

Project Description:

The project aims to use exploratory data analysis (EDA) to analyze loan application data and identify patterns that indicate if a client has difficulty paying their installments. The project aims to understand the driving factors behind loan default, i.e. the variables which are strong indicators of default, and utilize this knowledge for the company's portfolio and risk assessment.

Tech-Stack Used:

The project was performed using Excel version 2021 Excel was chosen for its powerful data analysis and visualization capabilities It is also widely used in the industry and provides a familiar environment for users.

Insights:

- 1. The mean income for the income is around 160000.
- 2. There is something wrong in the data the mean of the days_employeed in not acceptable.
- 3. 8 percent people have difficulties in payments.
- 4. It show that Amt_Goods_Price and Amt_Credit are the most correlated variable followed by Amt_Annuity & Amt_Goods_price and Amt_Credit & Amt_Annuity.
- 5. Although Females are more difficulties with payment but percentage wise it is male who face difficulties in payments.
- 6. Here are more females than males. The ratio of male and female in around 1:2.
- 7. The average of the income for male is more as compared to female.
- 8. Academic degree holder apply for the loan most.
- 9. Academic degree holder ask for more credit as compare to other educational background.
- 10. Male ask for greater money for loan as compare to the female.

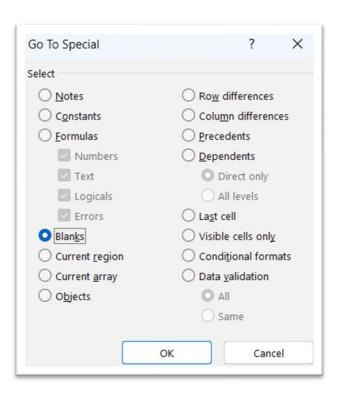
Result:

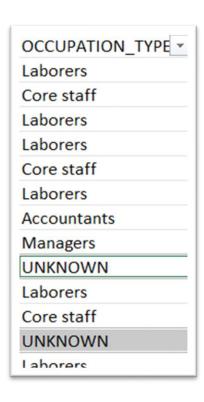
Through this project, I was able to gain valuable insights into the factors that influence loan default and develop a basic understanding of risk analytics in the banking and financial services sector. What I find is on next pages.

Present the overall approach of the analysis. Mention the problem statement and the analysis approach briefly

The project was executed by first performing data cleaning and preprocessing, followed by exploratory data analysis (EDA). During EDA, various statistical and visualization techniques were used to gain insights into the data and identify patterns related to loan default. The data was also analyzed in conjunction with the data dictionary and domain research on risk analytics.

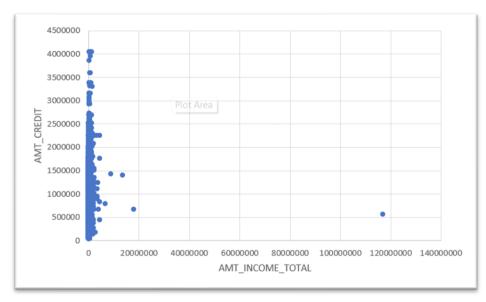
Identify the missing data and use appropriate method to deal with it. (Remove columns/or replace it with an appropriate value):

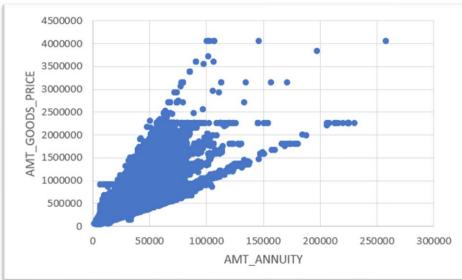




- Delete the row which has blank vlues.
- In some column such as occupation_type with blank values in replaced with unkown.
- Delete some columns which is not required in the analysis.

Identify if there are outliers in the dataset. Also, mention why do you think it is an outlier:





- SK_ID_CURR 114967 is the outlier because his credit is very less as compare to his income.
- In second graph SK_ID_CURR 120926 is the outlier with respect to amount goods price and amount annuity.
- Scatter Plot is used for getting the outliers.

Identify if there is data imbalance in the data. Find the ratio of data imbalance:

Row Labels	Count of NAME_INCO	ME_TYPE	Ratio
Working		158774	31754.8
Commercial associat	e	71617	14323.4
Pensioner		55362	11072.4
State servant		21703	4340.6
Unemployed		22	4.4
Student		18	3.6
Businessman		10	2.0
Maternity leave		5	1.0
Grand Total		307511	

 Pivot table is used for checking the data imbalance. Here data is imbalanced as most of the data is from the working Income Type where as very less for business. The ration for working and business income type is 31000:2

Row Labels 📝 Count of HOUSE	TYPE_MODE	Ratio
block of flats	150503	124.2
specific housing	1499	1.2
terraced house	1212	1.0
Grand Total	153214	

•	Here data is imbalanced because most of the House type is
	block of flats and others are very less. The ratio of block of
	flats and other house type is 124:2

- Row Labels Count of NAME_EDUCATION_TYPE Ratio Secondary / secondary special 1332 218391 456 Higher education 74863 Incomplete higher 10277 Lower secondary 3816 Academic degree 164 **Grand Total** 307511
- Here data is imbalanced because most of the data is has education type and very less for academic degree their ratio is 1332:1

Row Labels 🔻	Count of CODE_GENDER
F	202448
M	105059
XNA	4
Grand Total	307511

Here are more females than males. The ratio of male and female in around 1:2.

Univariate:

AMT_INCOM	ME_TOTAL	AMT_CREDIT		DAYS_EMPLO	YED In	Years
Mean	168798	Mean	599025.9997	Mean	63815	175
Median	147150	Median	513531	Median	-1213	-3
Mode	135000	Mode	450000	Mode	365243	1001
AMT_GOOD	S_PRICE	REGION_POPU	LATION_RELATIVE	DAYS_BIRTH	Ąį	ge in Years
Mean	538396	Mean	0.02087	Mean	-16037	44
Median	450000	Median	0.01885	Median	-15750	43
Mode	450000	Mode	0.03579	Mode	-13749	38
AMT_ANNU	ITY	DAYS_REGISTR	ATION	DAYS_ID_PUB	SLISH in	years
Mean	27109	Mean	-4986	Mean	-2994	8
Median	24903	Median	-4504	Median	-3254	9
Mode	9000	Mode	-1	Mode	-4053	11

- In first table AMT_INCOME_TATAL mean is amount 168798, median is 147150 where as mode is 135000
- In second table AMT_CREDIT mean is amount 599025, median is 513531 and mode is 450000.
- In DAYS_EMPLOYED mean is 175 years and mode is 1001 years which not acceptable as data.

=AVERAGE(application_data!H:H)

=MEDIAN(application_data!H:H)

=MODE.MULT(application data!H:H)

- These three formulae are used for the univariate Analysis.
- In Days_Employed Column there is some mistake in data because mean of the is 174 years which is wrong and mode is 1001 years when days are converted into years.

Segmented Univariate:

Average Amount of Income on Gender Basis								
Row Labels 🕶 Average of AMT_INCOME_TOTAL								
F	156032							
М	193396							
XNA	186750							
Grand Total	168798							

Average Amount of Credit on Education Basis							
Row Labels	Average of AMT_CREDIT						
Academic degree	723516						
Higher education	689950						
Incomplete higher	566731						
Lower secondary	489749						
Secondary / secondary speci-	al 571193						
Grand Total	599026						

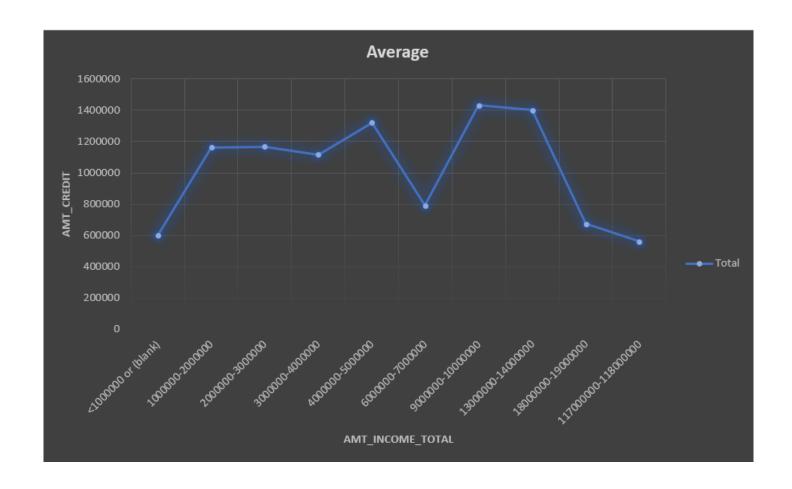
Average Amount	of	Credit on Education Basis
Row Labels	Ţ,	Average of AMT_INCOME_TOTAL
Academic degree		240009
Higher education		208652
Incomplete higher		181564
Lower secondary		130079
Secondary / secondary spec	ial	155159
Grand Total		168798
Average Amount o	of	Credit on Gender Basis
Row Labels 🕶 Ave	era	age of AMT_CREDIT

- The average of the income for male is more as compared to female.
- Academic degree holder apply for the loan most.
- Academic degree holder ask for more credit as compare to other educational background.
- Male ask for greater money for loan as compare to the female.

Bivariate Analysis (Part-1):

Row Labels	▼ Average of AMT_CREDIT
<1000000 or (blank)	598568
1000000-2000000	1163950
2000000-3000000	1167367
3000000-4000000	1116665
4000000-5000000	1322595
6000000-7000000	790830
9000000-10000000	1431531
13000000-14000000	1400504
18000000-19000000	675000
117000000-118000000	562491
Grand Total	599026

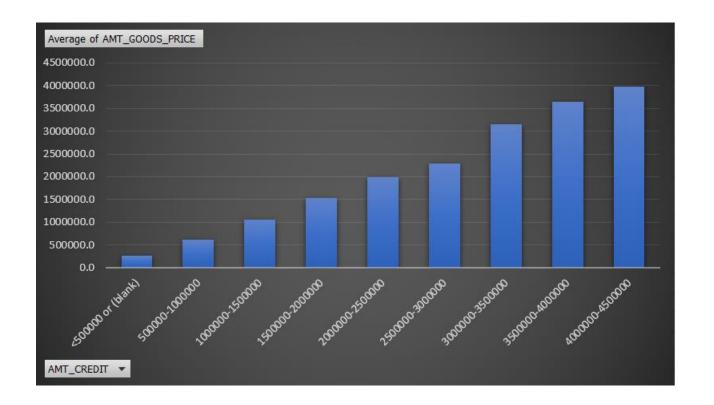
- This is Bivariate Analysis because it show how the one parameter is behave with the change in other.
- The adjacent line graph show that people with income 9000000 to 14000000 is ask for more loan.
- Where as income between 6000000 to 7000000 is lesser than the nearest income class.



Bivariate Analysis (Part-2):

Row Labels	¥	Average of AMT_GOODS_PRICE
<500000 or (blank	k)	259932.3
500000-1000000		622558.8
1000000-1500000	0	1064841.0
1500000-2000000	0	1531780.4
2000000-2500000	0	1989482.6
2500000-3000000	0	2285808.5
3000000-3500000	0	3157500.0
3500000-4000000	0	3645000.0
4000000-4500000	0	3971250.0
Grand Total		538396

• The adjacent bar graph show that Credit amount is increases with the increase of Good Price.



Find the top 10 correlation for the Clients other than payment difficulties:

	AMT_INCOM	AMT_CRE	AMT_ANN	AMT_GOO	REGION_PO	DAYS_BIRT	DAYS_EMPL	DAYS_REGIS	DAYS_ID_PU
AMT_INCOME_TOTAL	1.000								
AMT_CREDIT	0.343	1.000							
AMT_ANNUITY	0.419	0.771	1.000						
AMT_GOODS_PRICE	0.349	0.987	0.777	1.000					
REGION_POPULATION_RELATIVE	0.168	0.100	0.121	0.104	1.000				
DAYS_BIRTH	0.063	-0.047	0.013	-0.045	-0.025	1.000			
DAYS_EMPLOYED	-0.141	-0.073	-0.107	-0.071	-0.007	-0.618	1.000		
DAYS_REGISTRATION	0.065	0.014	0.039	0.016	-0.052	0.333	-0.210	1.000	
DAYS_ID_PUBLISH	0.023	-0.001	0.014	-0.004	-0.001	0.271	-0.274	0.100	1.000

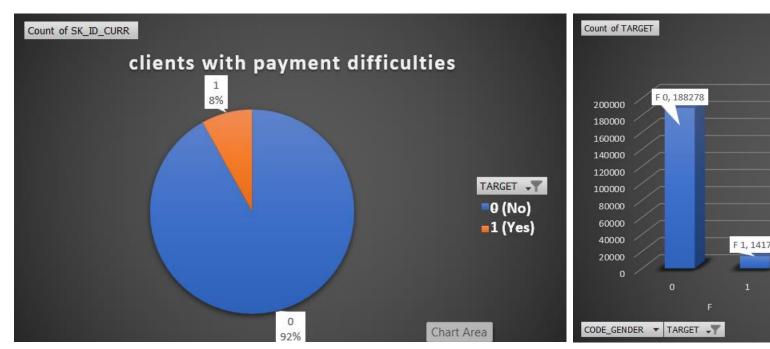
- This is the correlation matrix for clients other than payment difficulties.
- It show that Amt_Goods_Price and Amt_Credit are the most correlated variable followed by Amt_Annuity & Amt_Goods_price and Amt_Credit & Amt_Annuity.

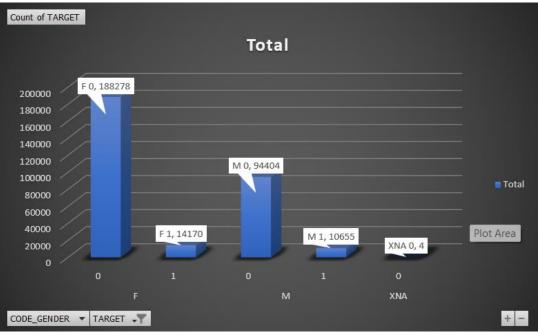
Find the top 10 correlation for the Client with payment difficulties:

	AMT_INCOME_	AMT_CR	AMT_AN	AMT_GO	REGION_	DAYS_BIR	DAYS_E	DAYS_RE	DAYS_ID
	TOTAL	EDIT	NUITY	ODS_PRI	POPULAT	TH	MPLOYE	GISTRATI	_PUBLIS
AMT_INCOME_TOTAL	1.000								
AMT_CREDIT	0.038	1.000							
AMT_ANNUITY	0.046	0.752	1.000						
AMT_GOODS_PRICE	0.038	0.983	0.753	1.000					
REGION_POPULATION_RELATIVE	0.009	0.069	0.072	0.076	1.000				
DAYS_BIRTH	0.003	-0.135	-0.014	-0.136	-0.048	1.000			
DAYS_EMPLOYED	-0.015	-0.001	-0.083	0.004	0.015	-0.575	1.000		
DAYS_REGISTRATION	0.000	-0.026	0.034	-0.026	-0.056	0.289	-0.189	1.000	
DAYS_ID_PUBLISH	-0.004	-0.052	-0.017	-0.056	-0.016	0.253	-0.226	0.097	1.000

- This is the correlation matrix for clients with <u>payment</u> difficulties.
- It show that Amt_Goods_Price and Amt_Credit are the most correlated variable followed by Amt_Annuity & Amt_Goods_price and Amt_Credit & Amt_Annuity.

Include visualizations and summarize the most important results in the presentation:

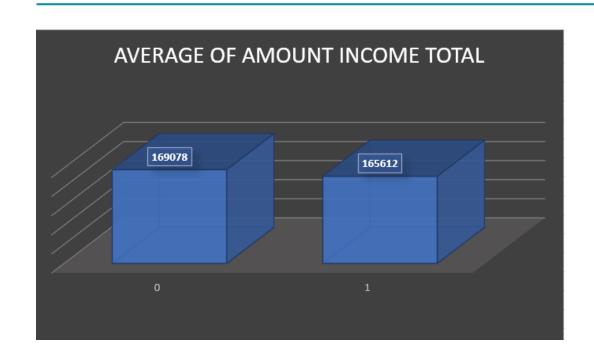


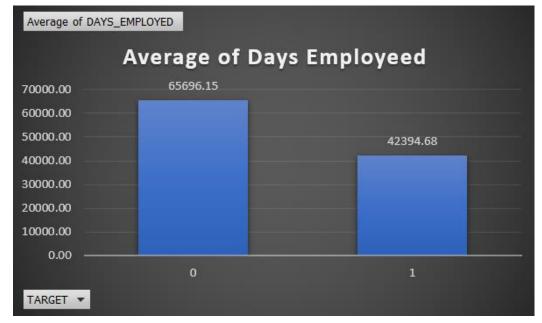


• In above pie chart we can see that on 8% of clients have payments difficulties.

• According to the above graph it is shown that although Females are more difficulties with payment but percentage wise it is male who face difficulties in payments.

Include visualizations and summarize the most important results in the presentation





• In above bar chart it can be seen that client with payment difficulties and other clients have almost same income amount.

• In above bar chart it can be seen that client with payment difficulties has lesser number of days employeed.

