## **Bank Loan Case Study**

## **Project Description**

#### **Objective:**

To analyze patterns in loan application data to identify factors influencing loan defaults.

## **Key Risks:**

- Approving loans for customers who cannot repay.
- Rejecting loans for customers who can repay.

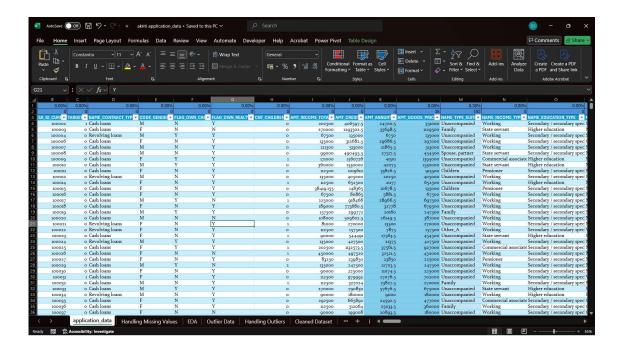
#### Goal:

Enable data-driven decisions for loan approval based on customer and loan attributes.

## **Approach**

### **Data Preparation:**

• **Dataset Inspection:** Downloaded and reviewed the dataset structure.



## • Handling Missing Values:

o Used mean/median imputation for numerical variables.

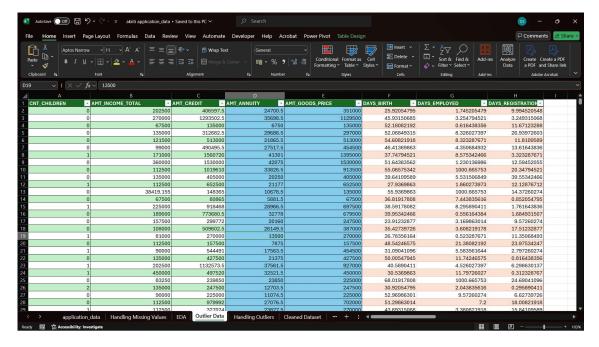
	Dealing with Missing Values							
S No.	Columns Replacing Missing Values	Mean Median		Mode				
1	AMT_ANNUITY	27107.14918	24939					
2	AMT_GOODS_PRICE	539052.277	450000					
3	NAME_TYPE_SUITE			Unaccompanied				
4	OCCUPATION_TYPE			Laborers				
5	CNT_FAM_MEMBERS	2.158726349	2					
	OBS_30_CNT_SOCIAL_CIRCLE	1.419396776	0					
7	DEF_30_CNT_SOCIAL_CIRCLE	0.141345654	0					
8	OBS_60_CNT_SOCIAL_CIRCLE	1.402336093	0					
9	DEF_60_CNT_SOCIAL_CIRCLE	0.09800392	0					
10	DAYS_LAST_PHONE_CHANGE	-964.3153985	-755					
11	AMT_REQ_CREDIT_BUREAU_HOUR	0.006140246	0					
12	AMT_REQ_CREDIT_BUREAU_DAY	0.00650026	0					
13	AMT_REQ_CREDIT_BUREAU_WEEK	0.028021121	0					
14	AMT_REQ_CREDIT_BUREAU_MON	0.233889356	0					
15	AMT_REQ_CREDIT_BUREAU_QRT	0.225829033	0					
16	AMT_REQ_CREDIT_BUREAU_YEAR	1.627725109	1					

o Applied mode or introduced an 'Unknown' category for categorical variables.

Name Type:	suite	OCCUPATION_TYPE			
Category	Count	Category	Count		
Unaccompanied	40626	Laborers	24606		
Family	6549	Core staff	4434		
Spouse, partner	1849	Accountants	1621		
Children	542	Managers	3488		
Other_A	137	Drivers	3044		
Other_B	259	Sales staff	5160		
Group of people	36	Cleaning staff	739		
Blanks	192	Cooking staff	963		
		Private service staff	447		
		Medicine staff	1403		
		Security staff	1140		
		High skill tech staff	1852		
		Waiters/barmen staff	228		
		Low-skill Laborers	357		
		Realty agents	123		
		Secretaries	212		
		IT staff	80		
		HR staff	101		

#### • Outlier Detection:

- o Identified outliers using the Interquartile Range (IQR) method.
- Adjusted outliers within calculated bounds to maintain data integrity.



#### **Analysis:**

- Conducted univariate, segmented univariate, and bivariate analyses.
- Assessed data imbalance and class proportions.
- Computed correlations between variables and target outcomes within segmented data.

#### **Visualization:**

• Created histograms, bar charts, pie charts, box plots, and scatter plots.

#### Reporting:

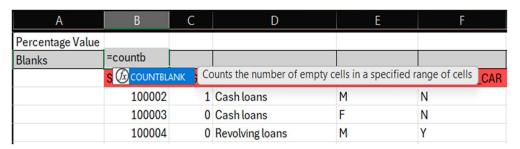
- Documented insights and findings.
- Linked detailed analysis in Excel sheets for reference.

#### **Tech-Stack Used**

- Microsoft Excel 2022: Data preparation, analysis, and visualization.
- Google Drive: Hosted Excel files for sharing and integration.
- **Ms Word:** Designed this report.

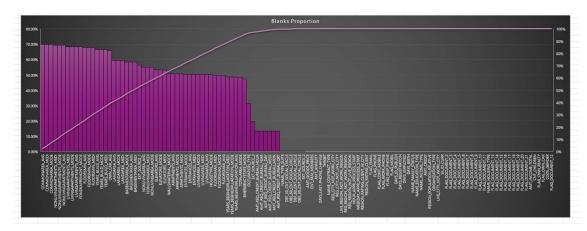
## A. <u>Identify Missing Data and Deal with it Appropriately:</u>

• Identification: Counted blank values using the COUNTBLANK() function.

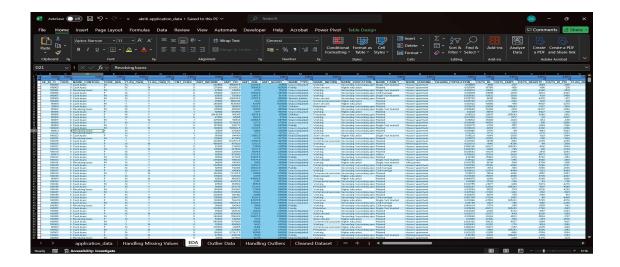


#### • Resolution:

- o Numerical variables: Imputed using mean/median.
- o Categorical variables: Used mode or created 'Unknown' category.
- **Visualization:** Created Pareto charts showing the percentage of missing data by column.



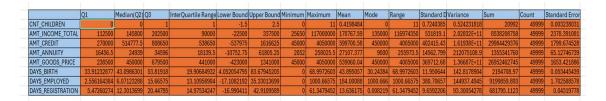
• Columns with more than 50% missing data were deleted.



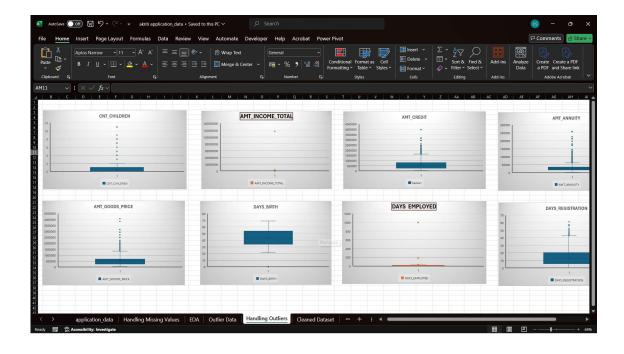
## **B.** Identify Outliers in the Dataset:

#### **Method:**

- o Calculated Q1, Q3, and IQR.
- Defined bounds: Q1 1.5IQR and Q3 + 1.5IQR.
- o Adjusted values beyond bounds.

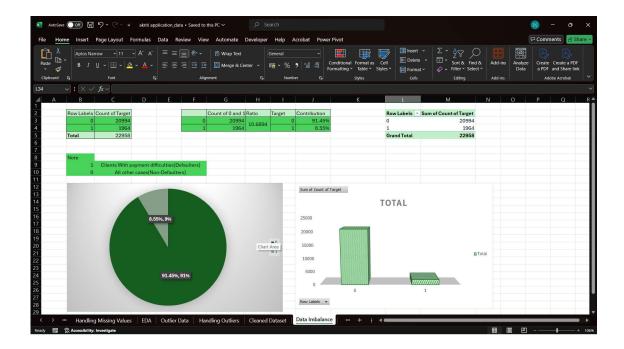


- **Example:** A record with 1000 years of employment was identified as an error and removed.
- Visualization: Used box plots to identify and address outliers.



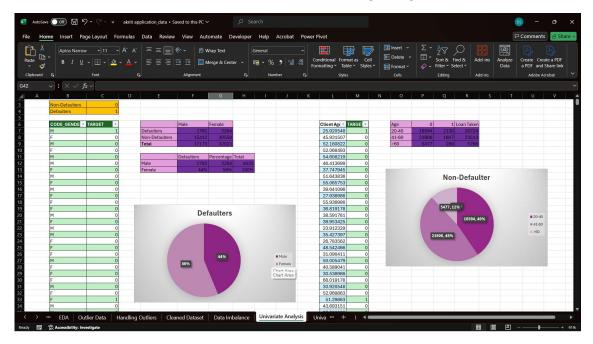
## C. Analyze Data Imbalance:

- Assessed class distribution of target variables using the COUNTIF () function.
- Identified significant gaps in male-to-female loan outcomes.
- Visualization: Pie charts displaying class distribution.



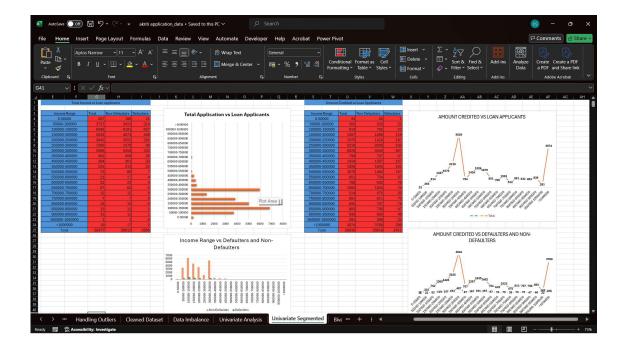
# D. Perform Univariate, Segmented Univariate, and Bivariate Analysis:

Examined distributions of individual variables using histograms and bar charts.



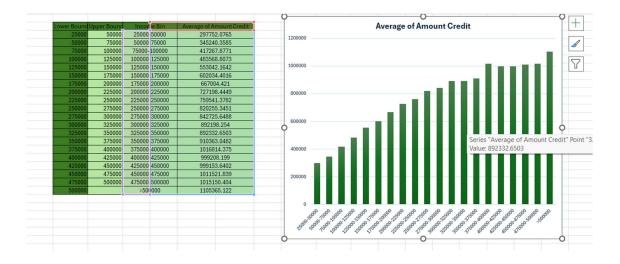
## **Segmented Univariate Analysis:**

• Compared distributions for defaulters and non-defaulters.



## **Bivariate Analysis:**

• Analyzed relationships between variables using scatter plots and pivot tables.



## **E. Identify Top Correlations for Different Scenarios:**

Calculated correlation coefficients using the CORREL function.

	_			_		_			
					REGION_POPULATION_RELATIVE -				
0	0	270000	1293502.5	35698.5	0.003541			-1186	
0	0	67500	135000	6750	0.010032		-225	-4260	
0	0	135000	312682.5	29686.5	0.008019		-3039	-9833	
0	0	121500	513000	21865.5	0.028663		-3038	-4311	
0	0	99000	490495.5	27517.5	0.035792		-1588	-4970	
0	1	171000	1560726	41301	0.035792		-3130	-1213	
0	0	360000	1530000	42075	0.003122		-449	-4597	
0	0	112500	1019610	33826.5	0.018634		365243	-7427	
0	0	135000	405000	20250	0.019689		-2019	-14437	
0	1	112500	652500	21177	0.0228	-10197	-679	-4427	
0	0	38419.155	148365	10678.5	0.015221	-20417	365243	-5246	
0	0	67500	80865	5881.5	Plot Area   1329		-2717	-311	
0	1	225000	918468	28966.5	0.016612			-643	
0	0	189000	773680.5	32778	0.010006		-203	-615	
0	0	157500	299772	20160	0.020713		-1157	-3494	
0	0	108000	509602.5	26149.5	0.018634		-1317	-6392	
0	1	81000	270000	13500	0.010966			-4143	
0	0	112500	157500	7875	0.04622		-7804	-8751	
0	1	90000	544491	17563.5	0.015221	-11348	-2038	-1021	
0	0	135000	427500	21375	0.015221	-18252	-4286	-298	
0	1	202500	1132573.5	37561.5	0.025164	-14815	-1652	-2299	
0	1	450000	497520	32521.5	0.020713	-11146	-4306	-114	
0	0	83250	239850	23850	0.006296	-24827	365243	-9012	
0	2	135000	247500	12703.5	0.026392	-11286	-746	-108	
0	0	00000	225000	11074 5	0.000662	10224	2404	2/10	
					REGION_POPULATION_RELATIVE				REGION_RATING_CLIENT
1	0	202500	406597.5	24700.5	0.018801	-9461	-637	-3648	
1	0	112500	979992	27076.5	0.018029	-18724	-2628	-6573	
1	0	202500	1193580	35028	0.025164	-17482	-1262	-1182	
1	0	135000	288873	16258.5	0.007305	-13384	-3597	-45	
1	0	81000	252000	14593.5	0.028663	-24794	365243	-5391	
1	0	315000	953460	64107	0.030755	-10199	-2015	-4802	
1	1	157500	723006	02.5	0.007274	-10526	-267	-387	

1	0	202500	406597.5	24/00.5	0.018801	-9461	-63/	-3648	2
1	0	112500	979992	27076.5	0.018029	-18724	-2628	-6573	3
1	0	202500	1193580	35028	0.025164	-17482	-1262	-1182	2
1	0	135000	288873	16258.5	0.007305	-13384	-3597	-45	3
1	0	81000	252000	14593.5	0.028663	-24794	365243	-5391	2
1	0	315000	953460	64107	0.030755	-10199	-2015	-4802	2
1	1	157500	723996	Plot Area   02.5	0.007274	-10526	-267	-387	2
1	0	292500	675000	36747	0.035792	-16667	-200	-5239	2
1	0	157500	245619	12667.5	0.022625	-17538	-7676	-774	2
1	0	111915	225000	21037.5	0.018801	-7989	-150	-2570	2
1	3	180000	540000	27000	0.02461	-15326	-1038	-782	2
1	1	202500	436032	28516.5	0.032561	-10127	-111	-1724	1
1	0	135000	495216	26995.5	0.008019	-12824	-154	-6764	2
1	0	157500	1710000	66262.5	0.004849	-23127	-9274	-817	2
1	0	73341	135000	6750	0.018801	-9157	-146	-2926	2
1	1	121500	263686.5	17298	0.014464	-17847	-448	-3595	2
1	1	225000	1019205	31032	0.072508	-11356	-602	-335	1
1	0	63000	426645	22468.5	0.018634	-12209	-1572	-6348	2
1	1	112500	571486.5	31131	0.022625	-8292	-166	-3062	2
1	0	36000	284400	10849.5	0.018209	-22078	365243	-6748	3
1	0	157500	497520	28692	0.02461	-13989	-1350	-8112	2
1	0	112500	417024	25330.5	0.025164	-13707	365243	-7809	2
1	0	202500	343683	16663.5	0.010966	-20200	365243	-973	2
1	0	112500	117162	12433.5	0.01885	-18302	-136	-4005	2
- 1	0	91000	212940	22901 5	0.01005	9029	600	9020	2

- Identified top predictors influencing loan defaults.
- Visualization: Heatmaps showcasing correlations.



## **Insights**

#### **Key Insights:**

- 1. High-income customers are less likely to default.
- 2. Loan amounts exceeding a certain threshold increase default risk.
- 3. Data imbalance in loan approvals impacts analysis.

#### **Recommendations:**

- Implement stricter checks for high-risk applicants.
- Offer loans with adjusted terms for risky applicants (e.g., higher interest rates).
- Improve data collection processes to minimize missing values.

### **Result**

The analysis revealed actionable insights for loan approval strategies. By addressing high-risk attributes, the company can:

- Deny loans to high-risk applicants.
- Adjust loan terms based on risk levels.
- Develop policies for shorter loan tenures for specific customer segments.

#### **Drive Link**

https://drive.google.com/drive/folders/1tYHOD8kxxyXugeXr8f6fHzRcNaNcXnm\_?usp=sharing