

PROJECT DESCRIPTION:

The objective of this project was to analyze a dataset containing loan applications from urban customers. Our company faces the challenge of accurately predicting loan defaults to minimize financial losses while maximizing business opportunities. By identifying patterns and factors that influence loan default, we aim to improve our decision-making process regarding loan approvals.

APPROACH:

The approach I have used here is Exploratory Data Analysis (EDA) to understand the distribution and relationships between customer attributes and loan attributes. I began by identifying missing data and outliers in the dataset, ensuring data integrity through appropriate handling techniques.

	A	B	C	D	E	F	G	H	I	J	K	L	M	N
1	SR_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODES_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	NAME_TYPE_SUITE	NAME_INCOME_TYPE	NAME_EDUCATION_TYPE
2	100002	1	Cash loans	M	N	Y	0	202300	406097.5	24700.5	351000	Unaccompanied	Working	Secondary / secondary spec
3	100003	0	Cash loans	F	N	N	0	270000	1293002.5	30608.5	1129000	Family	State servant	Higher education
4	100004	0	Revolving loans	M	Y	Y	0	67000	135000	6700	135000	Unaccompanied	Working	Secondary / secondary spec
5	100006	0	Cash loans	F	N	Y	0	135000	312602.5	29086.5	257900	Unaccompanied	Working	Secondary / secondary spec
6	100007	0	Cash loans	M	N	Y	0	121500	513000	21865.5	513000	Unaccompanied	Working	Secondary / secondary spec
7	100008	0	Cash loans	M	N	Y	0	99000	406405.5	27517.5	434500	Spouse, partner	State servant	Secondary / secondary spec
8	100009	0	Cash loans	F	Y	Y	1	171000	1566726	41301	1395000	Unaccompanied	Commercial associate	Higher education
9	100010	0	Cash loans	M	Y	Y	0	360000	1530000	42075	1530000	Unaccompanied	State servant	Higher education
10	100011	0	Cash loans	F	N	Y	0	112500	3019010	33026.5	912500	Children	Pensioner	Secondary / secondary spec
11	100012	0	Revolving loans	M	N	Y	0	135000	405000	20250	405000	Unaccompanied	Working	Secondary / secondary spec
12	100014	0	Cash loans	F	N	Y	1	112500	652500	21177	652500	Unaccompanied	Working	Higher education
13	100015	0	Cash loans	F	N	Y	0	36415.155	148363	10678.5	135000	Children	Pensioner	Secondary / secondary spec
14	100016	0	Cash loans	F	N	Y	0	67000	80863	5081.5	67000	Unaccompanied	Working	Secondary / secondary spec
15	100017	0	Cash loans	M	Y	N	1	225000	918404	28966.5	607500	Unaccompanied	Working	Secondary / secondary spec
16	100018	0	Cash loans	F	N	Y	0	389000	773600.5	32778	679000	Unaccompanied	Working	Secondary / secondary spec
17	100019	0	Cash loans	M	Y	Y	0	157500	299772	20160	247500	Family	Working	Secondary / secondary spec
18	100020	0	Cash loans	M	N	N	0	108000	509602.5	26149.5	387000	Unaccompanied	Working	Secondary / secondary spec
19	100021	0	Revolving loans	F	N	Y	1	81000	270000	13500	270000	Unaccompanied	Working	Secondary / secondary spec
20	100022	0	Revolving loans	F	N	Y	0	112500	197500	7675	157500	Other_A	Working	Secondary / secondary spec
21	100023	0	Cash loans	F	N	Y	1	90000	544401	17562.5	454500	Unaccompanied	State servant	Higher education
22	100024	0	Revolving loans	M	Y	Y	0	135000	427500	21375	427500	Unaccompanied	Working	Secondary / secondary spec
23	100025	0	Cash loans	F	Y	Y	1	202500	1132573.5	37961.5	927000	Unaccompanied	Commercial associate	Secondary / secondary spec
24	100026	0	Cash loans	F	N	N	1	450000	497520	32521.5	450000	Unaccompanied	Working	Secondary / secondary spec
25	100027	0	Cash loans	F	N	Y	0	83250	239850	23850	225000	Unaccompanied	Pensioner	Secondary / secondary spec
26	100029	0	Cash loans	M	Y	N	2	135000	247500	12701.5	247500	Unaccompanied	Working	Secondary / secondary spec
27	100030	0	Cash loans	F	N	Y	0	90000	225000	11074.5	225000	Unaccompanied	Working	Secondary / secondary spec
28	100031	1	Cash loans	F	N	Y	0	112500	979992	27076.5	762000	Unaccompanied	Working	Secondary / secondary spec
29	100032	0	Cash loans	M	N	Y	1	112500	327024	23827.5	270000	Family	Working	Secondary / secondary spec
30	100033	0	Cash loans	M	Y	Y	0	279000	790530	5767.5	670000	Unaccompanied	State servant	Higher education
31	100034	0	Revolving loans	M	N	Y	0	90000	180000	9000	180000	Unaccompanied	Working	Higher education
32	100035	0	Cash loans	F	N	Y	0	292500	965092	24302.5	477000	Unaccompanied	Commercial associate	Secondary / secondary spec
33	100036	0	Cash loans	F	N	Y	0	112500	512064	25033.5	360000	Family	Working	Secondary / secondary spec
34	100037	0	Cash loans	F	N	N	0	90000	199008	20893.5	180000	Unaccompanied	Working	Secondary / secondary spec
35	100039	0	Cash loans	M	Y	N	1	360000	733315.5	39069	670000	Unaccompanied	Commercial associate	Secondary / secondary spec
36	100040	0	Cash loans	F	N	Y	0	135000	1125000	32095	1125000	Unaccompanied	State servant	Higher education
37	100041	0	Cash loans	F	N	N	0	112500	450000	44505.5	450000	Unaccompanied	Working	Higher education

TECH-STACK USED:

Here, I have used Microsoft Excel for various operations, while using Excel functions like count, quartile, correl and many more. The Excel functionalities also helped in getting accurate analysis by analyzing the data in the form of graphs like pivot tables, charts and conditional formatting.

Insights:

Here, I will be sharing the insights for the tasks provided in the project. This goes as follows:

TASK A: Identifying the missing data and handling them.

Missing values were identified using the 'ISBLANK' function and imputed with 'AVERAGE' or 'MEDIAN', ensuring that the dataset remained unbiased and reliable. The screenshot of the following task is shown below:

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O							
1	Blank Count	0	0	0	0	0	0	0	0	0	1	38	192	0	0							
2	Blank %	0	0	0	0	0	0	0	0	0	0.00200004	0.07600182	0.38400768	0	0							
3		SK ID	CURR	TARGET	NAME	CONTRACT	TYPE	CODE	GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT CHILDREN	AMT INCOME TOTAL	AMT CREDIT	AMT ANNUITY	AMT GOODS PRICE	NAME	TYPE	SURTE	NAME	INCOME	TYPE
4		100002	1	Cash loans	M	N	N	Y				0	202500	406597.5	24700.5	351000	Unaccompanied	Working				
5		100003	0	Cash loans	F	N	N	N				0	270000	1293502.5	35098.5	1125000	Family	State servant				
6		100004	0	Revolving loans	M	N	Y	Y				0	67500	155000	6750	135000	Unaccompanied	Working				
7		100006	0	Cash loans	F	N	Y	Y				0	135000	312682.5	29686.5	297000	Unaccompanied	Working				
8		100007	0	Cash loans	M	N	N	Y				0	121500	513000	21865.5	513000	Unaccompanied	Working				
9		100008	0	Cash loans	M	N	Y	Y				0	99000	404935.5	27517.5	454500	Spouse, partner	State servant				
10		100009	0	Cash loans	F	Y	Y	Y				1	171000	1560726	41301	1395000	Unaccompanied	Commercial associate				
11		100010	0	Cash loans	M	Y	Y	Y				0	360000	1553000	42075	1530000	Unaccompanied	State servant				
12		100011	0	Cash loans	F	N	Y	Y				0	112500	1019610	33265.5	913500	Children	Pensioner				
13		100012	0	Revolving loans	M	N	N	Y				0	135000	405000	20250	405000	Unaccompanied	Working				
14		100014	0	Cash loans	F	N	N	Y				1	112500	652500	21177	652500	Unaccompanied	Working				
15		100015	0	Cash loans	F	N	N	Y				0	38419.155	148365	10678.5	135000	Children	Pensioner				
16		100016	0	Cash loans	F	N	Y	Y				0	67500	80965	5881.5	67500	Unaccompanied	Working				
17		100017	0	Cash loans	M	Y	N	N				1	225000	919468	28966.5	67500	Unaccompanied	Working				
18		100018	0	Cash loans	F	N	Y	Y				0	189000	773680.5	32778	675000	Unaccompanied	Working				
19		100019	0	Cash loans	M	Y	Y	Y				0	157500	299772	20160	247500	Family	Working				
20		100020	0	Cash loans	M	N	Y	N				0	108000	509602.5	26149.5	387000	Unaccompanied	Working				
21		100021	0	Revolving loans	F	N	N	Y				1	81000	270000	13500	270000	Unaccompanied	Working				
22		100022	0	Revolving loans	F	N	N	Y				0	112500	157500	7875	137500	Other A	Working				
23		100023	0	Cash loans	F	N	Y	Y				0	90000	544491	17563.5	454500	Unaccompanied	State servant				
24		100024	0	Revolving loans	M	Y	Y	Y				0	135000	427500	21375	427500	Unaccompanied	Working				
25		100025	0	Cash loans	F	Y	Y	Y				1	202500	1132573.5	37561.5	927000	Unaccompanied	Commercial associate				
26		100026	0	Cash loans	F	N	N	N				1	450000	497520	32521.5	450000	Unaccompanied	Working				
27		100027	0	Cash loans	F	N	N	Y				0	82500	239650	23850	225000	Unaccompanied	Pensioner				
28		100029	0	Cash loans	M	Y	N	N				2	135000	247500	12763.5	247500	Unaccompanied	Working				
29		100030	0	Cash loans	F	N	Y	Y				0	90000	225000	11074.5	225000	Unaccompanied	Working				
30		100031	1	Cash loans	F	N	Y	Y				0	112500	979992	27076.5	702000	Unaccompanied	Working				
31		100032	0	Cash loans	M	N	Y	Y				1	112500	327024	23827.5	270000	Family	Working				
32		100033	0	Cash loans	M	Y	Y	Y				0	270000	790830	57676.5	675000	Unaccompanied	State servant				
33		100034	0	Revolving loans	M	N	Y	Y				0	90000	180000	9000	100000	Unaccompanied	Working				
34		100035	0	Cash loans	F	N	Y	Y				0	292500	665892	24892.5	477000	Unaccompanied	Commercial associate				
35		100036	0	Cash loans	F	N	N	Y				0	112500	512064	25033.5	360000	Family	Working				
36		100037	0	Cash loans	F	N	N	N				0	90000	199008	20893.5	180000	Unaccompanied	Working				
37		100039	0	Cash loans	M	Y	N	N				1	360000	733315.5	35069	675000	Unaccompanied	Commercial associate				

HANDLING MISSING VALUES

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
1																	
2	NO. OF NULL	1564	192														
3																	
4		OCCUPATION TYPE	NAME TYPE SURTE														
5		Unaccompanied															
6		Core staff	Family														
7		Laborers	Unaccompanied														
8		Laborers	Unaccompanied														
9		Core staff	Unaccompanied														
10		Laborers	Spouse, partner														
11		Accountants	Unaccompanied														
12		Managers	Unaccompanied														
13		Children															
14		Laborers	Unaccompanied														
15		Core staff	Unaccompanied														
16		Children															
17		Laborers	Unaccompanied														
18		Drivers	Unaccompanied														
19		Laborers	Unaccompanied														
20		Laborers	Family														
21		Drivers	Unaccompanied														
22		Laborers	Unaccompanied														
23		Laborers	Other A														
24		Core staff	Unaccompanied														
25		Laborers	Unaccompanied														
26		Sales staff	Unaccompanied														
27		Sales staff	Unaccompanied														
28		Drivers	Unaccompanied														
29		Cleaning staff	Unaccompanied														
30		Cooking staff	Unaccompanied														
31		Laborers	Family														
32		Laborers	Unaccompanied														
33		Laborers	Unaccompanied														
34		Laborers	Unaccompanied														
35		Private service staff	Family														

FULL DATA WITHOUT MISSING VALUES

#	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	SK ID CURR	TARGET	NAME CONTRACT TYPE	CODE GENDER	FLAG OWN CAR	FLAG OWN REALTY	CNT CHILDREN	AMT INCOME TOTAL	AMT CREDIT	AMT ANNUITY	AMT GOODS PRICE	NAME TYPE SUITE	NAME INCOME TYPE	NAME EDUCATION TYPE	
2	100002	1	Cash loans	M	N	Y	0	202500	406597.5	24700.5	351000	Unaccompanied	Working	Secondary / secondary speci	
3	100003	0	Cash loans	F	N	N	0	270000	1293502.5	34908.5	1129500	Family	State servant	Higher education	
4	100004	0	Revolving loans	M	Y	Y	0	67500	130000	6750	135000	Unaccompanied	Working	Secondary / secondary speci	
5	100006	0	Cash loans	F	N	Y	0	135000	312682.5	29606.5	297000	Unaccompanied	Working	Secondary / secondary speci	
6	100007	0	Cash loans	M	N	Y	0	121500	513000	21865.5	513000	Unaccompanied	Working	Secondary / secondary speci	
7	100008	0	Cash loans	M	N	Y	0	99000	490495.5	27517.5	434500	Spouse, partner	State servant	Secondary / secondary speci	
8	100009	0	Cash loans	F	Y	Y	1	171000	1560726	41361	1365600	Unaccompanied	Commercial associate	Higher education	
9	100010	0	Cash loans	M	Y	Y	0	265000	1350000	42075	1530000	Unaccompanied	State servant	Higher education	
10	100011	0	Cash loans	F	N	Y	0	112500	1019610	33826.5	913500	Children	Pensioner	Secondary / secondary speci	
11	100012	0	Revolving loans	M	N	Y	0	135000	405000	20250	405000	Unaccompanied	Working	Secondary / secondary speci	
12	100014	0	Cash loans	F	N	Y	1	112500	652500	21177	652500	Unaccompanied	Working	Higher education	
13	100015	0	Cash loans	F	N	Y	0	38419.155	148365	10676.5	159000	Children	Pensioner	Secondary / secondary speci	
14	100016	0	Cash loans	F	N	Y	0	67500	80965	3691.5	67500	Unaccompanied	Working	Secondary / secondary speci	
15	100017	0	Cash loans	M	Y	N	1	225000	918468	26966.5	697500	Unaccompanied	Working	Secondary / secondary speci	
16	100018	0	Cash loans	F	N	Y	0	189000	773680.5	32776	679500	Unaccompanied	Working	Secondary / secondary speci	
17	100019	0	Cash loans	M	Y	Y	0	157500	299772	20160	247500	Family	Working	Secondary / secondary speci	
18	100020	0	Cash loans	M	N	N	0	108000	509602.5	26149.5	367000	Unaccompanied	Working	Secondary / secondary speci	
19	100021	0	Revolving loans	F	N	Y	1	81000	270000	13500	270000	Unaccompanied	Working	Secondary / secondary speci	
20	100022	0	Revolving loans	F	N	Y	0	112500	157500	7675	157500	Other A	Working	Secondary / secondary speci	
21	100023	0	Cash loans	F	N	Y	1	90000	544491	17563.5	454500	Unaccompanied	State servant	Higher education	
22	100024	0	Revolving loans	M	Y	Y	0	135000	427500	21375	427500	Unaccompanied	Working	Secondary / secondary speci	
23	100025	0	Cash loans	F	Y	Y	1	202500	1132573.5	37561.5	927000	Unaccompanied	Commercial associate	Secondary / secondary speci	
24	100026	0	Cash loans	F	N	N	1	450000	497520	32521.5	450000	Unaccompanied	Working	Secondary / secondary speci	
25	100027	0	Cash loans	F	N	Y	0	83250	239550	23950	225000	Unaccompanied	Pensioner	Secondary / secondary speci	
26	100029	0	Cash loans	M	Y	N	2	135000	247500	12703.5	247500	Unaccompanied	Working	Secondary / secondary speci	
27	100030	0	Cash loans	F	N	Y	0	90000	225000	11074.5	225000	Unaccompanied	Working	Secondary / secondary speci	
28	100031	1	Cash loans	F	N	Y	0	112500	979992	27076.5	702000	Unaccompanied	Working	Secondary / secondary speci	
29	100032	0	Cash loans	M	N	Y	1	112500	32724	23627.5	270000	Family	Working	Secondary / secondary speci	
30	100033	0	Cash loans	M	Y	Y	1	270000	790630	57676.5	675000	Unaccompanied	State servant	Higher education	
31	100034	0	Revolving loans	M	N	Y	0	90000	180000	9000	180000	Unaccompanied	Working	Higher education	
32	100035	0	Cash loans	F	N	Y	0	292500	660892	24592.5	477000	Unaccompanied	Commercial associate	Secondary / secondary speci	
33	100036	0	Cash loans	F	N	Y	0	112500	512064	25033.5	360000	Family	Working	Secondary / secondary speci	
34	100037	0	Cash loans	F	N	N	0	90000	190908	26999.5	150000	Unaccompanied	Working	Secondary / secondary speci	
35	100039	0	Cash loans	M	Y	N	1	360000	733315.5	39069	675000	Unaccompanied	Commercial associate	Secondary / secondary speci	
36	100040	0	Cash loans	F	N	Y	0	135000	1125000	32895	1125000	Unaccompanied	State servant	Higher education	
37	100041	0	Cash loans	F	N	N	0	112500	450000	44509.5	450000	Unaccompanied	Working	Higher education	

TASK B:

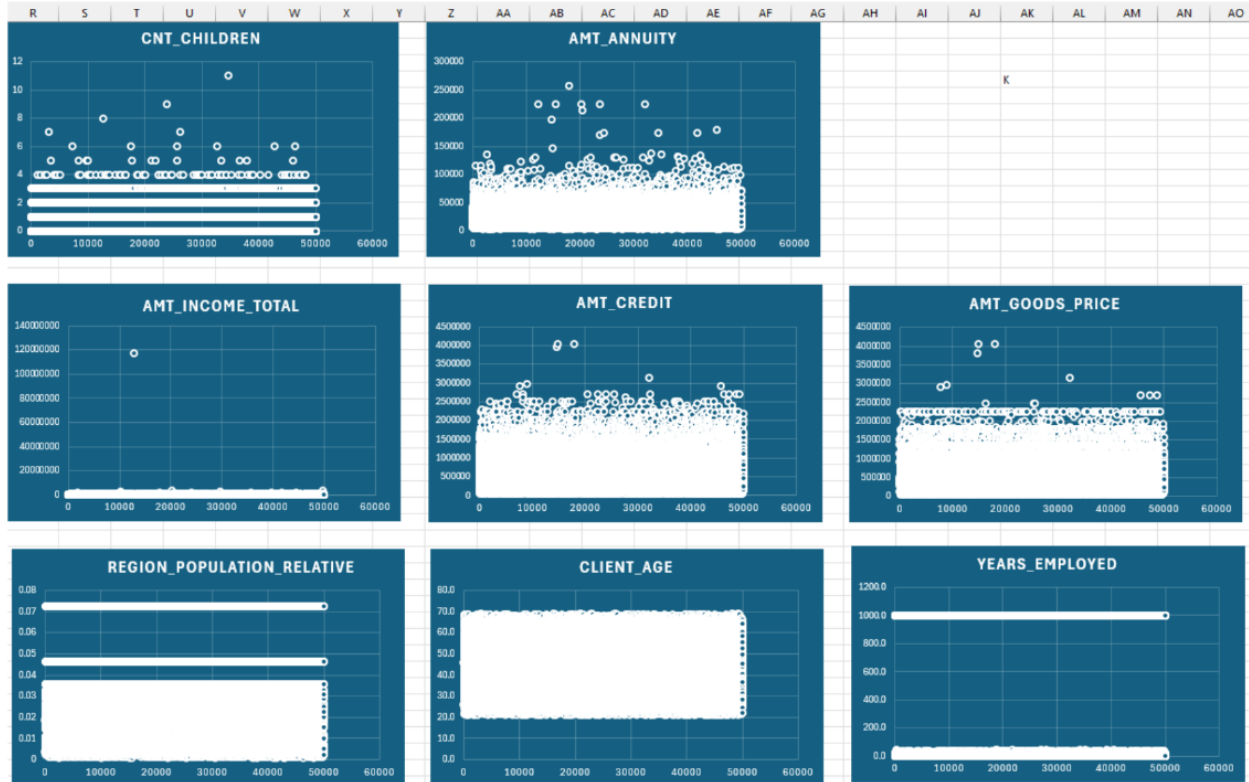
OUTLIER DETECTION: Identify outliers in the dataset which can affect the analysis.

Outliers were detected using the IQR method. Validity assessments ensured that the outliers did not affect our analysis.

FINDING QUARTILE 1, QUARTILE 3, IQR and others

K	L	M	N	O	P
COLUMN NAME	Q1	Q3	IQR	UPPER LIMIT	LOWER LIMIT
CNT_CHILDREN	0	1	1	2.5	-1.5
AMT_INCOME_TOTAL	112500	202500	90000	337500	-22500
AMT_CREDIT	270000	808650	538650	1616625	-537975
AMT_ANNUITY	16456.5	34596	18139.5	61805.25	-10752.75
AMT_GOODS_PRICE	238500	679500	441000	1341000	-423000
REGION_POPULATION_RELATIVE	0.010006	0.028663	0.018657	0.0566485	-0.0179795
client_age	33.91233	53.81918	19.90685	83.67945205	4.052054795
Years_employed	2.556164	15.66575	13.10959	35.33013699	-17.10821918
Years_Registration	5.473973	20.44795	14.97397	42.90890411	-16.9869863

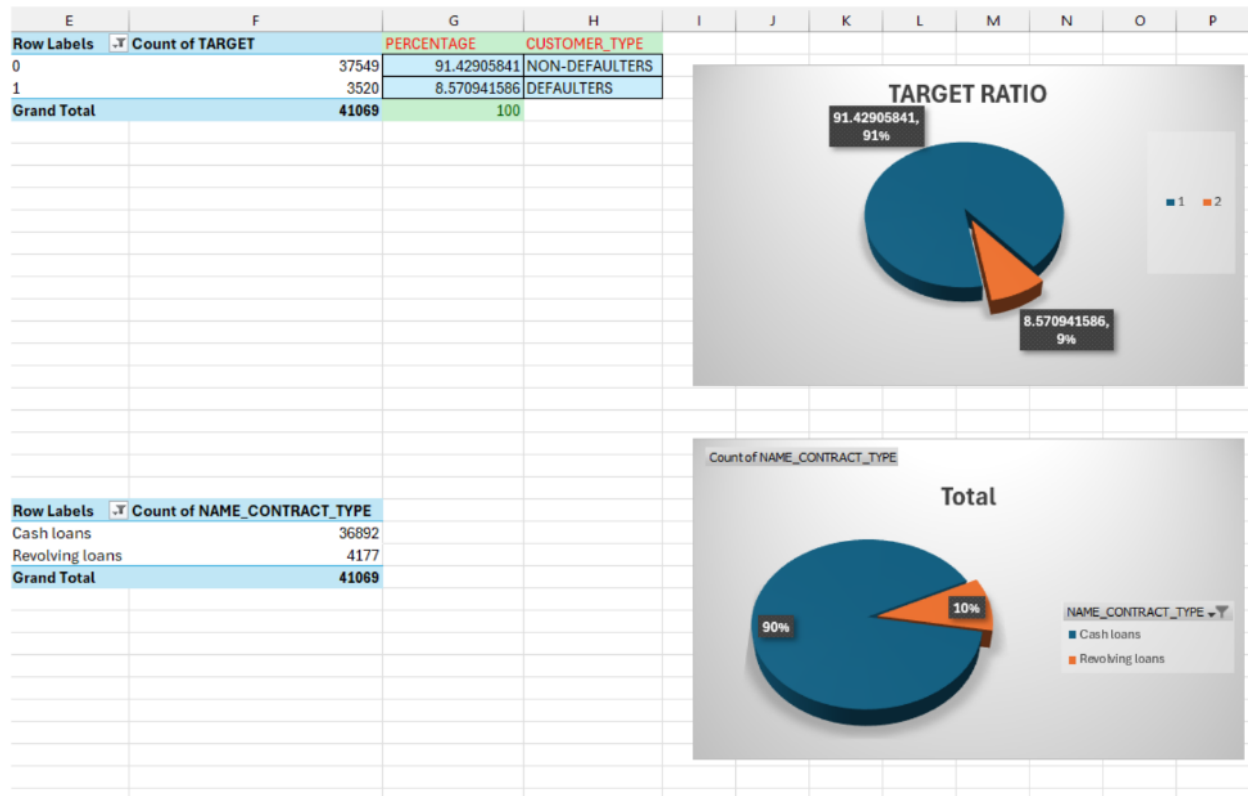
OUTLIER SCATER PLOT



TASK C: Analyse data imbalance

Data imbalance can affect the analysis to a lot extent, therefore it needs to be handled carefully. The screenshot of the tasks are shown below:

RATIO OF DEFAULTERS AND NON DEFAULTERS



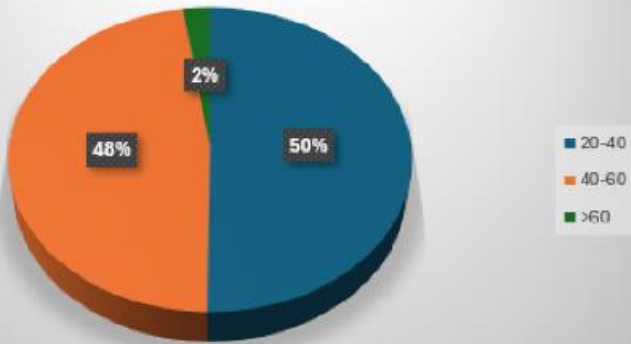
TASK D: Univariate, Segmented Univariate and Bivariate Analysis

Univariate analysis helped identify key attributes such as income levels and credit history, which emerged as significant indicators of loan defaults. Segmented univariate analysis allowed for comparisons across different customer scenarios, revealing patterns that correlate with default likelihood.

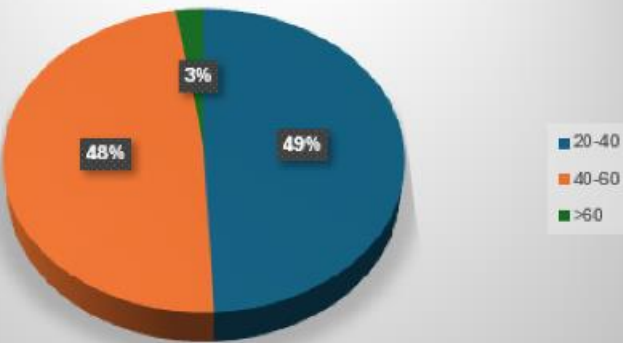
UNIVARIATE ANALYSIS

CLIENT_AGE_RANGE	LOAN_TAKEN	DEFAULTER	NON-DEFAULTER
20-40	20616	2114	18502
40-60	19472	1359	18113
>60	981	47	934

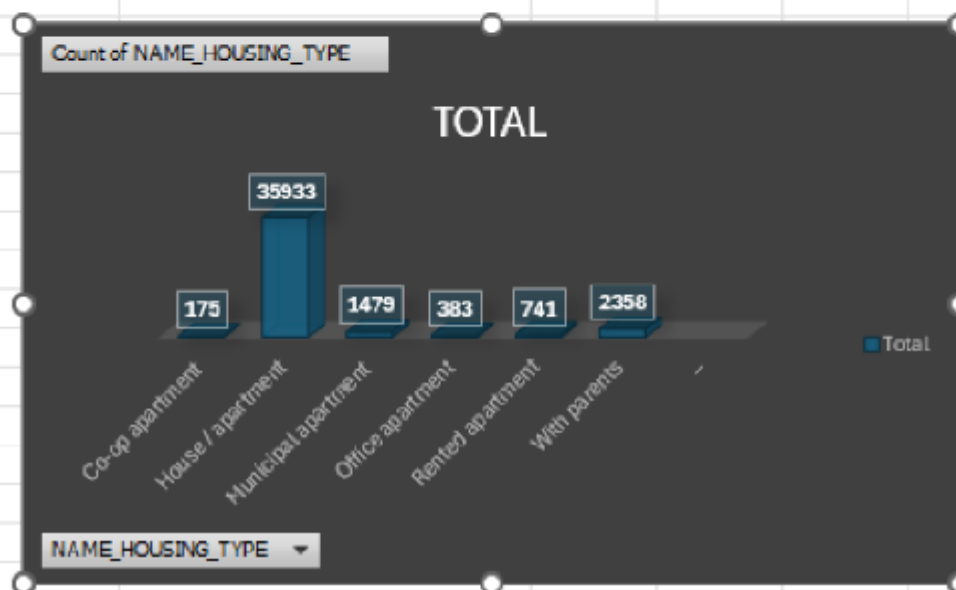
LOAN_TAKEN BY AGE_RANGE



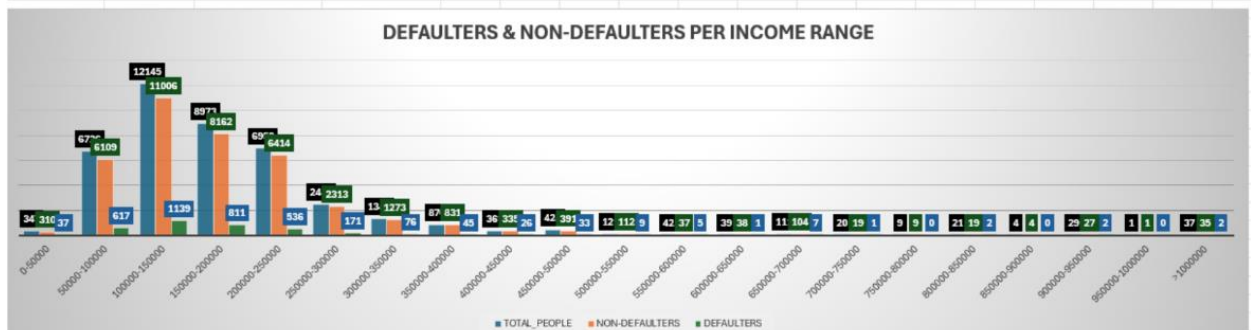
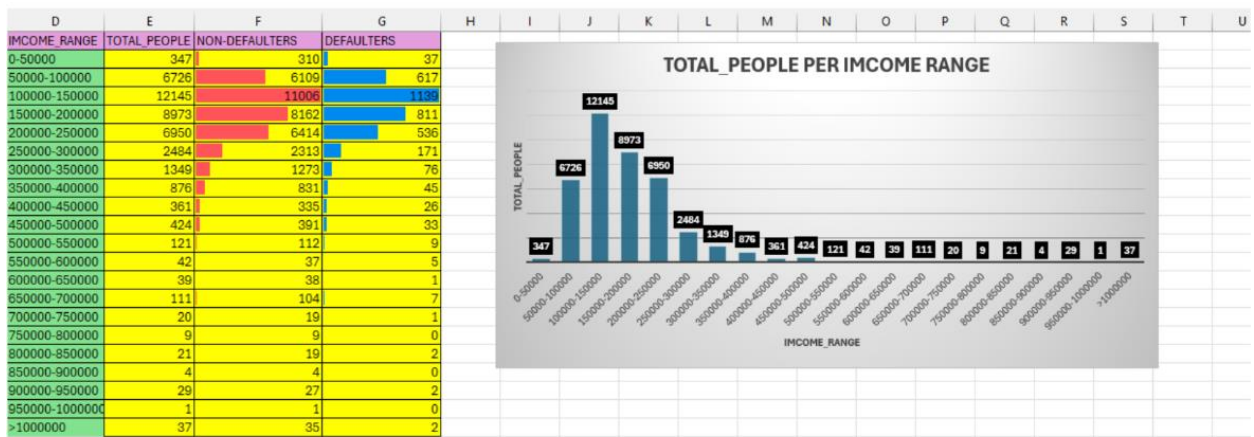
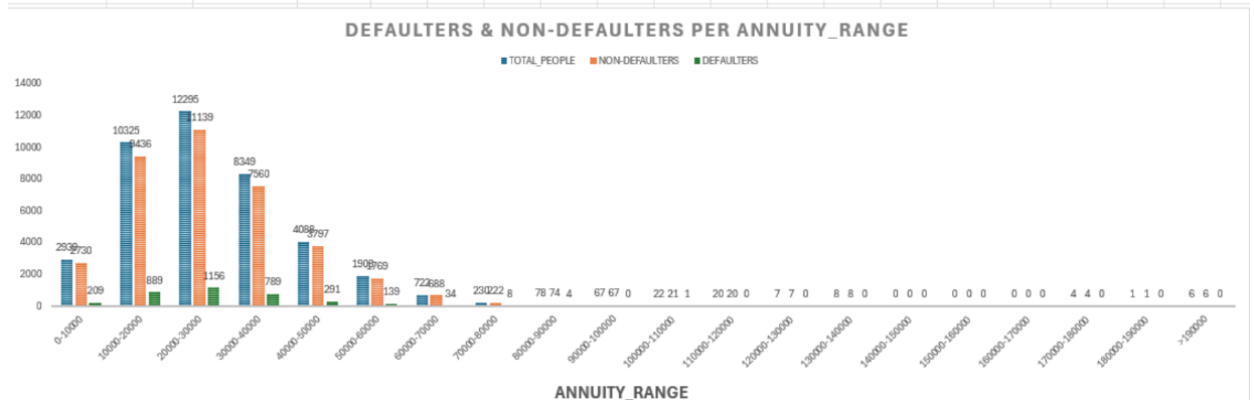
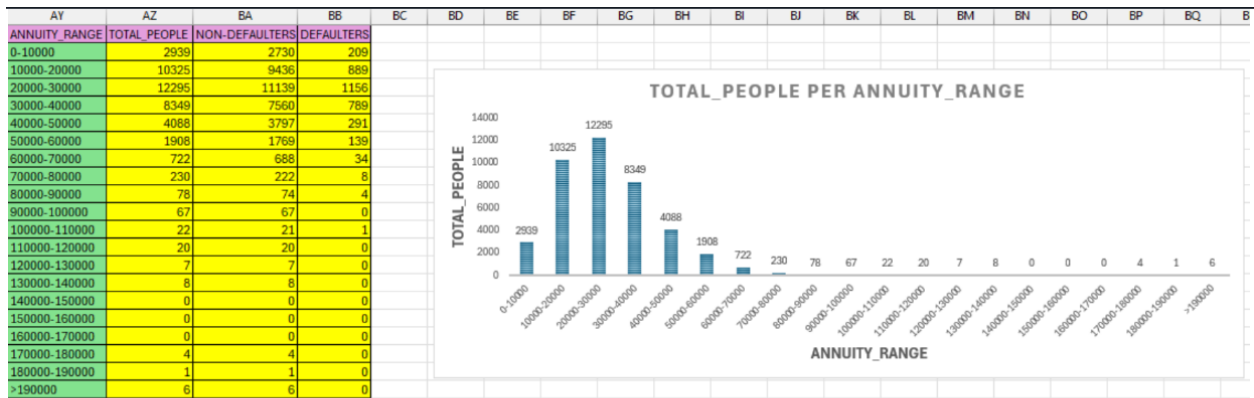
NON-DEFAULTER BY AGE_RANGE

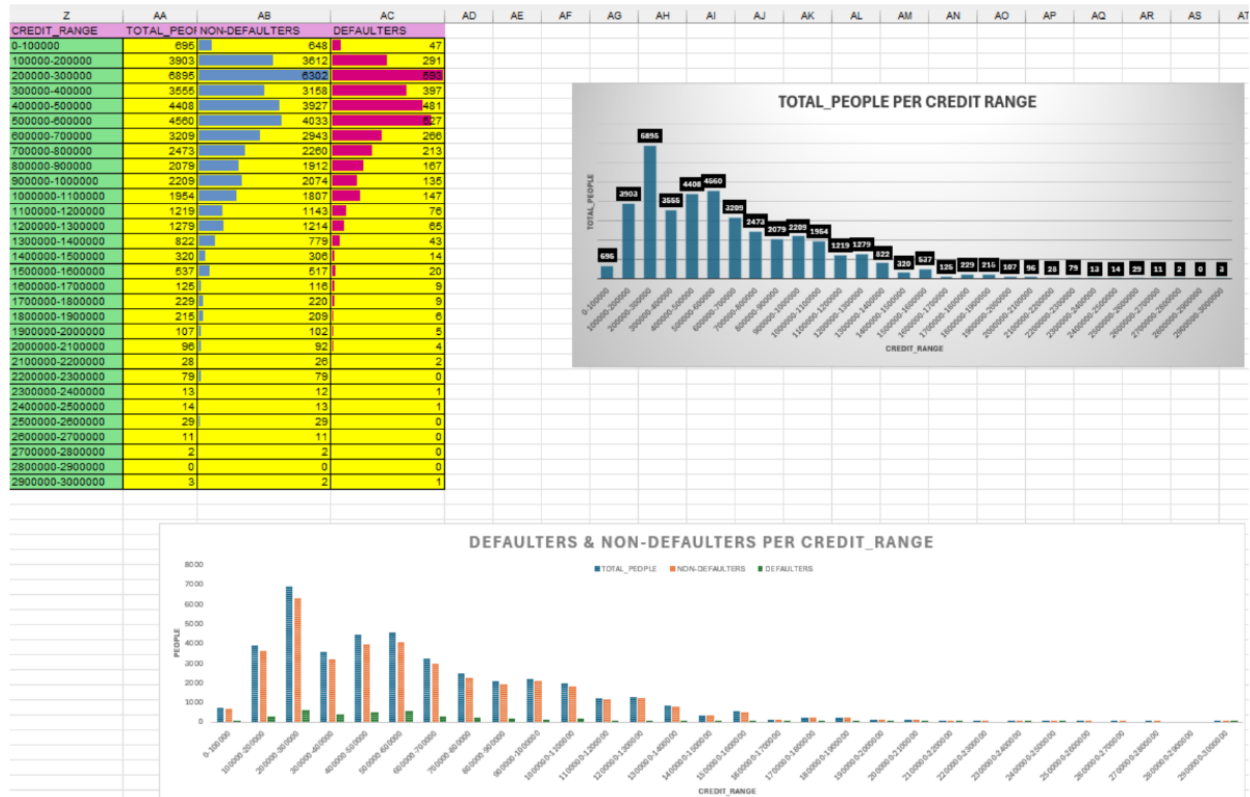


AB	AC	AD	AE	AF	AG
Row Labels	Count of NAME_HOUSING_TYPE				
Co-op apartment	175				
House / apartment	35933				
Municipal apartment	1479				
Office apartment	383				
Rented apartment	741				
With parents	2358				
--					
Grand Total	41069				



SEGMENTED UNIVARIATE ANALYSIS





TASK E: CORRELATION ANALYSIS

Identifying the correlation between variables and target variable can prove insights into strong indicators of loan default. We used the CORREL function here to get the analysis done. The screenshot of the analysis is shown below:

	A	B	C	D	E	F	G	H	I
1	Top correlation of Non-Defaulters					Top Correlation of Defaulters			
2	Rank	Variable 1	Variable 2	Correlation		Rank	Variable 1	Variable 2	Correlation
3	1	AMT_GOODS_PRICE	AMT_CREDIT	0.98635817		1	AMT_GOODS_PRICE	AMT_CREDIT	0.981928143
4	2	REGION_RATING_CLIENT_W_CITY	REGION_RATING_CLIENT	0.950286525		2	REGION_RATING_CLIENT	REGION_RATING_CLIENT_W_CITY	0.948020808
5	3	CNT_CHILDREN	CNT_FAM_MEMBERS	0.893735596		3	CNT_FAM_MEMBERS	CNT_CHILDREN	0.895600339
6	4	REG_REGION_NOT_WORK_REGION	LIVE_REGION_NOT_WORK_REGION	0.860167703		4	DEF_60_CNT_SOCIAL_CIRCLE	DEF_30_CNT_SOCIAL_CIRCLE	0.891467244
7	5	DEF_30_CNT_SOCIAL_CIRCLE	DEF_60_CNT_SOCIAL_CIRCLE	0.853040752		5	LIVE_REGION_NOT_WORK_REGION	REG_REGION_NOT_WORK_REGION	0.805583225
8	6	REG_CITY_NOT_WORK_CITY	LIVE_CITY_NOT_WORK_CITY	0.815604978		6	LIVE_CITY_NOT_WORK_CITY	REG_CITY_NOT_WORK_CITY	0.773107352
9	7	REGION_RATING_CLIENT	AMT_GOODS_PRICE	0.765201743		7	AMT_ANNUITY	AMT_GOODS_PRICE	0.746422447
10	8	AMT_ANNUITY	AMT_GOODS_PRICE	0.765201743		8	AMT_ANNUITY	AMT_CREDIT	0.745132112
11	9	AMT_CREDIT	AMT_ANNUITY	0.760827873					

RESULTS:

This project enhanced my learning about the factors contributing to loan defaults. The analysis done for the tasks assigned helps in identifying high risk applicants, adjust loan offerings and set interest rates which will strengthen our company's financial status and performance. I now feel confident to apply these data analytics skill to real world problems.

DRIVE LINK:

<https://drive.google.com/drive/folders/1LMCb93VETA-2UxNcJDJOfb2X5NZpmJ-J?usp=sharing>