

Intro to Data Science - EDA - Correlation

March 27, 2020

```
[13]: # This Python 3 environment comes with many helpful analytics libraries
      ↪ installed
      # It is defined by the kaggle/python docker image: https://github.com/kaggle/
      ↪ docker-python
      # For example, here's several helpful packages to load in

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the "../input/" directory.
# For example, running this (by clicking run or pressing Shift+Enter) will list
↪ all files under the input directory

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

# Any results you write to the current directory are saved as output.
```

/kaggle/input/default-of-credit-card-clients-dataset/UCI_Credit_Card.csv

0.1 ### Correlation and Heatmaps

Correlation is a mathematical technique that shows how strongly two variables are linked. It quantifies the strength of the relationship. For instance, we know that the weight and height of a person are correlated. Taller people tend to have more weight. Hence, we say that height and weight are correlated.

Correlation is measured in terms of a number called correlation coefficient, which ranges from -1 to 1. The value of 1 or -1 denotes complete correlation, while 0 indicates that no correlation is present between the two variables. Negative values mean there is an inverse relationship between the two variables, while a positive value denotes a direct relationship.

```
[14]: df = pd.read_csv('/kaggle/input/default-of-credit-card-clients-dataset/
      ↪ UCI_Credit_Card.csv')
      correlation = df.corr()
      print(correlation)
```

	ID	LIMIT_BAL	SEX	EDUCATION	\
ID	1.000000	0.026179	0.018497	0.039177	
LIMIT_BAL	0.026179	1.000000	0.024755	-0.219161	
SEX	0.018497	0.024755	1.000000	0.014232	
EDUCATION	0.039177	-0.219161	0.014232	1.000000	
MARRIAGE	-0.029079	-0.108139	-0.031389	-0.143464	
AGE	0.018678	0.144713	-0.090874	0.175061	
PAY_0	-0.030575	-0.271214	-0.057643	0.105364	
PAY_2	-0.011215	-0.296382	-0.070771	0.121566	
PAY_3	-0.018494	-0.286123	-0.066096	0.114025	
PAY_4	-0.002735	-0.267460	-0.060173	0.108793	
PAY_5	-0.022199	-0.249411	-0.055064	0.097520	
PAY_6	-0.020270	-0.235195	-0.044008	0.082316	
BILL_AMT1	0.019389	0.285430	-0.033642	0.023581	
BILL_AMT2	0.017982	0.278314	-0.031183	0.018749	
BILL_AMT3	0.024354	0.283236	-0.024563	0.013002	
BILL_AMT4	0.040351	0.293988	-0.021880	-0.000451	
BILL_AMT5	0.016705	0.295562	-0.017005	-0.007567	
BILL_AMT6	0.016730	0.290389	-0.016733	-0.009099	
PAY_AMT1	0.009742	0.195236	-0.000242	-0.037456	
PAY_AMT2	0.008406	0.178408	-0.001391	-0.030038	
PAY_AMT3	0.039151	0.210167	-0.008597	-0.039943	
PAY_AMT4	0.007793	0.203242	-0.002229	-0.038218	
PAY_AMT5	0.000652	0.217202	-0.001667	-0.040358	
PAY_AMT6	0.003000	0.219595	-0.002766	-0.037200	
default.payment.next.month	-0.013952	-0.153520	-0.039961	0.028006	

	MARRIAGE	AGE	PAY_0	PAY_2	PAY_3	\
ID	-0.029079	0.018678	-0.030575	-0.011215	-0.018494	
LIMIT_BAL	-0.108139	0.144713	-0.271214	-0.296382	-0.286123	
SEX	-0.031389	-0.090874	-0.057643	-0.070771	-0.066096	
EDUCATION	-0.143464	0.175061	0.105364	0.121566	0.114025	
MARRIAGE	1.000000	-0.414170	0.019917	0.024199	0.032688	
AGE	-0.414170	1.000000	-0.039447	-0.050148	-0.053048	
PAY_0	0.019917	-0.039447	1.000000	0.672164	0.574245	
PAY_2	0.024199	-0.050148	0.672164	1.000000	0.766552	
PAY_3	0.032688	-0.053048	0.574245	0.766552	1.000000	
PAY_4	0.033122	-0.049722	0.538841	0.662067	0.777359	
PAY_5	0.035629	-0.053826	0.509426	0.622780	0.686775	
PAY_6	0.034345	-0.048773	0.474553	0.575501	0.632684	
BILL_AMT1	-0.023472	0.056239	0.187068	0.234887	0.208473	
BILL_AMT2	-0.021602	0.054283	0.189859	0.235257	0.237295	
BILL_AMT3	-0.024909	0.053710	0.179785	0.224146	0.227494	
BILL_AMT4	-0.023344	0.051353	0.179125	0.222237	0.227202	
BILL_AMT5	-0.025393	0.049345	0.180635	0.221348	0.225145	
BILL_AMT6	-0.021207	0.047613	0.176980	0.219403	0.222327	
PAY_AMT1	-0.005979	0.026147	-0.079269	-0.080701	0.001295	
PAY_AMT2	-0.008093	0.021785	-0.070101	-0.058990	-0.066793	

PAY_AMT3	-0.003541	0.029247	-0.070561	-0.055901	-0.053311
PAY_AMT4	-0.012659	0.021379	-0.064005	-0.046858	-0.046067
PAY_AMT5	-0.001205	0.022850	-0.058190	-0.037093	-0.035863
PAY_AMT6	-0.006641	0.019478	-0.058673	-0.036500	-0.035861
default.payment.next.month	-0.024339	0.013890	0.324794	0.263551	0.235253

	PAY_4	...	BILL_AMT4	BILL_AMT5	BILL_AMT6	\
ID	-0.002735	...	0.040351	0.016705	0.016730	
LIMIT_BAL	-0.267460	...	0.293988	0.295562	0.290389	
SEX	-0.060173	...	-0.021880	-0.017005	-0.016733	
EDUCATION	0.108793	...	-0.000451	-0.007567	-0.009099	
MARRIAGE	0.033122	...	-0.023344	-0.025393	-0.021207	
AGE	-0.049722	...	0.051353	0.049345	0.047613	
PAY_0	0.538841	...	0.179125	0.180635	0.176980	
PAY_2	0.662067	...	0.222237	0.221348	0.219403	
PAY_3	0.777359	...	0.227202	0.225145	0.222327	
PAY_4	1.000000	...	0.245917	0.242902	0.239154	
PAY_5	0.819835	...	0.271915	0.269783	0.262509	
PAY_6	0.716449	...	0.266356	0.290894	0.285091	
BILL_AMT1	0.202812	...	0.860272	0.829779	0.802650	
BILL_AMT2	0.225816	...	0.892482	0.859778	0.831594	
BILL_AMT3	0.244983	...	0.923969	0.883910	0.853320	
BILL_AMT4	0.245917	...	1.000000	0.940134	0.900941	
BILL_AMT5	0.242902	...	0.940134	1.000000	0.946197	
BILL_AMT6	0.239154	...	0.900941	0.946197	1.000000	
PAY_AMT1	-0.009362	...	0.233012	0.217031	0.199965	
PAY_AMT2	-0.001944	...	0.207564	0.181246	0.172663	
PAY_AMT3	-0.069235	...	0.300023	0.252305	0.233770	
PAY_AMT4	-0.043461	...	0.130191	0.293118	0.250237	
PAY_AMT5	-0.033590	...	0.160433	0.141574	0.307729	
PAY_AMT6	-0.026565	...	0.177637	0.164184	0.115494	
default.payment.next.month	0.216614	...	-0.010156	-0.006760	-0.005372	

	PAY_AMT1	PAY_AMT2	PAY_AMT3	PAY_AMT4	PAY_AMT5	\
ID	0.009742	0.008406	0.039151	0.007793	0.000652	
LIMIT_BAL	0.195236	0.178408	0.210167	0.203242	0.217202	
SEX	-0.000242	-0.001391	-0.008597	-0.002229	-0.001667	
EDUCATION	-0.037456	-0.030038	-0.039943	-0.038218	-0.040358	
MARRIAGE	-0.005979	-0.008093	-0.003541	-0.012659	-0.001205	
AGE	0.026147	0.021785	0.029247	0.021379	0.022850	
PAY_0	-0.079269	-0.070101	-0.070561	-0.064005	-0.058190	
PAY_2	-0.080701	-0.058990	-0.055901	-0.046858	-0.037093	
PAY_3	0.001295	-0.066793	-0.053311	-0.046067	-0.035863	
PAY_4	-0.009362	-0.001944	-0.069235	-0.043461	-0.033590	
PAY_5	-0.006089	-0.003191	0.009062	-0.058299	-0.033337	
PAY_6	-0.001496	-0.005223	0.005834	0.019018	-0.046434	
BILL_AMT1	0.140277	0.099355	0.156887	0.158303	0.167026	
BILL_AMT2	0.280365	0.100851	0.150718	0.147398	0.157957	

BILL_AMT3	0.244335	0.316936	0.130011	0.143405	0.179712
BILL_AMT4	0.233012	0.207564	0.300023	0.130191	0.160433
BILL_AMT5	0.217031	0.181246	0.252305	0.293118	0.141574
BILL_AMT6	0.199965	0.172663	0.233770	0.250237	0.307729
PAY_AMT1	1.000000	0.285576	0.252191	0.199558	0.148459
PAY_AMT2	0.285576	1.000000	0.244770	0.180107	0.180908
PAY_AMT3	0.252191	0.244770	1.000000	0.216325	0.159214
PAY_AMT4	0.199558	0.180107	0.216325	1.000000	0.151830
PAY_AMT5	0.148459	0.180908	0.159214	0.151830	1.000000
PAY_AMT6	0.185735	0.157634	0.162740	0.157834	0.154896
default.payment.next.month	-0.072929	-0.058579	-0.056250	-0.056827	-0.055124

	PAY_AMT6	default.payment.next.month
ID	0.003000	-0.013952
LIMIT_BAL	0.219595	-0.153520
SEX	-0.002766	-0.039961
EDUCATION	-0.037200	0.028006
MARRIAGE	-0.006641	-0.024339
AGE	0.019478	0.013890
PAY_0	-0.058673	0.324794
PAY_2	-0.036500	0.263551
PAY_3	-0.035861	0.235253
PAY_4	-0.026565	0.216614
PAY_5	-0.023027	0.204149
PAY_6	-0.025299	0.186866
BILL_AMT1	0.179341	-0.019644
BILL_AMT2	0.174256	-0.014193
BILL_AMT3	0.182326	-0.014076
BILL_AMT4	0.177637	-0.010156
BILL_AMT5	0.164184	-0.006760
BILL_AMT6	0.115494	-0.005372
PAY_AMT1	0.185735	-0.072929
PAY_AMT2	0.157634	-0.058579
PAY_AMT3	0.162740	-0.056250
PAY_AMT4	0.157834	-0.056827
PAY_AMT5	0.154896	-0.055124
PAY_AMT6	1.000000	-0.053183
default.payment.next.month	-0.053183	1.000000

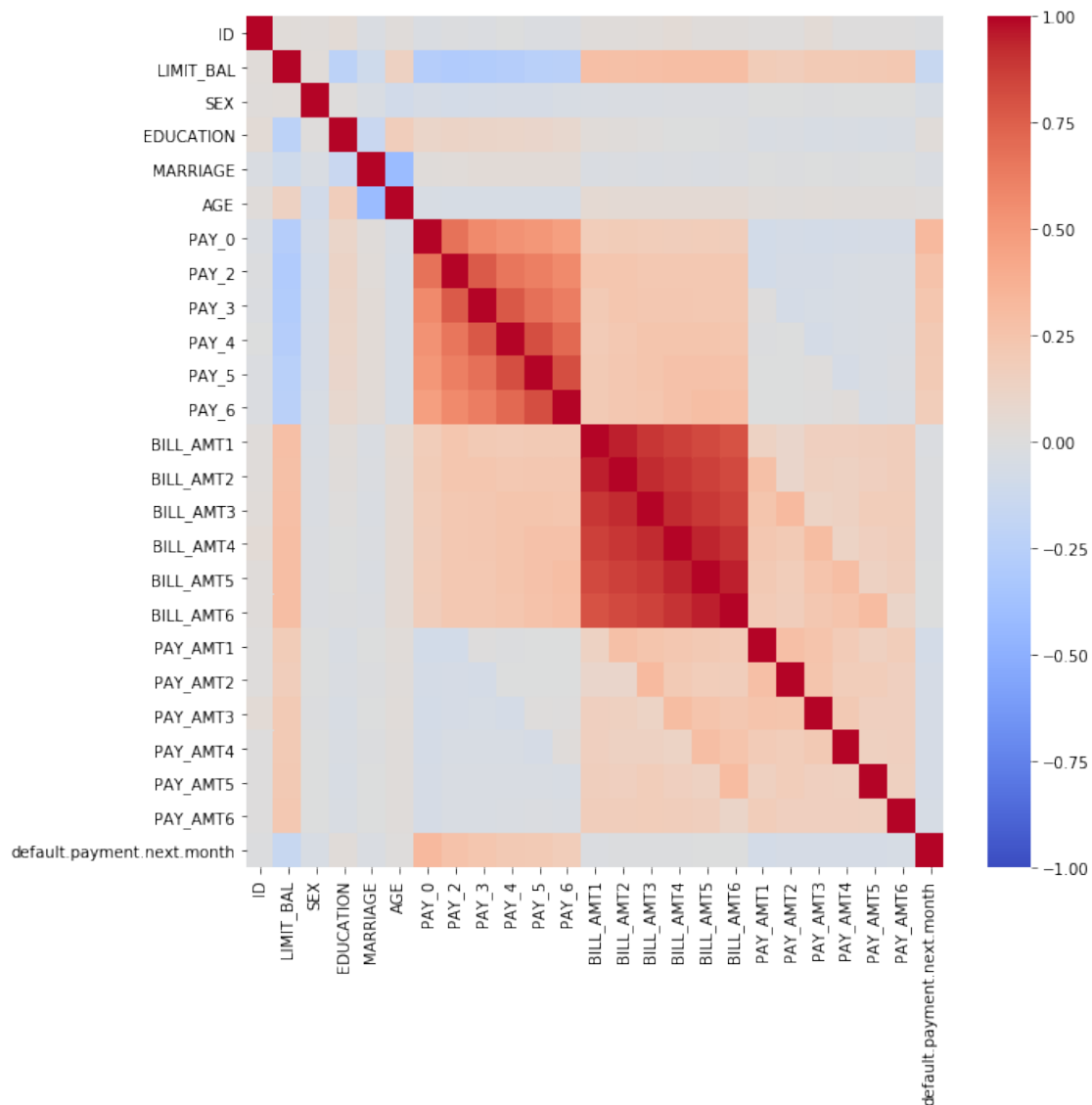
[25 rows x 25 columns]

we can see here a list of correlation value, it is very difficult to study let's create a **Heatmap** of correlation that is much easier to study. A **heatmap** is a graphical representation of data where individual values are represented as colors. The intensity of the colors indicates the values.

```
[18]: import matplotlib.pyplot as plt
import seaborn as sns
```

```
# correlation table
corr = df.corr()
#plot heatmap
#vmin = minimum number of color
#vmax = maximum number of color
plt.figure(figsize=(10,10))
sns.heatmap(data=corr, vmin=-1, vmax=1, cmap='coolwarm')
```

[18]: <matplotlib.axes._subplots.AxesSubplot at 0x7f6320db22e8>



let's study on this heatmap,

1. we can see **BILL_AMT1**, **BILL_AMT2** **BILL_AMT6** are positively correlated with each other.
2. **LIMIT_BAL** is has positive correlation with **BILL_AMT1**, **BILL_AMT2** **BILL_AMT6**, which implies that people who were given more credit (higher values of **LIMIT_BAL**) tend to have larger bills.
3. **LIMIT_BAL** has negative correlation with payment delay variables **PAY_0**, **PAY_6** which implies higher limite balance tends to fewer payment delay.
4. There are not correlation in **SEX**, **MARRIAGE**.