CREDIT EDA CASE STUDY

Presenters:

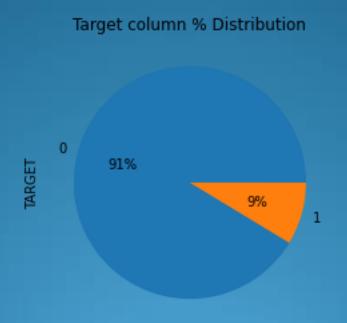
- ☐ Pradhan Nayak
- ☐ Vijay Kumar Singh

Categorical Univariate analysis for overall population vs. Target 0 vs. Target 1 (Application Dataset)

Target column % Distribution

Points to be concluded from the graph.

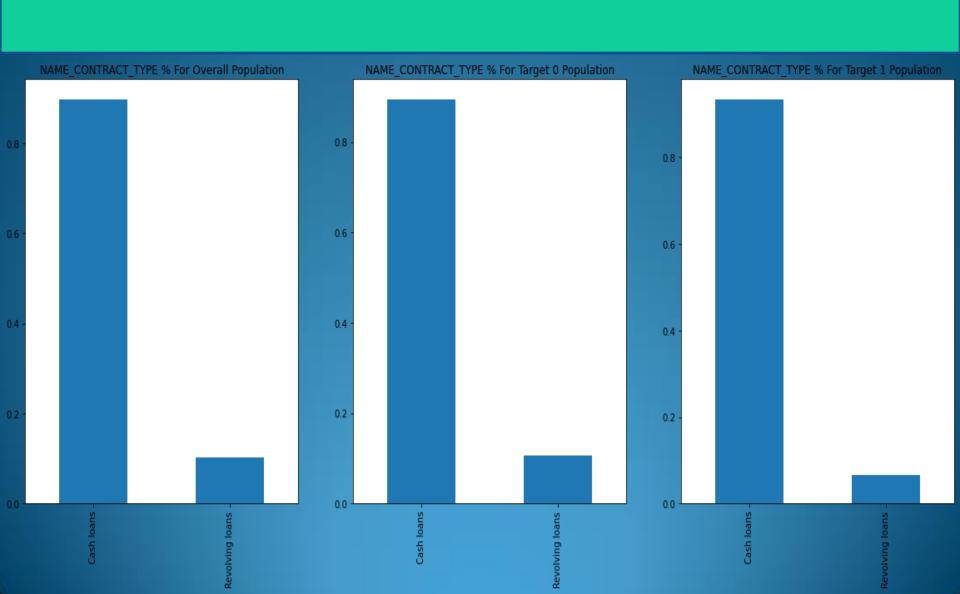
1. Of the overall population in application dataset, close to 91% applicants had no payment difficulties and remaining others had.



Contract Type % Distribution

Points to be concluded from the graph.

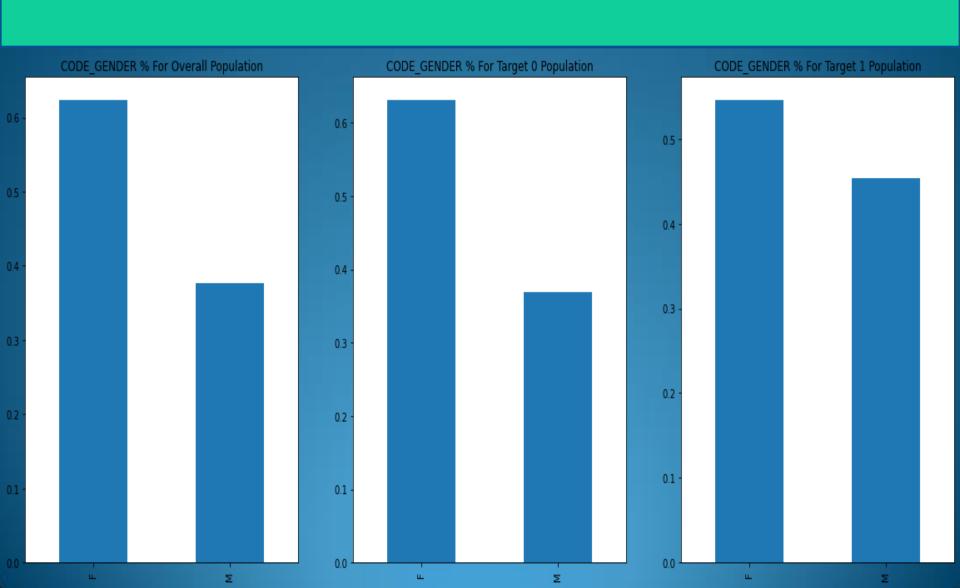
1. Credit was given mainly for cash loan



Gender % Distribution

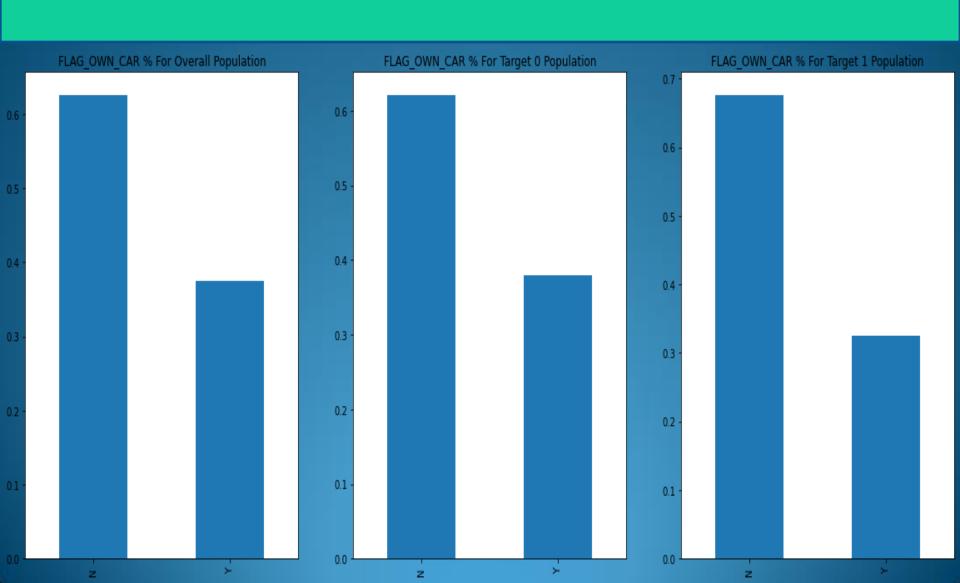
- Credit was given mainly to females

 Male population have higher chance of defaulted than female



Car Owner % Distribution

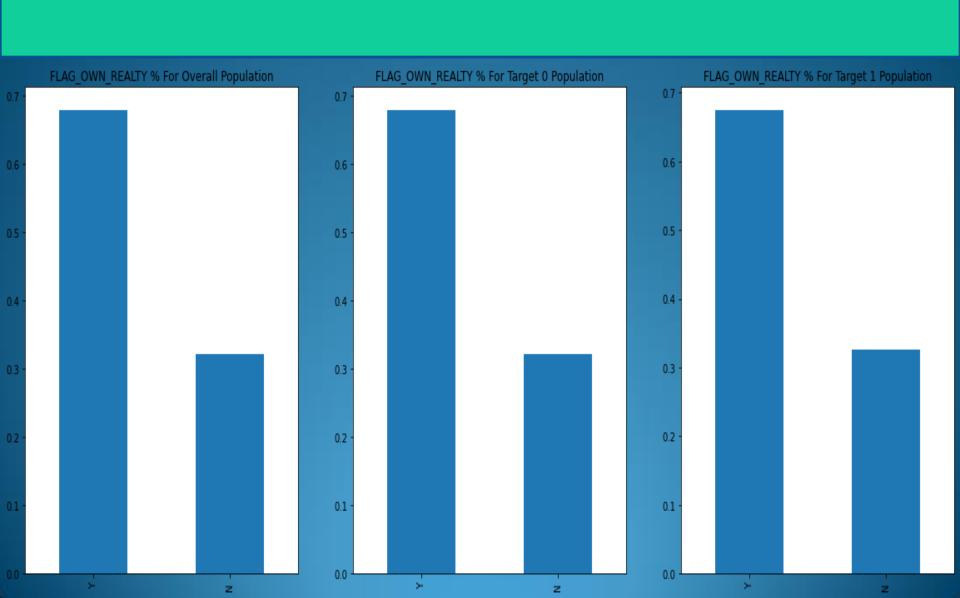
- 1. Applicants without own car represents higher % of overall credit then with own car
- 2. Applicants without own car have higher chance of defaulted than with own car



Realty Owner % Distribution

Points to be concluded from the graph.

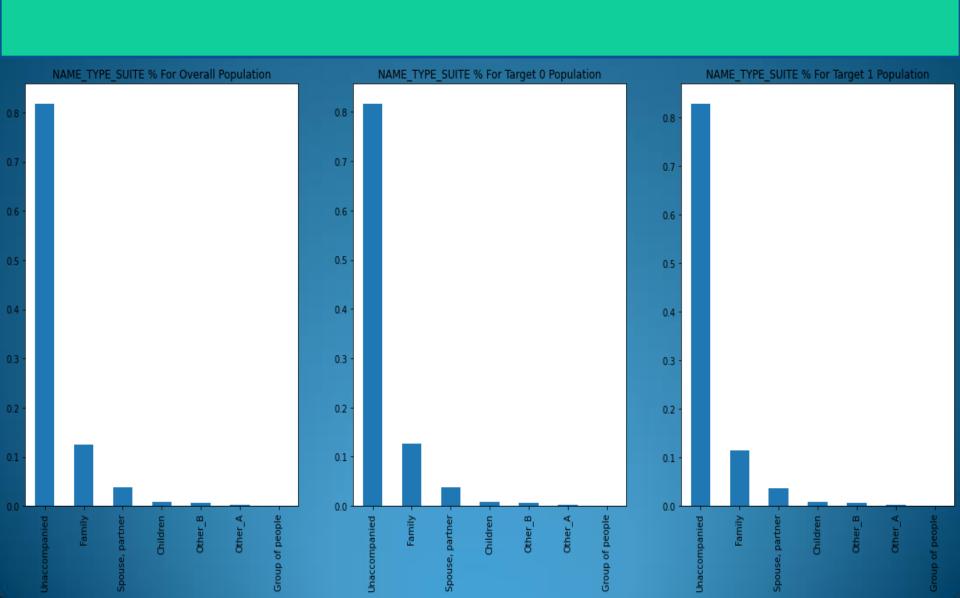
1. Applicants with realty represents higher % of overall credit then without realty



Suite Type % Distribution

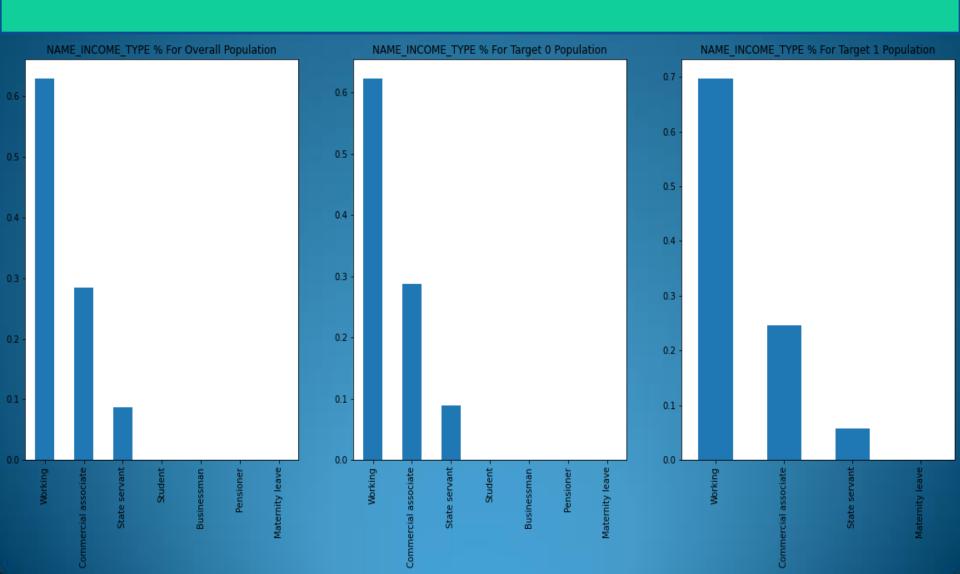
Points to be concluded from the graph.

1. "Unaccompanied" represents higher % credit



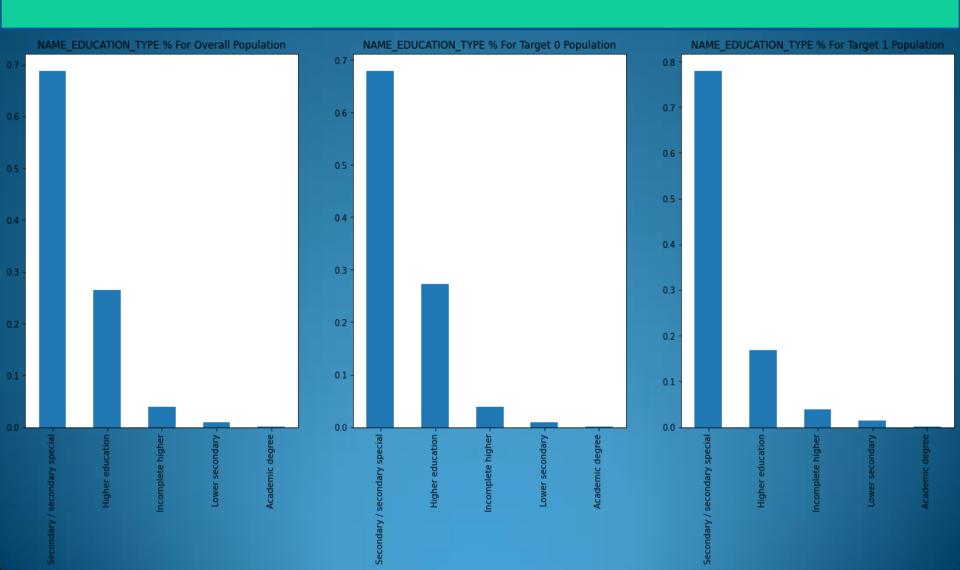
Income Type % Distribution

- 1. Applicants with "Working", "Commercial Associate" & State Servant" status represents higher % of overall credit then others
- 2. Applicants with "Student", "Businessman" & "Pensioner" status have less chance to being defaulted



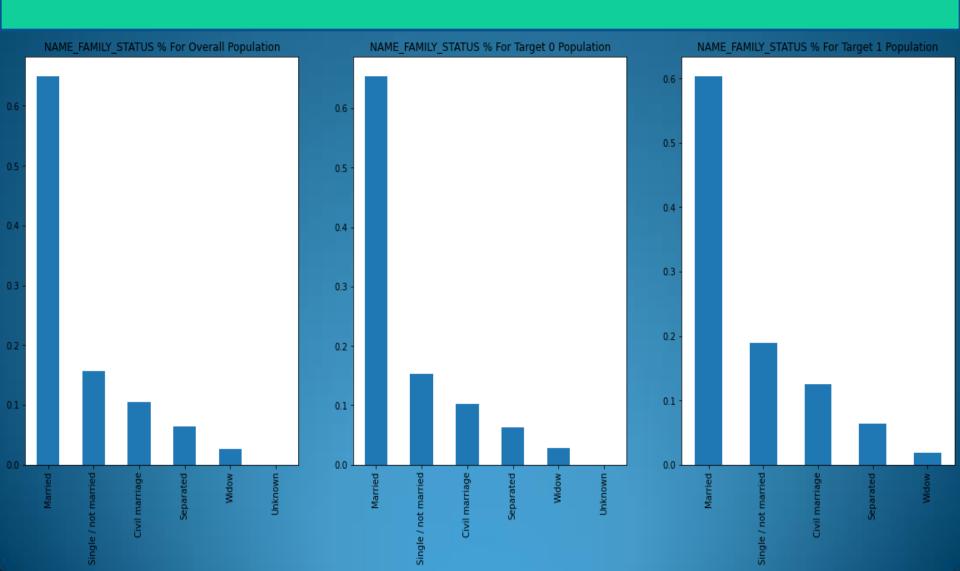
Education Type % Distribution

- 1. Applicants with "Secondary/Secondary Special" & "Higher Education" status represents higher % of overall credit then others
- 2. Applicants with "Secondary/Secondary Special" have higher chance to being defaulted compare to other



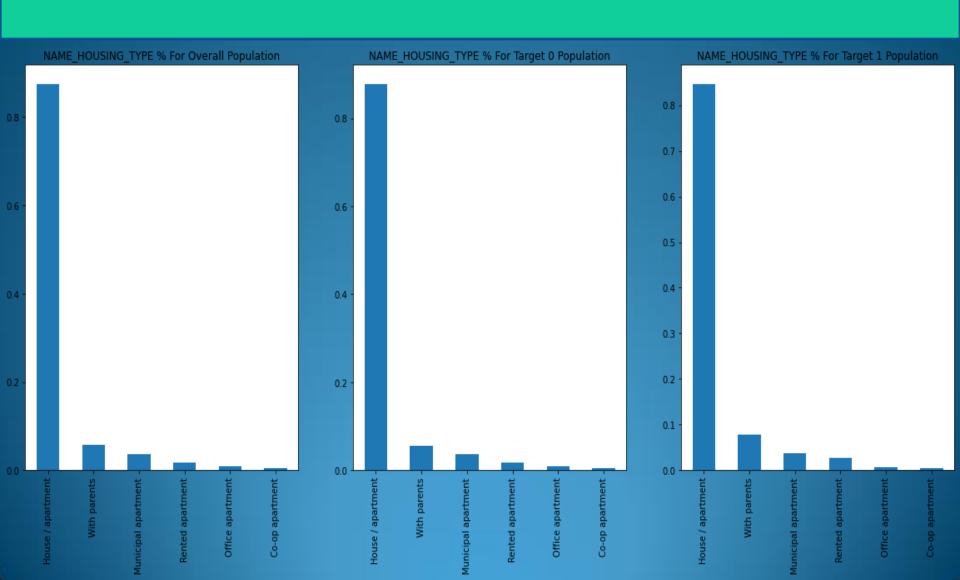
Family Status % Distribution

- 1. Applicants with "Married" & "Single/Not Married" status represents higher % of overall credit then others
- 2. Applicants with "Married" status have lesser chance to being defaulted compare to other



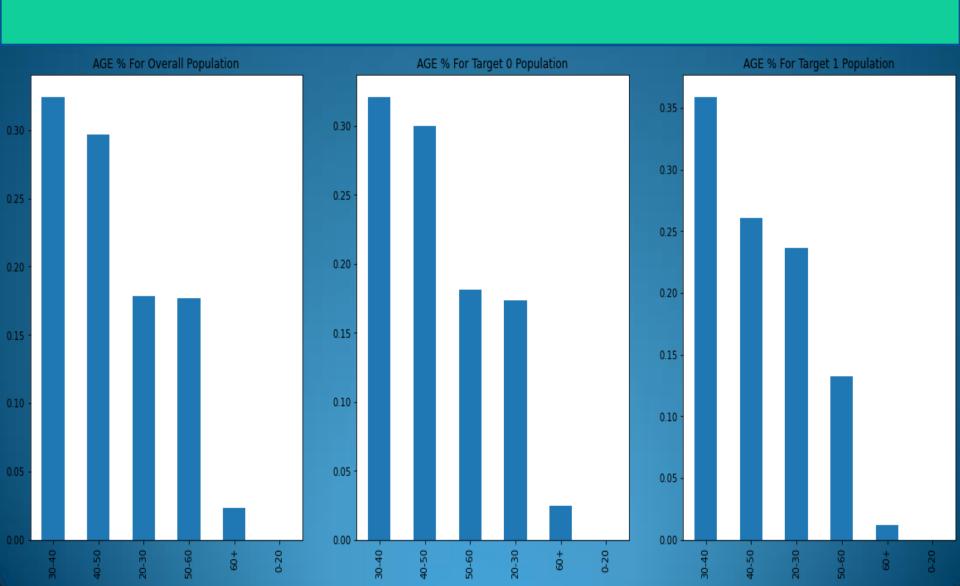
Housing Type % Distribution

- 1. Applicants with "House/Apartment" status represents higher % of overall credit then others
- 2. Applicants "With Parent" status have higher chance to being defaulted compare to other



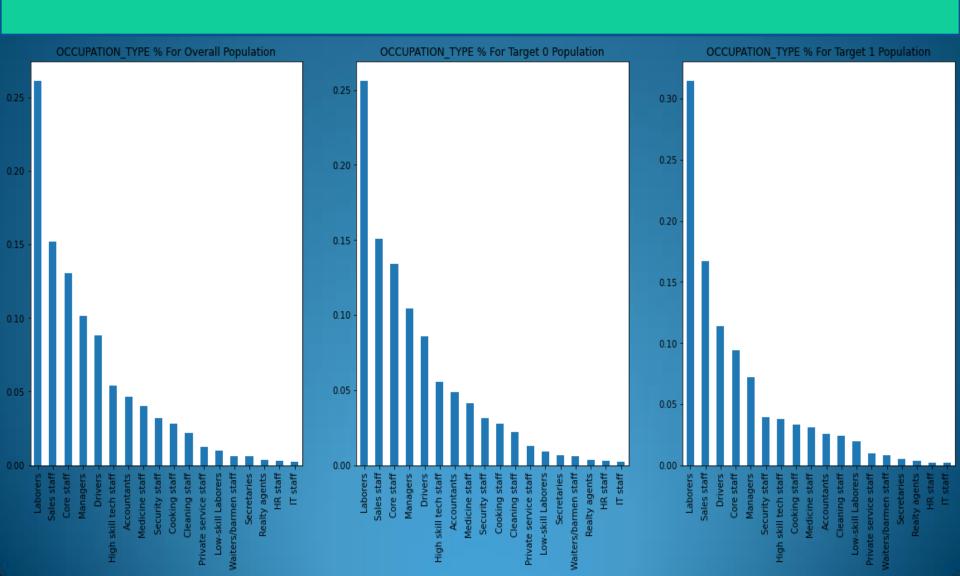
Age % Distribution

- 1. The max credit is given to applicants above age 30
- 2. The age group 20-30 have higher chance of default than others



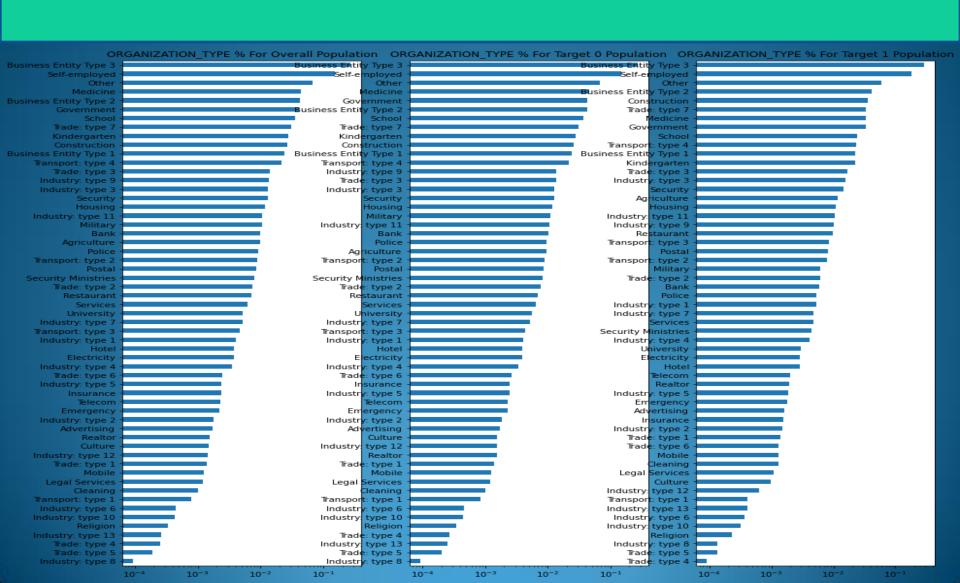
Occupation Type % Distribution

- 1. "Laborers", "Sales Staff", "Core Staff", "Managers" & "Drivers" represents higher % of overall credit then others
- 2. "Laborers" & "Sales Staff" have higher chance to being defaulted compare to other



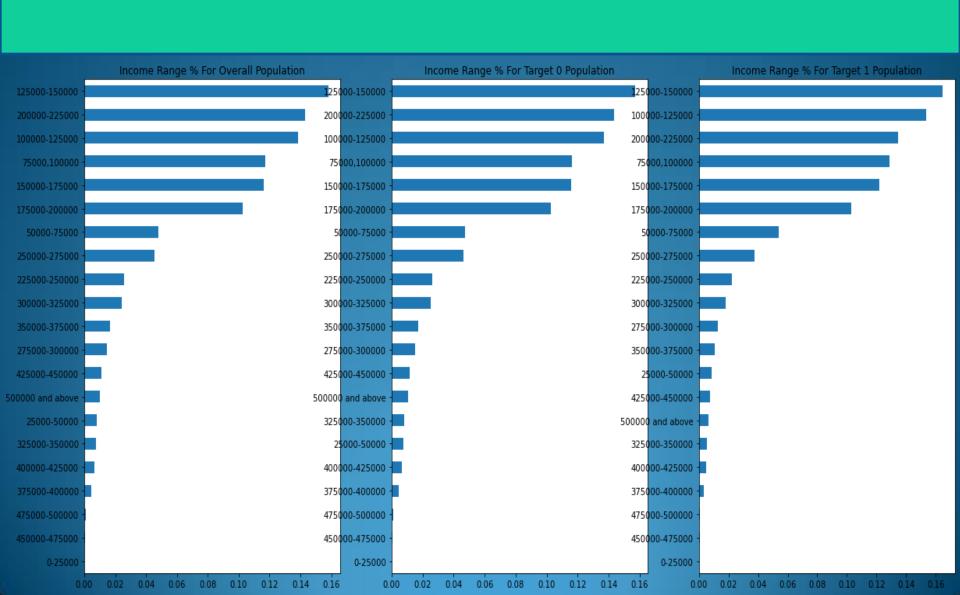
Organization Type % Distribution

- 1. Applicants with "Business Entity Type 3" status represents higher % of overall credit then others
- 2. Applicants "Business Entity Type 3" & "Self-employed" status have higher chance of being defaulted compare to other



Income Range % Distribution

- 1. Applicants with income range "100000-200000" has maximum credit
- 2. Applicants with income range 400000 and above has less credit



Credit Range % Distribution

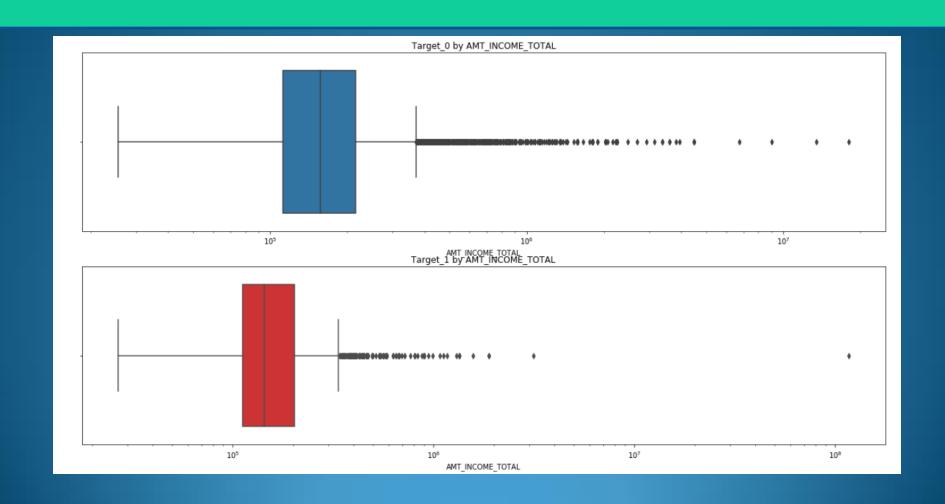
- 1. The max credit is of the range of 900000 and above
- 2. The credit range "500000-550000" have higher chance of default than others



Checking Outliers For Target0 vs Target1

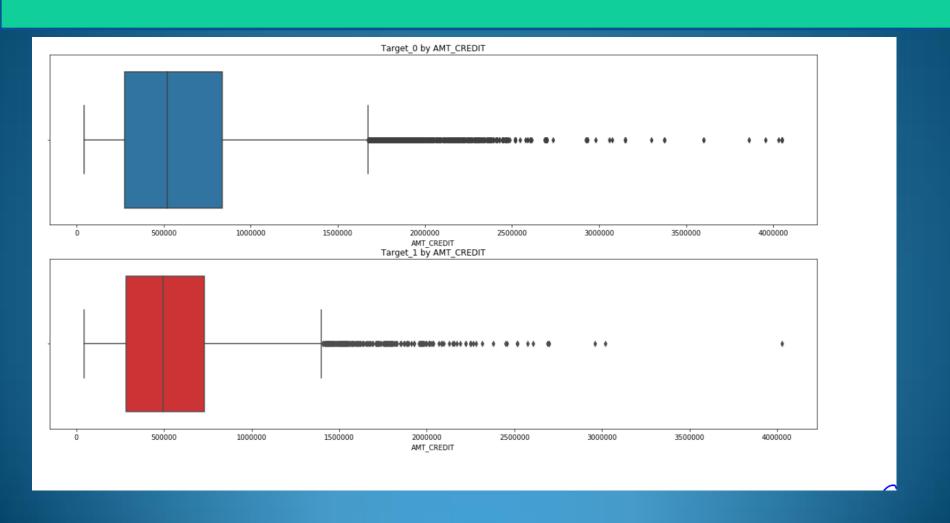
Target 0 vs Target 1: Amount Income Total

- 1. Some outliers are noticed for Target 1
- 2. Target 1 has the maximum income amount value compared to Target 0.
- 3. For Target 1: most of the clients income are in between median and third quartile region.



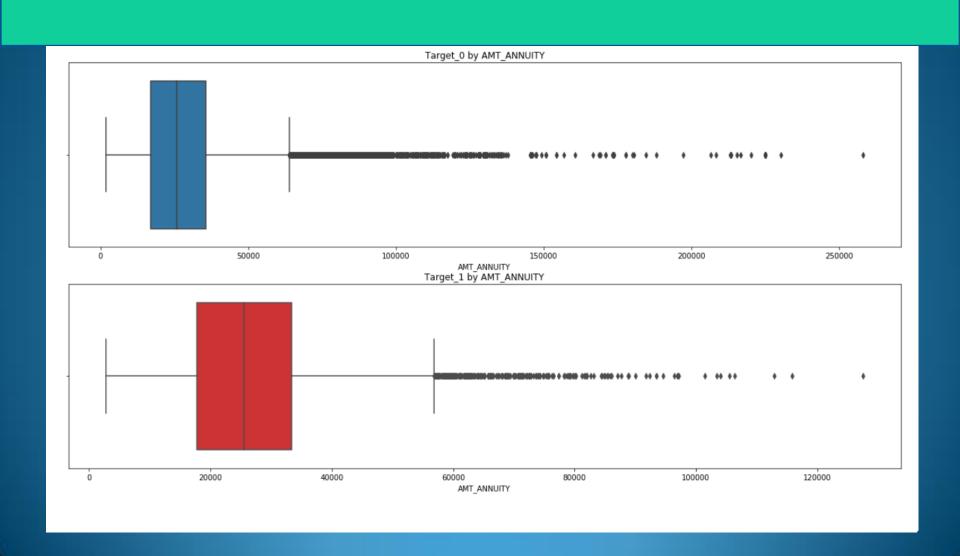
Target 0 vs Target 1 : Amount Credit

- 1. Some outliers are noticed for Target 1.
- 2.For Target 0 and Target 1: most of the credit amounts of clients are present in between maximum to median compared to values between minimum and median.
- 3. For Target 0 and Target 1: most of the clients have been given higher credit amount



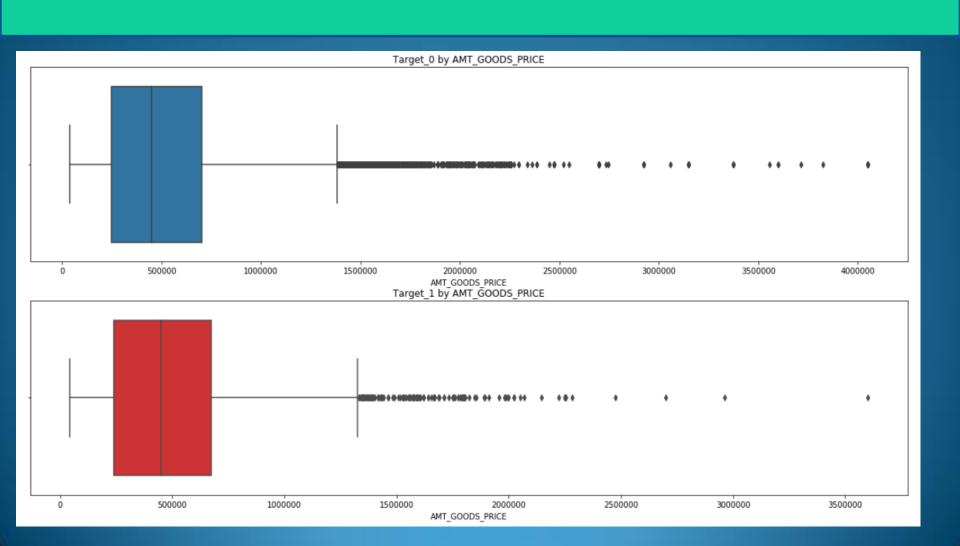
Target 0 vs Target 1: Amount Annuity

- 1. Some outliers are noticed for Target 1 and Target 0
- 2. Target 0 has a higher AMT ANNUITY' value compared to Target 1.
- 3. For Target 1 and Target 0: most of the clients amount annuity lies between maximum to median compared to values between minimum to median.



Target 0 vs Target 1: Amount Goods Price

- 1. Some outliers are noticed for Target 1 and Target 0.
- 2. For Target 1 and Target 0: most of the clients amount good's price lies between maximum to median compared to values between minimum to median.
- 3. Target 0 has received the maximum amount value for the goods compared to Target 1.



Correlation for numerical columns for Target 0

Correlation for numerical columns for Target 0.

- 1. Amount annuity has a higher correlation with respect to Amount credit meaning clients higher credit amount tend to take higher EMI.
- 2.client's permanent address does not match contact address(REG_REGION_NOT_WORK_REGION) has a higher correlation with client's contact address does not match work address(LIVE_REGION_NOT_WORK_REGION).
- 3.client's permanent address does not match work address(REG_CITY_NOT_WORK_CITY) has a higher correlation with client's contact address does not match work address(LIVE_CITY_NOT_WORK_CITY).
- 4. Amount annuity has a higher correlation with respect to Amount Goods price meaning clients with higher good's amount tend to take higher EMI.
- 5. Amount credit has a higher correlation with respect to Amount Goods price meaning clients with higher credit amount tend to take higher consumer good's amount.
- 6.Region rating client is inversely proportional to the client lives in more populated region.
- 7. Region rating client has a higher correlation to region where client lives with taking city into account.
- 8.family members clients have a higher correlation to Number of children the client has.

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CNT CHILDREN

AMT CREDIT

DAYS_REGISTRATION

REG REGION NOT LIVE REGION -

REG_REGION_NOT_WORK_REGION

LIVE REGION NOT WORK REGION -

REG CITY NOT LIVE CITY -

REG CITY NOT WORK CITY -

LIVE_CITY_NOT_WORK_CITY

REGION_RATING_CLIENT

REGION_RATING_CLIENT_W_CITY -

- 0.2
- 0.0
0.2
0.4

Correlation for numerical columns for Target 1

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- 1.Amount annuity has a higher correlation with respect to Amount credit meaning clients higher credit amount tend to take higher EMI.
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- 3.client's permanent address does not match work address(REG_CITY_NOT_WORK_CITY) has a higher correlation with client's contact address does not match work address(LIVE_CITY_NOT_WORK_CITY).
- 4.Amount annuity has a higher correlation with respect to Amount Goods price meaning clients with higher good's amount tend to take higher EMI.
- 5. Amount credit has a higher correlation with respect to Amount Goods price meaning clients with higher credit amount tend to take higher consumer good's price.
- 6. Region rating client is inversely proportional to the client lives in more populated region.
- 7. Region rating client has a higher correlation to region where client lives with taking city into account.
- 8.family members clients have a higher correlation to Number of children the client has.
- 9.region where client lives with taking city into account is inversely proportional to the client lives in more populated region.

CNT_CHILDREN -	1	0.0019	-0.0021	0.016	-0.0077	-0.032	0.18	0.033	0.13		-0.039	-0.032	-0.039	-0.028	-0.019	-0.00088	0.016	0.89	0.047	0.049
AMT_INCOME_TOTAL	0.0019	1	0.036	0.043	0.036	0.0085	-0.0078	-3.9e-05	-0.004	-0.0089	0.013	0.007	0.013	0.012	-0.0037	-0.0069	-0.0044	0.0037	-0.021	-0.022
AMT_CREDIT -	-0.0021	0.036	1	0.75	0.98	0.069	-0.19	-0.11	-0.033	-0.062	0.029	0.02	0.036	0.036	-0.035	-0.041	-0.018	0.045	-0.061	-0.074
AMT_ANNUITY -	0.016	0.043	0.75	1	0.75	0.074	-0.09	-0.049	0.014	-0.044	0.023	0.033	0.063	0.062	-0.014	-0.02	-0.0068	0.055	-0.076	-0.092
AMT_GOODS_PRICE -	-0.0077	0.036	0.98	0.75	1	0.076	-0.19	-0.11	-0.032	-0.065	0.042	0.024	0.039	0.039	-0.036	-0.042	-0.017	0.042	-0.069	-0.081
REGION_POPULATION_RELATIVE -	-0.032	0.0085	0.069	0.074	0.076	1	-0.05	-0.016	-0.056	-0.012	0.14	-0.022	0.025	0.051	-0.06	-0.054	-0.019	-0.031	-0.44	-0.45
DAYS_BIRTH -	0.18	-0.0078	-0.19	-0.09	-0.19	-0.05	1	0.31	0.24	0.14	0.035	0.045	0.022	0.00048	0.14	0.096	0.012	0.11	0.049	0.049
DAYS_EMPLOYED =	0.033	-3.9e-05	-0.11	-0.049	-0.11	-0.016	0.31	1	0.14	0.097	9.2e-05	0.059	0.075	0.055	0.11	0.13	0.071	-0.0016	0.0069	0.0038
days_registration -	0.13	-0.004	-0.033	0.014	-0.032	-0.056	0.24	0.14	1	0.05	-0.048	0.011	0.0046	0.00077	0.029	0.049	0.029	0.12	0.11	0.11
DAYS_ID_PUBLISH -	-0.09	-0.0089	-0.062	-0.044	-0.065	-0.012	0.14	0.097	0.05	1	0.0088	0.025	0.013	0.0027	0.048	0.016	-0.015	-0.089	0.0048	0.0014
HOUR_APPR_PROCESS_START -	-0.039	0.013	0.029	0.023	0.042	0.14	0.035	9.2e-05	-0.048	0.0088	1	0.055	0.067	0.052	0.0051	0.0053	0.0033	-0.04	-0.29	-0.27
REG_REGION_NOT_LIVE_REGION -	-0.032	0.007	0.02	0.033	0.024	-0.022	0.045	0.059	0.011	0.025	0.055	1	0.51	0.068	0.32	0.15	-0.014	-0.034	-0.034	-0.031
REG_REGION_NOT_WORK_REGION -	-0.039	0.013	0.036	0.063	0.039	0.025	0.022	0.075	0.0046	0.013	0.067	0.51	1	0.85	0.14	0.22	0.18	-0.046		-0.11
LIVE_REGION_NOT_WORK_REGION -	-0.028	0.012	0.036	0.062	0.039	0.051	0.00048	0.055	0.00077	0.0027	0.052	0.068	0.85	1	-0.007	0.17	0.23	-0.033		-0.12
REG_CITY_NOT_LIVE_CITY -	-0.019	-0.0037	-0.035	-0.014	-0.036	-0.06	0.14	0.11	0.029	0.048	0.0051	0.32	0.14	-0.007	1	0.48	-0.029	-0.024	0.044	0.055
REG_CITY_NOT_WORK_CITY -	-0.00088	-0.0069	-0.041	-0.02	-0.042	-0.054	0.096	0.13	0.049	0.016	0.0053	0.15	0.22	0.17	0.48	1	0.77	0.00069	0.012	0.034
LIVE_CITY_NOT_WORK_CITY =	0.016	-0.0044	-0.018	-0.0068	-0.017	-0.019	0.012	0.071	0.029	-0.015	0.0033	-0.014	0.18	0.23	-0.029	0.77	1	0.026	-0.023	-0.0063
CNT_FAM_MEMBERS -	0.89	0.0037	0.045	0.055	0.042	-0.031	0.11	-0.0016	0.12	-0.089	-0.04	-0.034	-0.046	-0.033	-0.024	0.00069	0.026	1	0.047	0.051
	0.047	-0.021	-0.061	-0.076	-0.069	-0.44	0.049	0.0069	0.11	0.0048	-0.29	-0.034	-0.12	-0.12	0.044	0.012	-0.023	0.047	1	0.96
REGION_RATING_CLIENT -																				
REGION_RATING_CLIENT_W_CITY -	0.049	-0.022	-0.074	-0.092	-0.081	-0.45	0.049	0.0038	0.11	0.0014	-0.27	-0.031	-0.11	-0.12	0.055	0.034	-0.0063	0.051	0.96	1
	ONT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	GION_POPULATION_RELATIVE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	HOUR_APPR_PROCESS_START	SG_REGION_NOT_LIVE_REGION	REGION_NOT_WORK_REGION	REGION_NOT_WORK_REGION	REG_CITY_NOT_LIVE_CITY	REG_CITY_NOT_WORK_CITY	LIVE_CITY_NOT_WORK_CITY	CNT_FAM_MEMBERS	REGION_RATING_CLIENT	SION_RATING_CLIENT_W_CTTY

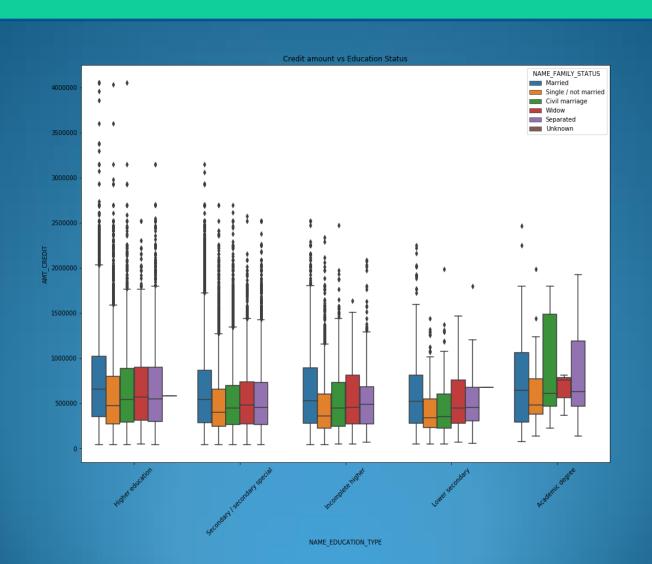
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Bivariate Analysis for Target 0

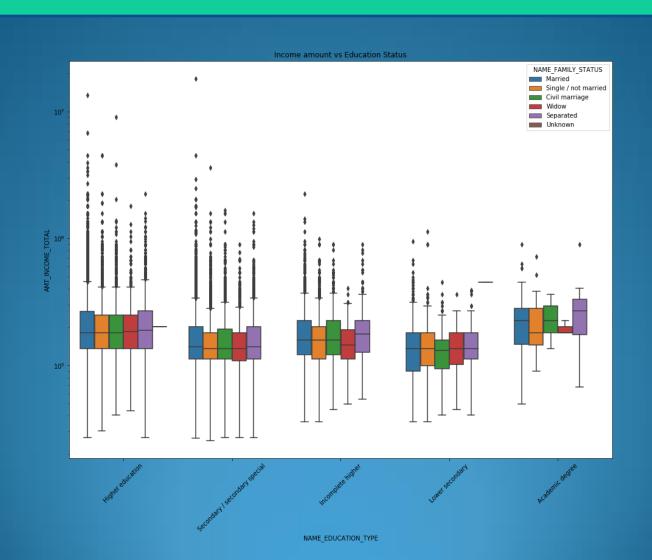
Target 0 : Credit amount vs Education type

- 1. Family status of 'civil marriage', 'marriage' and 'separated' of Academic degree education are having higher number of credits than others.
- 2.higher education of family status of 'marriage', 'single', 'civil marriage', separated are having more outliers.
- 3. Civil marriage for Academic degree is having most of the credits in the third quartile.



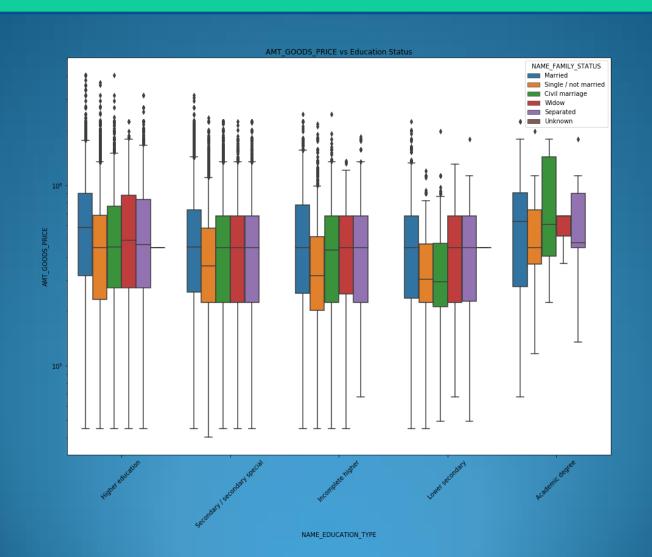
Target 0: Income amount vs Education Type

- 1.for Education type 'Higher education' the income amount is mostly equal with family status. It does contain many outliers.
- 2.Less outlier are having for Academic degree but there income amount is little higher that Higher education.
- 3.Lower secondary of civil marriage family status are have less income amount than others.



Target 0: Amount Goods Price vs Education Type

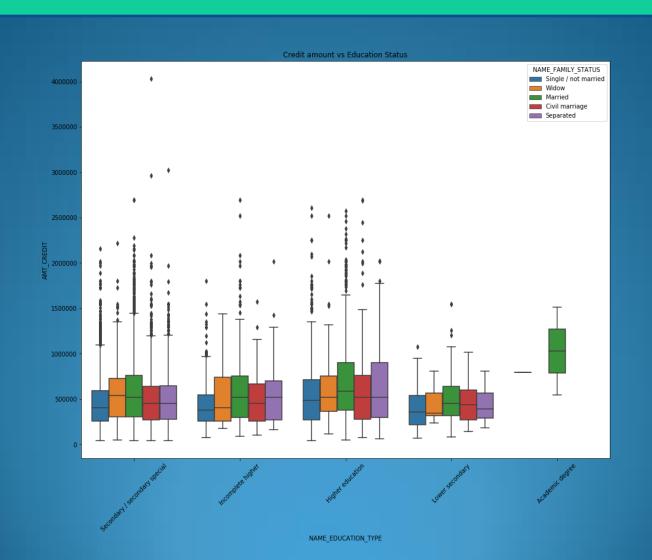
- 1. For Education type Higher education tend to have more outliers.
- 2. For education type Academic degree and family status of civil marriage tend to have opted for higher amount Good's price.
- 3. For education type Academic degree all family status seem to have less number of outliers.



Bivariate Analysis for Target 1

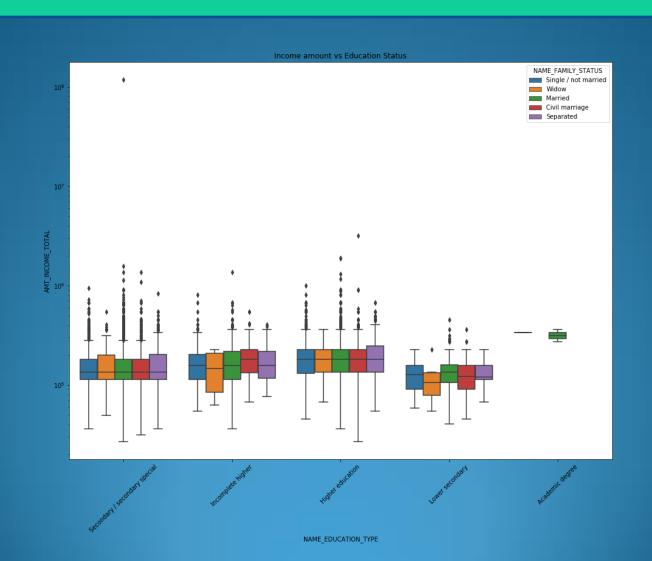
Target 1: Credit amount vs Education type

- 1. From education type Academic degree the family status with single, widow, civil marriage, separated have not opted for Amount credit.
- 2. Most of the outliers are from Education type 'Higher education' and 'Secondary'.
- 3.married for Academic degree is having most of the credits compared to other education type and family status.



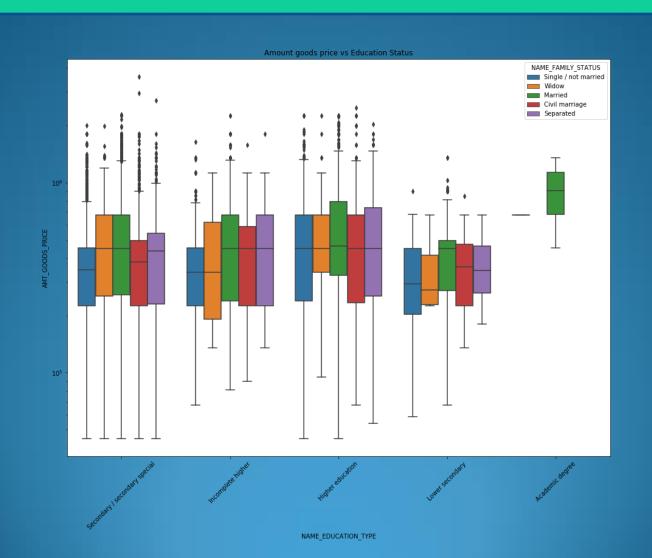
Target 1: Income amount vs Education Type

- 1. For Education type 'Higher education' seem to have the income amount is mostly equal with family status.
- 2.Less outlier are having for Academic degree but there income amount for family status 'married' is little higher that Higher education.
- 3.Lower secondary are have less income amount than others.



Target 1: Amount Goods Price vs Education Type

- 1. From above boxplot for Education type 'Higher education' and "secondary special seem to have more outliers than the rest."
- 2.Less outlier are having for Academic degree but there income amount for family status 'married' is higher than rest of the education type and rest of the family status.
- 3.Lower secondary are have less income amount than others.

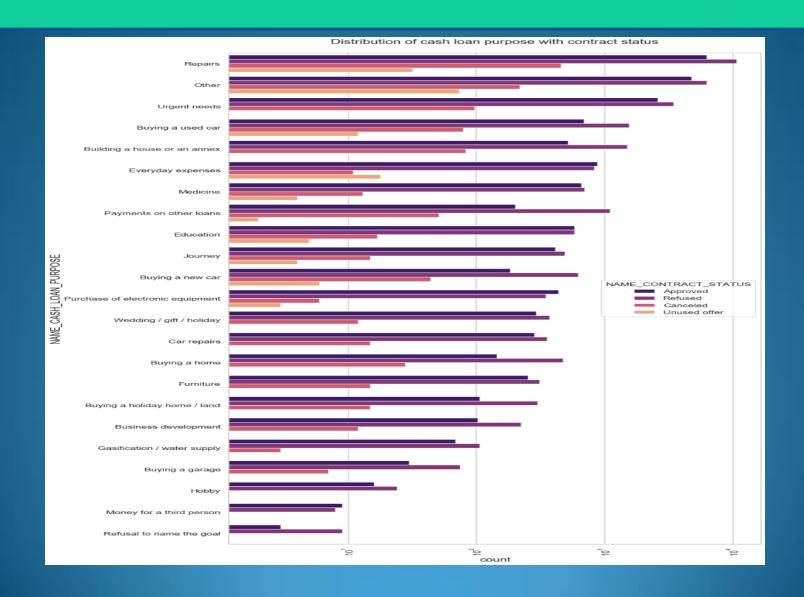


Previous application data

Univariate analysis on combined dataset

Distribution of cash loan purpose with contract status

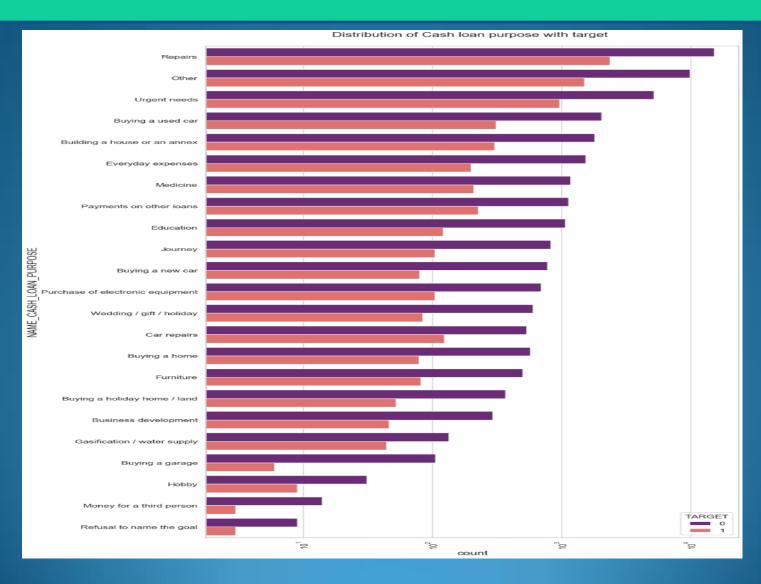
- 1. Most refusal are for the loans came from purpose 'repairs'.
- 2. For education purposes we have equal number of approves and refusal.
- 3. Paying other loans and buying a new car is having significant higher refusals than approvals.
- 4. Most of the cancellation are for the loan purpose repairs'.



Distribution of cash loan purpose with Target column

- 1.Loan purposes with 'Repairs' are facing more difficulties in payment on time.
- 2.Following Loan purposes that are 'Buying a garage', 'Business development', 'Buying land', 'Buying a new car' and 'Education' where loan payment on time is significant higher than facing difficulties.

Hence the Banks can focus on these purposes for which the client is having significant higher chances of repaying the loan amount on time.

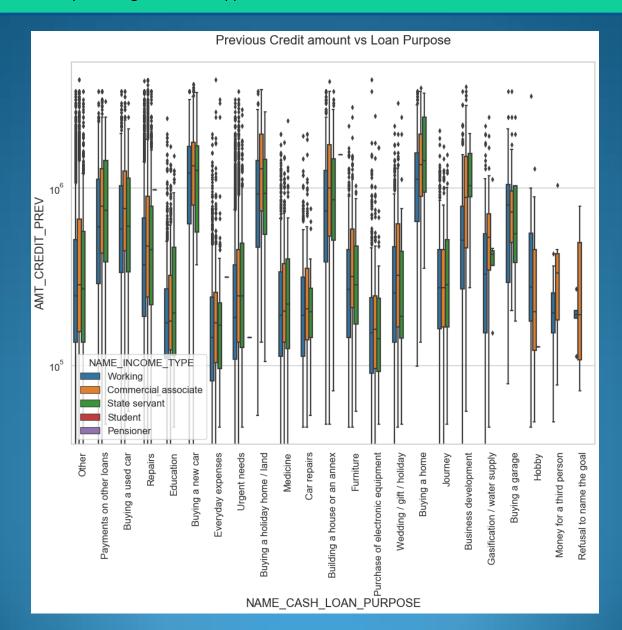


Previous application data

Bivariate analysis on combined dataset

Previous Credit amount vs Loan Purpose

- 1. The credit amount of Loan purposes like 'Buying a new car', 'Buying a land', 'Buying a home' and 'Building a house' is higher.
- 2. Income type of state servants have a significant amount of credit applied
- 3. Money for third person or a Hobby is having less credits applied for.



Previous application data

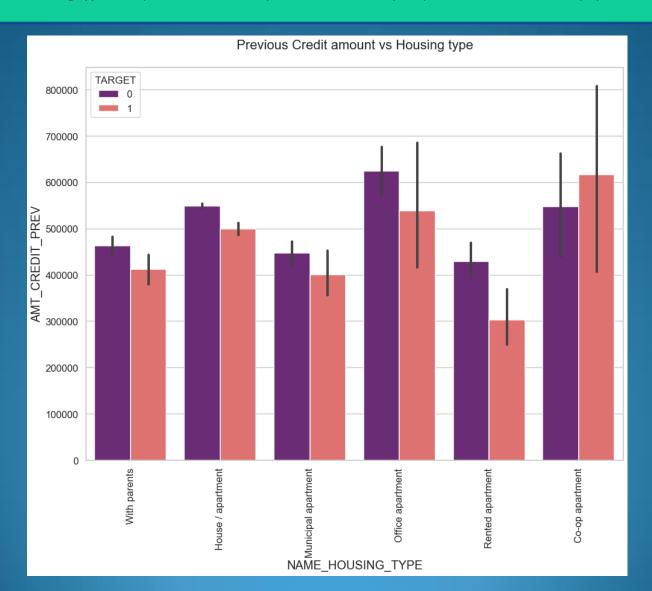
Categorical vs Numeric Bivariate Analysis

Previous Credit amount vs Housing type

- 1.Office apartment is having higher credit of target 0
- 2.Co-op apartment is having higher credit of target 1.

From the above inferences we can conclude that bank should avoid giving loans to the housing type of co-op apartment as they are having difficulties in payment.

Bank can focus mostly on housing type with parents or House apartment or municipal apartment for successful payments.



Final Inferences from the Case study

- 1. Banks should focus more on contract type 'pensioner', 'Businessman' and 'Student' with housing 'type other than 'Co-op apartment' for successful payments.
- 2. Banks should focus less on income type 'Working' as they are having most number of unsuccessful payments.
- 3. Also bank should avoid giving loans with purpose 'Repair' as it is having a higher number of unsuccessful payments on time.
- 4. Get as much as clients from housing type 'With Rented apartment' as they are having least number of unsuccessful payments.