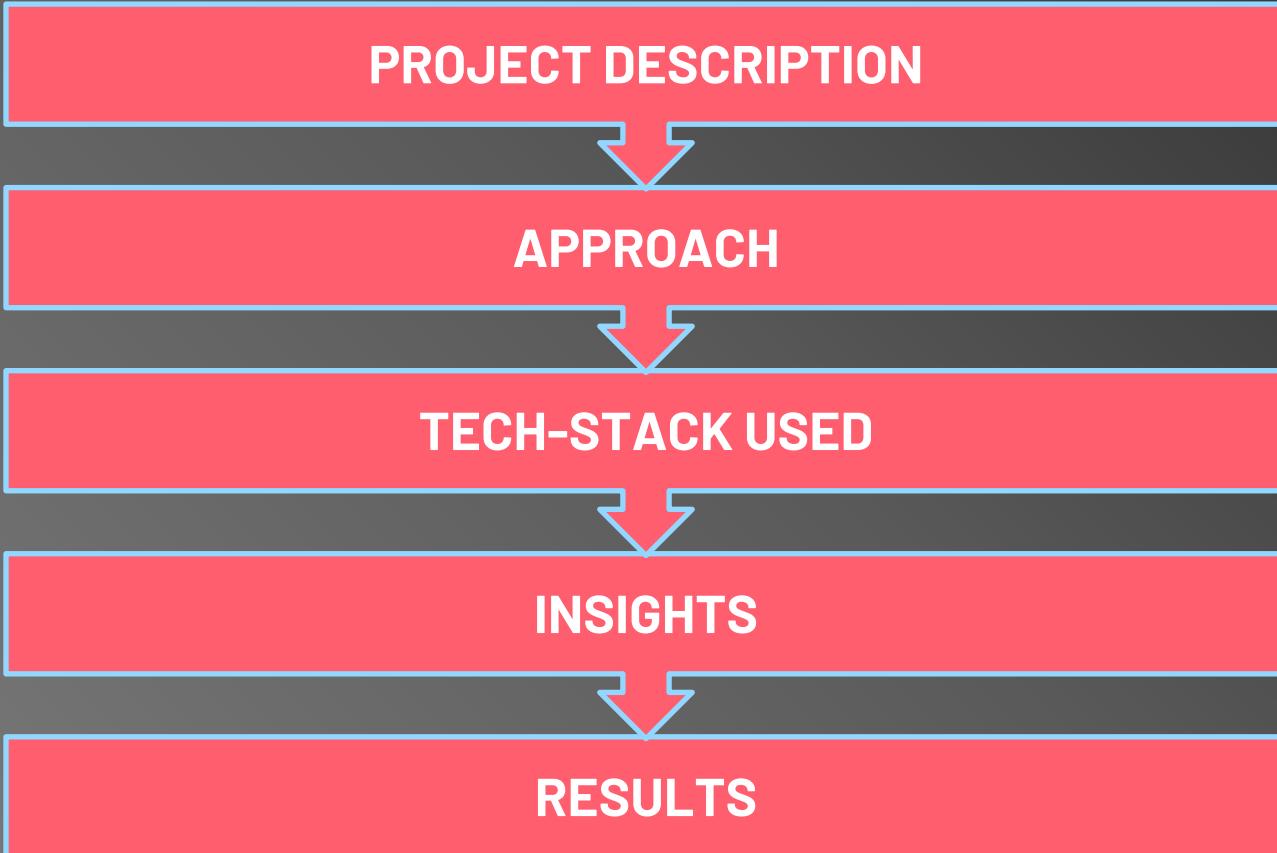




# Bank Loan Case Study

Made by- Anurag John Phillips

# ROADMAP



# Project Description

In this project, we're diving into the world of "loan defaults" using Exploratory Data Analysis (EDA).

Our main goal is to give loans to good applicants and not turn them away.

A finance company has two big risks:

- They might say "no" to someone who actually deserves the loan, and that means losing business.
- They could give loan to someone who might not be able to pay it back, and that's a financial risk.

The data we're working with has two different situations:

- People who are having a hard time making their payments. This means they're paying late more often, especially in the beginning.
- People who are making their payments on time.

The objective of this project is to use EDA to comprehend how customer characteristics and loan-related factors impact the probability of loan default.



# Approach

## Understanding the Dataset

application\_data.csv

## Imputing Data

Using Median/ Mode

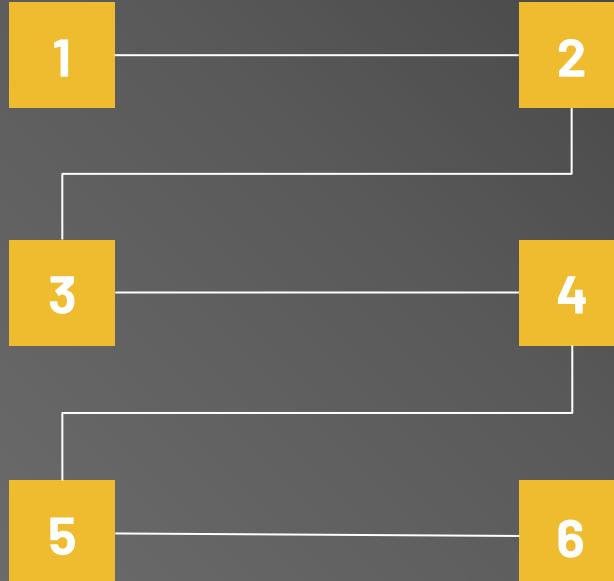
## Visualizing Dataset

## Cleaning Dataset

Removing Duplicates and Columns, Rows which are not needed

## Analysing Dataset

## Gathering Insights



# Working Files

[https://drive.google.com/drive/folders/1r6bM200g8DAag94qym5\\_\\_Dhdz8aVdh0\\_?usp=drive\\_link](https://drive.google.com/drive/folders/1r6bM200g8DAag94qym5__Dhdz8aVdh0_?usp=drive_link)





# Tech-Stack Used

Microsoft Excel for Mac Version 16.74

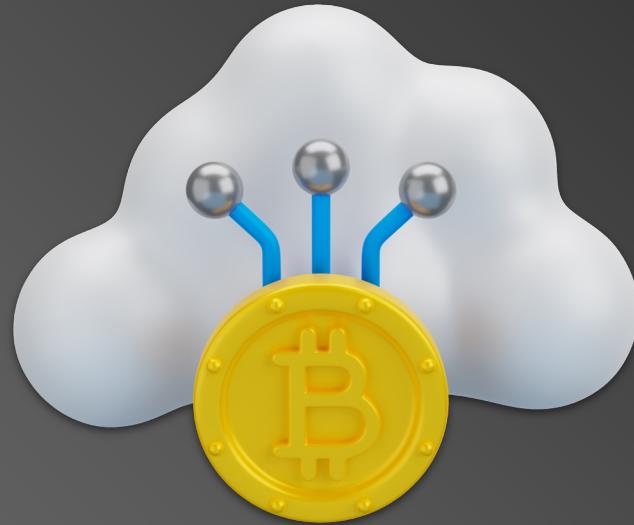
The screenshot shows a Microsoft Excel window titled "Book1 - Excel". The ribbon menu includes File, Home, Insert, Page Layout, Formulas, Data, Review, View, Tell me..., Javier Flores, and Share. The Home tab is selected, displaying tools for Clipboard, Font, Alignment, and Number. The worksheet area has rows labeled 1 through 15 and columns labeled A through K. Cell A1 is selected. A yellow emoji of a money bag with a white "B" on it is overlaid on the sheet. The status bar at the bottom right shows "Ready", "Sheet3", and "100%".



01

# Identify Missing Data and Deal with it appropriately

Identify the missing data in the dataset and decide on an appropriate method to deal with it.



# Original Dataset

SK_ID_CURR	TARGET	NAME_CONTRACT_TYP	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUIT	AMT_GOODS_PRICE	NAME_TYPE_SUITE	NAME_INCOME_TYP	NAME_EDUCATION_TYP	NAME_FAMILY_STATUS	...
100002	1	Cash loans	M	N	Y	0	202500	406597.5	24700.5	351000	Unaccompanied	Working	Secondary / secondary speci	Single / not married	
100003	0	Cash loans	F	N	N	0	270000	1293502.5	35480.5	1129500	Family	State servant	Higher education	Married	
100004	0	Revolving loans	M	Y	Y	0	67000	135000	6700.0	150000	Unaccompanied	Working	Secondary / secondary speci	Single / not married	
100005	0	Cash loans	F	N	Y	0	330000	318250.5	29684.5	297000	Unaccompanied	Working	Secondary / secondary speci	Civil marriage	
100007	0	Cash loans	M	N	Y	0	121500	513000	21865.5	513000	Unaccompanied	Working	Secondary / secondary speci	Single / not married	
100008	0	Cash loans	M	N	Y	0	99000	490495.5	27517.5	454500	Spouse+partner	State servant	Secondary / secondary speci	Married	
100009	0	Cash loans	F	Y	Y	1	171000	1560726	41301	139500	Unaccompanied	Commercial associate	Higher education	Married	
100010	0	Cash loans	M	Y	Y	0	360000	1520000	42075	1520000	Unaccompanied	State servant	Higher education	Married	
100011	0	Cash loans	F	N	Y	0	112500	1019610	33826.5	913500	Children	Pensioner	Secondary / secondary speci	Married	
100012	0	Revolving loans	M	N	Y	0	135000	405000	20250	405000	Unaccompanied	Working	Secondary / secondary speci	Single / not married	
100014	0	Cash loans	F	N	Y	1	112500	652500	21177	652500	Unaccompanied	Working	Higher education	Married	
100015	0	Cash loans	F	N	Y	0	38419.155	148365	10678.5	135000	Children	Pensioner	Secondary / secondary speci	Married	
100016	0	Cash loans	F	N	Y	0	67500	80865	5881.5	67500	Unaccompanied	Working	Secondary / secondary speci	Married	
100017	0	Cash loans	M	Y	N	1	225000	918468	28966.5	697500	Unaccompanied	Working	Secondary / secondary speci	Married	
100018	0	Cash loans	F	N	Y	0	189000	773680.5	32778	679500	Unaccompanied	Working	Secondary / secondary speci	Married	
100019	0	Cash loans	M	Y	Y	0	157500	299772	20160	247500	Family	Working	Secondary / secondary speci	Single / not married	
100020	0	Cash loans	M	N	N	0	108000	509602.5	26149.5	387000	Unaccompanied	Working	Secondary / secondary speci	Married	
100021	0	Revolving loans	F	N	Y	1	81000	270000	13500	270000	Unaccompanied	Working	Secondary / secondary speci	Married	
100022	0	Revolving loans	F	N	Y	0	112500	157500	7875	157500	Other_A	Working	Secondary / secondary speci	Widow	
100023	0	Cash loans	F	N	Y	1	90000	544491	17563.5	454500	Unaccompanied	State servant	Higher education	Single / not married	
100024	0	Revolving loans	M	Y	Y	0	135000	427500	21375	427500	Unaccompanied	Working	Secondary / secondary speci	Married	
100025	0	Cash loans	F	Y	Y	1	202500	1132573.5	37561.5	927000	Unaccompanied	Commercial associate	Secondary / secondary speci	Married	
100026	0	Cash loans	F	N	N	1	450000	497520	32521.5	450000	Unaccompanied	Working	Secondary / secondary speci	Married	
100027	0	Cash loans	F	N	Y	0	83250	239850	23850	225000	Unaccompanied	Pensioner	Secondary / secondary speci	Married	
100029	0	Cash loans	M	Y	N	2	135000	125000	12215.5	297000	Unaccompanied	Working	Secondary / secondary speci	Married	
100030	0	Cash loans	F	N	Y	0	90000	232000	11074.5	225000	Unaccompanied	Working	Secondary / secondary speci	Married	
100031	1	Cash loans	F	N	Y	0	112500	799992	27076.5	702000	Unaccompanied	Working	Secondary / secondary speci	Widow	
100032	0	Cash loans	M	N	Y	1	112500	327024	23827.5	270000	Family	Working	Secondary / secondary speci	Married	
100033	0	Cash loans	M	Y	Y	0	270000	799830	57676.5	675000	Unaccompanied	State servant	Higher education	Single / not married	
100034	0	Revolving loans	M	N	Y	0	90000	180000	9000	180000	Unaccompanied	Working	Higher education	Single / not married	
100035	0	Cash loans	F	N	Y	0	295000	665892	24592.5	477000	Unaccompanied	Commercial associate	Secondary / secondary speci	Civil marriage	
100036	0	Cash loans	F	N	Y	0	112500	512064	25033.5	360000	Family	Working	Secondary / secondary speci	Civil marriage	
100037	0	Cash loans	F	N	N	0	90000	199008	20893.5	180000	Unaccompanied	Working	Secondary / secondary speci	Civil marriage	
100039	0	Cash loans	M	Y	N	1	360000	733315.5	39069	679500	Unaccompanied	Commercial associate	Secondary / secondary speci	Married	
100040	0	Cash loans	F	N	Y	0	135000	1125000	32895	1125000	Unaccompanied	State servant	Higher education	Married	
100041	0	Cash loans	F	N	N	0	112500	450000	44509.5	450000	Unaccompanied	Working	Higher education	Married	
100043	0	Cash loans	F	N	Y	2	198000	641173.5	23157	553500	Unaccompanied	Commercial associate	Secondary / secondary speci	Married	
100044	0	Cash loans	M	N	Y	0	121500	454500	15151.5	454500	Unaccompanied	Working	Secondary / secondary speci	Married	
100045	0	Cash loans	F	N	Y	0	99000	247275	17383.5	225000	Unaccompanied	Pensioner	Secondary / secondary speci	Married	
100046	0	Revolving loans	M	Y	Y	0	180000	540000	27000	540000	Unaccompanied	Working	Higher education	Married	
100047	1	Cash loans	M	N	Y	0	202500	1193580	35028	855000	Unaccompanied	Commercial associate	Secondary / secondary speci	Married	
100048	0	Cash loans	F	N	Y	0	202500	604152	29196	540000	Unaccompanied	Working	Secondary / secondary speci	Married	
100049	1	Cash loans	F	N	N	0	135000	288873	16258.5	238500	Unaccompanied	Working	Secondary / secondary speci	Civil marriage	

Before Cleaning- 122 Columns, some columns have 50,003 rows while others have 307478 Rows

For my analysis, to make the data uniform across all the columns I took first 50,003 rows and deleted the other rows, then made it into a table and then removed Duplicates which in under **Table Tab**

## Calculated Percentage of Missing Values in each Column

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	SK_ID_PREV
0	0	0	0	0	0	0	0	0	0	0	38	192	0
49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999
100002	1	Cash loans	M	N	Y	0	202500	40557.95	24700.5	35100	Unaccompanied	Working	100002
100000	0	Cash loans	F	N	Y	0	270000	129350.2	35698.5	1129500	Family	State servant	100000
100004	0	Revolving loans	M	M	Y	0	67500	135000	6785	135000	Unaccompanied	Working	100004
100006	0	Cash loans	M	N	Y	0	135000	135000	26050.5	135000	Unaccompanied	Working	100006
100007	0	Cash loans	M	N	Y	0	211500	115000	23816.5	115000	Unaccompanied	Working	100007
100008	0	Cash loans	M	N	Y	0	99000	490495.5	27517.5	454500	Spoiled partner	State servant	100008
100009	0	Cash loans	F	Y	Y	1	171000	1560726	41301	139500	Unaccompanied	Commercial associate	100009
100010	0	Cash loans	M	Y	Y	0	360000	153000	42075	153000	Unaccompanied	State servant	100010
100011	0	Cash loans	F	N	Y	0	112500	180100	33855	913500	Unaccompanied	Pensioner	100011
100012	0	Revolving loans	M	N	Y	0	112500	65000	20250	40550	Unaccompanied	Working	100012
100014	0	Cash loans	F	N	Y	1	112500	652500	21177	652500	Unaccompanied	Working	100014
100015	0	Cash loans	F	N	Y	0	38419.155	148365	10678.5	13500	Children	Pensioner	100015
100016	0	Cash loans	M	N	Y	0	67500	80865	5881.5	67500	Unaccompanied	Working	100016
100017	0	Cash loans	M	Y	Y	1	180000	118000	28685.5	60750	Unaccompanied	Working	100017
100018	0	Cash loans	F	N	Y	0	180000	77360.5	32778	67500	Unaccompanied	Working	100018
100019	0	Cash loans	M	Y	Y	0	157500	299772	2014	247500	Family	Working	100019
100020	0	Cash loans	M	N	N	0	108000	50960.5	26149.5	387000	Unaccompanied	Working	100020
100021	0	Revolving loans	F	N	Y	1	81000	27000	1396	27000	Unaccompanied	Working	100021
100022	0	Revolving loans	F	N	Y	0	108000	40000	7675	108000	Other_A.	Working	100022
100023	0	Cash loans	F	N	Y	1	90000	544491	17583.5	454500	Unaccompanied	State servant	100023
100024	0	Revolving loans	M	Y	Y	0	135000	427500	21375	427500	Unaccompanied	Working	100024
100025	0	Cash loans	F	Y	Y	1	202500	112355.5	37561.5	97200	Unaccompanied	Commercial associate	100025
100026	0	Cash loans	F	N	N	1	45000	49750	32521.5	45000	Unaccompanied	Working	100026
100027	0	Cash loans	M	N	Y	0	120000	25000	25800	25000	Unaccompanied	Working	100027
100029	0	Cash loans	M	Y	N	2	135000	247500	12703.5	147500	Unaccompanied	Working	100029
100030	0	Cash loans	F	N	Y	0	90000	252000	11074.5	225000	Unaccompanied	Working	100030
100031	1	Cash loans	F	N	Y	0	112500	97992	27076.5	7076.5	Unaccompanied	Working	100031
100032	0	Cash loans	M	N	Y	1	135000	327024	23827.5	23827.5	Unaccompanied	Working	100032
100033	0	Cash loans	M	Y	Y	0	202500	78500	5751.5	67500	Unaccompanied	State servant	100033
100034	0	Revolving loans	M	N	Y	0	90000	180000	9000	180000	Unaccompanied	Working	100034
100035	0	Cash loans	F	N	Y	0	202500	655892	42592.5	427000	Unaccompanied	Commercial associate	100035

**Yellow Row- Count of  
number of missing  
values**

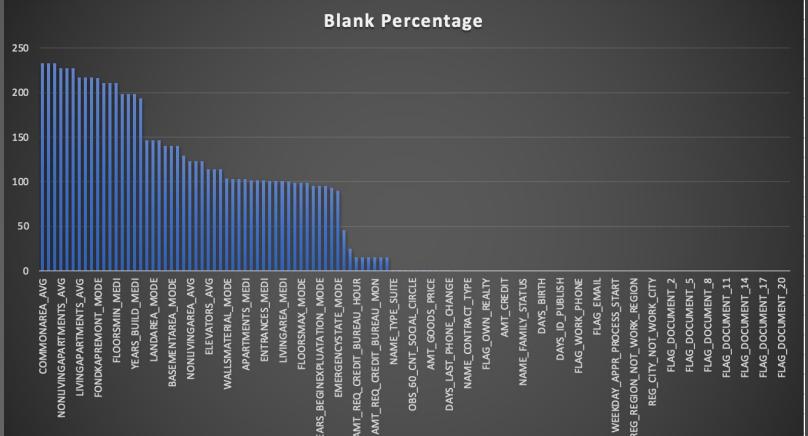
=COUNTBLANK(A5:A500)  
D3)

**Red Row- Count the number of cells that have value in it**

=COUNTA(A5:A50003)

**Green Row- Calculated**  
the percentage of  
missing values

=COUNTBLANK(A:A) /  
COUNTA(A:A) \* 100



# Before Imputation

	H	I	J	K	L	M	N	O	P	Q
1	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	NAME_TYPE_SUITE	NAME_INCOME_TYPE	NAME_EDUCATION_TYPE	NAME_FAMILY_STATUS	NAME_HOUSING_TYPE	REGION_POPULATION_RELATIVE
2	0	0	1	38	192	0	0	0	0	0
3	49999	49999	49998	49961	49807	49999	49999	49999	49999	49999
4	0	0	0.0020008	0.076059326	0.385487984	0	0	0	0	0
5	202500	406597.5	24700.5	351000	Unaccompanied	Working	Secondary / secondary specia	Single / not married	House / apartment	0.018801
6	270000	1293502.5	35698.5	1129500	Family	State servant	Higher education	Married	House / apartment	0.003541
7	67500	135000	6750	135000	Unaccompanied	Working	Secondary / secondary specia	Single / not married	House / apartment	0.010032
8	135000	312682.5	29686.5	297000	Unaccompanied	Working	Secondary / secondary specia	Civil marriage	House / apartment	0.008019
9	121500	513000	21865.5	513000	Unaccompanied	Working	Secondary / secondary specia	Single / not married	House / apartment	0.028663
10	99000	490495.5	27517.5	454500	Spouse, partner	State servant	Secondary / secondary specia	Married	House / apartment	0.035792
11	171000	1560726	41301	1395000	Unaccompanied	Commercial associate	High education	Married	House / apartment	0.035792
12	360000	1530000	42075	1530000	Unaccompanied	State servant	Higher education	Married	House / apartment	0.003122
13	112500	1019610	33826.5	913500	Children	Pensioner	Secondary / secondary specia	Married	House / apartment	0.018634
14	135000	405000	20250	405000	Unaccompanied	Working	Secondary / secondary specia	Single / not married	House / apartment	0.019689
15	112500	652500	21177	652500	Unaccompanied	Working	Higher education	Married	House / apartment	0.0228
16	38419.155	148365	10678.5	135000	Children	Pensioner	Secondary / secondary specia	Married	House / apartment	0.015221
17	67500	80865	5881.5	67500	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.031329
18	225000	918468	28966.5	697500	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.016612
19	189000	773680.5	32778	679500	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.010006
20	157500	299772	20160	247500	Family	Working	Secondary / secondary specia	Single / not married	Rented apartment	0.020713
21	108000	509602.5	26149.5	387000	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.018634
22	81000	270000	13500	270000	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.010966
23	112500	157500	7875	157500	Other_A	Working	Secondary / secondary specia	Widow	House / apartment	0.04622
24	90000	544491	17563.5	454500	Unaccompanied	State servant	Higher education	Single / not married	House / apartment	0.015221
25	135000	427500	21375	427500	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.015221
26	202500	1132573.5	37561.5	927000	Unaccompanied	Commercial associate	Secondary / secondary specia	Married	House / apartment	0.025164
27	450000	497520	32521.5	450000	Unaccompanied	Working	Secondary / secondary specia	Married	Rented apartment	0.020713
28	83250	239850	23850	225000	Unaccompanied	Pensioner	Secondary / secondary specia	Married	House / apartment	0.006296
29	135000	247500	12703.5	247500	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.026392
30	90000	225000	11074.5	225000	Unaccompanied	Working	Secondary / secondary specia	Married	House / apartment	0.028663
31	112500	979992	27076.5	702000	Unaccompanied	Working	Secondary / secondary specia	Widow	House / apartment	0.018029
32	112500	327024	23827.5	270000	Family	Working	Secondary / secondary specia	Married	House / apartment	0.019101
33	270000	790830	57676.5	675000	Unaccompanied	State servant	Higher education	Single / not married	House / apartment	0.04622
34	90000	180000	9000	180000	Unaccompanied	Working	Higher education	Single / not married	With parent	0.030755
35	292500	665892	24592.5	477000	Unaccompanied	Commercial associate	Secondary / secondary specia	Civil marriage	House / apartment	0.025164

Numerical Columns- Median  
 Text Columns - Mode

# After Imputation

SK_ID_CURR	TARGET	NAME_CTRY	CODE_GEO	FLAG_OVW	FLAG_OVW	CNT_CHIL	AMT_INC	AMT_CRE	AMT_ANI	AMT_GOV	NAME_TYPE_SUITE	NAME_INM	NAME_EIN	NAME_FIN	NAME_HIN	REGION	DAYS_BII	DAYS_ENI	DAYS_REV
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	49999	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
100002	1	Cash loans	M	N	Y		0	202500	406597.5	24700.5	351000 Unaccompanied	Working	Secondary / Single / not i House / apar	0.018801	-9461	-637	-3648		
100003	0	Cash loans	F	N	N		0	270000	1293502.5	35698.5	1129500 Family	State servan	Higher educ Married	House / apar	0.003541	-16765	-1188	-1186	
100004	0	Revolving lo	M	Y	Y		0	67500	135000	6750	135000 Unaccompanied	Working	Secondary / Single / not i House / apar	0.010032	-19046	-225	-4260		
100006	0	Cash loans	F	N	Y		0	135000	312682.5	29686.5	297000 Unaccompanied	Working	Secondary / Civil marria	House / apar	0.008019	-19005	-3039	-9833	
100007	0	Cash loans	M	N	Y		0	121500	513000	21865.5	513000 Unaccompanied	Working	Secondary / Single / not i House / apar	0.028663	-19932	-3038	-4311		
100008	0	Cash loans	M	N	Y		0	99000	490495.5	27517.5	454500 Spouse, partner	State servan	Secondary / Married	House / apar	0.035792	-16941	-1588	-4970	
100009	0	Cash loans	F	Y	Y		1	171000	1560726	41301	1395000 Unaccompanied	Commercial	Higher educ Married	House / apar	0.035792	-13778	-3130	-1213	
100010	0	Cash loans	M	Y	Y		0	360000	1530000	42075	1530000 Unaccompanied	State servan	Higher educ Married	House / apar	0.003122	-18850	-449	-4597	
100011	0	Cash loans	F	N	Y		0	112500	1019610	33826.5	913500 Children	Pensioner	Secondary / Married	House / apar	0.018634	-20099	365243	-7427	
100012	0	Revolving lo	M	N	Y		0	135000	405000	20250	405000 Unaccompanied	Working	Secondary / Single / not i House / apar	0.019689	-14469	-2019	-14437		
100014	0	Cash loans	F	N	Y		1	112500	652500	21177	652500 Unaccompanied	Working	Higher educ Married	House / apar	0.0228	-10197	-679	-4427	
100015	0	Cash loans	F	N	Y		0	38419.155	148365	10678.5	135000 Children	Pensioner	Secondary / Married	House / apar	0.015221	-20417	365243	-5246	
100016	0	Cash loans	F	N	Y		0	67500	80865	5881.5	67500 Unaccompanied	Working	Secondary / Married	House / apar	0.031329	-13439	-2717	-311	
100017	0	Cash loans	M	Y	N		1	225000	918468	28966.5	697500 Unaccompanied	Working	Secondary / Married	House / apar	0.016612	-14086	-3028	-643	
100018	0	Cash loans	F	N	Y		0	189000	773680.5	32778	679500 Unaccompanied	Working	Secondary / Married	House / apar	0.010006	-14583	-203	-615	
100019	0	Cash loans	M	Y	Y		0	157500	299772	20160	247500 Family	Working	Secondary / Single / not i Rented apart	0.020713	-8728	-1157	-3494		
100020	0	Cash loans	M	N	N		0	108000	509602.5	26149.5	387000 Unaccompanied	Working	Secondary / Married	House / apar	0.018634	-12931	-1317	-6392	
100021	0	Revolving lo	F	N	Y		1	81000	270000	13500	270000 Unaccompanied	Working	Secondary / Married	House / apar	0.010966	-9776	-191	-4143	
100022	0	Revolving lo	F	N	Y		0	112500	157500	7875	157500 Other_A	Working	Secondary / Widow	House / apar	0.04622	-17718	-7804	-8751	
100023	0	Cash loans	F	N	Y		1	90000	544491	17563.5	454500 Unaccompanied	State servan	Higher educ Single / not i House / apar	0.015221	-11348	-2038	-1021		
100024	0	Revolving lo	M	Y	Y		0	135000	427500	21375	427500 Unaccompanied	Working	Secondary / Married	House / apar	0.015221	-18252	-4286	-298	
100025	0	Cash loans	F	Y	Y		1	202500	1132573.5	37561.5	927000 Unaccompanied	Commercial	Secondary / Married	House / apar	0.025164	-14815	-1652	-2299	
100026	0	Cash loans	F	N	N		1	450000	497520	32521.5	450000 Unaccompanied	Working	Secondary / Married	Rented apart	0.020713	-11146	-4306	-114	
100027	0	Cash loans	F	N	Y		0	83250	239850	23850	225000 Unaccompanied	Pensioner	Secondary / Married	House / apar	0.006296	-24827	365243	-9012	
100029	0	Cash loans	M	Y	N		2	135000	247500	12703.5	247500 Unaccompanied	Working	Secondary / Married	House / apar	0.026392	-11286	-746	-108	
100030	0	Cash loans	F	N	Y		0	90000	225000	11074.5	225000 Unaccompanied	Working	Secondary / Married	House / apar	0.028663	-19334	-3494	-2419	
100031	1	Cash loans	F	N	Y		0	112500	979992	27076.5	702000 Unaccompanied	Working	Secondary / Widow	House / apar	0.018029	-18724	-2628	-6573	
100032	0	Cash loans	M	N	Y		1	112500	327024	23827.5	270000 Family	Working	Secondary / Married	House / apar	0.019101	-15948	-1234	-5782	
100033	0	Cash loans	I	Y	Y		0	270000	790830	57676.5	675000 Unaccompanied	State servan	Higher educ Single / not i House / apar	0.04622	-9994	-1796	-4668		
100034	0	Revolving lo	M	N	Y		0	90000	180000	9000	180000 Unaccompanied	Working	Higher educ Single / not i With parents	0.030755	-10341	-1010	-4799		

# Working Datasheet

SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUIT	AMT_GOODS	NAME_TYPE_SUITE
100002	1	Cash loans	M	N	Y	0	202500	406597.5	24700.5	351000	Unaccompanied
100003	0	Cash loans	F	N	N	0	270000	1293502.5	35698.5	1129500	Family
100004	0	Revolving loans	M	Y	Y	0	67500	135000	6750	135000	Unaccompanied
100006	0	Cash loans	F	N	Y	0	135000	312682.5	29686.5	297000	Unaccompanied
100007	0	Cash loans	M	N	Y	0	121500	513000	21865.5	513000	Unaccompanied
100008	0	Cash loans	M	N	Y	0	99000	490495.5	27517.5	454500	Spouse, partner
100009	0	Cash loans	F	Y	Y	1	171000	1560726	41301	1395000	Unaccompanied
100010	0	Cash loans	M	Y	Y	0	360000	1530000	42075	1530000	Unaccompanied
100011	0	Cash loans	F	N	Y	0	112500	1019610	33826.5	913500	Children
100012	0	Revolving loans	M	N	Y	0	135000	405000	20250	405000	Unaccompanied
100014	0	Cash loans	F	N	Y	1	112500	652500	21177	652500	Unaccompanied
100015	0	Cash loans	F	N	Y	0	38419.155	148365	10678.5	135000	Children
100016	0	Cash loans	F	N	Y	0	67500	80865	5881.5	67500	Unaccompanied
100017	0	Cash loans	M	Y	N	1	225000	918468	28966.5	697500	Unaccompanied
100018	0	Cash loans	F	N	Y	0	189000	773680.5	32778	679500	Unaccompanied
100019	0	Cash loans	M	Y	Y	0	157500	299772	20160	247500	Family
100020	0	Cash loans	M	N	N	0	108000	509602.5	26149.5	387000	Unaccompanied
100021	0	Revolving loans	F	N	Y	1	81000	270000	13500	270000	Unaccompanied
100022	0	Revolving loans	F	N	Y	0	112500	157500	7875	157500	Other_A
100023	0	Cash loans	F	N	Y	1	90000	544491	17563.5	454500	Unaccompanied
100024	0	Revolving loans	M	Y	Y	0	135000	427500	21375	427500	Unaccompanied
100025	0	Cash loans	F	Y	Y	1	202500	1132573.5	37561.5	927000	Unaccompanied
100026	0	Cash loans	F	N	N	1	450000	497520	32521.5	450000	Unaccompanied
100027	0	Cash loans	F	N	Y	0	83250	239850	23850	225000	Unaccompanied
100029	0	Cash loans	M	Y	N	2	135000	247500	12703.5	247500	Unaccompanied
100030	0	Cash loans	F	N	Y	0	90000	225000	11074.5	225000	Unaccompanied
100031	1	Cash loans	F	N	Y	0	112500	979992	27076.5	702000	Unaccompanied
100032	0	Cash loans	M	N	Y	1	112500	327024	23827.5	270000	Family
100033	0	Cash loans	M	Y	Y	0	270000	790830	57676.5	675000	Unaccompanied
100034	0	Revolving loans	M	N	Y	0	90000	180000	9000	180000	Unaccompanied
100035	0	Cash loans	F	N	Y	0	292500	665892	24592.5	477000	Unaccompanied
100036	0	Cash loans	F	N	Y	0	112500	512064	25033.5	360000	Family
100037	0	Cash loans	F	N	N	0	90000	199008	20893.5	180000	Unaccompanied
100039	0	Cash loans	M	Y	N	1	360000	733315.5	39059	679500	Unaccompanied

After Cleaning- 73 Columns and 50,000 Rows

★ This completes the first task, here I have identified the missing data in the dataset and used Median and Mode imputation to deal with it.

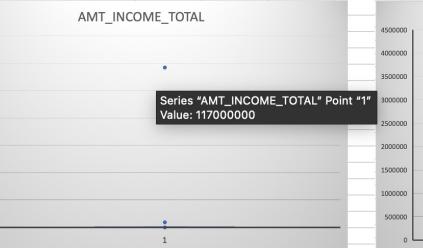
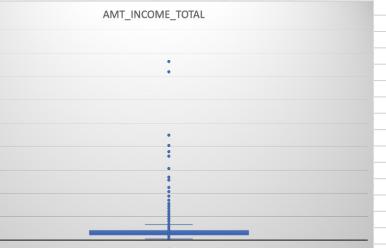
# 02

## Identify Outliers in the Dataset

Detect and identify outliers in the dataset using Excel statistical functions and features, focusing on numerical variables.



# Outlier - Amt\_Income\_Total

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	AMT_INCOME_TOTAL		202500	Quartile 1	112500	=QUARTILE(A2:A50000,1)								
2			270000	Quartile 3	202500	=QUARTILE(A2:A50000,3)								
3			67500											
4			135000											
5			1212500	Inter Quartile Range IQR	90000	=D4-D2								
6			99000											
7			171000	Lower Threshold	-22500	=D2-(1.5*D6)								
8			3600000											
9			112500	Upper Threshold	337500	=D4+(1.5*D6)								
10			135000											
11			112500											
12			38419155											
13			67500	AMT_INCOME_TOTAL				AMT_INCOME_TOTAL						
14			225000											
15			1890000											
16			157500											
17			108000											
18			81000											
19			112500											
20			90000											
21			135000											
22			202500											
23			450000											
24			83250											
25			135000											
26			90000											
27			112500											
28			112500											
29			270000											
30			90000											
31			292500											
32			112500											
33														

**QUARTILE 1 (Q1):** Q1 is 112,500. This means that 25% of the "AMT\_INCOME\_TOTAL" values are less than or equal to 112,500.

**QUARTILE 3 (Q3):** Q3 is 202,500. This means that 75% of the "AMT\_INCOME\_TOTAL" values are less than or equal to 202,500.

**INTERQUARTILE RANGE (IQR):** IQR is 90,000, calculated as Q3 - Q1. This means that the middle 50% of the "AMT\_INCOME\_TOTAL" values fall within the range of 112,500 to 202,500.

**LOWER THRESHOLD:** Lower Threshold is -22,500. This means that if there are any negative values or extremely low income values in the dataset (less than -22,500), they would be flagged as potential outliers.

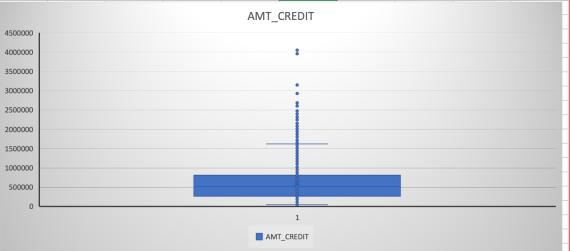
**UPPER THRESHOLD:** Upper Threshold is 337,500. This means that if there are any extremely high income values in the dataset (greater than 337,500), they would be flagged as potential outliers.

The Lower Threshold and Upper Threshold are calculated based on the interquartile range (IQR) and data points that fall below the Lower Threshold or above the Upper Threshold are considered potential outliers and may require further investigation to determine if they are valid data points or data entry errors.

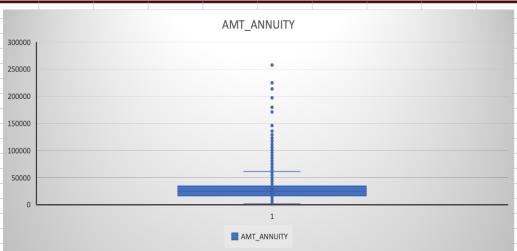
# Outliers



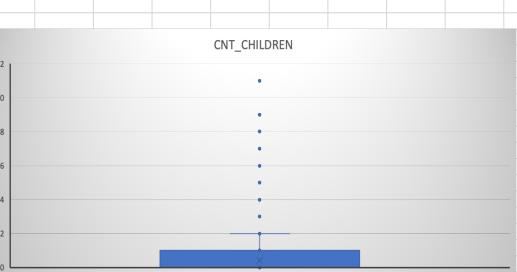
AMT_CREDIT
604957.9
1293502.5
135000
312682.5
513000
490495.9
1560726
1530000
1019610
405000
652500
148365
80865
918468
773680.5
299772
509602.5
270000
157500
544491
427500
1132573.5
497520
239850
247500
225000
979992
327024
790830



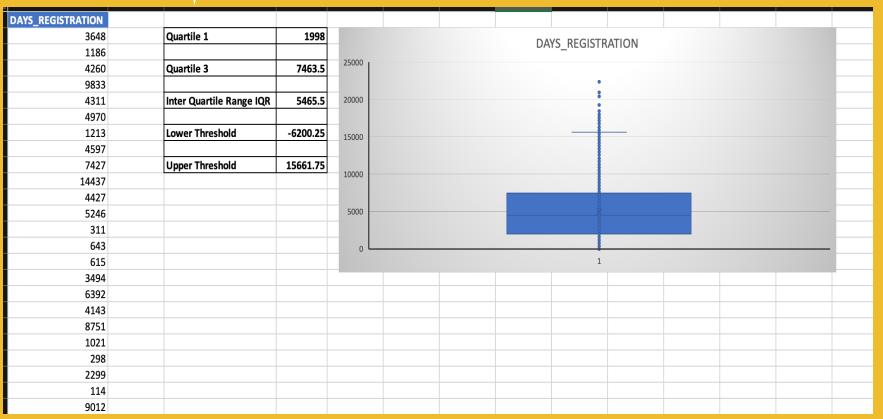
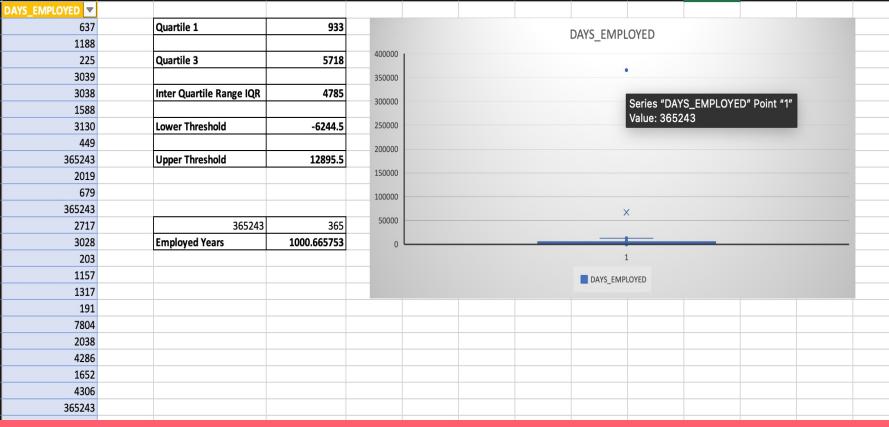
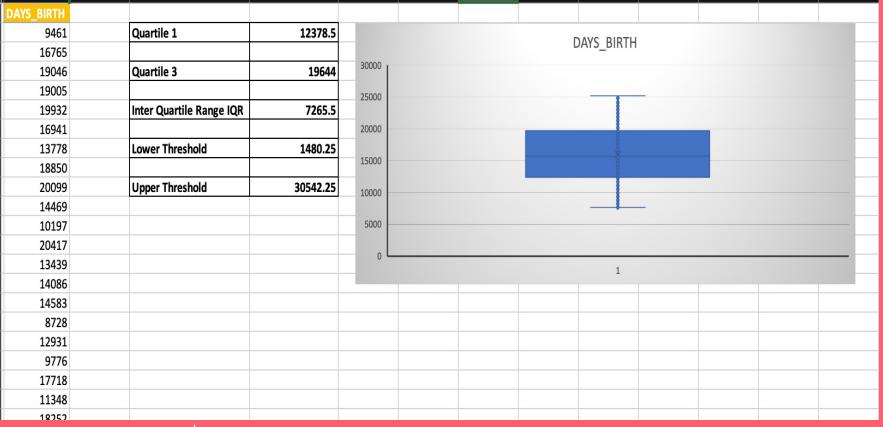
AMT_ANNUITY			
24700.5			
35698.5	Quartile 1		16456.5
6750	Quartile 3		34596
29686.5			
21865.5	Inter Quartile Range IQR		18139.5
27517.5			
41301	Lower Threshold		-10752.75
42075			
33826.5	Upper Threshold		61805.25
20250			
21177			
10678.5			
5881.5			
28966.5			
32778			
20160			
26149.5			
13500			
7875			
17563.5			
21375			
37561.5			
32521.5			
23850			
12703.5			
11074.5			
20706.5			



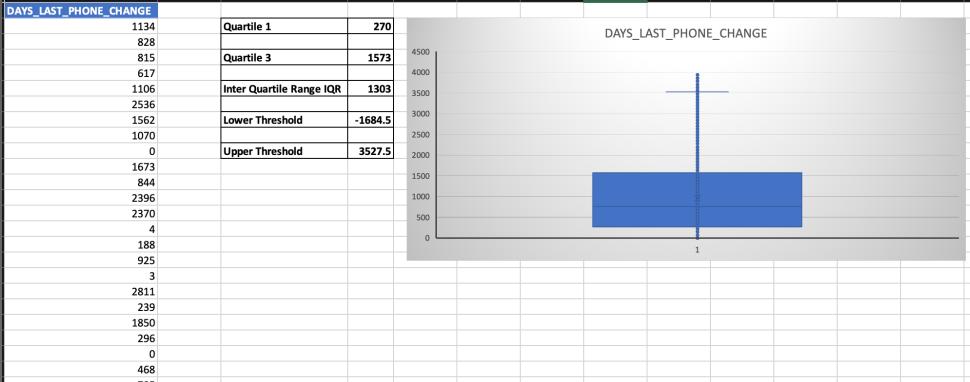
AMT_GOODS_PRICE		
351000		
1129500	Quartile 1	238500
135000	Quartile 3	679500
297000	Inter Quartile Range IQR	441000
513000		
454500	Lower Threshold	-423000
1395000		
1530000	Upper Threshold	1341000
913500		
405000		
652500		
135000		
67500		
697500		
679500		
247500		
387000		
270000		
157500		
454500		
427500		
927000		
450000		
225000		
247500		
235000		
702000		
372000		



# Outliers



# Outliers



We observe here that most of the columns have outliers in them except Days\_Birth and Days\_Id\_Publish

# 03

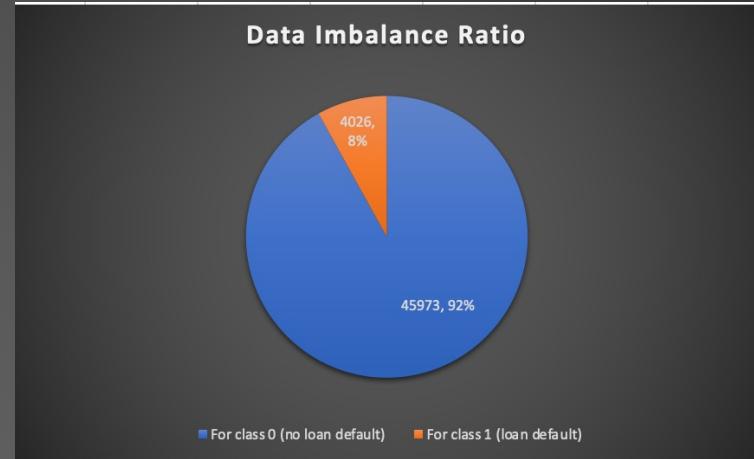
## Analyse Data Imbalance

Determine if there is data imbalance in the loan application dataset and calculate the ratio of data imbalance using Excel functions.



# Data Imbalance Ratio

For class 0 (no loan default)	45973	=COUNTIF(A:A, 0)
For class 1 (loan default)	4026	=COUNTIF(A:A, 1)
<b>Data Imbalance Ratio</b>	<b>8.75%</b>	<b>=C4/C3</b>



The data imbalance ratio of approximately 8.75% indicates that the 'loan default class' is about 8.75% of the 'no loan default class', highlighting the data imbalance in the target variable.

# 04

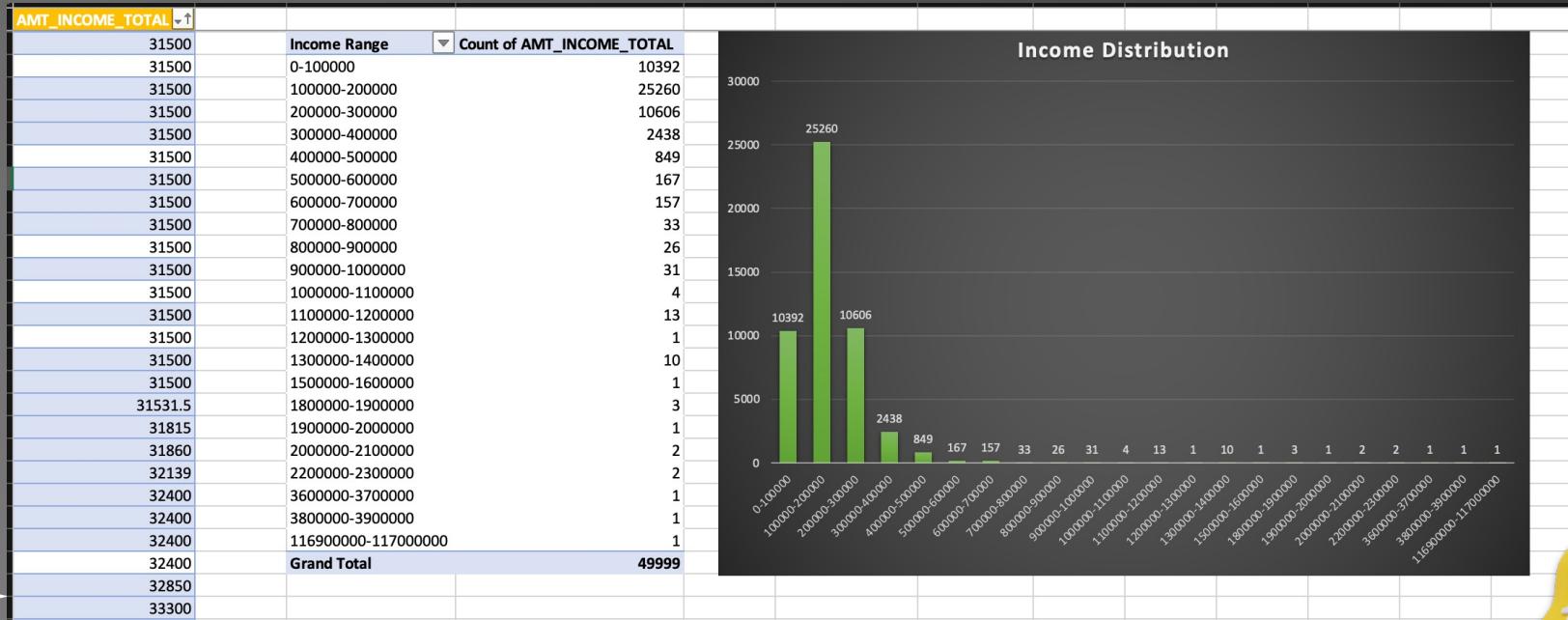
## Perform Univariate, Segmented Univariate, and Bivariate Analysis

Perform univariate analysis to understand the distribution of individual variables, segmented univariate analysis to compare variable distributions for different scenarios, and bivariate analysis to explore relationships between variables and the target variable using Excel functions and features



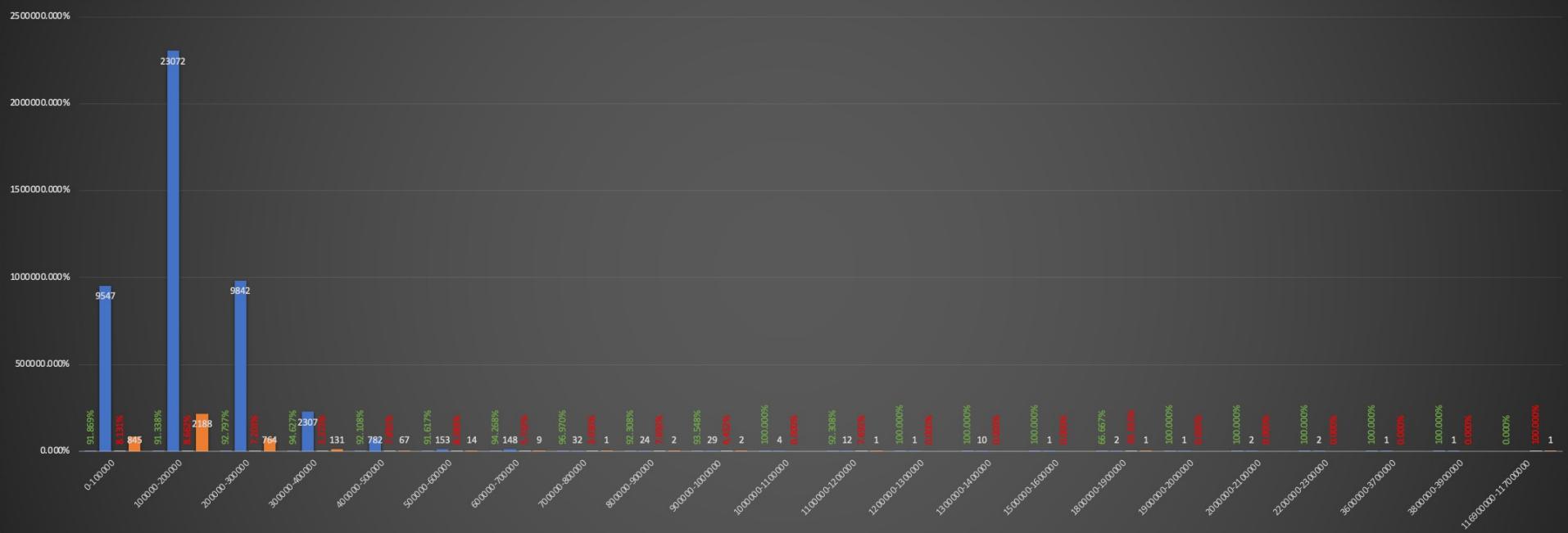
# Univariate Analysis

Univariate analysis is a statistical technique used to analyse and summarize data from a single variable. It focuses on understanding the distribution and characteristics of that particular variable in isolation, without considering any other variables. The primary goal of univariate analysis is to describe and gain insights into the behaviour and patterns of the individual variable.



# Segmented Univariate Analysis

Amt\_Income\_Total - Target



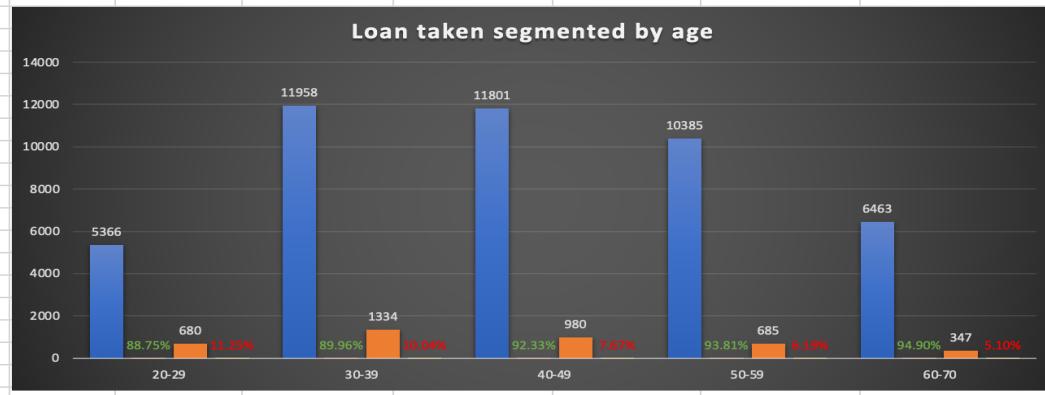
Generally as Income is Increasing the loan default percentage is decreasing.



# Segmented Univariate Analysis

## Age - Target

B2	A	B	C	D	E	F	G	H	I	J	K
1	TARGET	Age	Year_BIRTH								
2	1	26	-9461								
3	0	46	-16765								
4	0	53	-19046								
5	0	53	-19005								
6	0	55	-19932	20-29	5366	88.75%	680	11.25%	6046	100.00%	
7	0	47	-16941	30-39	11958	89.96%	1334	10.04%	13292	100.00%	
8	0	38	-13778	40-49	11801	92.33%	980	7.67%	12781	100.00%	
9	0	52	-18850	50-59	10385	93.81%	685	6.19%	11070	100.00%	
10	0	56	-20099	60-70	6463	94.90%	347	5.10%	6810	100.00%	
11	0	40	-14469	Grand Total	45973	91.95%	4026	8.05%	49999	100.00%	
12	0	28	-10197								
13	0	56	-20417								
14	0	37	-13439								
15	0	39	-14086								
16	0	40	-14583								
17	0	24	-8728								
18	0	36	-12931								
19	0	27	-9776								
20	0	49	-17718								
21	0	32	-11348								
22	0	51	-18252								
23	0	41	-14815								
24	0	31	-11146								
25	0	69	-24827								
26	0	31	-11286								
27	0	53	-19334								
28	1	52	-18724								
29	0	44	-15948								
30	0	28	-9994								



Generally as Age is Increasing the loan default percentage is decreasing.

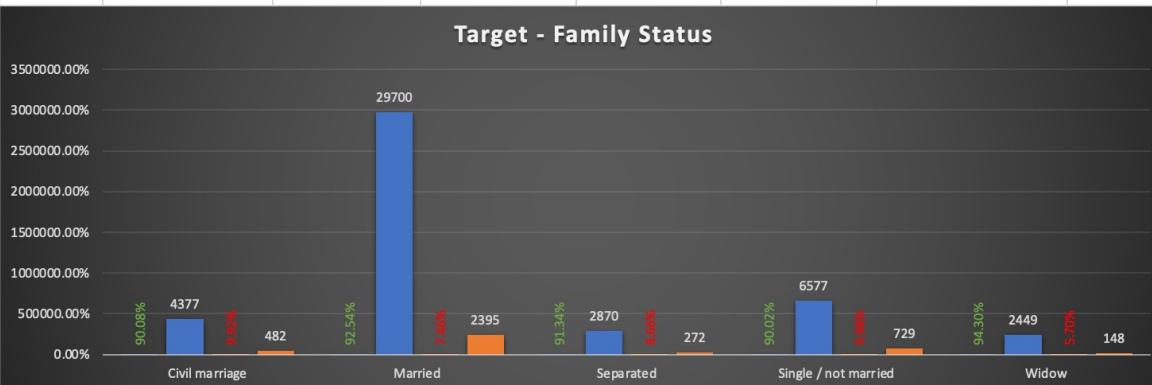


# Segmented Univariate Analysis

## Family Status - Target

TARGET	NAME_FAMILY_STATUS					Total Count of TARGET	Total Count of TARGET2
		0	1	Total Count of TARGET	Total Count of TARGET2		
Row Label	Count of TARGET	Count of TARGET2	Count of TARGET	Count of TARGET2			
1 Single / not married							
0 Married							
0 Single / not married							
0 Civil marriage	Civil marriage	90.08%	4377	9.92%	482	100.00%	4859
0 Single / not married	Married	92.54%	29700	7.46%	2395	100.00%	32095
0 Married	Separated	91.34%	2870	8.66%	272	100.00%	3142
0 Married	Single / not i	90.02%	6577	9.98%	729	100.00%	7306
0 Married	Widow	94.30%	2449	5.70%	148	100.00%	2597
0 Married	<b>Grand Total</b>	<b>91.95%</b>	<b>45973</b>	<b>8.05%</b>	<b>4026</b>	<b>100.00%</b>	<b>49999</b>
0 Single / not married							
0 Married							
0 Married							
0 Married		3500000.00%					
0 Married		3000000.00%					
0 Married		2500000.00%					
0 Married		2000000.00%					
0 Widower		1500000.00%					
0 Single / not married		1000000.00%					
0 Married		500000.00%					
0 Married		0.00%					
0 Married	Civil marriage	90.08%	4377	9.92%	482	100.00%	4859
0 Married	Married	92.54%	29700	7.46%	2395	100.00%	32095
0 Married	Separated	91.34%	2870	8.66%	272	100.00%	3142
0 Married	Single / not i	90.02%	6577	9.98%	729	100.00%	7306
0 Married	Widow	94.30%	2449	5.70%	148	100.00%	2597

Target - Family Status



People who are Married or Widow have lower loan default percentage.



# Segmented Univariate Analysis

## Years Employed - Target

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	TARGET	Years Employed	Days Employed												
2	1	2	-637												
3	0	4	-1188												
4	0	1	-225												
5	0	9	-3039	0-4	16626	89.38%	1975	10.62%	18601	100.00%					
6	0	9	-3038	5-9	11534	91.75%	1037	8.25%	12571	100.00%					
7	0	5	-1588	10-14	4946	94.35%	296	5.65%	5242	100.00%					
8	0	9	-3130	15-19	2123	94.82%	116	5.18%	2239	100.00%					
9	0	2	-449	20-24	1164	95.18%	59	4.82%	1223	100.00%					
10	0	1001	365243	25-29	595	96.12%	24	3.88%	619	100.00%					
11	0	6	-2019	30-34	336	96.83%	11	3.17%	347	100.00%					
12	0	2	-679	35-39	169	97.13%	5	2.87%	174	100.00%					
13	0	1001	365243	40-44	52	100.00%		0.00%	52	100.00%					
14	0	8	-2717	45-49	7	100.00%		0.00%	7	100.00%					
15	0	9	-3028	Grand Total	37552	91.42%	3523	8.58%	41075	100.00%					
16	0	1	-203												
17	0	4	-1157												
18	0	4	-1317												
19	0	1	-191												
20	0	22	-7804												
21	0	6	-2038		16626										
22	0	12	-4286		11534										
23	0	5	-1652		4946										
24	0	12	-4306		2123										
25	0	1001	365243		1164										
26	0	3	-746		595										
27	0	10	-3494		336										
28	1	8	-2628		169										
29	0	4	-1234		174										
30	0	5	-1796		169										
31	0	3	-1010		169										
32	0	8	-2668		169										
33	0	4	-1104		169										
34	0	13	-4404		169										
35	0	6	-2060		169										
36	0	13	-4585		169										
37	0	4	-1275		169										
38	0	3	-768		169										
39	0	4	-1288		169										
40	0	1001	365243		169										
41	0	5	-1761		169										
42	1	4	-1262		169										
43	0	2	-475		169										
44	1	10	-3597		169										
45	0	1001	365243		169										

People who are employed for more years have lower loan default percentage.

# Segmented Univariate Analysis

## Education\_Type - Target

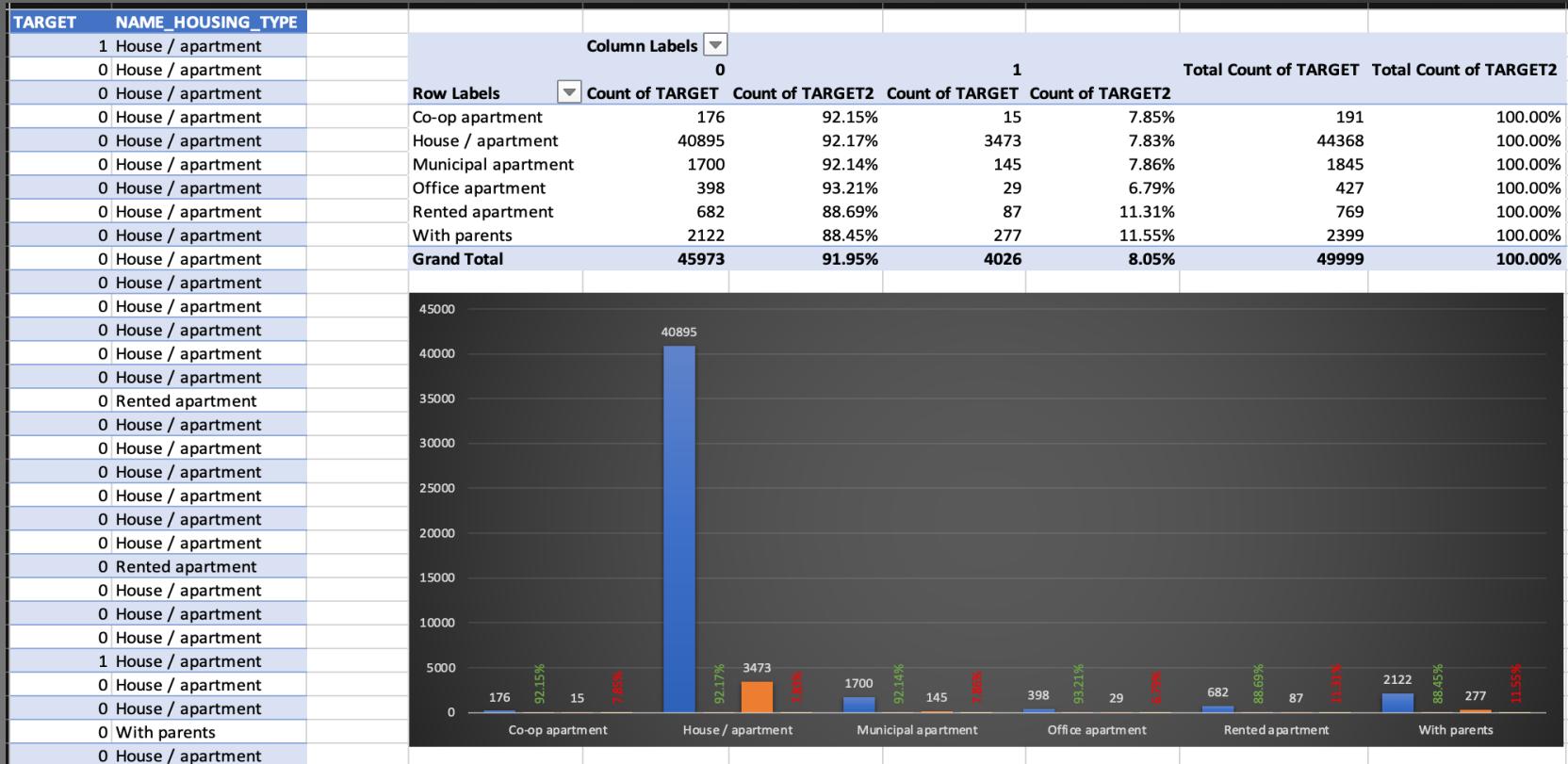
TARGET	NAME_EDUCATION_TYPE	Column Labels				Total Count of TARGET	Total Count of TARGET2
		0	1	Count of TARGET	Count of TARGET2		
Row Labels		Count of TARGET	Count of TARGET2	Count of TARGET	Count of TARGET2		
0 Secondary / secondary special	Secondary / secondary special	32363	90.98%	3209	9.02%	35572	100.00%
0 Higher education	Higher education	11561	95.02%	606	4.98%	12167	100.00%
0 Secondary / secondary special	Incomplete higher	1482	91.48%	138	8.52%	1620	100.00%
0 Higher education	Lower secondary	547	88.23%	73	11.77%	620	100.00%
0 Higher education	Academic degree	20	100.00%		0.00%	20	100.00%
0 Secondary / secondary special	Grand Total	45973	91.95%	4026	8.05%	49999	100.00%
0 Secondary / secondary special							
0 Higher education							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Higher education							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
0 Secondary / secondary special							
1 Secondary / secondary special							
0 Secondary / secondary special							

As education level increases, loan default percentage decreases. People who have higher education tend to default less and less on their loan.



# Segmented Univariate Analysis

## Housing-Type - Target

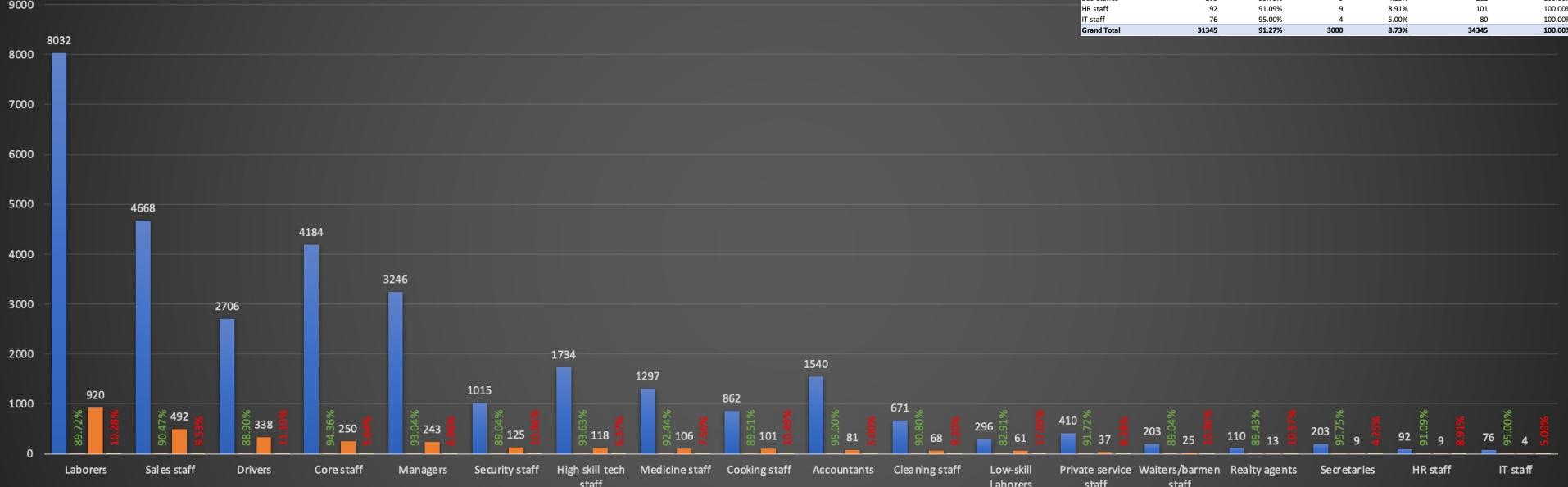


People living in a Rented Apartment or with Parents have higher loan default percentage as compared to people living in apartments or houses.



# Segmented Univariate Analysis

## Occupation Type- Target

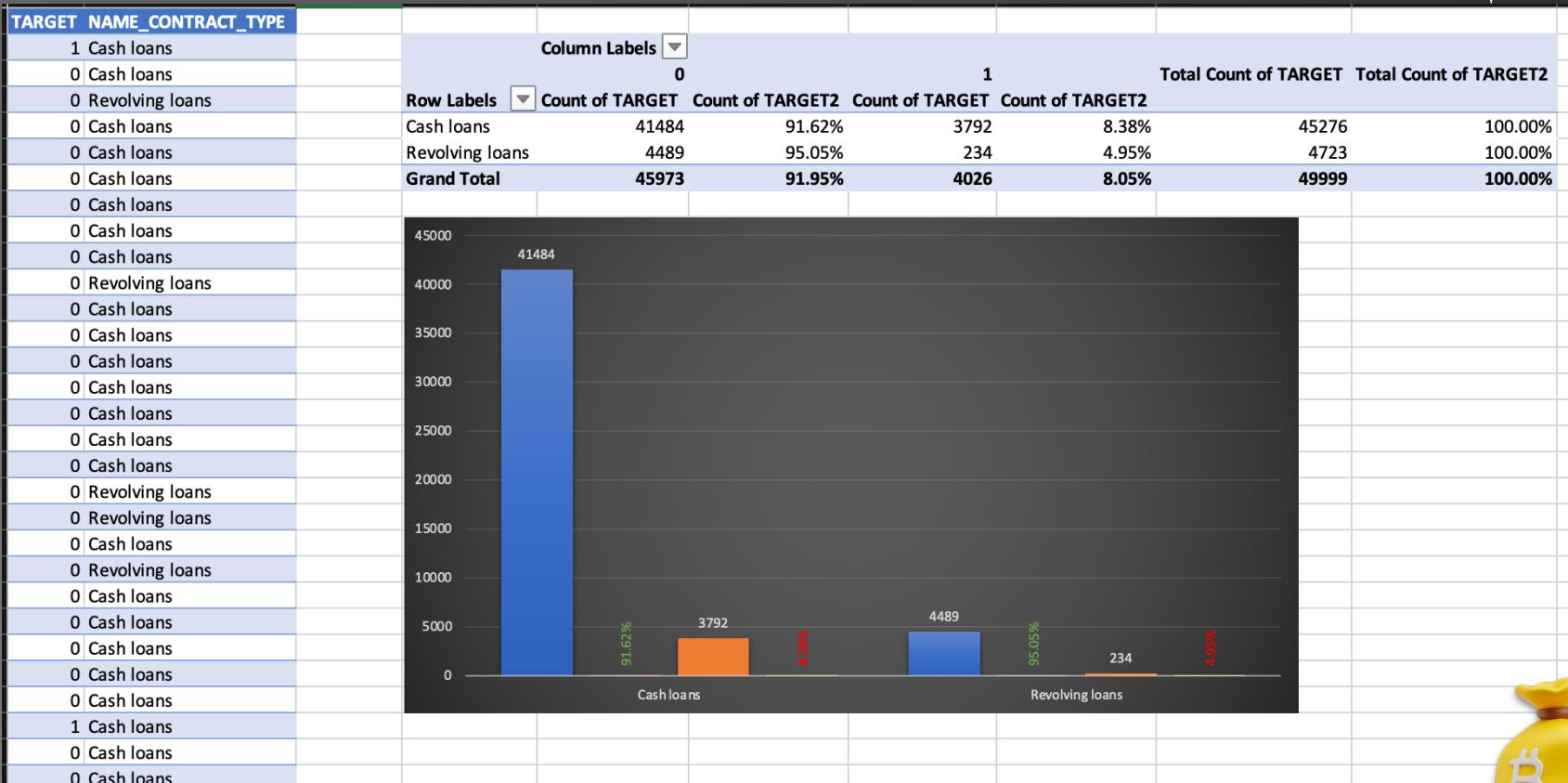


People working as Low Skill Labourers, Drivers, Labourers, Cleaning Staff, Waiters/Barmen Staff etc tend to have higher loan default percentage as compared to people working as IT Staff, Secretaries, Accountants, Core Staff and Managers.



# Segmented Univariate Analysis

## Contract\_Type - Target

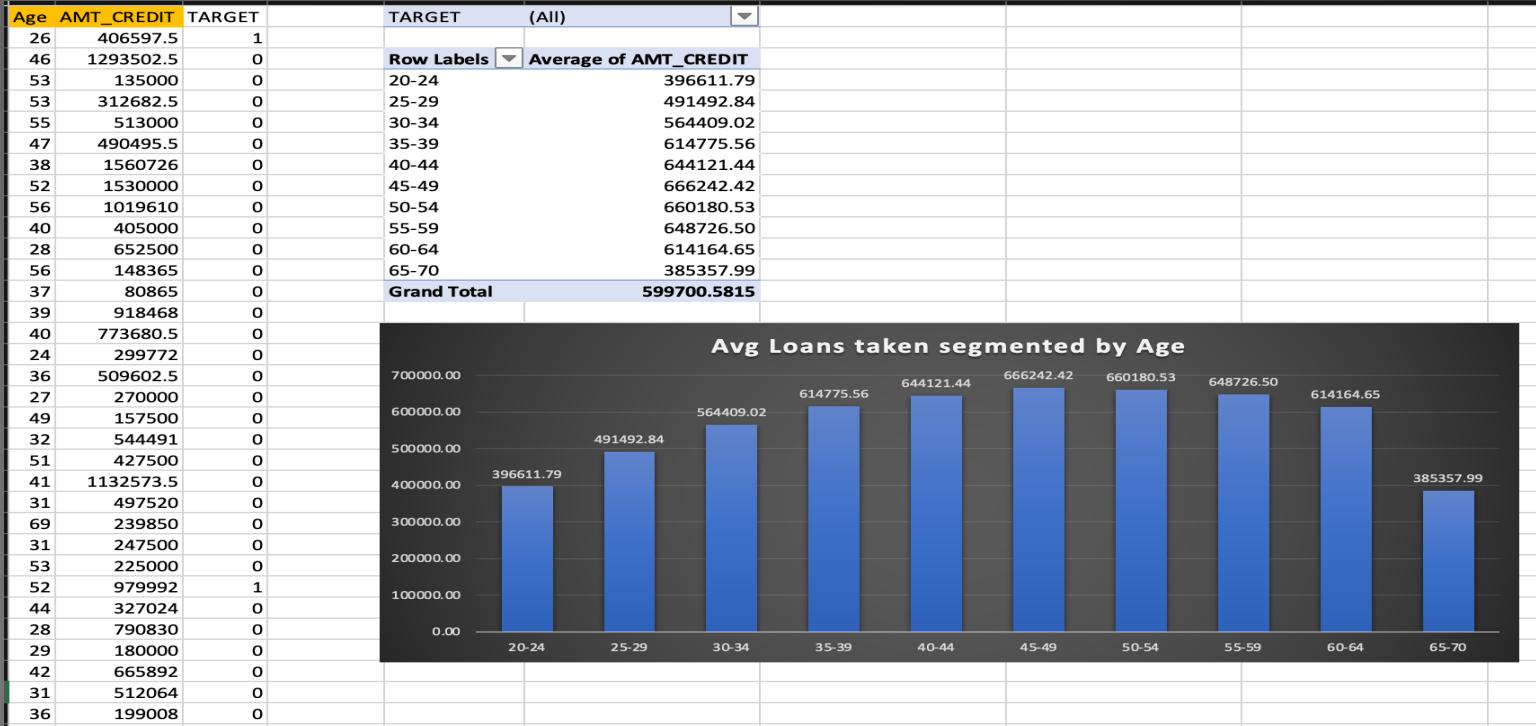


Although most people took cash loans but people who took revolving loans have lower loan default percentage.



# Bivariate Analysis

## Age-Amt\_Credit

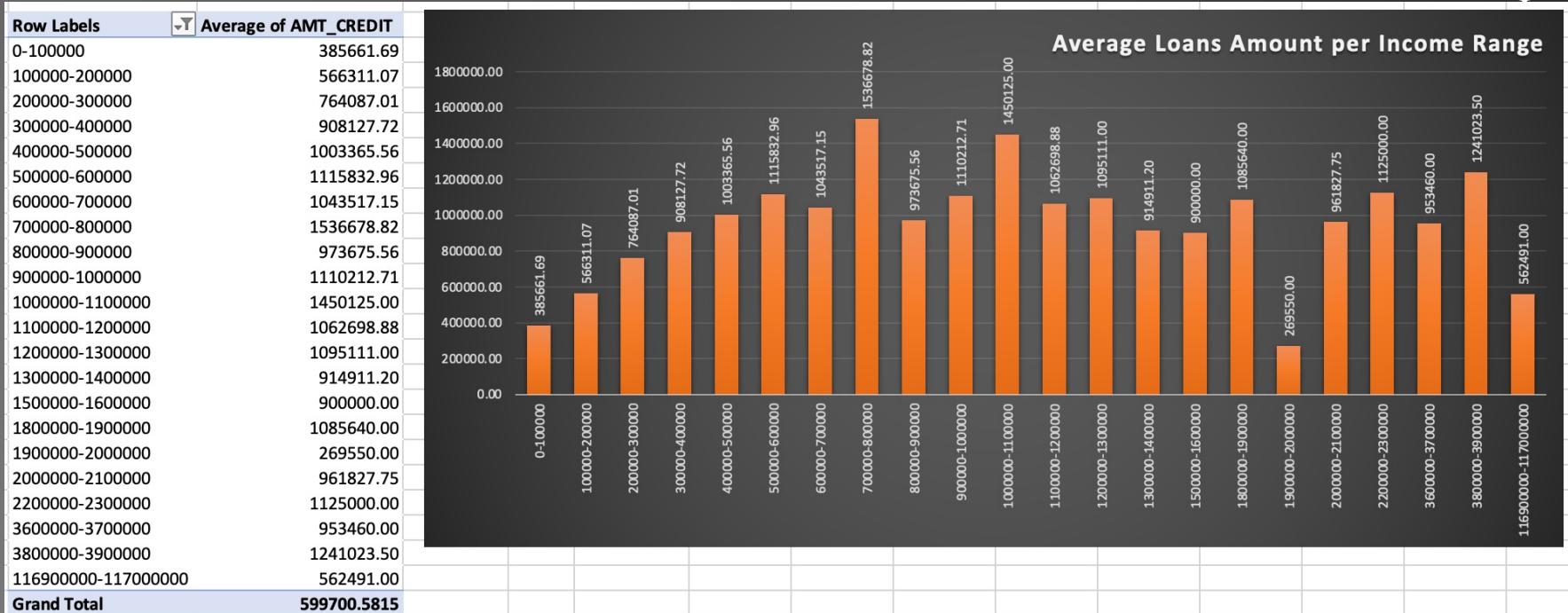


People in the age between 35 to 65 tend to take higher amount loan



# Bivariate Analysis

## Income - Amt\_Credit



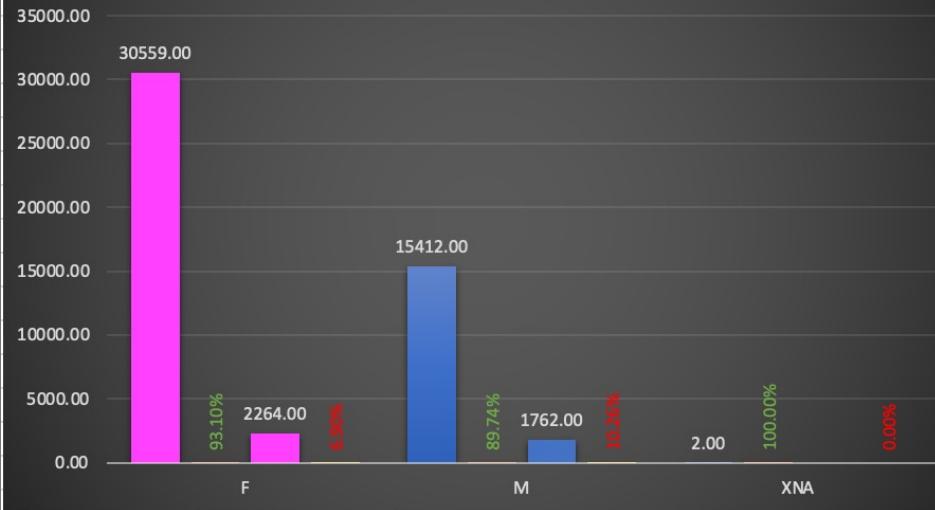
Highest average Loan amount taken is by People whose Income is in 700K-800K. Here we cant say that as their income increases the amount of loan also increases



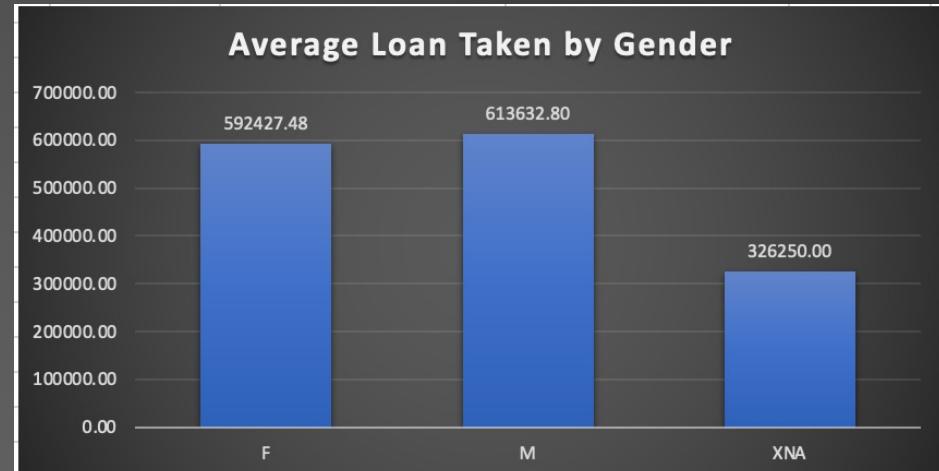
# Segmented and Bivariate Analysis

Gender-Amt\_Credit

Loan Taken Segmented by Gender



Average Loan Taken by Gender



More Females took loan compared to Males and their Loan Default Percentage is Less.

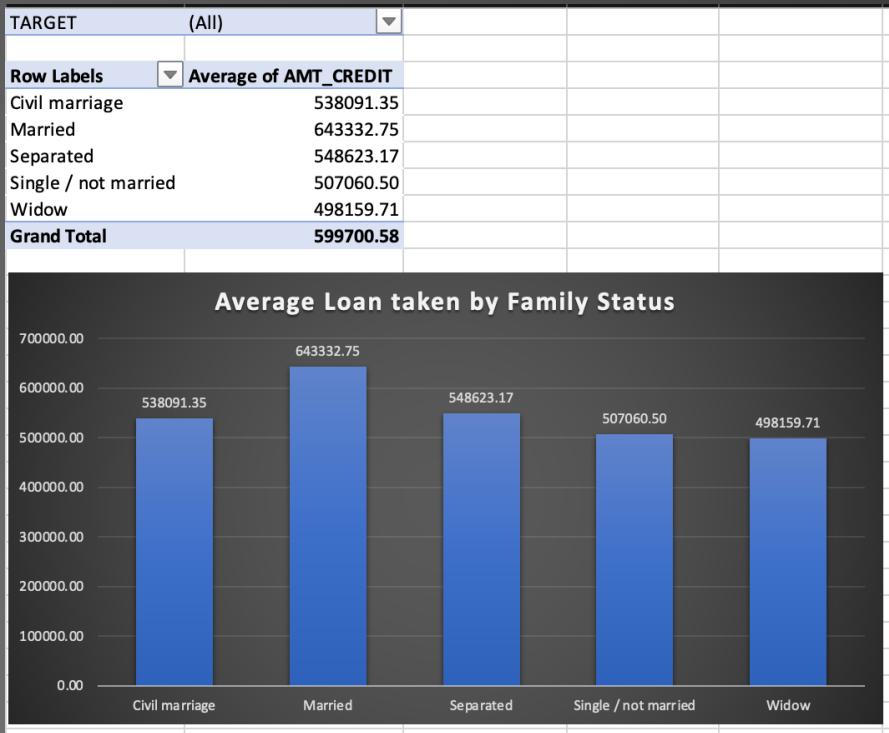
Although fewer Males take loan but their average Loan Amount is more.

This data shows that Females should be preferred over males when giving the loan.



# Bivariate Analysis

## Family\_Status - Amt\_Credit



Married People tend to take higher amount loan compared to others





# 05



## Identify Top Correlations for Different Scenarios



Segment the dataset based on different scenarios and identify the top correlations for each segmented data using Excel functions.



# Correlation

## Excel WorkBook

A	B	C	D	E	F	G	H	I	J	K	L	M	D
1	TARGET	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	REGION_POPULATION_RELATIVE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	EXT_SOURCE_2	EXT_SOURCE_3	D
2	1	202500	406597.5	24700.5	351000	0.018801	-9461	-637	-3648	-2120	0.262948593	0.13937578	
3	0	270000	1293502.5	35698.5	1129500	0.003541	-16765	-1188	-1186	-291	0.622245775	0.53527625	
4	0	67500	135000	6750	135000	0.010032	-19046	-225	-4260	-2531	0.555912083	0.729566691	
5	0	135000	312682.5	29686.5	297000	0.008019	-19005	-3039	-9833	-2437	0.65044169	0.53527625	
6	0	121500	513000	21865.5	513000	0.028663	-19932	-3038	-4311	-3458	0.322738287	0.53527625	
7	0	99000	490495.5	27517.5	454500	0.035792	-16941	-1588	-4970	-477	0.354224732	0.621226338	
8	0	171000	1560726	41301	1395000	0.035792	-13778	-3130	-1213	-619	0.723999852	0.492060094	
9	0	360000	1530000	42075	1530000	0.003122	-18850	-449	-4597	-2379	0.714279286	0.54065445	
10	0	112500	1019610	33826.5	913500	0.018634	-20099	365243	-7427	-3514	0.205747288	0.751723715	
11	0	135000	405000	20250	405000	0.019689	-14469	-2019	-14437	-3992	0.746643629	0.53527625	
12	0	112500	652500	21177	652500	0.0228	-10197	-679	-4427	-738	0.651862333	0.363945239	
13	0	38419.155	148365	10678.5	135000	0.015221	-20417	365243	-5246	-2512	0.555183162	0.652896552	
14	0	67500	80865	5881.5	67500	0.031329	-13439	-2717	-311	-3227	0.715041819	0.176652579	
15	0	225000	918468	28966.5	697500	0.016612	-14086	-3028	-643	-4911	0.566906613	0.77008707	
16	0	189000	773680.5	32778	679500	0.010006	-14583	-203	-615	-2056	0.642656205	0.53527625	
17	0	157500	299772	20160	247500	0.020713	-8728	-1157	-3494	-1368	0.346633981	0.678567689	
18	0	108000	509602.5	26149.5	387000	0.018634	-12931	-1317	-6392	-3866	0.23637784	0.062103038	
19	0	81000	270000	13500	270000	0.010966	-9776	-191	-4143	-2427	0.683513346	0.53527625	
20	0	112500	157500	7875	157500	0.04622	-17718	-7804	-8751	-1259	0.706428403	0.556727426	
21	0	90000	544491	17563.5	454500	0.015221	-11348	-2038	-1021	-3964	0.58661714	0.477649155	
22	0	135000	427500	21375	427500	0.015221	-18252	-4286	-298	-1800	0.113374513	0.53527625	
23	0	202500	1132573.5	37561.5	927000	0.025164	-14815	-1652	-2299	-2299	0.233766958	0.542445144	
24	0	450000	497520	32521.5	450000	0.020713	-11146	-4306	-114	-2518	0.457142972	0.358951229	
25	0	83250	239850	23850	225000	0.006296	-24827	365243	-9012	-3684	0.624304737	0.669056695	
26	0	135000	247500	12703.5	247500	0.026392	-11286	-746	-108	-3729	0.786179309	0.565607981	
27	0	90000	225000	11074.5	225000	0.028663	-19334	-3494	-2419	-2893	0.651405637	0.461482391	
28	1	112500	979992	27076.5	702000	0.018029	-18724	-2628	-6573	-1827	0.54847716	0.190705948	
29	0	112500	327024	23827.5	270000	0.019101	-15948	-1234	-5782	-3153	0.541123702	0.659405532	
30	0	270000	790830	57676.5	675000	0.04622	-9994	-1796	-4668	-2661	0.68501099	0.524496446	
31	0	90000	180000	9000	180000	0.030755	-10341	-1010	-4799	-3015	0.502779038	0.53527625	
32	0	292500	665892	24592.5	477000	0.025164	-15280	-2668	-5266	-3787	0.479987342	0.0102573	
33	0	112500	512064	25033.5	360000	0.008575	-11144	-1104	-7846	-2904	0.627300401	0.53527625	
34	0	90000	199008	20893.5	180000	0.010032	-12974	-4404	-7123	-4464	0.559466792	0.2137231	
35	0	360000	733315.5	39069	679500	0.015221	-11694	-2060	-3557	-3557	0.321744896	0.8556	
36	0	112500	112500	22005	1125000	0.010000	-15007	-1505	-5725	-1067	0.172467500	0.172467500	

Corr- Main Sheet

Correlation-Target 0

Target 0

Correlation-Target 1

Target 1

+



# Correlation

## Target - 0 (Loan Payees)

	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	REGION_POPULATION_RELATIVE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	EXT_SOURCE_2	EXT_SOURCE_23	EXT_SOURCE_24	DAYS_LAST_PHONE_CHANGE	CNT_CHILDREN
AMT_INCOME_TOTAL	1													
AMT_CREDIT	0.377965752	1												
AMT_ANNUITY	0.451135103	0.770772908	1											
AMT_GOODS_PRICE	0.384675888	0.987235417	0.77613285	1										
REGION_POPULATION_RELATIVE	0.181941261	0.095539444	0.117279106	0.09896058	1									
DAYS_BIRTH	0.073769425	-0.051084182	0.099910977	-0.04869335	-0.030435419	1								
DAYS_EMPLOYED	-0.162702675	-0.077367219	-0.113005115	-0.075172	-0.066610653	-0.61528998	1							
DAYS_REGISTRATION	0.068993375	0.008053758	0.034609001	0.01129847	-0.058501361	0.33502805	-0.204370881	1						
DAYS_ID_PUBLISH	0.032286356	-0.008290189	0.009427021	-0.00930065	-0.002236288	0.27007331	-0.27222439	0.103548902	1					
EXT_SOURCE_2	0.15617334	0.136258463	0.130022952	0.14329839	0.201089847	-0.0803327	-0.034096314	-0.053917105	-0.040900368	1				
EXT_SOURCE_23	-0.073654638	0.028831122	0.018547427	0.03105218	-0.013832665	-0.17888625	0.098594722	-0.099901082	-0.111944396	0.068846786	1			
DAYS_LAST_PHONE_CHANGE	-0.049497956	-0.071203379	-0.064450897	-0.07428594	-0.044127291	0.07253958	0.032951867	0.047780168	0.085063175	-0.184718008	-0.059902129	1		
CNT_CHILDREN	0.036319722	0.005705458	0.026384162	0.00155316	-0.024912809	0.33587627	-0.243591518	0.183072478	-0.032537221	-0.013466269	-0.039280661	-0.004822698	1	

★ Top 10 Correlations  
for Loan Payees

Bottom 10 Correlations  
for Loan Payees

AMT_GOODS_PRICE	AMT_CREDIT	0.987235417
AMT_GOODS_PRICE	AMT_ANNUITY	0.77613285
AMT_ANNUITY	AMT_CREDIT	0.770772908
AMT_ANNUITY	AMT_INCOME_TOTAL	0.451135103
AMT_GOODS_PRICE	AMT_INCOME_TOTAL	0.384675888
AMT_CREDIT	AMT_INCOME_TOTAL	0.377965752
CNT_CHILDREN	DAYS_BIRTH	0.335876269
DAYS_REGISTRATION	DAYS_BIRTH	0.335028046
DAYS_ID_PUBLISH	DAYS_BIRTH	0.270073313
EXT_SOURCE_2	REGION_POPULATION_RELATIVE	0.201089847

DAYS_EMPLOYED	DAYS_BIRTH	-0.615289978
DAYS_ID_PUBLISH	DAYS_EMPLOYED	-0.27222439
CNT_CHILDREN	DAYS_EMPLOYED	-0.243591518
DAYS_REGISTRATION	DAYS_EMPLOYED	-0.204370881
DAYS_LAST_PHONE_CHANGE	EXT_SOURCE_2	-0.184718008
EXT_SOURCE_23	DAYS_BIRTH	-0.17888625
DAYS_EMPLOYED	AMT_INCOME_TOTAL	-0.162702675
DAYS_EMPLOYED	AMT_ANNUITY	-0.113005115
EXT_SOURCE_23	DAYS_ID_PUBLISH	-0.111944396
EXT_SOURCE_24	DAYS_REGISTRATION	-0.099901082

# Correlation

## Target - 1 (Loan Defaulters)

	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	REGION_POPULATION_RELATIVE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	EXT_SOURCE_2	EXT_SOURCE_23	DAYS_LAST_PHONE_CHANGE	CNT_CHILDREN
AMT_INCOME_TOTAL	1												
AMT_CREDIT	0.015271444	1											
AMT_ANNUITY	0.018004594	0.749665201	1										
AMT_GOODS_PRICE	0.013266279	0.982432318	0.749705184	1									
REGION_POPULATION_RELATIVE	-0.006180303	0.067775624	0.073123998	0.076596242	1								
DAYS_BIRTH	0.009033662	-0.142506035	-0.008751713	-0.140996151	-0.016468731	1							
DAYS_EMPLOYED	-0.011555963	0.016039571	-0.079556008	0.020213912	0.007742909	-0.58147904	1						
DAYS_REGISTRATION	-0.009561152	-0.042844404	0.021581654	-0.043371319	-0.046130288	0.28843784	-0.188718437	1					
DAYS_ID_PUBLISH	-0.009122006	-0.043771901	-0.02132109	-0.049784603	-0.005118563	0.24789657	-0.230063668	0.09029149	1				
EXT_SOURCE_2	-0.016228175	0.119184882	0.113693247	0.133319809	0.159220129	-0.11141366	-0.018800184	-0.067763159	-0.03748574	1			
EXT_SOURCE_23	-0.026469291	0.045724291	0.017777631	0.04751535	-0.021648936	-0.13943364	0.083171459	-0.052718825	-0.073389069	0.047324503	1		
DAYS_LAST_PHONE_CHANGE	0.012457111	-0.124539343	-0.100470941	-0.128807068	-0.067105681	0.12460949	-0.015732544	0.078604652	0.138087781	-0.204067856	-0.041135827	1	
CNT_CHILDREN	0.010110177	0.007601905	0.029172977	-0.001116682	-0.020359154	0.2496732	-0.189324184	0.152113117	-0.042360717	-0.01541713	-0.015321808	0.011339334	1

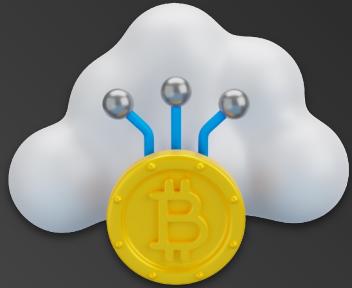
★ Top 10 Correlations  
for Loan Defaulters

AMT_GOODS_PRICE	AMT_CREDIT	0.982432318
AMT_GOODS_PRICE	AMT_ANNUITY	0.749705184
AMT_ANNUITY	AMT_CREDIT	0.749665201
DAYS_REGISTRATION	DAYS_BIRTH	0.288437837
CNT_CHILDREN	DAYS_BIRTH	0.2496732
DAYS_ID_PUBLISH	DAYS_BIRTH	0.247896571
EXT_SOURCE_2	REGION_POPULATION_RELATIVE	0.159220129
CNT_CHILDREN	DAYS_REGISTRATION	0.152113117
DAYS_LAST_PHONE_CHANGE	DAYS_ID_PUBLISH	0.138087781
EXT_SOURCE_2	AMT_GOODS_PRICE	0.133319809

Bottom 10 Correlations  
for Loan Defaulters

DAYS_EMPLOYED	DAYS_BIRTH	-0.581479041
DAYS_ID_PUBLISH	DAYS_EMPLOYED	-0.230063668
DAYS_LAST_PHONE_CHANGE	EXT_SOURCE_2	-0.204067856
CNT_CHILDREN	DAYS_EMPLOYED	-0.189324184
DAYS_REGISTRATION	DAYS_EMPLOYED	-0.188718437
DAYS_BIRTH	AMT_CREDIT	-0.142506035
DAYS_BIRTH	AMT_GOODS_PRICE	-0.140996151
EXT_SOURCE_23	DAYS_BIRTH	-0.139433636
DAYS_LAST_PHONE_CHANGE	AMT_GOODS_PRICE	-0.128807068
DAYS_LAST_PHONE_CHANGE	AMT_CREDIT	-0.124539343

# Insights & Results



# KEY INSIGHTS FROM OUR ANALYSIS

- **Outliers and Data Distribution:** Most columns display outliers, except for "Days\_Birth" and "Days\_ID\_Publish."
- **Data Imbalance:** The dataset has a substantial data imbalance ratio of 8.75%, emphasizing the need to address class imbalance.
- **Loan Amount and Income:** People with incomes between 100k to 200k tend to apply for loans more frequently. As income increases, loan default percentages decrease.
- **Age and Loan Default:** Loan default likelihood decreases as age increases, implying older applicants have better repayment behaviour.
- **Marital Status:** Married or widowed applicants show lower loan default percentages, suggesting a connection between marital status and loan repayment.
- **Employment Duration:** Longer job tenures associate with lower loan default rates, indicating stable employment positively influences repayments.
- **Education Level:** Higher education correlates with reduced default rates, implying educated individuals prioritize loan repayment.
- **Housing Situation:** Rented apartments or living with parents link to higher default rates, possibly reflecting financial stability differences.
- **Occupation Impact:** Occupations like IT Staff, Secretaries, Accountants, Core Staff, and Managers show lower default percentages than Low Skill Labourers, Drivers, and Cleaning Staff.
- **Loan Type:** Revolving loans have lower defaults, though cash loans are more common.
- **Gender Dynamics:** Despite more female applicants, their loan default rate is lower than males. Males, however, ask for higher average loan amounts.

These insights provide a deep understanding of loan default contributors and inform practical decisions to manage lending risks.

# RESULTS

- The main challenge was managing the large dataset. I learned how to handle large datasets.
- The dataset had a lot of missing data and outliers, which I learned how to manage effectively.
- I learned how to use data analysis add-ins in Excel.
- This project helped me in improving my Excel skills and gaining a better understanding of how to navigate complex datasets.



# Thank You!

