

ads-phase3

October 26, 2023

```
[ ]: #Uploading and Displaying Dataset
import pandas as pd
df = pd.read_csv('/content/creditcard.csv')
print(df)
```

	Time	V1	V2	V3	V4	V5	V6	\
0	0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	
1	0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	
2	1	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	
3	1	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	
4	2	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	
...	
61486	49862	-0.987401	1.056011	1.184880	-0.738908	-0.080874	-0.167563	
61487	49862	0.910338	-0.976578	1.308543	0.352233	-1.228617	1.053119	
61488	49863	1.033813	-0.261495	1.329732	1.820041	-0.974670	0.384867	
61489	49863	1.446884	-0.263871	-0.192448	-0.660946	-0.598712	-1.387964	
61490	49863	-1.256173	0.190250	0.486835	0.174933	0.615589	0.667585	
	V7	V8	V9	...	V21	V22	V23	\
0	0.239599	0.098698	0.363787	...	-0.018307	0.277838	-0.110474	
1	-0.078803	0.085102	-0.255425	...	-0.225775	-0.638672	0.101288	
2	0.791461	0.247676	-1.514654	...	0.247998	0.771679	0.909412	
3	0.237609	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321	
4	0.592941	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458	
...	
61486	0.224570	0.650254	-0.454139	...	-0.069511	-0.149341	-0.022932	
61487	-1.189630	0.422435	-0.756288	...	0.081774	0.571815	-0.057231	
61488	-0.580740	0.317767	1.384514	...	-0.398950	-0.748977	0.126709	
61489	-0.001654	-0.430535	-1.207218	...	0.194922	0.511846	-0.170092	
61490	0.983123	0.049528	0.077344	...	-0.024591	0.195936	-0.204945	
	V24	V25	V26	V27	V28	Amount	Class	
0	0.066928	0.128539	-0.189115	0.133558	-0.021053	149.62	0.0	
1	-0.339846	0.167170	0.125895	-0.008983	0.014724	2.69	0.0	
2	-0.689281	-0.327642	-0.139097	-0.055353	-0.059752	378.66	0.0	
3	-1.175575	0.647376	-0.221929	0.062723	0.061458	123.50	0.0	
4	0.141267	-0.206010	0.502292	0.219422	0.215153	69.99	0.0	
...	

61486	0.040407	-0.291335	0.307241	0.190307	0.104660	1.00	0.0
61487	-0.292605	0.065126	-0.226463	0.114948	0.054243	129.50	0.0
61488	0.367238	0.342699	-0.527094	0.082915	0.031702	22.02	0.0
61489	0.445334	0.805625	-0.086404	-0.026795	0.002982	15.00	0.0
61490	-0.850277	-0.393942	-0.584722	-0.117240	0.287411	NaN	NaN

[61491 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

# creating a list of column names
Column_name = pd.DataFrame(df.columns)
print('List of column names :',Column_name)
```

```
List of column names :          0
0      Time
1       V1
2       V2
3       V3
4       V4
5       V5
6       V6
7       V7
8       V8
9       V9
10      V10
11      V11
12      V12
13      V13
14      V14
15      V15
16      V16
17      V17
18      V18
19      V19
20      V20
21      V21
22      V22
23      V23
24      V24
25      V25
26      V26
27      V27
28      V28
29  Amount
30   Class
```

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Displaying First 5 rows
print(df.head())
```

	Time	V1	V2	V3	V4	V5	V6	V7	\
0	0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	
1	0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	
2	1	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	
3	1	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	
4	2	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	

	V8	V9	...	V21	V22	V23	V24	V25	\
0	0.098698	0.363787	...	-0.018307	0.277838	-0.110474	0.066928	0.128539	
1	0.085102	-0.255425	...	-0.225775	-0.638672	0.101288	-0.339846	0.167170	
2	0.247676	-1.514654	...	0.247998	0.771679	0.909412	-0.689281	-0.327642	
3	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321	-1.175575	0.647376	
4	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458	0.141267	-0.206010	

	V26	V27	V28	Amount	Class
0	-0.189115	0.133558	-0.021053	149.62	0.0
1	0.125895	-0.008983	0.014724	2.69	0.0
2	-0.139097	-0.055353	-0.059752	378.66	0.0
3	-0.221929	0.062723	0.061458	123.50	0.0
4	0.502292	0.219422	0.215153	69.99	0.0

[5 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Displaying Last 5 rows
print(df.tail())
```

	Time	V1	V2	V3	V4	V5	V6	\
75352	56019	-0.330203	1.265306	0.703968	0.943073	0.020692	-0.334221	
75353	56020	-0.824159	0.689632	0.238364	0.843827	-1.723679	2.003802	
75354	56020	0.929735	-0.299633	0.394750	0.540756	0.272645	1.656454	
75355	56021	-1.657683	-0.426294	1.687198	-1.454421	-2.383477	1.305648	
75356	56021	-1.711121	-0.369377	1.440787	0.136071	0.517265	0.196718	

	V7	V8	V9	...	V21	V22	V23	\
75352	0.406527	0.313829	-0.924290	...	0.222834	0.653481	-0.024697	
75353	3.194546	-0.727345	-0.680086	...	-0.092779	0.491220	-0.376126	
75354	-0.410577	0.600046	0.184701	...	-0.038120	0.102605	0.094452	
75355	0.658390	0.543973	-1.068176	...	0.166907	0.460901	-0.059376	
75356	-0.006951	0.325528	-1.817459	...	-0.365864	-0.524456	0.085383	

	V24	V25	V26	V27	V28	Amount	Class
75352	0.123168	-0.154517	-0.303343	-0.020581	-0.017914	17.45	0.0
75353	-0.768536	-0.152191	-0.255962	0.014022	-0.227837	570.00	0.0
75354	-0.985089	0.094890	0.376660	0.038140	-0.005591	44.43	0.0
75355	0.079649	0.353229	-0.323458	0.023944	-0.176926	375.01	0.0
75356	-0.342706	0.281718	-0.280851	NaN	NaN	NaN	NaN

[5 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Data Manipulation
print("shape of dataframe", df.shape)
print("number of rows : ", df.shape[0])
print("number of columns : ", df.shape[1])
print("DataFrame Size : ", df.size)
print('Info: ', df.info())
print('Correlation: ', df.corr())
```

```
shape of dataframe (81299, 31)
number of rows : 81299
number of columns : 31
DataFrame Size : 2520269
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 81299 entries, 0 to 81298
Data columns (total 31 columns):
#   Column  Non-Null Count  Dtype
---  -
0   Time    81299 non-null    int64
1   V1      81299 non-null    float64
2   V2      81299 non-null    float64
3   V3      81299 non-null    float64
4   V4      81299 non-null    float64
5   V5      81299 non-null    float64
6   V6      81298 non-null    float64
7   V7      81298 non-null    float64
8   V8      81298 non-null    float64
9   V9      81298 non-null    float64
10  V10     81298 non-null    float64
11  V11     81298 non-null    float64
12  V12     81298 non-null    float64
13  V13     81298 non-null    float64
14  V14     81298 non-null    float64
15  V15     81298 non-null    float64
16  V16     81298 non-null    float64
17  V17     81298 non-null    float64
```

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18 V18      81298 non-null float64
19 V19      81298 non-null float64
20 V20      81298 non-null float64
21 V21      81298 non-null float64
22 V22      81298 non-null float64
23 V23      81298 non-null float64
24 V24      81298 non-null float64
25 V25      81298 non-null float64
26 V26      81298 non-null float64
27 V27      81298 non-null float64
28 V28      81298 non-null float64
29 Amount   81298 non-null float64
30 Class    81298 non-null float64

```

dtypes: float64(30), int64(1)

memory usage: 19.2 MB

Info: None

```

Correlation:          Time      V1      V2      V3      V4      V5
V6 \
Time      1.000000 -0.013925 -0.082910 -0.043661 -0.032350 -0.055022 -0.009466
V1      -0.013925  1.000000  0.057876  0.254591 -0.084183  0.033767  0.081312
V2      -0.082910  0.057876  1.000000 -0.102670  0.053015 -0.065598  0.014305
V3      -0.043661  0.254591 -0.102670  1.000000 -0.105437  0.243695 -0.049882
V4      -0.032350 -0.084183  0.053015 -0.105437  1.000000 -0.015254 -0.046189
V5      -0.055022  0.033767 -0.065598  0.243695 -0.015254  1.000000  0.034152
V6      -0.009466  0.081312  0.014305 -0.049882 -0.046189  0.034152  1.000000
V7      -0.001258  0.105001 -0.048966  0.236764 -0.044037 -0.027642  0.088006
V8       0.019882 -0.027238  0.043323 -0.175038  0.060421 -0.047753 -0.054060
V9      -0.294523 -0.042588 -0.017073  0.136133  0.037307  0.054927  0.043671
V10     0.075641  0.043634 -0.015647  0.153577 -0.095702  0.111052 -0.001510
V11     -0.189116 -0.009836  0.037342 -0.092946  0.014902 -0.030613 -0.094684
V12     0.412812  0.008630 -0.065674  0.099716 -0.011019  0.036390  0.017711
V13     -0.301344 -0.001360  0.044314 -0.039569 -0.005934  0.047681 -0.003970
V14     -0.235872  0.078422 -0.020341  0.124337 -0.035336  0.055604  0.037352
V15     0.123733  0.049936  0.055277 -0.177173 -0.055727  0.095087 -0.116123
V16     0.008486  0.045083 -0.015888 -0.027388 -0.103932  0.114180 -0.011983
V17     -0.109927  0.071585 -0.070864  0.138581 -0.016371  0.003875  0.018168
V18     -0.025804  0.007413 -0.017925  0.018758 -0.038497  0.061610  0.053234
V19     0.021510 -0.002274  0.002709 -0.040243  0.004703  0.016228  0.085536
V20     -0.002366 -0.015352 -0.092247 -0.091617  0.019320 -0.063414  0.045452
V21     0.008656 -0.011930 -0.012513  0.031584  0.013247 -0.055142  0.014881
V22     0.027918 -0.050380 -0.029431  0.233243  0.028736 -0.068673  0.036349
V23     0.001374 -0.072715  0.008274  0.054219 -0.012430  0.022209 -0.018858
V24     -0.002721 -0.011388 -0.025602  0.023076  0.000122 -0.017287 -0.003973
V25     0.015738  0.184268 -0.087388 -0.183585 -0.001243 -0.011857  0.044941
V26     -0.001105  0.021676 -0.022378  0.044984 -0.013533 -0.039077  0.005438
V27     -0.015549 -0.051987  0.005089 -0.091089  0.050494 -0.037045 -0.011696
V28     -0.006571  0.127100  0.002739  0.038353 -0.005517  0.012376 -0.016829
Amount  0.054356 -0.253333 -0.554004 -0.225941  0.093923 -0.400681  0.215518

```

Class -0.013894 -0.162574 0.137032 -0.333239 0.180734 -0.158985 -0.078666

	V7	V8	V9	...	V21	V22	V23	\
Time	-0.001258	0.019882	-0.294523	...	0.008656	0.027918	0.001374	
V1	0.105001	-0.027238	-0.042588	...	-0.011930	-0.050380	-0.072715	
V2	-0.048966	0.043323	-0.017073	...	-0.012513	-0.029431	0.008274	
V3	0.236764	-0.175038	0.136133	...	0.031584	0.233243	0.054219	
V4	-0.044037	0.060421	0.037307	...	0.013247	0.028736	-0.012430	
V5	-0.027642	-0.047753	0.054927	...	-0.055142	-0.068673	0.022209	
V6	0.088006	-0.054060	0.043671	...	0.014881	0.036349	-0.018858	
V7	1.000000	-0.051467	0.054032	...	0.022815	-0.040019	0.069559	
V8	-0.051467	1.000000	-0.038433	...	0.054271	0.015439	-0.017122	
V9	0.054032	-0.038433	1.000000	...	-0.002439	0.031733	-0.045455	
V10	0.096803	-0.078103	-0.047429	...	-0.014369	-0.045958	-0.007092	
V11	-0.040472	0.009657	0.011211	...	0.032944	0.064967	0.025463	
V12	0.098922	-0.061677	-0.087678	...	-0.001387	0.060334	-0.016007	
V13	0.008785	-0.006662	0.058764	...	-0.002050	-0.030179	-0.012066	
V14	0.054294	-0.055198	0.121347	...	-0.047152	-0.071733	0.021075	
V15	0.070667	-0.038075	-0.142897	...	0.018090	0.003075	0.019800	
V16	0.100726	-0.069454	-0.044932	...	-0.019515	0.028570	-0.014112	
V17	0.080897	-0.060492	0.139072	...	-0.035270	0.049571	0.031509	
V18	0.065408	-0.027215	0.014743	...	-0.051807	-0.108083	0.008027	
V19	-0.044750	0.030207	-0.002049	...	0.006943	0.015623	0.002206	
V20	0.065459	-0.002663	-0.002665	...	0.010473	-0.023015	-0.033039	
V21	0.022815	0.054271	-0.002439	...	1.000000	-0.113310	-0.024161	
V22	-0.040019	0.015439	0.031733	...	-0.113310	1.000000	0.071515	
V23	0.069559	-0.017122	-0.045455	...	-0.024161	0.071515	1.000000	
V24	0.000048	0.007096	0.011220	...	0.002170	-0.001147	0.001182	
V25	-0.098344	0.015210	0.111125	...	-0.015111	-0.008739	0.110282	
V26	-0.034239	0.005321	0.109188	...	-0.027972	-0.081681	0.034052	
V27	-0.099873	0.005151	-0.018785	...	-0.031751	-0.008068	-0.034120	
V28	-0.034944	0.029829	-0.008674	...	0.053049	-0.033607	0.040300	
Amount	0.387053	-0.098580	-0.022131	...	0.139633	-0.085768	-0.146106	
Class	-0.266607	0.118478	-0.133856	...	0.052548	-0.003990	-0.014415	

	V24	V25	V26	V27	V28	Amount	Class
Time	-0.002721	0.015738	-0.001105	-0.015549	-0.006571	0.054356	-0.013894
V1	-0.011388	0.184268	0.021676	-0.051987	0.127100	-0.253333	-0.162574
V2	-0.025602	-0.087388	-0.022378	0.005089	0.002739	-0.554004	0.137032
V3	0.023076	-0.183585	0.044984	-0.091089	0.038353	-0.225941	-0.333239
V4	0.000122	-0.001243	-0.013533	0.050494	-0.005517	0.093923	0.180734
V5	-0.017287	-0.011857	-0.039077	-0.037045	0.012376	-0.400681	-0.158985
V6	-0.003973	0.044941	0.005438	-0.011696	-0.016829	0.215518	-0.078666
V7	0.000048	-0.098344	-0.034239	-0.099873	-0.034944	0.387053	-0.266607
V8	0.007096	0.015210	0.005321	0.005151	0.029829	-0.098580	0.118478
V9	0.011220	0.111125	0.109188	-0.018785	-0.008674	-0.022131	-0.133856
V10	0.004753	0.011314	-0.019610	-0.067253	0.015601	-0.134701	-0.300631
V11	0.034763	-0.102636	0.000187	0.019101	0.016485	-0.015314	0.202480

V12	0.009741	0.012421	0.020705	-0.023450	-0.008842	0.029181	-0.308905
V13	-0.012051	0.038934	0.001649	-0.001168	-0.004819	-0.009559	0.000711
V14	0.024351	-0.063342	0.023317	-0.036010	-0.003872	0.004561	-0.393122
V15	-0.001744	-0.088926	-0.017775	0.017499	-0.006796	-0.035699	0.002375
V16	-0.003343	0.103633	0.023723	-0.014902	-0.028925	-0.005469	-0.275815
V17	-0.006534	-0.079071	-0.069495	-0.027731	-0.025454	0.016563	-0.437197
V18	-0.016236	0.018811	0.019509	-0.019647	0.000474	0.043009	-0.175555
V19	-0.008071	-0.010355	-0.005147	0.005555	-0.011828	-0.054247	0.049155
V20	-0.009344	-0.035345	-0.000856	-0.047559	0.123477	0.436861	0.021748
V21	0.002170	-0.015111	-0.027972	-0.031751	0.053049	0.139633	0.052548
V22	-0.001147	-0.008739	-0.081681	-0.008068	-0.033607	-0.085768	-0.003990
V23	0.001182	0.110282	0.034052	-0.034120	0.040300	-0.146106	-0.014415
V24	1.000000	-0.037993	-0.007921	0.002140	-0.006432	0.019843	-0.008330
V25	-0.037993	1.000000	-0.112286	0.043842	0.062729	-0.068349	0.010672
V26	-0.007921	-0.112286	1.000000	-0.002866	0.000062	-0.002276	0.006278
V27	0.002140	0.043842	-0.002866	1.000000	-0.084928	0.013230	0.071564
V28	-0.006432	0.062729	0.000062	-0.084928	1.000000	0.020746	0.007709
Amount	0.019843	-0.068349	-0.002276	0.013230	0.020746	1.000000	-0.000752
Class	-0.008330	0.010672	0.006278	0.071564	0.007709	-0.000752	1.000000

[31 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Dropping Empty Cells
Emp_drop = df.dropna(how= "all", axis=1)
print(df)
```

	Time	V1	V2	V3	V4	V5	V6 \
0	0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388
1	0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361
2	1	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499
3	1	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203
4	2	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921
...
83274	59766	-0.465927	0.740705	1.731343	0.036013	0.098876	0.055606
83275	59767	-0.725577	0.379742	1.468039	1.356958	0.546077	0.109718
83276	59768	-1.711347	1.308095	-0.232786	-1.607585	-1.299910	-0.627938
83277	59768	1.471591	-1.234024	0.279178	-1.595504	-1.344324	-0.101130
83278	59769	1.135220	0.124163	0.109082	0.902728	0.038998	0.004539
	V7	V8	V9	...	V21	V22	V23 \
0	0.239599	0.098698	0.363787	...	-0.018307	0.277838	-0.110474
1	-0.078803	0.085102	-0.255425	...	-0.225775	-0.638672	0.101288
2	0.791461	0.247676	-1.514654	...	0.247998	0.771679	0.909412
3	0.237609	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321
4	0.592941	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458

...
83274	0.472815	0.134921	-0.302919	...	0.032294	0.287631	-0.337062
83275	0.416071	0.180774	-0.273686	...	0.006224	0.017405	-0.101053
83276	-1.562053	-1.722859	-1.534618	...	-0.902970	1.269219	0.178712
83277	-1.249330	0.030445	-1.813220	...	0.000785	0.310947	-0.194016
83278	0.007332	0.113029	-0.000000	...	NaN	NaN	NaN

	V24	V25	V26	V27	V28	Amount	Class
0	0.066928	0.128539	-0.189115	0.133558	-0.021053	149.62	0.0
1	-0.339846	0.167170	0.125895	-0.008983	0.014724	2.69	0.0
2	-0.689281	-0.327642	-0.139097	-0.055353	-0.059752	378.66	0.0
3	-1.175575	0.647376	-0.221929	0.062723	0.061458	123.50	0.0
4	0.141267	-0.206010	0.502292	0.219422	0.215153	69.99	0.0
...
83274	0.079929	0.091760	0.396666	0.078020	0.093914	1.00	0.0
83275	-0.294967	0.164357	-0.166337	0.114452	0.115366	46.96	0.0
83276	0.527195	-0.313104	-0.365365	0.216308	0.181065	10.00	0.0
83277	-0.504003	0.513079	0.008417	0.029493	0.007596	40.00	0.0
83278	NaN	NaN	NaN	NaN	NaN	NaN	NaN

[83279 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Dropping Empty Cells
df = df.drop(columns=df.columns[1:3], inplace=True)
print(df)
```

None

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Dropping Empty Cells
df = df.dropna()
print(df)
```

	Time	V1	V2	V3	V4	V5	V6 \
0	0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388
1	0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361
2	1	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499
3	1	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203
4	2	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921
...
95156	65189	-0.327858	0.966095	1.541728	0.745525	0.042368	-1.097118
95157	65189	1.142364	-1.090177	0.511068	-0.792356	-1.160044	0.134166
95158	65190	-2.145283	1.280406	0.014577	-2.003358	1.479294	4.673049


```

95159 65190 -3.715715 3.870511 -1.525809 0.082535 -0.244009 -0.901579
95160 65190 -5.164795 4.510526 -0.994499 -1.110853 -0.913228 -0.889076

```

```

          V7          V8          V9 ...          V21          V22          V23 \
0      0.239599  0.098698  0.363787 ... -0.018307  0.277838 -0.110474
1     -0.078803  0.085102 -0.255425 ... -0.225775 -0.638672  0.101288
2      0.791461  0.247676 -1.514654 ...  0.247998  0.771679  0.909412
3      0.237609  0.377436 -1.387024 ... -0.108300  0.005274 -0.190321
4      0.592941 -0.270533  0.817739 ... -0.009431  0.798278 -0.137458
...      ...      ...      ...      ...      ...      ...
95156  0.559739 -0.171171 -0.762272 ... -0.125798 -0.371319  0.089315
95157 -0.997184  0.236721 -0.583865 ...  0.399798  0.815983 -0.172587
95158 -2.008023 -2.933663  0.036717 ...  0.276433 -0.038006  0.085076
95159  0.708830  0.070491  2.349423 ... -0.327180  0.573451  0.266379
95160  0.373572  0.361552  3.841062 ... -0.908623 -1.154210  0.300341

```

```

          V24          V25          V26          V27          V28 Amount Class
0      0.066928  0.128539 -0.189115  0.133558 -0.021053  149.62  0.0
1     -0.339846  0.167170  0.125895 -0.008983  0.014724   2.69  0.0
2     -0.689281 -0.327642 -0.139097 -0.055353 -0.059752  378.66  0.0
3     -1.175575  0.647376 -0.221929  0.062723  0.061458  123.50  0.0
4      0.141267 -0.206010  0.502292  0.219422  0.215153   69.99  0.0
...      ...      ...      ...      ...      ...
95156  0.695077 -0.671527  0.134372  0.129250  0.171650   2.97  0.0
95157 -0.298962  0.310993 -0.077793  0.011223  0.019004  106.00  0.0
95158  1.049870  0.489570  1.045371 -0.363337 -0.222526   34.61  0.0
95159  0.040564 -0.175983 -0.494220  0.257349 -0.309196    0.89  0.0
95160 -0.102776  0.817800  0.201861  2.384092  1.576142    7.18  0.0

```

[95161 rows x 31 columns]

```

[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Dropping Empty Cells
df=df.dropna(subset = ['V5'])
df=df.dropna(subset = ['Time'])
print(df)

```

```

          Time          V1          V2          V3          V4          V5          V6 \
0           0 -1.359807 -0.072781  2.536347  1.378155 -0.338321  0.462388
1           0  1.191857  0.266151  0.166480  0.448154  0.060018 -0.082361
2           1 -1.358354 -1.340163  1.773209  0.379780 -0.503198  1.800499
3           1 -0.966272 -0.185226  1.792993 -0.863291 -0.010309  1.247203
4           2 -1.158233  0.877737  1.548718  0.403034 -0.407193  0.095921
...      ...      ...      ...      ...      ...
124855  77448 -0.358736  1.088973  1.182810 -0.134401  0.274668 -0.469507
124856  77448 -0.835868  0.673600  1.904021 -0.263201  0.351435  0.236937

```

```

124857 77448 -2.817937 -0.107162 0.402607 -0.980962 0.682698 0.980801
124858 77449 -1.262504 1.434804 0.290138 -0.436888 0.801078 0.380510
124859 77449 1.087845 0.522904 0.275704 2.474583 0.213114 -0.225345

```

```

          V7          V8          V9  ...          V21          V22          V23  \
0      0.239599  0.098698  0.363787  ... -0.018307  0.277838 -0.110474
1     -0.078803  0.085102 -0.255425  ... -0.225775 -0.638672  0.101288
2      0.791461  0.247676 -1.514654  ...  0.247998  0.771679  0.909412
3      0.237609  0.377436 -1.387024  ... -0.108300  0.005274 -0.190321
4      0.592941 -0.270533  0.817739  ... -0.009431  0.798278 -0.137458
...
124855 0.604681  0.085536 -0.659111  ... -0.232968 -0.622835 -0.075680
124856 0.223675  0.334808 -0.232802  ...  0.114441  0.182323 -0.105240
124857 0.117625  1.387942 -0.779965  ...  0.280573  0.323453 -0.374319
124858 0.399066  0.554237 -0.531654  ...  0.195197  0.383922 -0.281029
124859 0.389247 -0.100848 -0.960723  ...  0.007535 -0.054694 -0.061887

```

```

          V24          V25          V26          V27          V28  Amount  Class
0      0.066928  0.128539 -0.189115  0.133558 -0.021053  149.62    0.0
1     -0.339846  0.167170  0.125895 -0.008983  0.014724    2.69    0.0
2     -0.689281 -0.327642 -0.139097 -0.055353 -0.059752  378.66    0.0
3     -1.175575  0.647376 -0.221929  0.062723  0.061458  123.50    0.0
4      0.141267 -0.206010  0.502292  0.219422  0.215153   69.99    0.0
...
124855 -0.048200 -0.140704  0.074554  0.243321  0.085024    0.89    0.0
124856 -0.357454 -0.357218 -0.704835  0.014996  0.144135    4.99    0.0
124857 -0.962394  0.422271  0.402743 -0.288024 -0.339470  115.00    0.0
124858 -1.152876  0.030816 -0.486364 -0.148485  0.121988    1.07    0.0
124859  0.083010  0.527815  0.030754 -0.017737  0.020217   45.95    0.0

```

[124860 rows x 31 columns]

```
[ ]: print('Correlation: ', df.corr())
```

```

Correlation:          Time          V1          V2          V3          V4          V5
V6  \
Time    1.000000 -0.001731 -0.016649 -0.027271 -0.026884 -0.032948 -0.015875
V1     -0.001731  1.000000  0.034695  0.207392 -0.057541  0.013500  0.055785
V2     -0.016649  0.034695  1.000000 -0.082420  0.037251 -0.034903  0.014035
V3     -0.027271  0.207392 -0.082420  1.000000 -0.074047  0.179836 -0.076819
V4     -0.026884 -0.057541  0.037251 -0.074047  1.000000  0.003631 -0.039537
V5     -0.032948  0.013500 -0.034903  0.179836  0.003631  1.000000  0.048400
V6     -0.015875  0.055785  0.014035 -0.076819 -0.039537  0.048400  1.000000
V7     -0.002300  0.074587 -0.020980  0.183199 -0.008168 -0.037217  0.040139
V8      0.018370 -0.013531  0.009422 -0.118408  0.033319 -0.020964 -0.041360
V9     -0.228759 -0.060808  0.000264  0.106021  0.050187  0.032609  0.043028
V10     0.057180  0.003614  0.017297  0.093117 -0.093751  0.057504 -0.014579
V11    -0.155421  0.007005  0.027791 -0.077015  0.004255 -0.016998 -0.090787

```

V12	0.359075	-0.011586	-0.021159	0.069677	0.036061	0.022932	0.009588
V13	-0.211731	-0.000922	0.034834	-0.054465	-0.015588	0.047990	-0.014092
V14	-0.182534	0.040881	-0.007710	0.061805	-0.009655	0.031442	0.019828
V15	0.109789	0.049571	0.076964	-0.192634	-0.029724	0.112446	-0.119873
V16	0.007012	0.026197	0.021932	-0.058463	-0.080473	0.106973	-0.019527
V17	-0.088058	0.040200	-0.061148	0.090980	0.007361	-0.043552	0.003437
V18	-0.003972	-0.001203	-0.019662	-0.023103	-0.033652	0.045294	0.045135
V19	0.012287	-0.000434	0.008021	-0.029651	0.013725	0.021349	0.097323
V20	-0.001224	-0.037519	-0.087105	-0.088143	0.014125	-0.034410	0.036281
V21	-0.006755	-0.024190	-0.019134	0.032217	0.024608	-0.049929	0.013744
V22	-0.004826	-0.054859	-0.021969	0.247679	0.035689	-0.088336	0.047566
V23	0.007458	-0.075960	0.024361	0.043088	-0.010482	0.010918	-0.016348
V24	0.003518	-0.017771	-0.018493	0.026454	-0.000922	-0.018686	-0.006352
V25	-0.001881	0.203496	-0.090811	-0.192733	0.000069	-0.012302	0.049216
V26	0.002957	0.016352	-0.029063	0.039059	-0.033655	-0.047352	0.006177
V27	-0.011496	-0.009894	-0.005468	-0.056625	0.039821	-0.014892	-0.011501
V28	-0.004022	0.158473	0.025303	0.034221	-0.006167	-0.008885	-0.005152
Amount	0.007181	-0.246225	-0.552406	-0.230229	0.095720	-0.398748	0.206234
Class	-0.015757	-0.134643	0.111500	-0.280966	0.150421	-0.128631	-0.055980

	V7	V8	V9	...	V21	V22	V23 \
Time	-0.002300	0.018370	-0.228759	...	-0.006755	-0.004826	0.007458
V1	0.074587	-0.013531	-0.060808	...	-0.024190	-0.054859	-0.075960
V2	-0.020980	0.009422	0.000264	...	-0.019134	-0.021969	0.024361
V3	0.183199	-0.118408	0.106021	...	0.032217	0.247679	0.043088
V4	-0.008168	0.033319	0.050187	...	0.024608	0.035689	-0.010482
V5	-0.037217	-0.020964	0.032609	...	-0.049929	-0.088336	0.010918
V6	0.040139	-0.041360	0.043028	...	0.013744	0.047566	-0.016348
V7	1.000000	0.006108	0.043743	...	-0.000291	-0.045846	0.067208
V8	0.006108	1.000000	-0.016782	...	0.024654	0.033789	-0.022804
V9	0.043743	-0.016782	1.000000	...	0.005793	0.047020	-0.051603
V10	0.055038	-0.038591	-0.018848	...	-0.027016	-0.048318	-0.012671
V11	-0.001759	-0.007690	-0.011560	...	0.039156	0.060644	0.031058
V12	0.061941	-0.037606	-0.023319	...	-0.012814	0.052172	-0.014010
V13	0.021356	-0.001678	-0.016396	...	-0.000373	-0.023536	-0.012013
V14	0.023922	-0.017589	0.027150	...	-0.051646	-0.068516	0.020988
V15	0.088714	-0.043979	-0.124072	...	0.005929	-0.021919	0.025834
V16	0.071073	-0.040617	-0.051200	...	-0.020259	0.032575	-0.021773
V17	0.017742	-0.022144	0.080219	...	-0.029595	0.078478	0.030268
V18	0.040412	-0.013631	-0.035553	...	-0.057267	-0.117504	0.002944
V19	-0.040987	0.028988	0.010052	...	0.012069	0.019012	0.000281
V20	0.061584	-0.021565	0.026754	...	-0.016810	-0.009024	-0.019887
V21	-0.000291	0.024654	0.005793	...	1.000000	-0.096912	-0.001906
V22	-0.045846	0.033789	0.047020	...	-0.096912	1.000000	0.051094
V23	0.067208	-0.022804	-0.051603	...	-0.001906	0.051094	1.000000
V24	-0.000040	0.006951	0.012873	...	0.004337	-0.000268	0.009821
V25	-0.098119	0.012248	0.121999	...	-0.012958	-0.011344	0.096890
V26	-0.031692	0.010040	0.111252	...	-0.027034	-0.091848	0.037266

V27	-0.087161	0.011159	-0.030038	...	-0.013856	-0.015287	-0.017093
V28	-0.043675	0.018065	-0.036978	...	0.049342	-0.035736	0.024326
Amount	0.380759	-0.091063	-0.018740	...	0.130499	-0.077105	-0.136018
Class	-0.226664	0.054860	-0.105946	...	0.082287	-0.014888	-0.005658

	V24	V25	V26	V27	V28	Amount	Class
Time	0.003518	-0.001881	0.002957	-0.011496	-0.004022	0.007181	-0.015757
V1	-0.017771	0.203496	0.016352	-0.009894	0.158473	-0.246225	-0.134643
V2	-0.018493	-0.090811	-0.029063	-0.005468	0.025303	-0.552406	0.111500
V3	0.026454	-0.192733	0.039059	-0.056625	0.034221	-0.230229	-0.280966
V4	-0.000922	0.000069	-0.033655	0.039821	-0.006167	0.095720	0.150421
V5	-0.018686	-0.012302	-0.047352	-0.014892	-0.008885	-0.398748	-0.128631
V6	-0.006352	0.049216	0.006177	-0.011501	-0.005152	0.206234	-0.055980
V7	-0.000040	-0.098119	-0.031692	-0.087161	-0.043675	0.380759	-0.226664
V8	0.006951	0.012248	0.010040	0.011159	0.018065	-0.091063	0.054860
V9	0.012873	0.121999	0.111252	-0.030038	-0.036978	-0.018740	-0.105946
V10	0.004571	0.022135	-0.023164	-0.071278	-0.038361	-0.129267	-0.248719
V11	0.037871	-0.103821	-0.008451	0.009765	-0.002398	-0.011857	0.169400
V12	0.011255	0.012831	0.027530	-0.018287	-0.005782	0.014459	-0.280580
V13	-0.020292	0.045123	0.001219	-0.001160	-0.004636	-0.004004	-0.004530
V14	0.029491	-0.068786	0.019904	-0.019526	0.019863	0.016120	-0.354402
V15	-0.002059	-0.094524	-0.014405	0.007960	-0.010206	-0.040925	0.001032
V16	-0.006498	0.104621	0.021533	-0.003670	-0.016621	-0.021677	-0.232216
V17	-0.006226	-0.077231	-0.071875	-0.006536	-0.011389	0.020056	-0.379155
V18	-0.019283	0.024451	0.017732	-0.012131	0.002233	0.051106	-0.141271
V19	-0.013462	-0.009305	-0.001971	0.000394	-0.004995	-0.056512	0.042667
V20	-0.005633	-0.022870	0.005235	-0.017929	0.097327	0.422955	0.012551
V21	0.004337	-0.012958	-0.027034	-0.013856	0.049342	0.130499	0.082287
V22	-0.000268	-0.011344	-0.091848	-0.015287	-0.035736	-0.077105	-0.014888
V23	0.009821	0.096890	0.037266	-0.017093	0.024326	-0.136018	-0.005658
V24	1.000000	-0.041055	-0.006681	-0.008176	-0.006570	0.018339	-0.009232
V25	-0.041055	1.000000	-0.112424	0.051270	0.058479	-0.061672	0.007702
V26	-0.006681	-0.112424	1.000000	0.000782	0.004712	0.006370	0.003130
V27	-0.008176	0.051270	0.000782	1.000000	0.031998	0.009893	0.058279
V28	-0.006570	0.058479	0.004712	0.031998	1.000000	0.011129	0.011325
Amount	0.018339	-0.061672	0.006370	0.009893	0.011129	1.000000	0.004205
Class	-0.009232	0.007702	0.003130	0.058279	0.011325	0.004205	1.000000

[31 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Create Duplicate Columns
pd.get_dummies(df)
```



```
[ ]:      Time      V1      V2      V3      V4      V5      V6      V7      V8      V9      ...      V21      \
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  True  True  True  True  True  True  True  True  True  True  True  ...  True

      V22      V23      V24      V25      V26      V27      V28      Amount      Class
0  True  True  True  True  True  True  True  True      True      True
1  True  True  True  True  True  True  True  True      True      True
2  True  True  True  True  True  True  True  True      True      True
3  True  True  True  True  True  True  True  True      True      True
4  True  True  True  True  True  True  True  True      True      True
```

[5 rows x 31 columns]

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

#Change data to Numpy
df.to_numpy()
```

<ipython-input-14-7a866dd08d87>:2: DtypeWarning: Columns (28) have mixed types.
Specify dtype option on import or set low_memory=False.

```
df = pd.read_csv('/content/creditcard.csv')
```

```
[ ]: array([[0, -1.3598071336738, -0.0727811733098497, ...,
          -0.0210530534538215, 149.62, 0.0],
          [0, 1.19185711131486, 0.26615071205963, ..., 0.0147241691924927,
          2.69, 0.0],
          [1, -1.35835406159823, -1.34016307473609, ...,
          -0.0597518405929204, 378.66, 0.0],
          ...,
          [81845, 1.06740445339533, 0.0871526496833023, ...,
          '0.0222817106909724', 85.9, 0.0],
          [81846, -0.833219444483895, 0.746825586488521, ...,
          '0.0384589819489235', 6.8, 0.0],
          [81847, -0.418804707094463, 0.740210337181224, ..., '-', nan, nan]],
          dtype=object)
```

```
[ ]: import pandas as pd
df = pd.read_csv('/content/creditcard.csv')

# Shuffle dataframe
df = df.sample(frac=1)
print(df)
```

```
Time      V1      V2      V3      V4      V5      V6      \
```

103292	68588	1.016117	-0.035096	0.307863	0.833742	0.559375	1.824413
26645	34181	-1.012508	2.345409	-3.203335	1.733481	-0.476341	-1.683988
28141	34871	-0.581294	1.327858	1.206441	0.007123	0.067904	-1.006753
99618	67202	-0.771040	-1.978799	0.819289	-0.569889	-2.304708	0.623287
44496	41971	-0.342552	1.193489	1.311542	0.054581	0.104088	-0.960848
...
54915	46685	0.971260	-0.254421	0.511950	1.146945	-0.618884	-0.229477
51650	45059	1.143427	0.235989	-0.076281	1.177647	-0.113112	-0.936325
100880	67655	1.229858	0.310747	0.193094	0.499643	-0.123778	-0.550250
11585	19910	-0.287009	0.250399	1.712981	-0.872371	-0.683187	-0.681445
32866	36996	1.232573	0.149492	0.479569	0.639037	-0.560774	-1.003719

	V7	V8	V9	...	V21	V22	V23	\
103292	-0.423823	0.666053	0.128384	...	-0.069706	0.020362	0.104289	
26645	0.144158	1.088745	-0.704785	...	-0.070167	-0.110774	0.402346	
28141	0.718263	-0.096415	-0.294980	...	-0.288499	-0.661535	-0.018665	
99618	0.072131	-0.019383	-1.931263	...	0.131476	1.184782	1.342933	
44496	0.777579	-0.124738	-0.531466	...	-0.252685	-0.615717	-0.022604	
...	
54915	-0.173476	0.150363	0.176163	...	-0.118085	-0.734289	-0.011911	
51650	0.417738	-0.155531	-0.065475	...	0.054307	0.091272	-0.086392	
100880	0.007009	-0.057952	-0.261950	...	-0.251004	-0.749065	0.068659	
11585	-0.074780	-0.139561	0.494561	...	-0.093508	-0.058258	-0.142823	
32866	0.028859	-0.165630	0.159376	...	-0.257898	-0.815284	0.132604	

	V24	V25	V26	V27	V28	Amount	Class
103292	-1.376160	0.214000	-0.379246	0.093677	-0.004731	1.00	0.0
26645	0.181389	-0.320410	-0.404628	0.226855	-0.221503	89.99	0.0
28141	0.348269	-0.089580	0.074641	0.359116	0.159540	1.78	0.0
99618	0.052859	-1.459349	-0.019029	0.481530	0.087172	345.00	0.0
44496	0.358666	-0.129821	0.068806	0.251077	0.100312	4.49	0.0
...
54915	-0.096452	0.197764	-0.609371	-0.002517	0.036314	116.00	0.0
51650	0.378940	0.669354	-0.285210	-0.008010	0.013372	35.90	0.0
100880	-0.026469	0.241188	0.095673	-0.025545	0.018746	5.37	0.0
11585	0.307129	0.022700	-0.331559	-0.100278	0.040143	5.00	0.0
32866	0.354537	0.186945	0.095908	-0.036743	0.019808	16.00	0.0

[140703 rows x 31 columns]