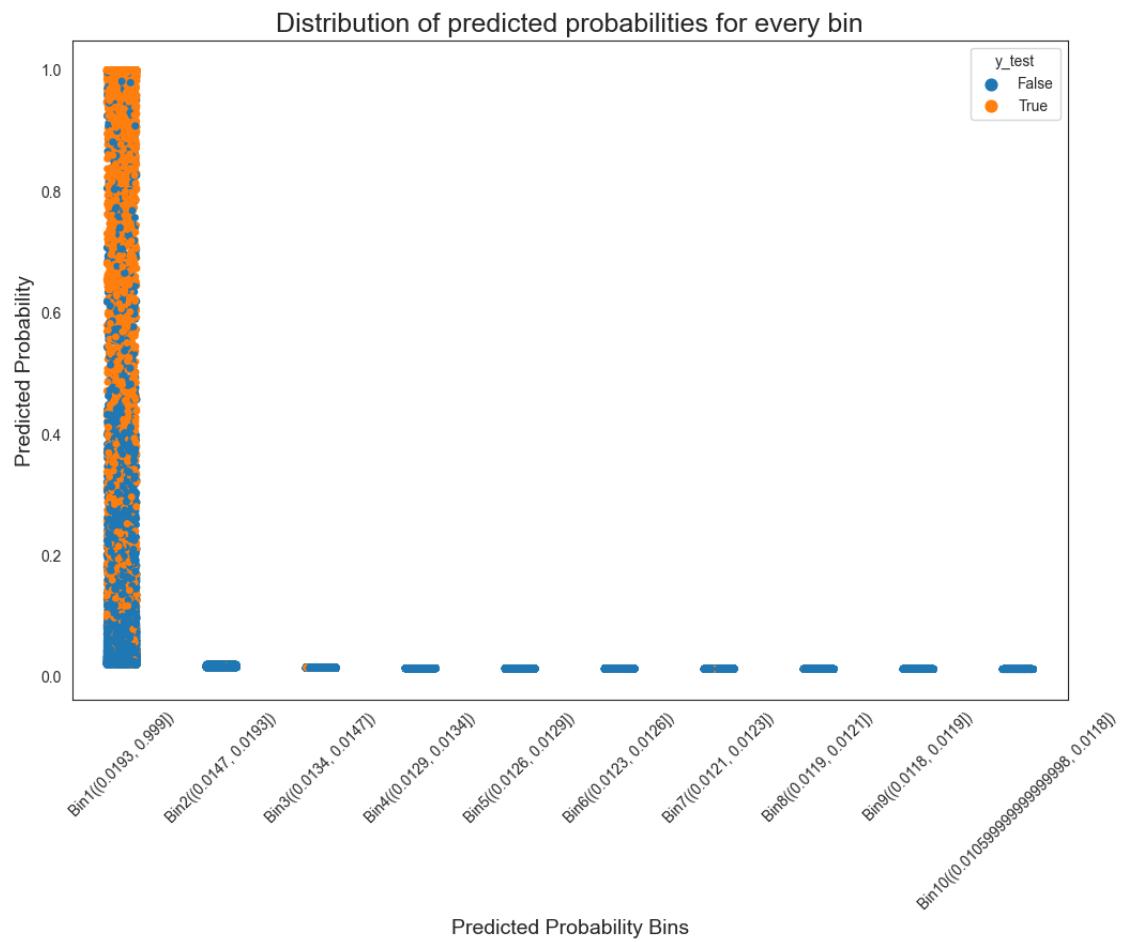
PR curve :

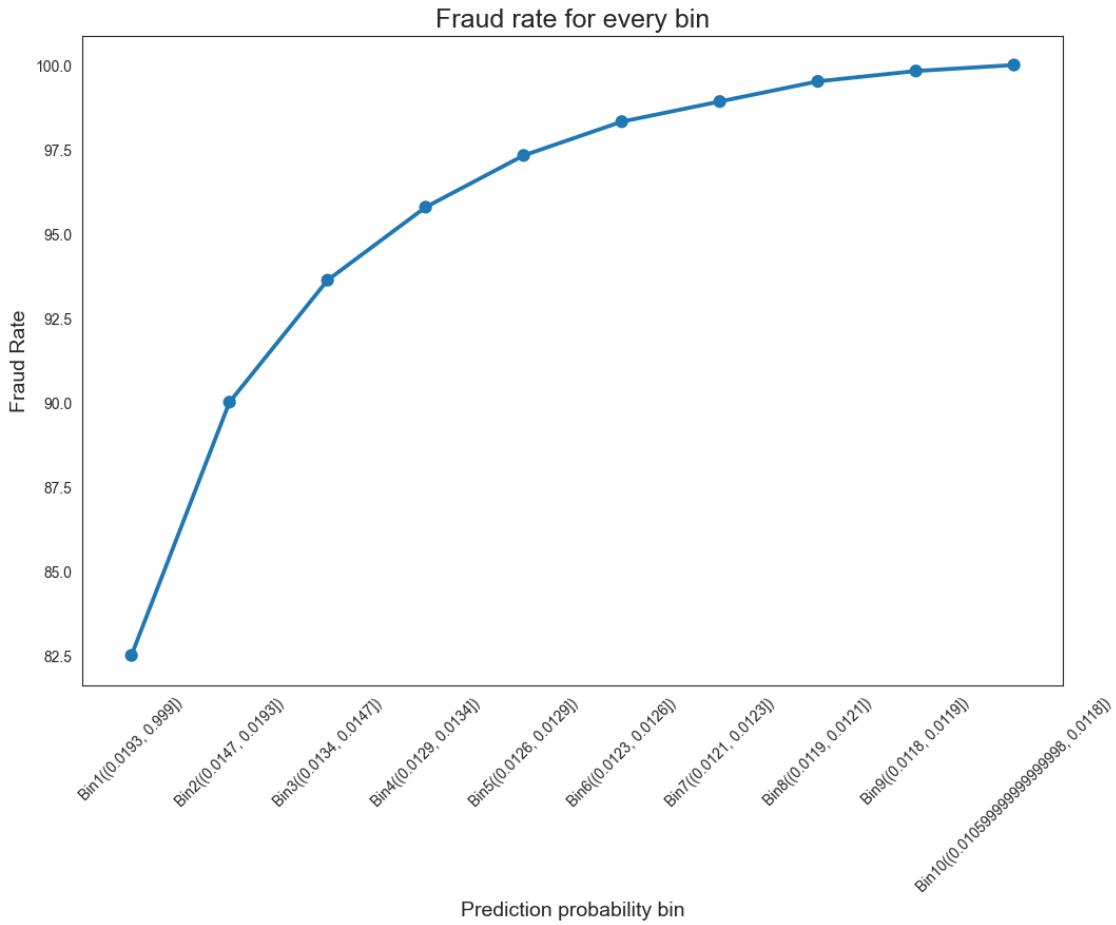


Additional Metrics:

TPR (Recall) : 0.5041
FPR : 0.0023
TNR (Specificity) : 0.9977
FNR : 0.4959

```
[160]: captures(y_test, y_pred_calib, y_prob_pred_calib)
```



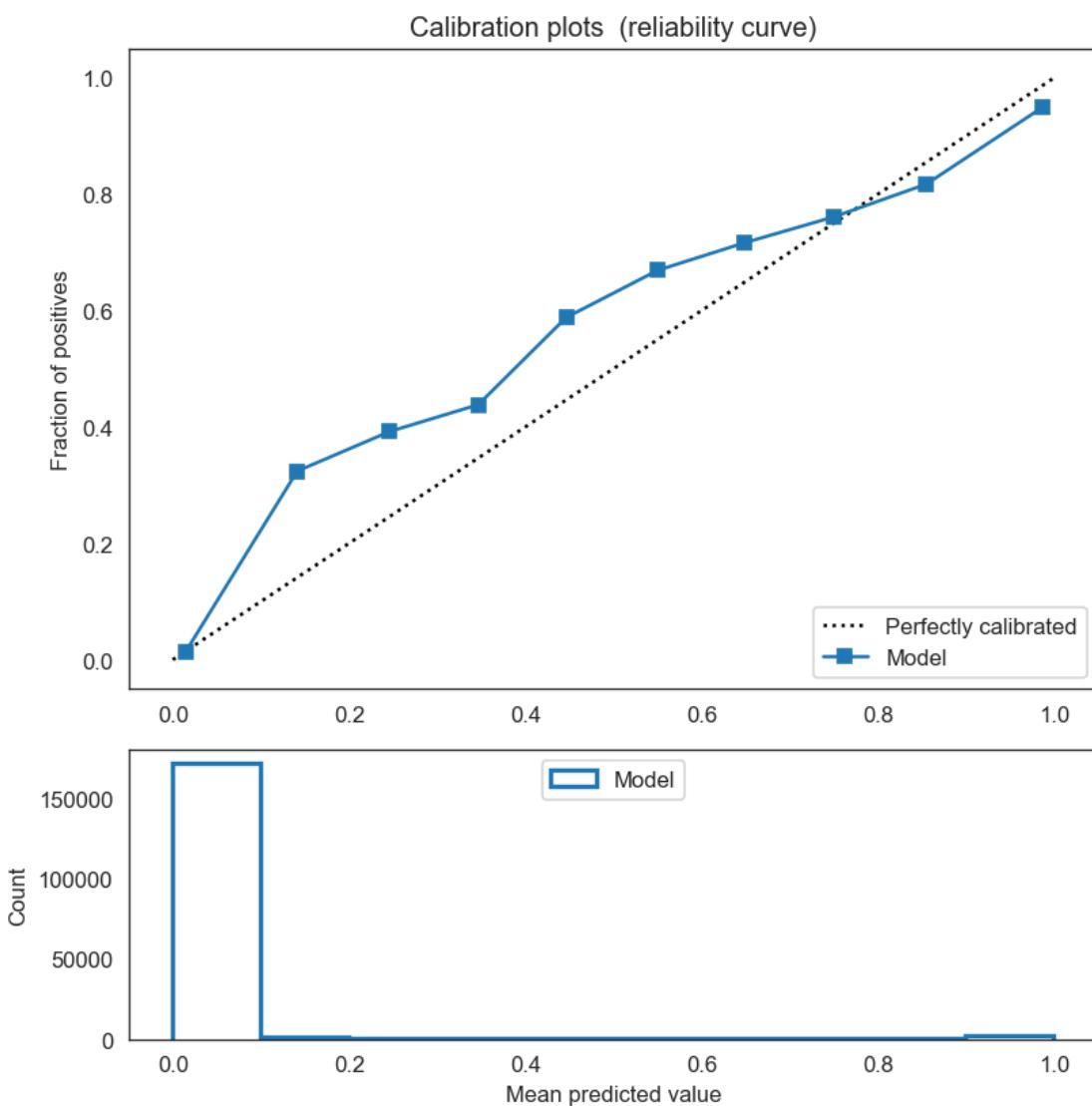


[160]:

	prob_bin	not_fraud	fraud	perc_fraud	\
0	Bin1((0.0193, 0.999])	12602	5115	0.825133	
1	Bin2((0.0147, 0.0193])	17251	465	0.075012	
2	Bin3((0.0134, 0.0147])	17492	224	0.036135	
3	Bin4((0.0129, 0.0134])	17582	134	0.021616	
4	Bin5((0.0126, 0.0129])	17621	95	0.015325	
5	Bin6((0.0123, 0.0126])	17654	62	0.010002	
6	Bin7((0.0121, 0.0123])	17679	37	0.005969	
7	Bin8((0.0119, 0.0121])	17679	37	0.005969	
8	Bin9((0.0118, 0.0119])	17697	19	0.003065	
9	Bin10((0.010599999999999998, 0.0118])	17706	11	0.001774	
	perc_not_fraud	cum_perc_fraud	cum_perc_not_fraud		
0	0.073712	82.513309	7.371186		
1	0.100905	90.014518	17.461673		
2	0.102315	93.628005	27.693127		
3	0.102841	95.789643	37.977223		
4	0.103069	97.322149	48.284132		

5	0.103262	98.322310	58.610343
6	0.103408	98.919181	68.951177
7	0.103408	99.516051	79.292011
8	0.103514	99.822552	89.643373
9	0.103566	100.000000	100.000000

```
[159]: draw_calibration_curve(y_test, y_prob_pred_calib, n_bins=10)
```



- Accuracy score is 0.98
- AUC score and Concordance index are 0.95, which are the best so far
- Classification report is also balanced between both the classes
- ROC curve and PR are the best so far

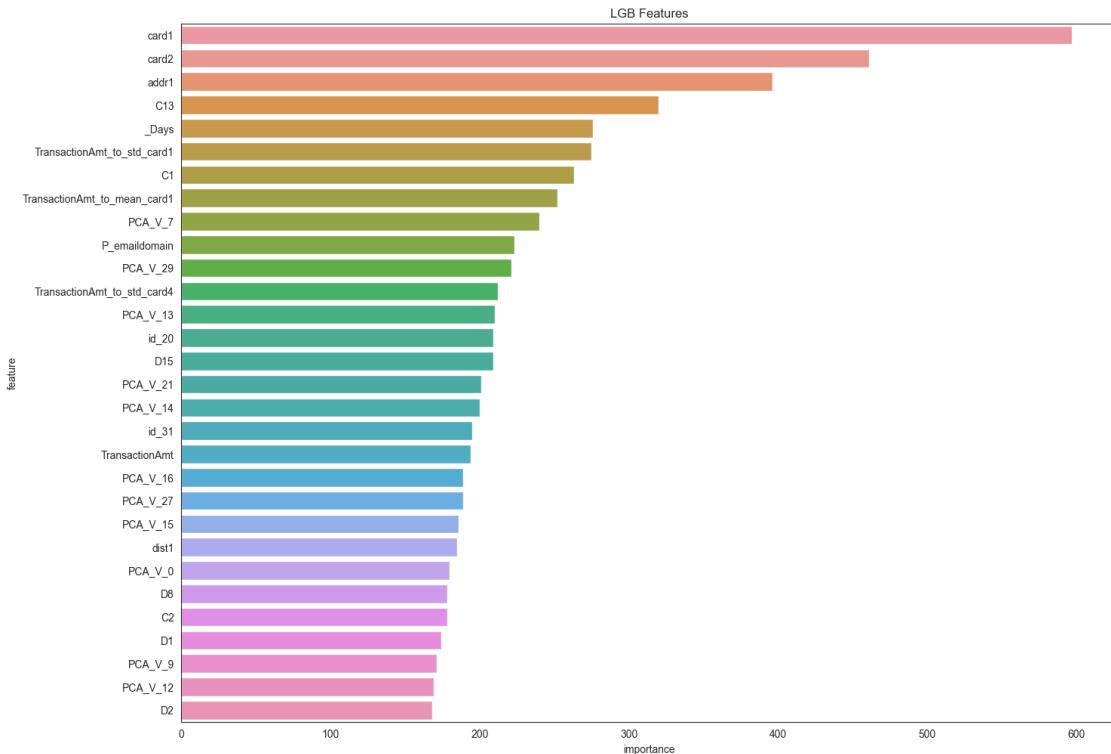
20 19. Feature Importance

```
[161]: feature_importance_df = pd.DataFrame({'feature' : X_train.columns, 'importance'_
    ↪: lgbmclassifier.feature_importances_ })
```

```
[162]: feature_importance_df = feature_importance_df.sort_values(by="importance",_
    ↪ascending=False)
feature_importance_df = feature_importance_df.iloc[:30,:]
feature_importance_df
```

```
[162]:          feature  importance
2                  card1      597
3                  card2      461
8                  addr1      396
25                 C13       320
496                _Days     276
501  TransactionAmt_to_std_card1     275
13                  C1       263
499  TransactionAmt_to_mean_card1     252
510                  PCA_V_7     240
11                  P_emaildomain     223
532                  PCA_V_29     221
502  TransactionAmt_to_std_card4     212
516                  PCA_V_13     210
66                  id_20      209
40                  D15       209
524                  PCA_V_21     201
517                  PCA_V_14     200
70                  id_31      195
0                   TransactionAmt     194
519                  PCA_V_16     189
530                  PCA_V_27     189
518                  PCA_V_15     186
10                  dist1      185
503                  PCA_V_0      180
33                  D8        178
14                  C2        178
27                  D1        174
512                  PCA_V_9      171
515                  PCA_V_12     169
28                  D2        168
```

```
[163]: plt.figure(figsize=(16, 12));
sns.barplot(x="importance", y="feature", data=feature_importance_df.
    ↪sort_values(by="importance", ascending=False));
plt.title('LGB Features');
```



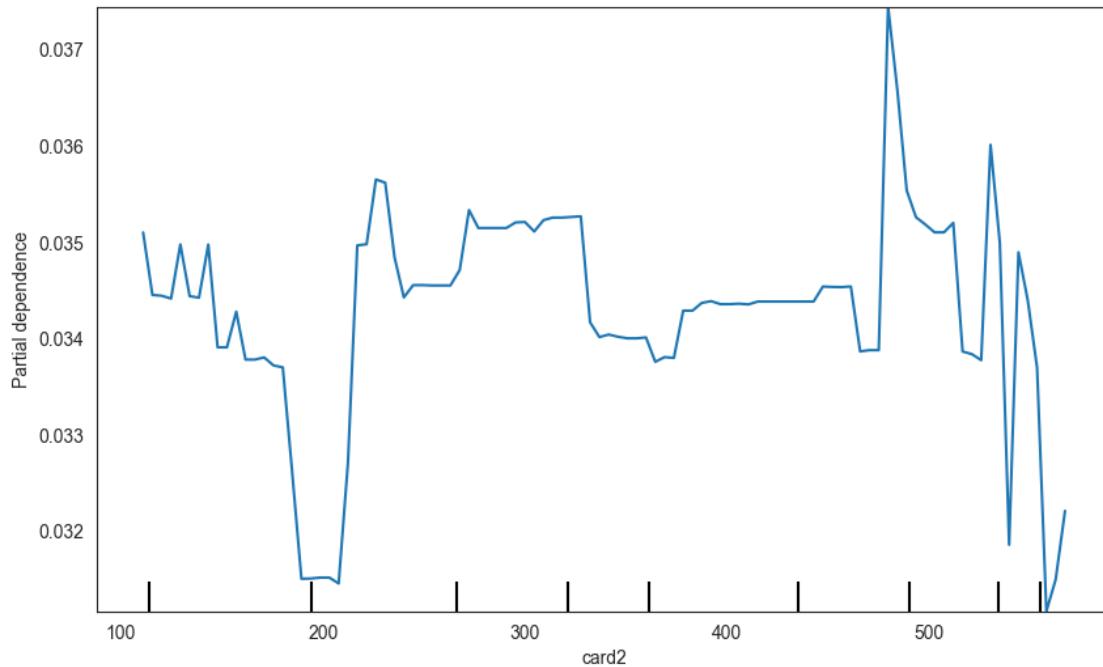
- card1 is contributing the most in predicting if a transaction is fraud or not
- card2, addr1, C13, P_emaildomain, C1,_days etc are some of the most important features in predicting the fraud
- Certain card types, addresses and emails are at high risk of fraud, so there is a need to monitor these carefully

21 20. Partial Dependence and Individual Conditional Expectations (ICE)

Plot Partial Dependence

```
[168]: from sklearn.inspection import PartialDependenceDisplay
import matplotlib.pyplot as plt

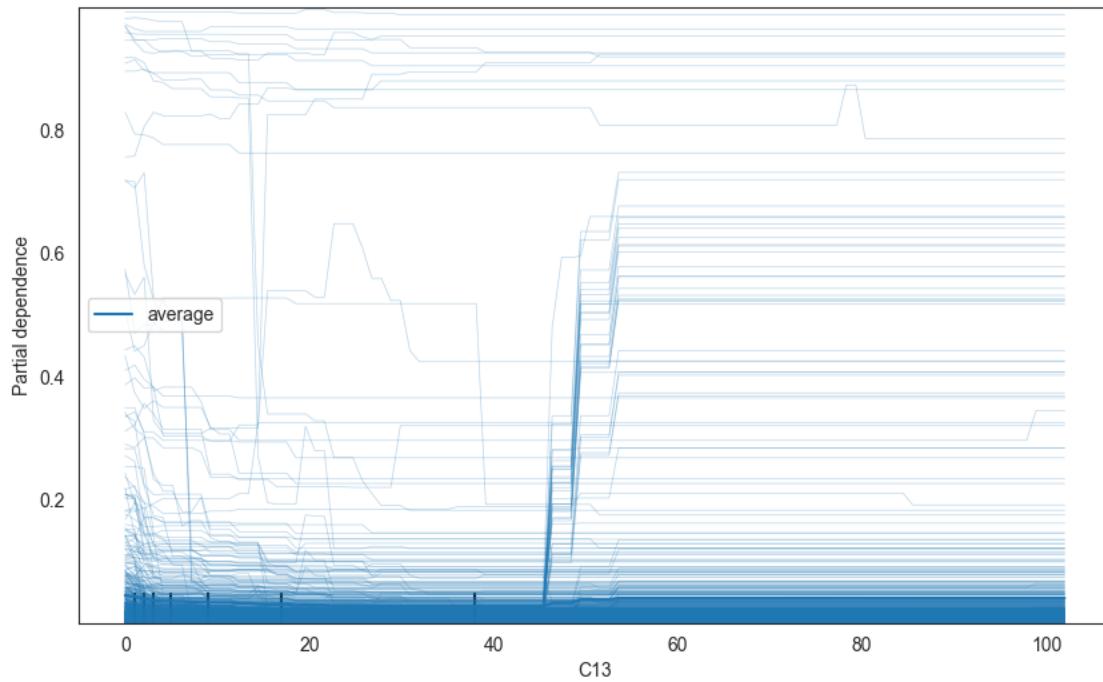
fig, ax = plt.subplots(figsize=(10, 6))
PartialDependenceDisplay.from_estimator(
    lgbmclassifier,          # trained LightGBM model
    X_train,                 # your training data
    features=['card2'],      # feature name as a list
    ax=ax
)
plt.show()
```



```
[170]: fig, ax = plt.subplots(figsize=(10, 6))

PartialDependenceDisplay.from_estimator(
    lgbmclassifier,
    X_train,
    features=['C13'],
    kind='both',
    ax=ax
)

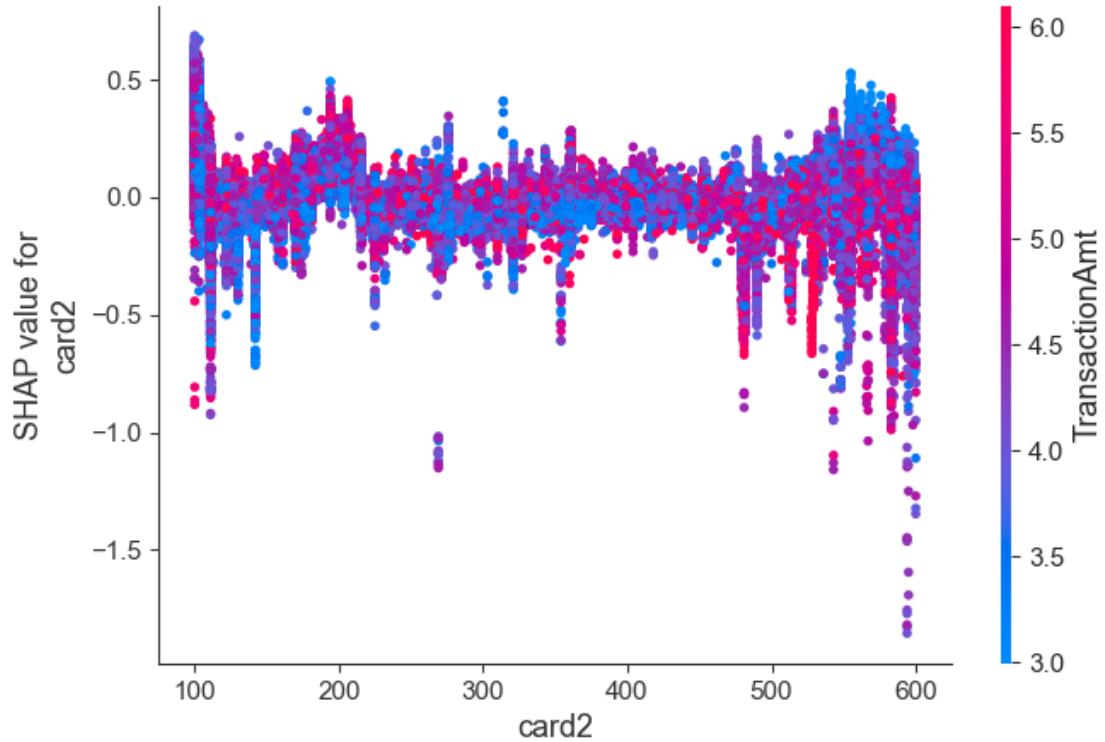
plt.show()
```



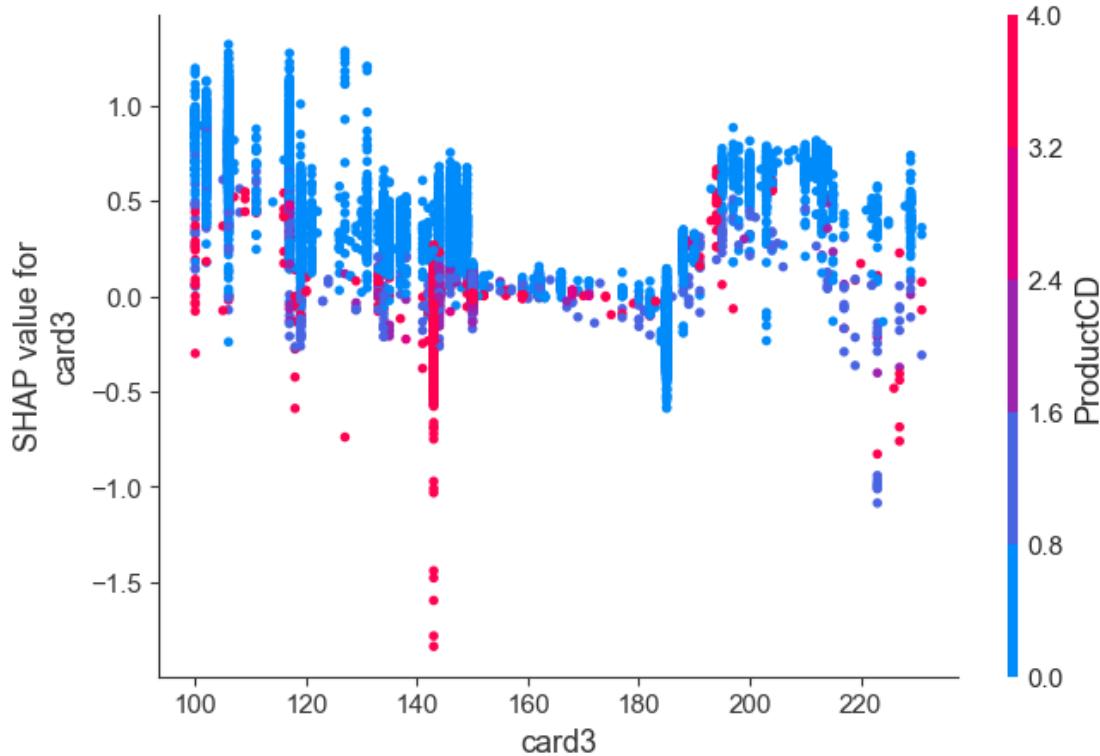
22 21. SHAP Values

```
[172]: import shap  
shap_model = shap.TreeExplainer(lgbmclassifier)  
shap_values = shap_model.shap_values(X_train)
```

```
[173]: shap.dependence_plot("card2", shap_values[0], X_train)
```



```
[174]: shap.dependence_plot("card3", shap_values[0], X_train)
```



```
[175]: shap.initjs()
shap.force_plot(shap_model.expected_value[1], shap_values[1][14], X_train.
    ↪iloc[14, :])
```

<IPython.core.display.HTML object>

```
[175]: <shap.plots._force.AdditiveForceVisualizer at 0x1e026cad048>
```

```
[176]: shap.initjs() # needed to show viz
shap.force_plot(shap_model.expected_value[1], shap_values[1][14], X_train.
    ↪iloc[14, :], link='logit')
```

<IPython.core.display.HTML object>

```
[176]: <shap.plots._force.AdditiveForceVisualizer at 0x1e02525f898>
```

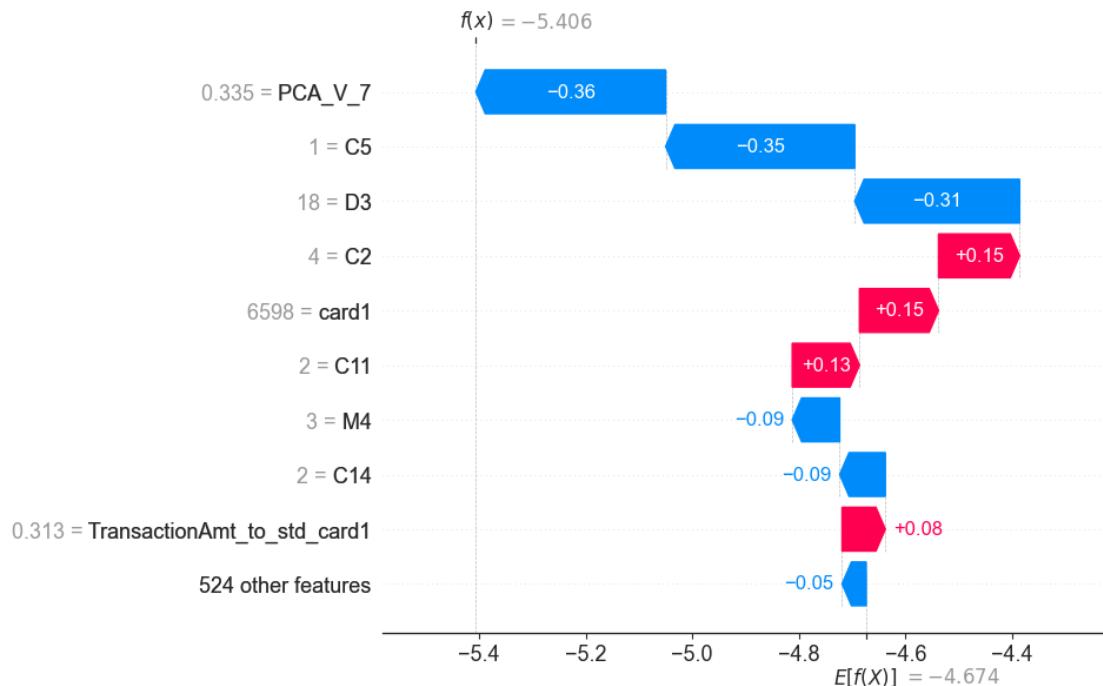
```
[183]: index = 0

shap.plots.waterfall(
    shap.Explanation(
        values=shap_values[1][index],
        base_values=shap_model.expected_value[1],
        data=X_train.iloc[index],
```

```

        feature_names=X_train.columns
    )
)

```



[]:

22.1 Author

MIT TRIVEDI

22.2 Project

Credit Card Fraud Detection

22.3 Date

October 2025

22.4 Contact

[LinkedIn](#) · [GitHub](#) · [Email](#)

22.5 Summary

This project presents an **end-to-end credit card fraud detection pipeline** featuring:

- **Models:** LightGBM and XGBoost
- **Feature Engineering:** Handling missing values, encoding, new features, etc.
- **Model Evaluation:** AUC, confusion matrix, precision, recall
- **Explainability:** SHAP visualizations for model interpretability

All code and notebooks are available in this repository for reproducibility and experimentation.

[]: