

The background is a solid blue gradient. Overlaid on this are numerous thin, white, curved lines that flow from the left side towards the right, creating a sense of movement and depth. These lines vary in density and curvature, some forming gentle waves while others are more pronounced.

# CREDIT EDA CASE STUDY

BY SANTHOSH GUDUR



The background features a solid blue gradient with a series of thin, white, wavy lines that flow from the left side towards the right, creating a sense of movement and depth. These lines are more densely packed in some areas, forming a wave-like pattern that peaks in the upper right quadrant.

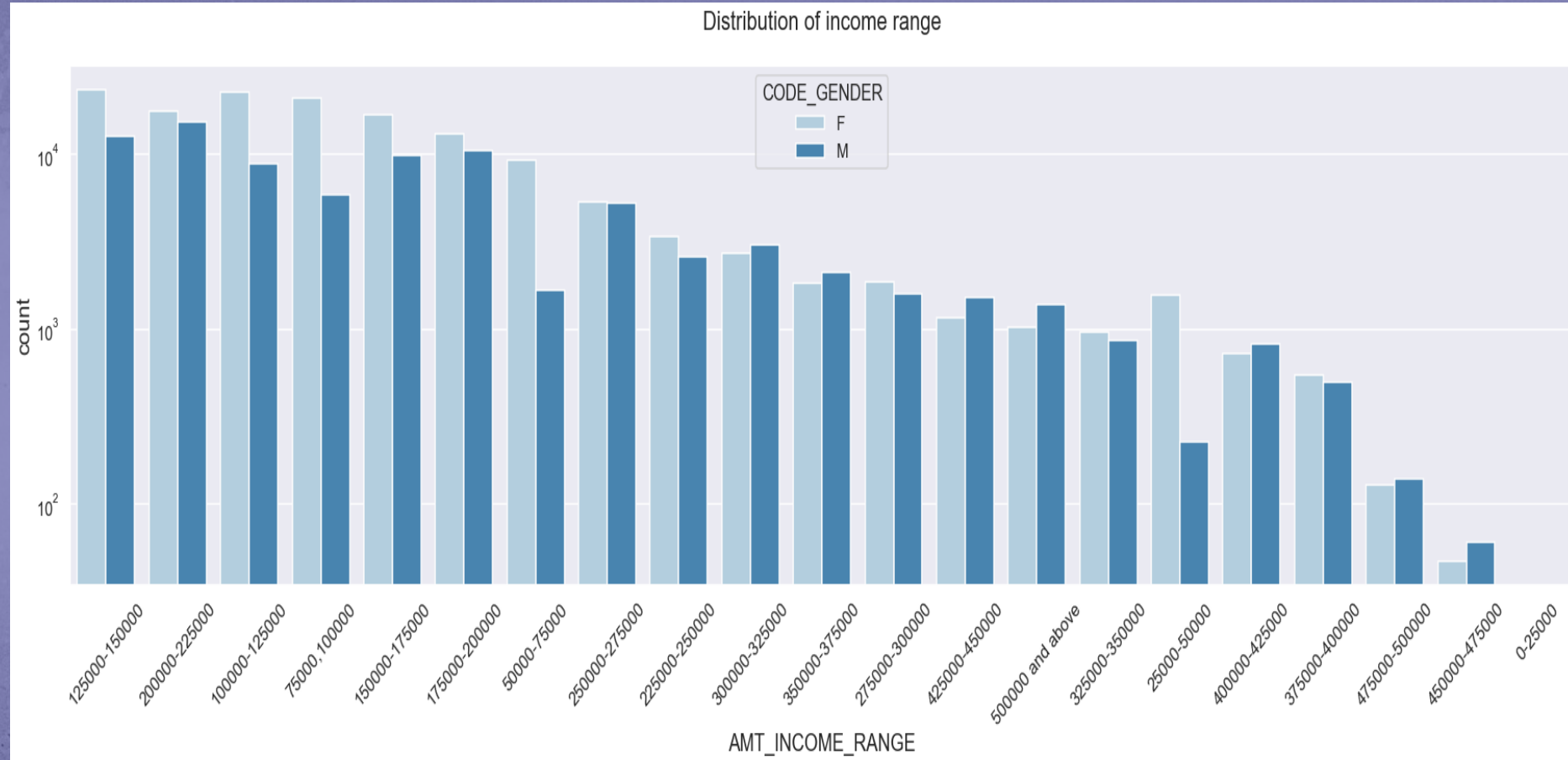
# CATEGORICAL UNIVARIATE ANALYSIS FOR TARGET 0



# DISTRIBUTION OF INCOME RANGE

Observations from the graph:

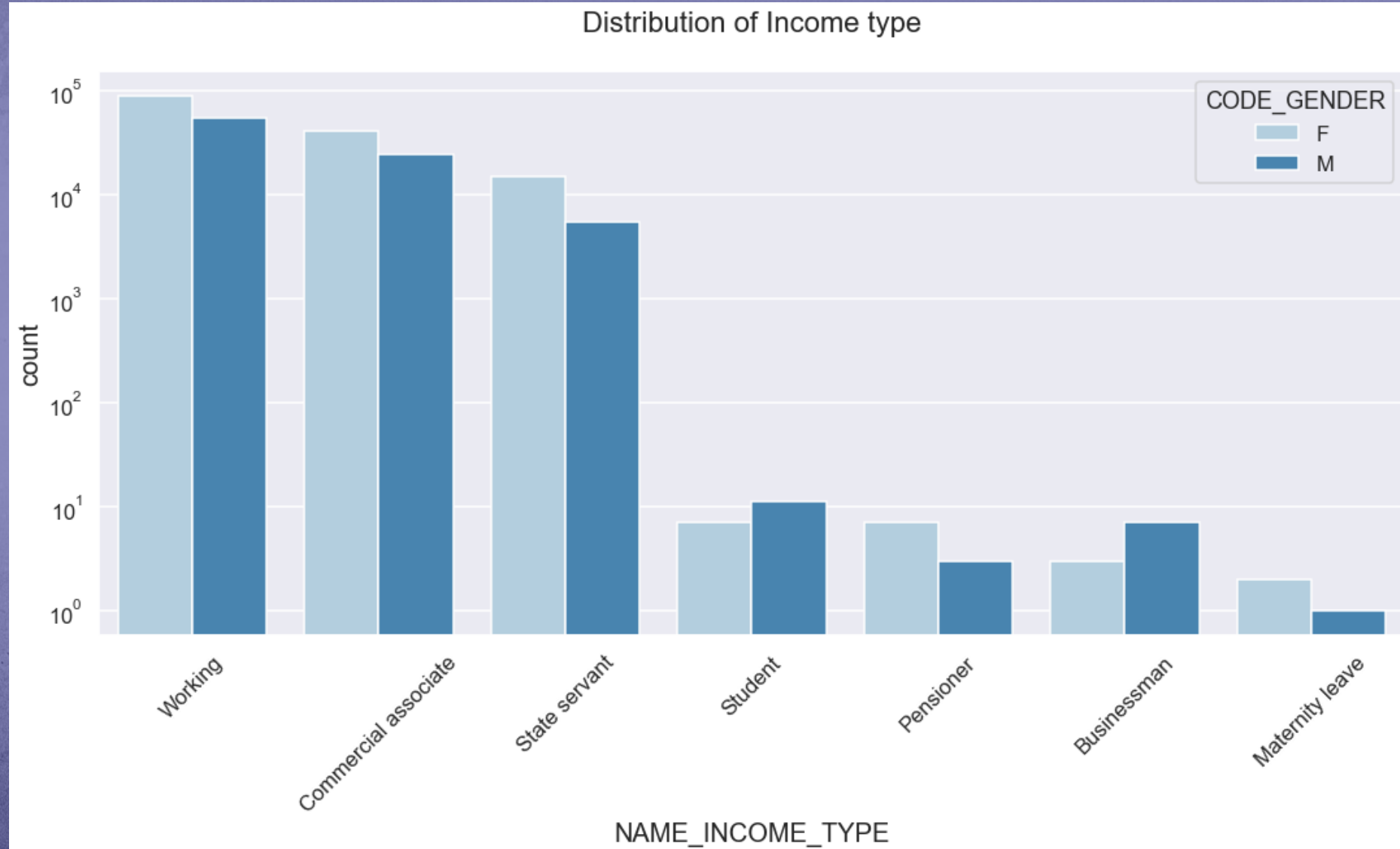
- Female counts are higher than male.
- Income range from 100000 to 200000 is having more number of credits.
- Females are more than male in having credits for the range 100000 to 200000
- Count is very less for income range 400000 and above.



# DISTRIBUTION OF INCOME TYPE

Observations from the graph:

- For income type 'working', 'commercial associate', and 'State Servant' the number of credits are higher than others.
- For income type 'working', 'commercial associate', and 'State Servant', Females are having more number of credits than male.
- Credits are less for income type 'student', 'pensioner', 'Businessman' and 'Maternity leave'.

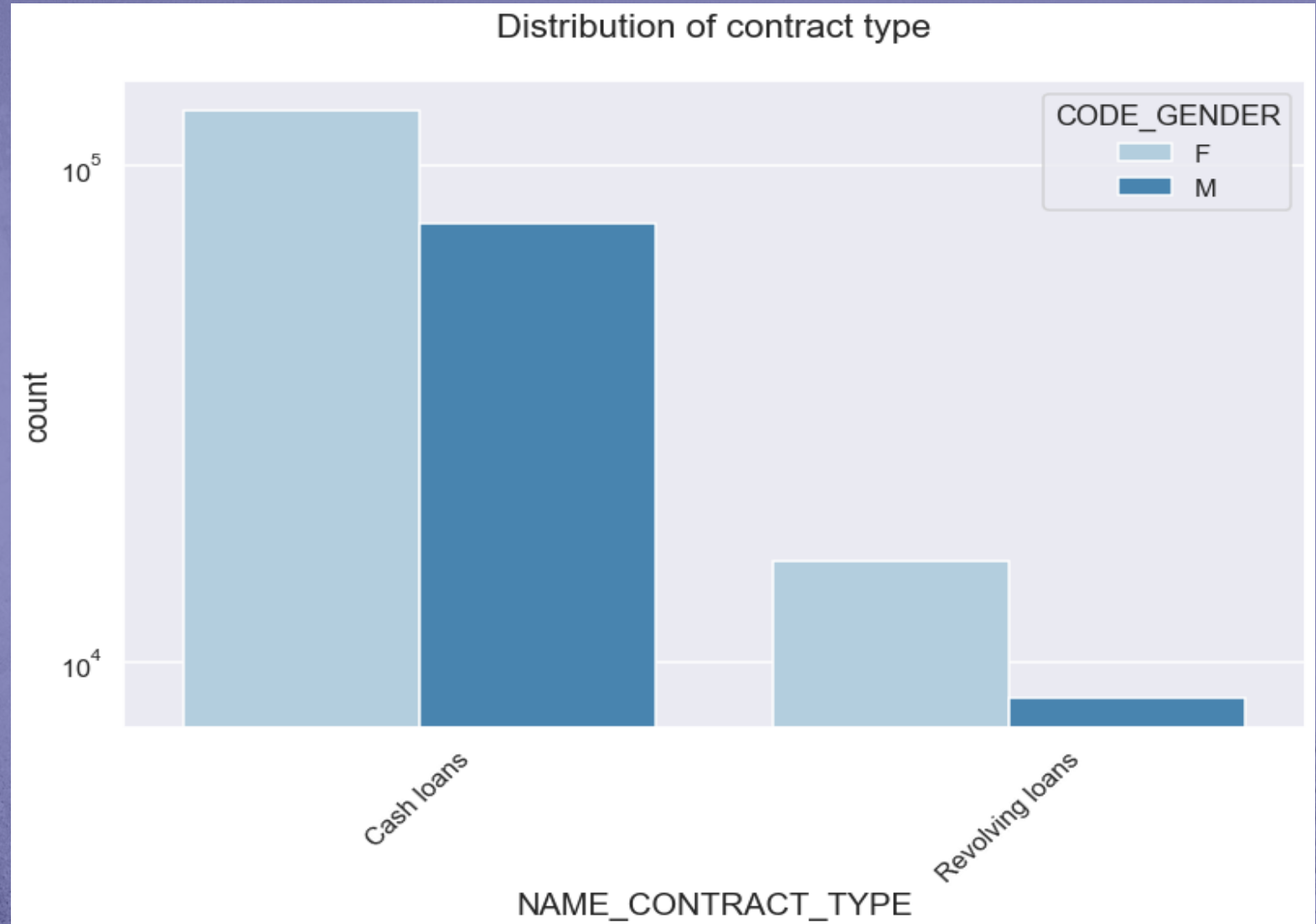




# DISTRIBUTION OF CONTRACT TYPE

Observations from the graph:

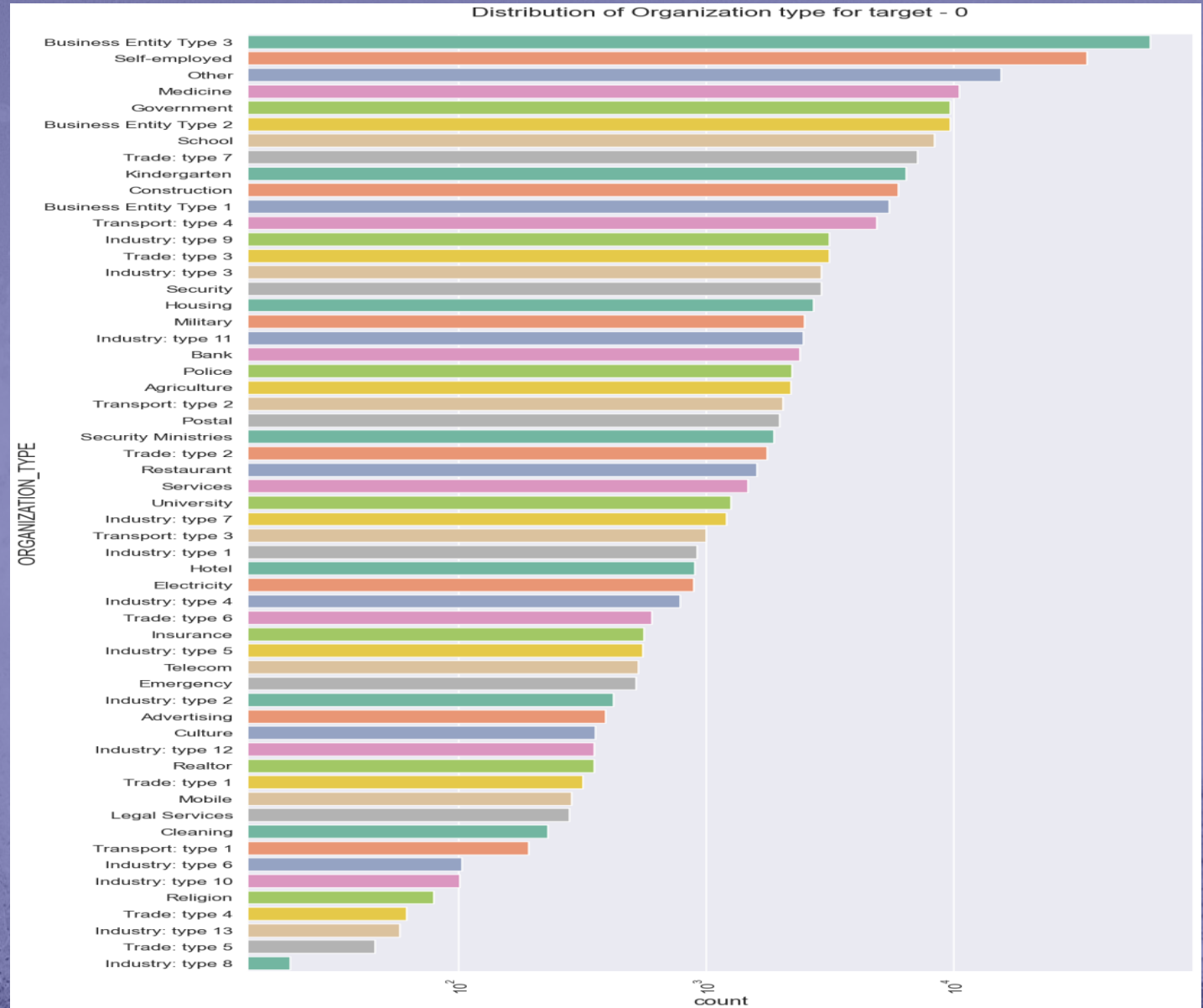
- For contract type 'cash loans' is having higher number of credits than 'Revolving loans' contract type.
- For this also Female is leading for applying credits.



# DISTRIBUTION OF ORGANIZATION TYPE

Observations from the graph:

- Clients which have applied for credits are from most of the organization type 'Business entity Type 3' , 'Self employed', 'Other' , 'Medicine' and 'Government'.
- Less clients are from Industry type 8, type 6, type 10, type 13, religion and trade type 5, type 4.





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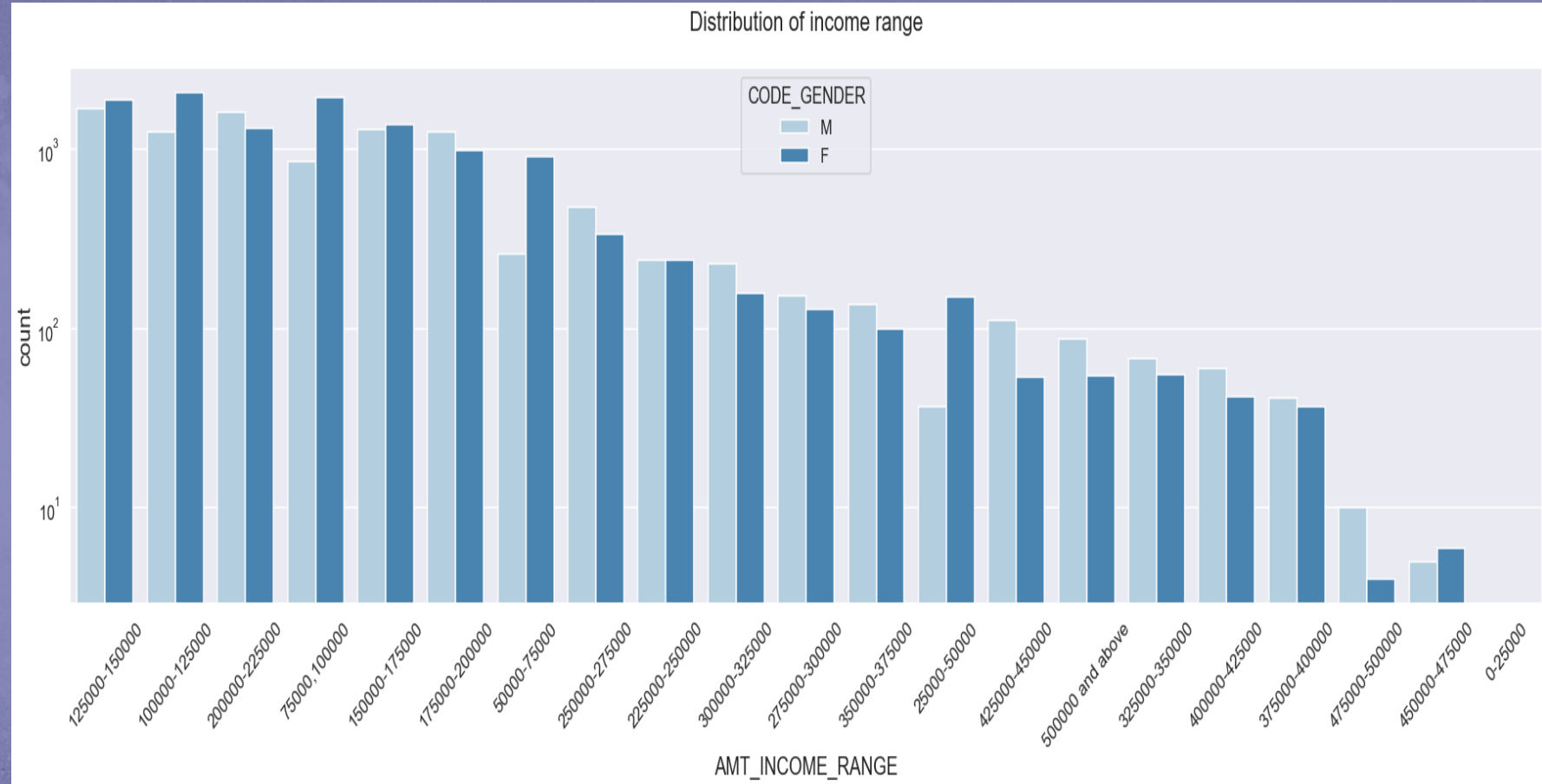
# CATEGORICAL UNIVARIATE ANALYSIS FOR TARGET 1



# DISTRIBUTION OF INCOME RANGE

Observations from the graph:

- Male counts are higher than female.
- Income range from 100000 to 200000 is having more number of credits.
- This graph show that males are more than female in having credits for that range.
- Very less count for income range 400000 and above.

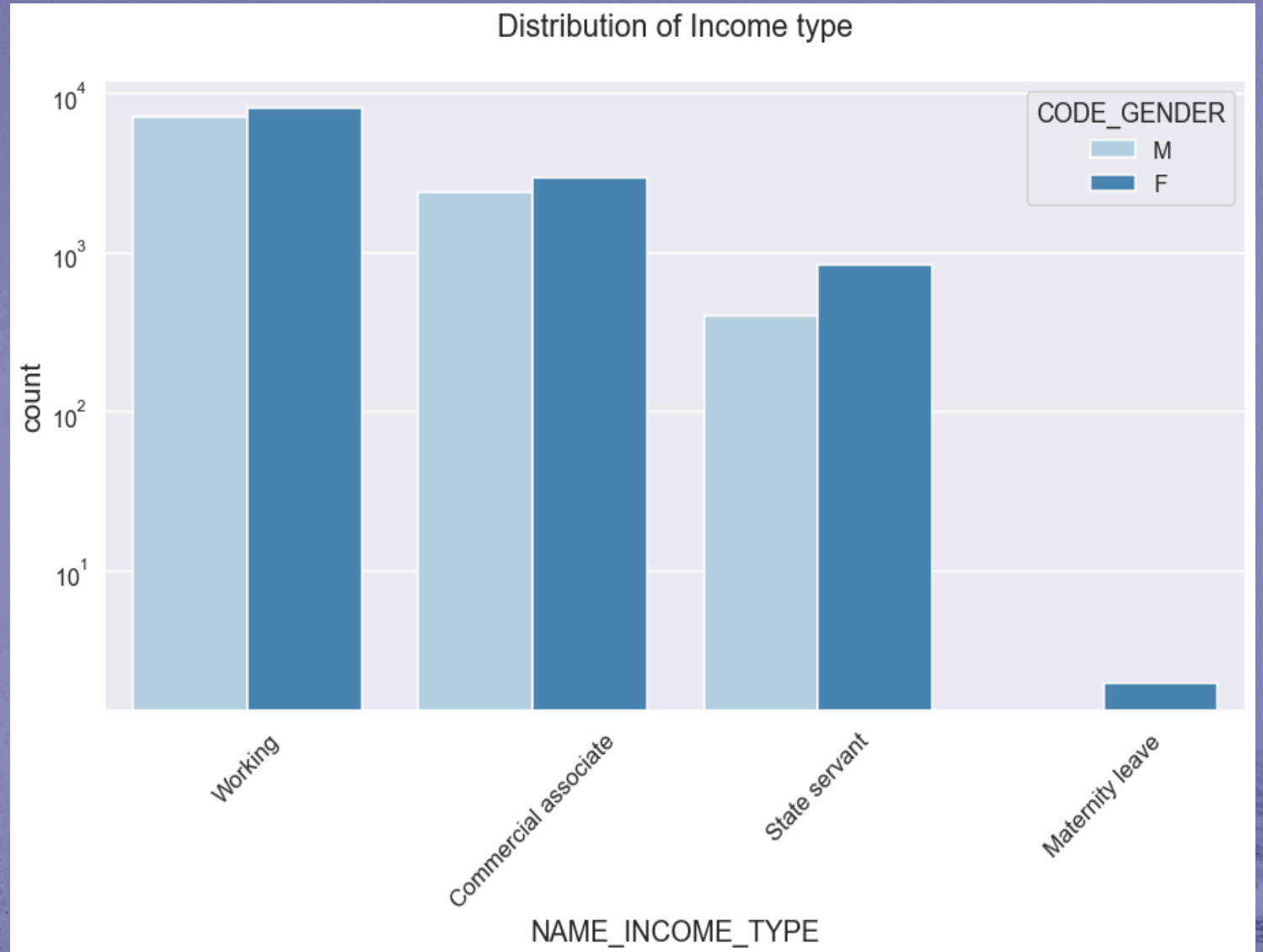




# DISTRIBUTION OF INCOME TYPE

Observations from the graph:

- For income type 'Working', 'Commercial associate', and 'State Servant' the number of credits are higher than other i.e. 'Maternity leave'.
- For this Females are having more number of credits than male.
- Less number of credits are for income type 'Maternity leave'.
- For type 1: There is no income type for 'Student', 'Pensioner' and 'Businessman' which means they don't do any late payments.

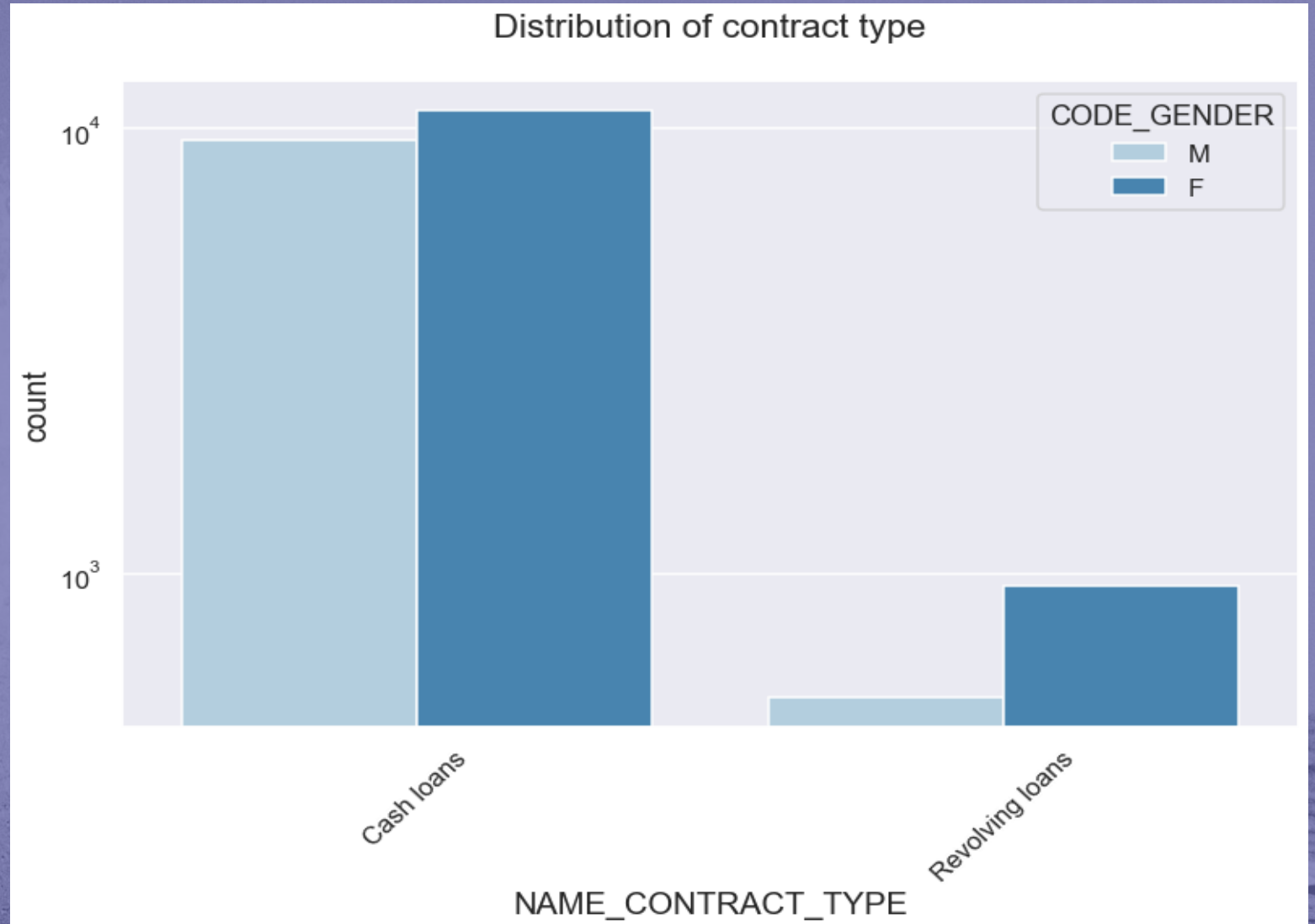




# DISTRIBUTION OF CONTRACT TYPE

Observations from the above graph:

- For contract type, 'cash loans' is having higher number of credits than 'Revolving loans' contract type.
- Female is leading for applying credits.
- For type 1: there is only Female Revolving loans.

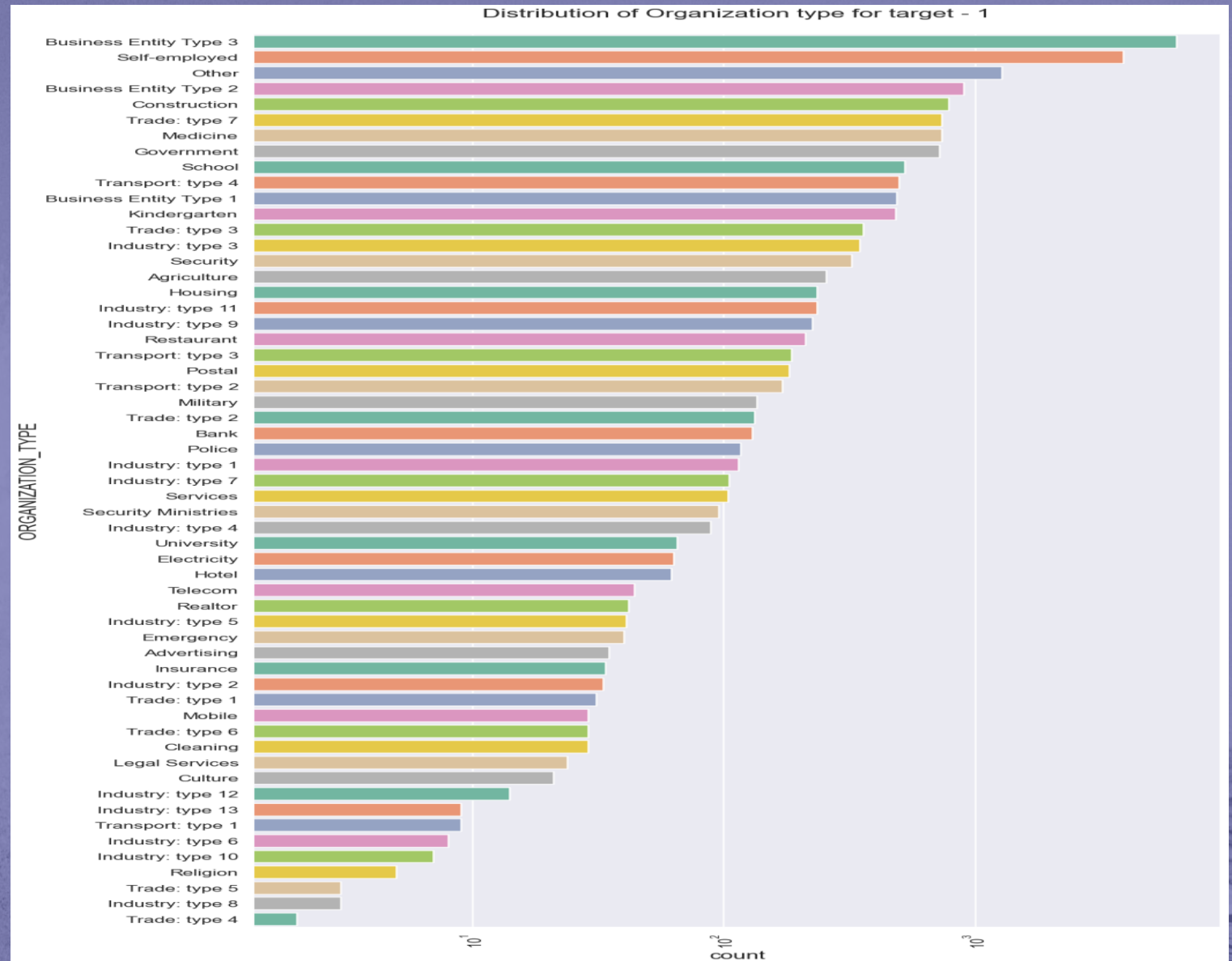




# DISTRIBUTION OF ORGANIZATION TYPE

Observations from the graph:

- Clients which have applied for credits are from most of the organization type 'Business entity Type 3', 'Self employed', 'Other', and 'Business entity Type 2'.
- Less clients are from Industry type 8, type 6, type 10, religion and trade type 5, type 4.







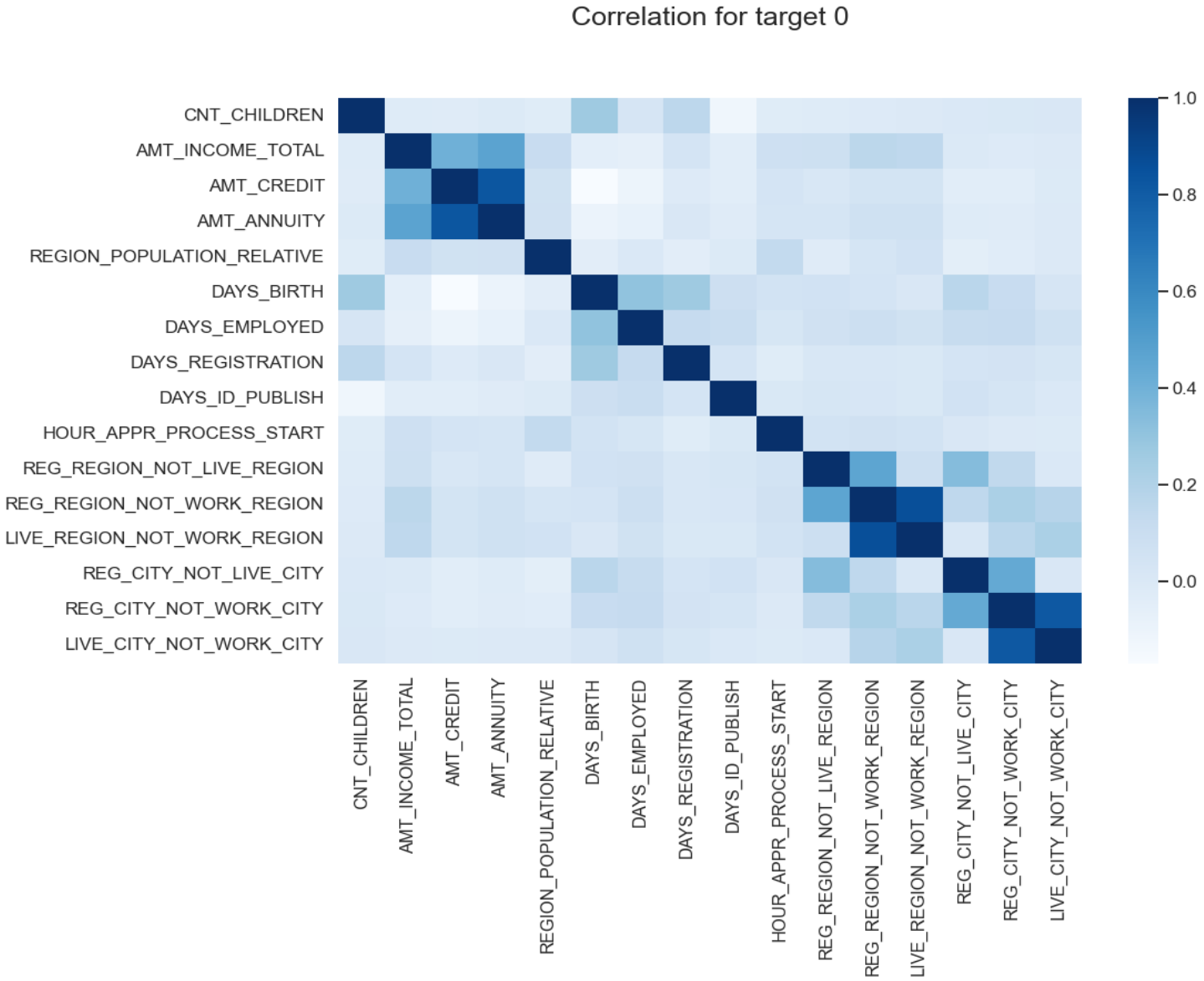
# CORRELATION OF TARGET 0 AND TARGET 1



# CORRELATION OF TARGET-0

Observations that can be derived from the correlation heatmap are:

- Credit amount is inversely proportional to the date of birth, which means Credit amount is higher for low age and vice-versa.
- Credit amount is inversely proportional to the number of children client have, means Credit amount is higher for less children count client have and vice-versa.
- Income amount is inversely proportional to the number of children client have, means more income for less children client have and vice-versa.
- Less children client are in densely populated area.
- Credit amount is higher in densely populated area.
- Income is also higher in densely populated area.

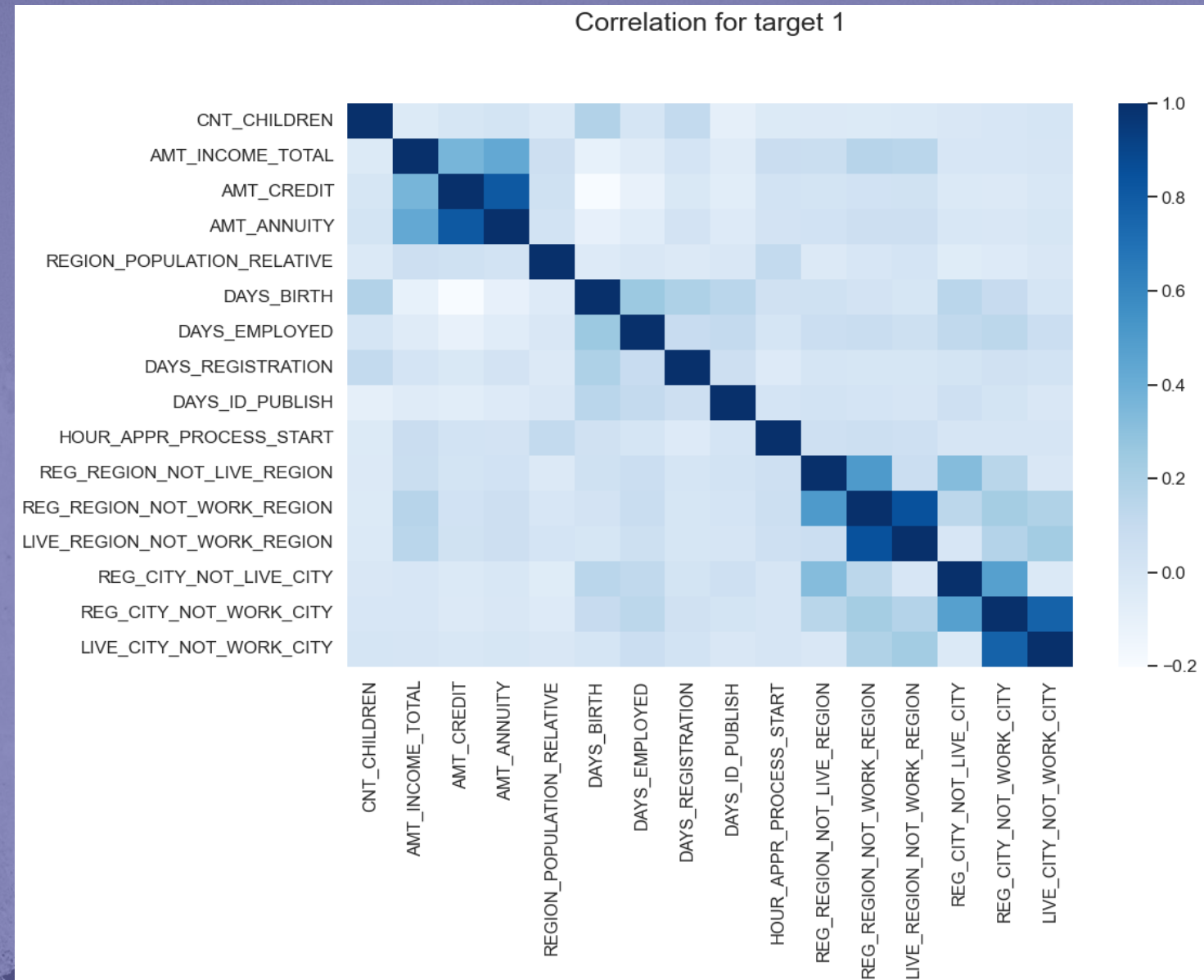




# CORRELATION OF TARGET-1

Observations that can be derived from the correlation heatmap are as follows:

- The client's permanent address does not match contact address are having less children and vice-versa
- The client's permanent address does not match work address are having less children and vice-versa





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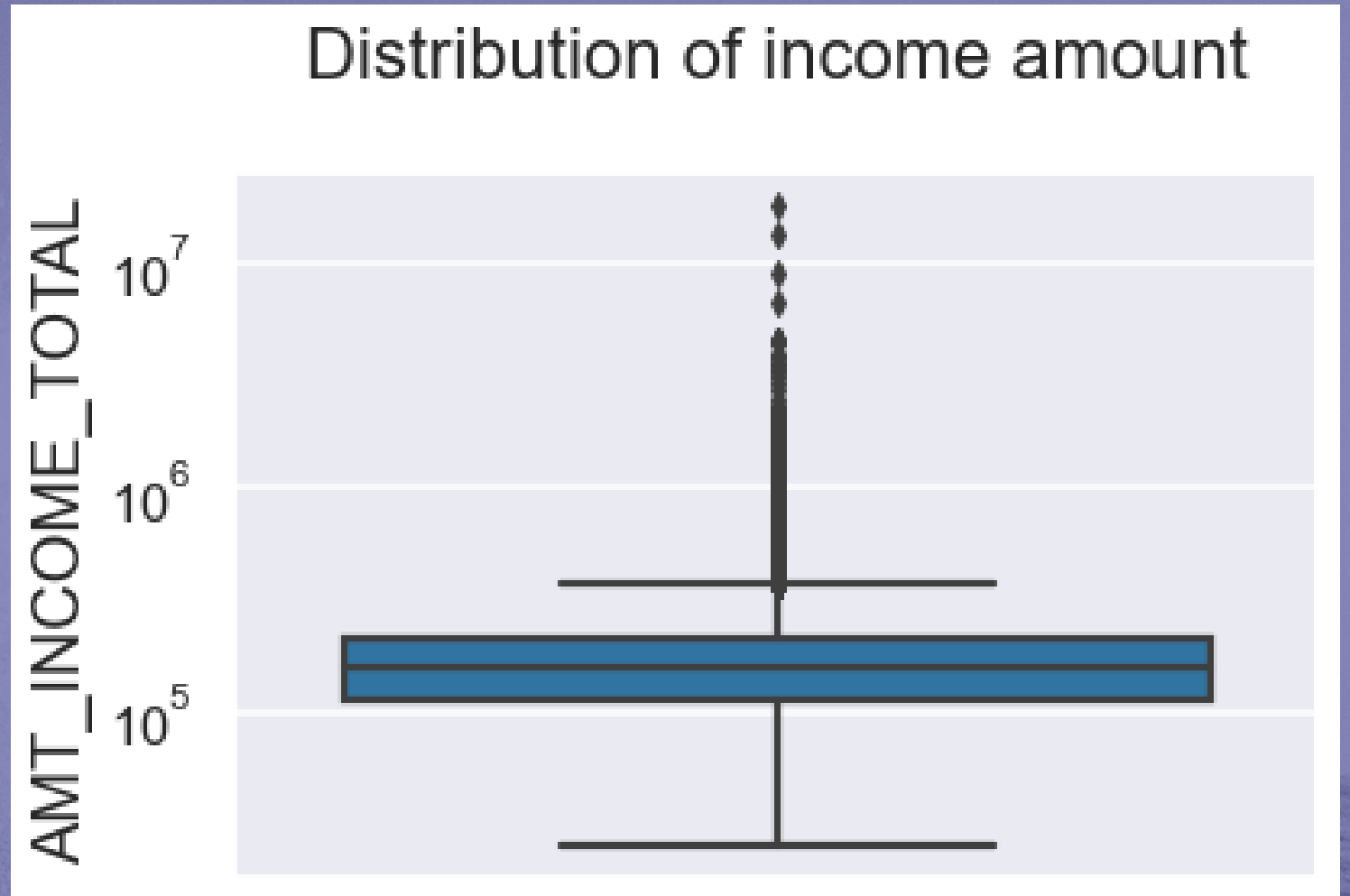
# CATEGORICAL UNIVARIATE ANALYSIS FOR VARIABLES TARGET 0



# BOX PLOT FOR INCOME AMOUNT

Observation from the graph:

- Outliers are noticed in income amount.
- Third Quartile is very slim for income amount.

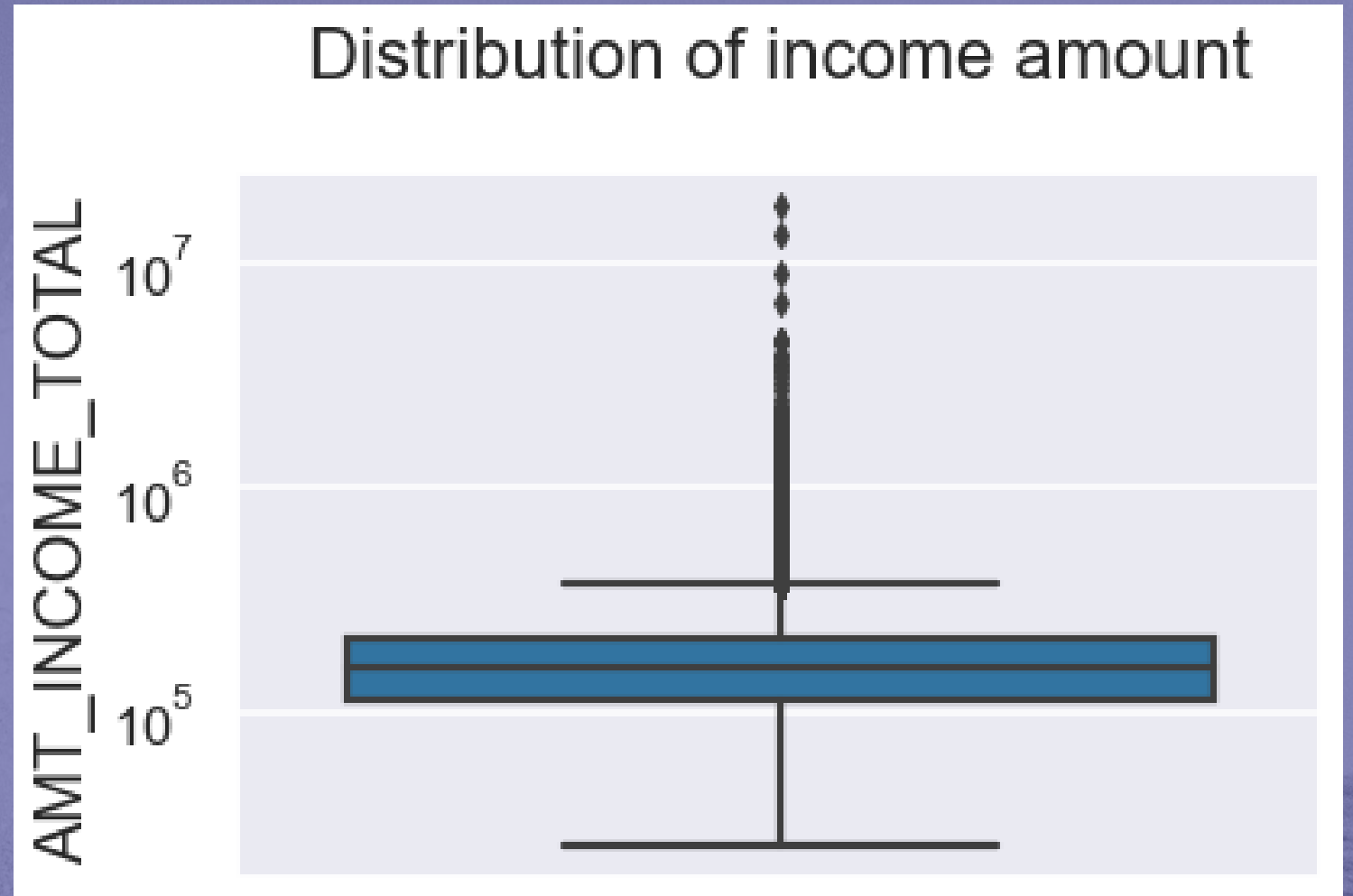




# BOX PLOT FOR CREDIT AMOUNT

Observation from the graph:

- Outliers are noticed in credit amount.
- First quartile is bigger than third quartile for credit amount which means most of the credits of clients are present in the first quartile.

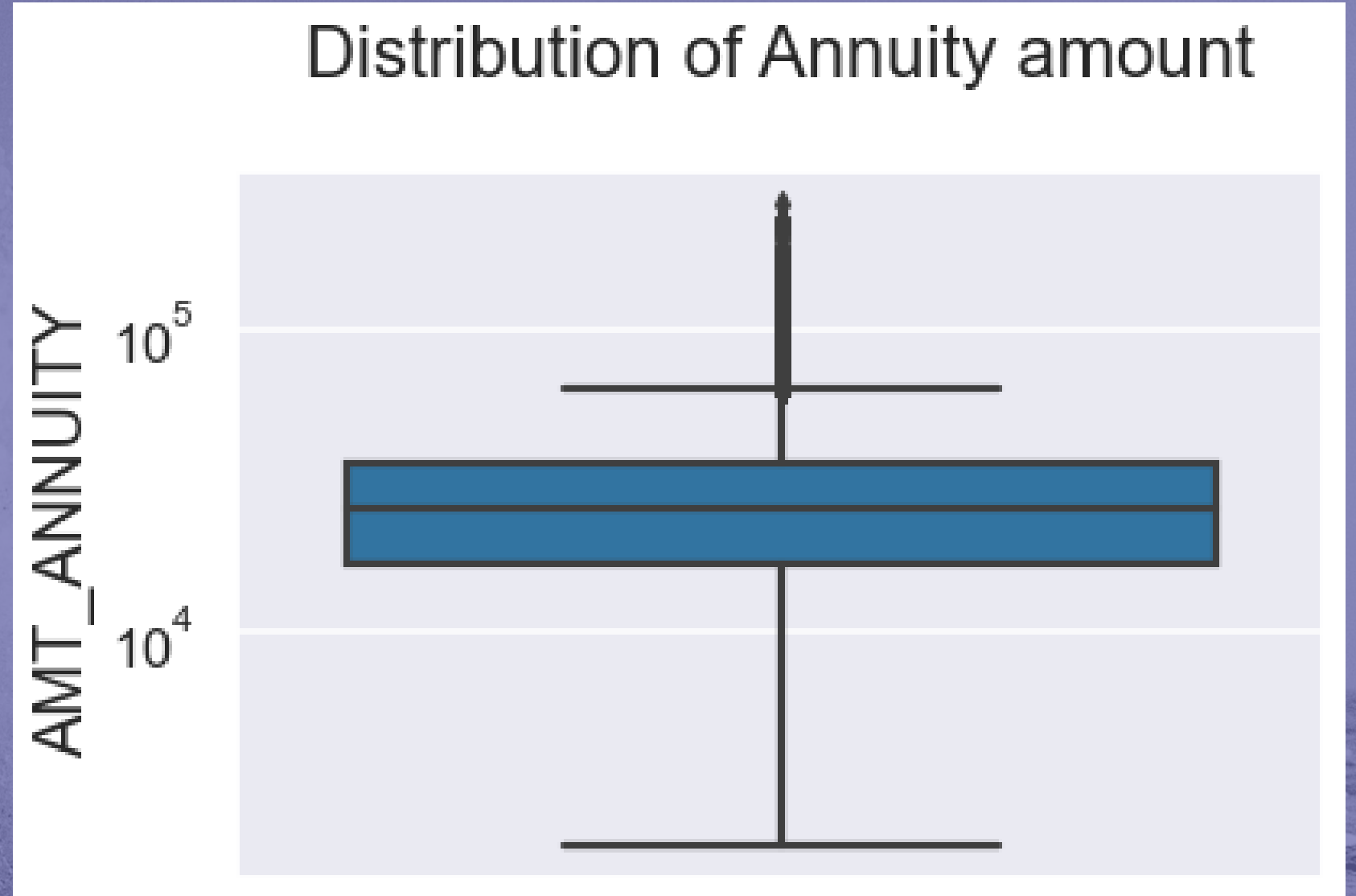




# BOX PLOT FOR ANNUITY AMOUNT

Observation from the graph:

- Outliers are noticed in annuity amount.
- First quartile is bigger than third quartile for annuity amount which means most of the annuity clients are from first quartile.





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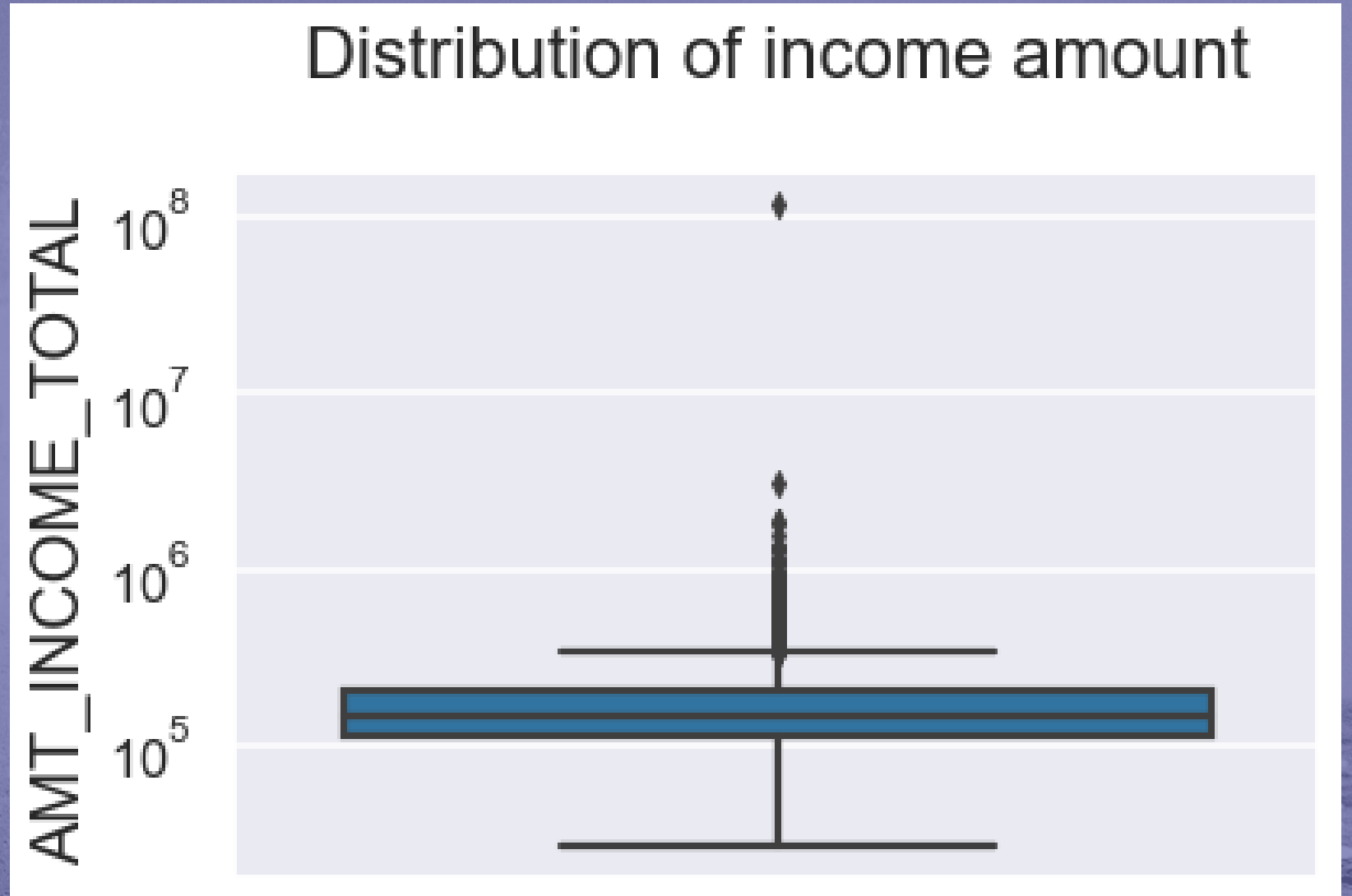
# CATEGORICAL UNIVARIATE ANALYSIS FOR VARIABLES TARGET 1



# BOX PLOT FOR INCOME AMOUNT

Observation from the graph:

- Outliers are noticed in income amount.
- First quartiles is very slim for income amount.
- Most of the clients of income are present in third quartile.

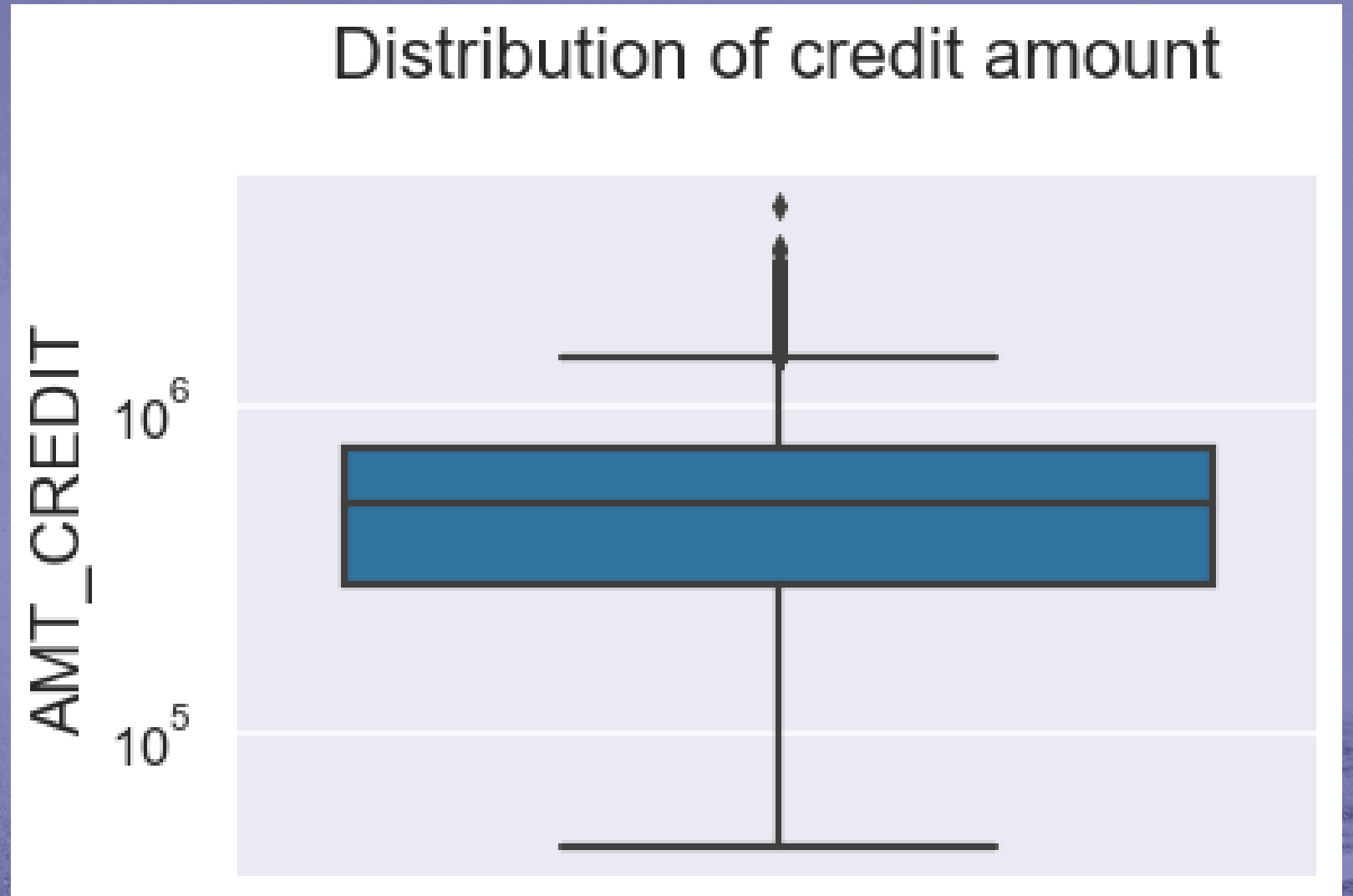




# BOX PLOT FOR CREDIT AMOUNT

Observation from the graph:

- Outliers are noticed in credit amount.
- The first quartile is bigger than third quartile for credit amount which means most of the credits of clients are present in the first quartile.

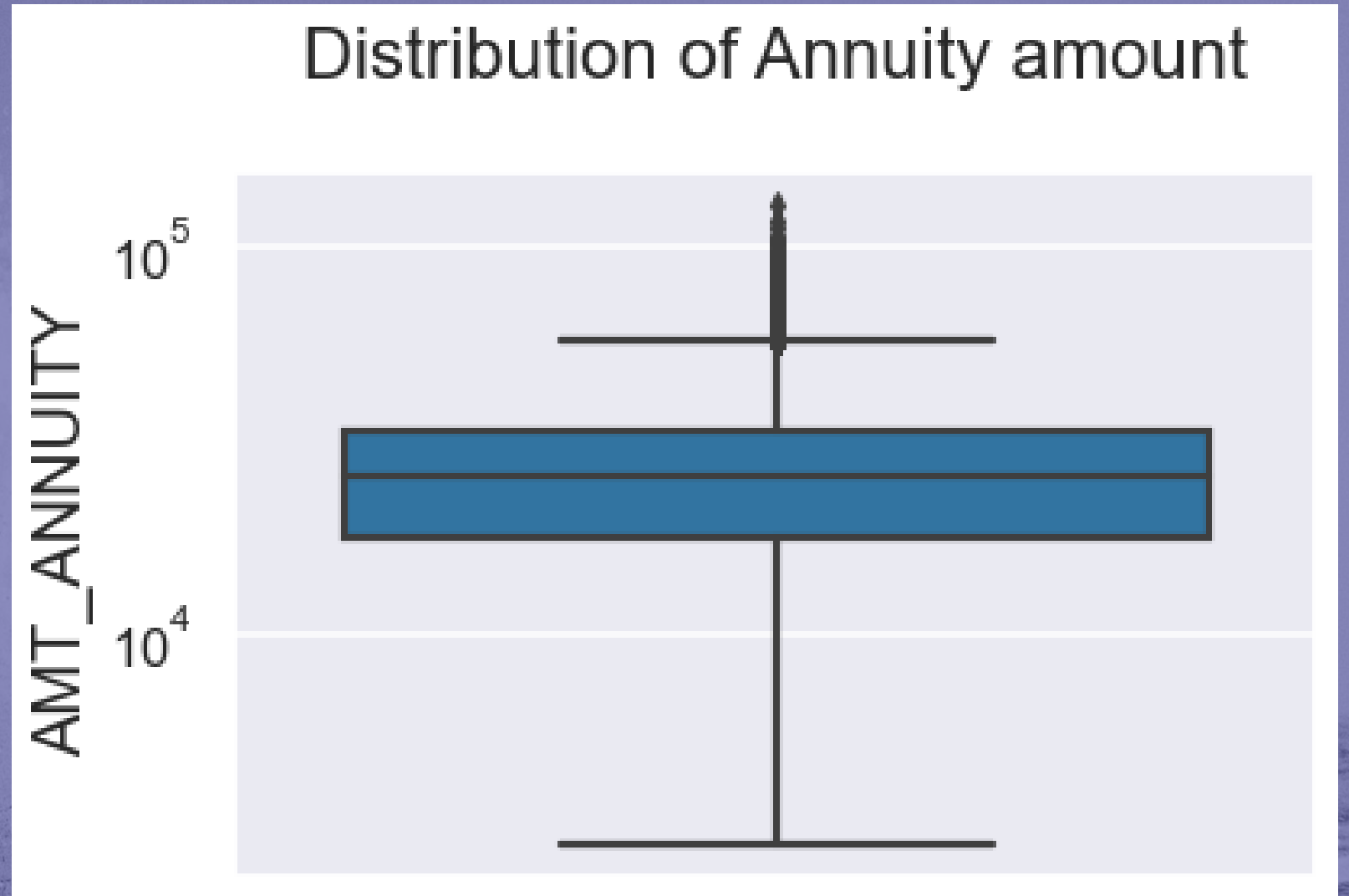




# BOX PLOT FOR ANNUITY AMOUNT

Observation from the graph:

- Outliers are noticed in annuity amount.
- The first quartile is bigger than third quartile for annuity amount which means most of the annuity clients are from first quartile.







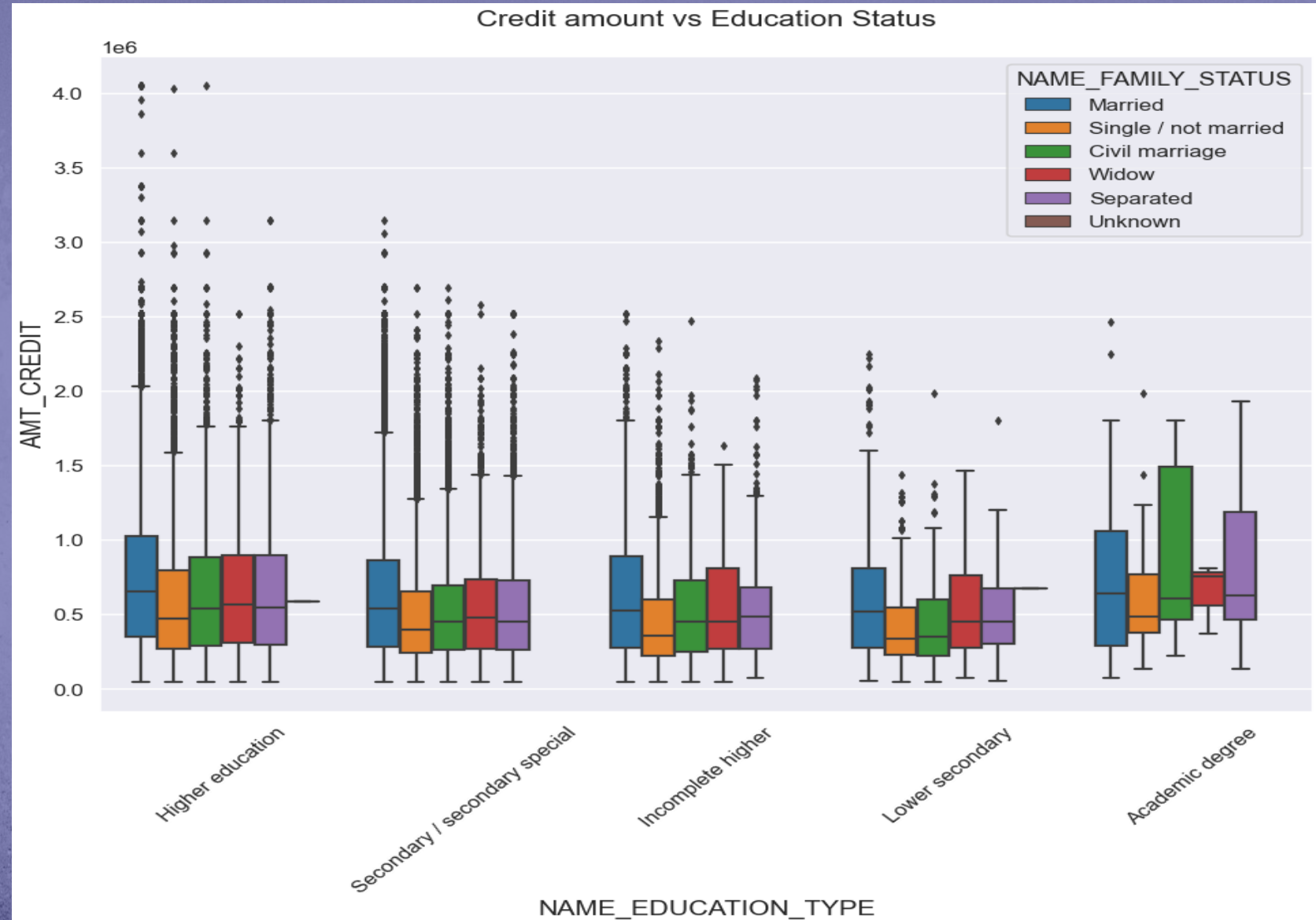
# BIVARIATE ANALYSIS FOR TYPE 0



# CREDIT AMOUNT VS. EDUCATION STATUS

Observation from the graph:

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

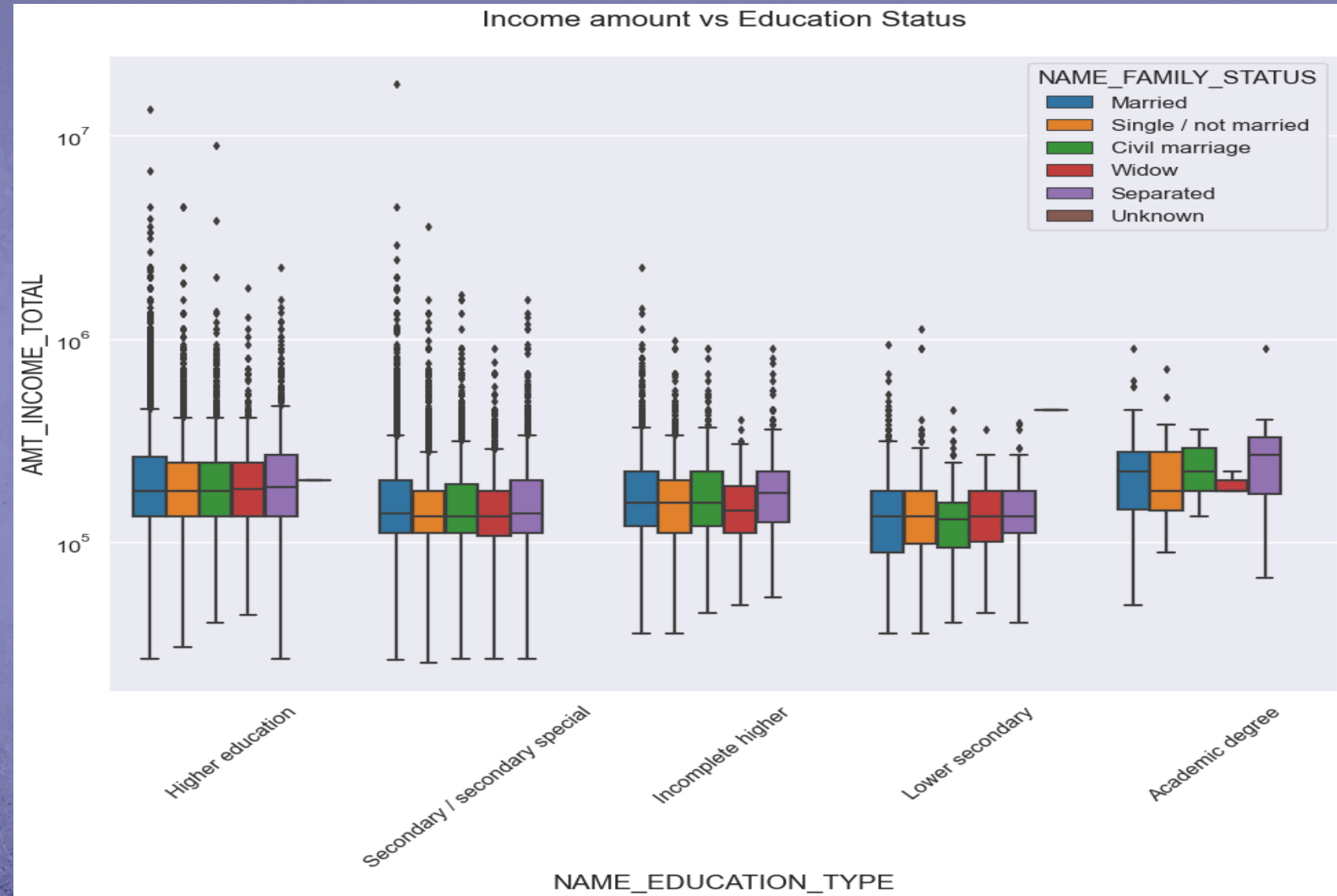




# INCOME AMOUNT VS. EDUCATION STATUS

Observation from the graph:

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







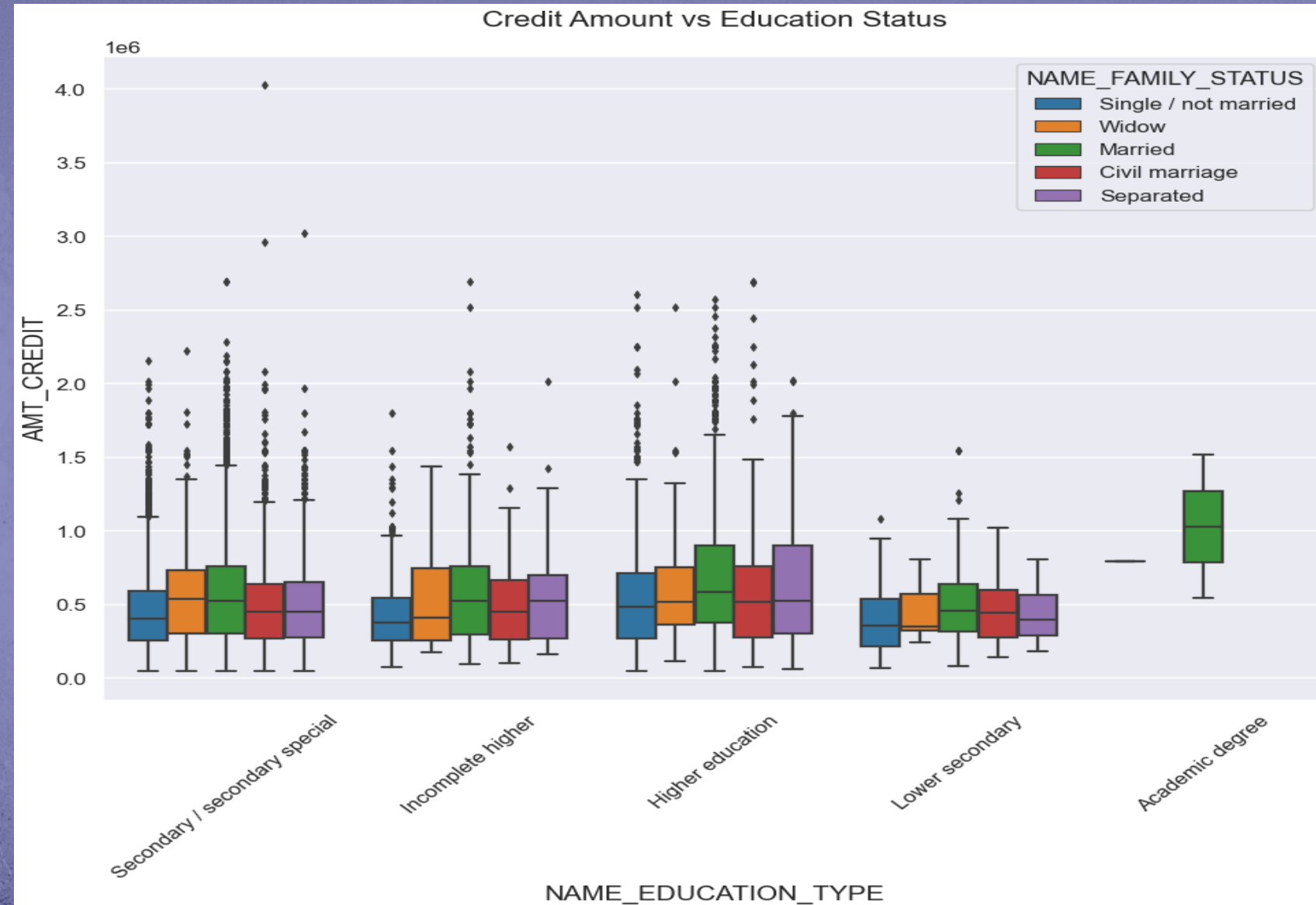
# BIVARIATE ANALYSIS FOR TYPE 1



# CREDIT AMOUNT VS. EDUCATION STATUS

Observation from the graph:

- Most of the outliers are present in education type 'Secondary' and 'Higher education'.
- Family status of 'Married', 'Separated' and 'civil marriage' of 'Higher education' are having higher number of credits than others.
- 'Separated' for 'Higher education' is having most of the credits in the third quartile.

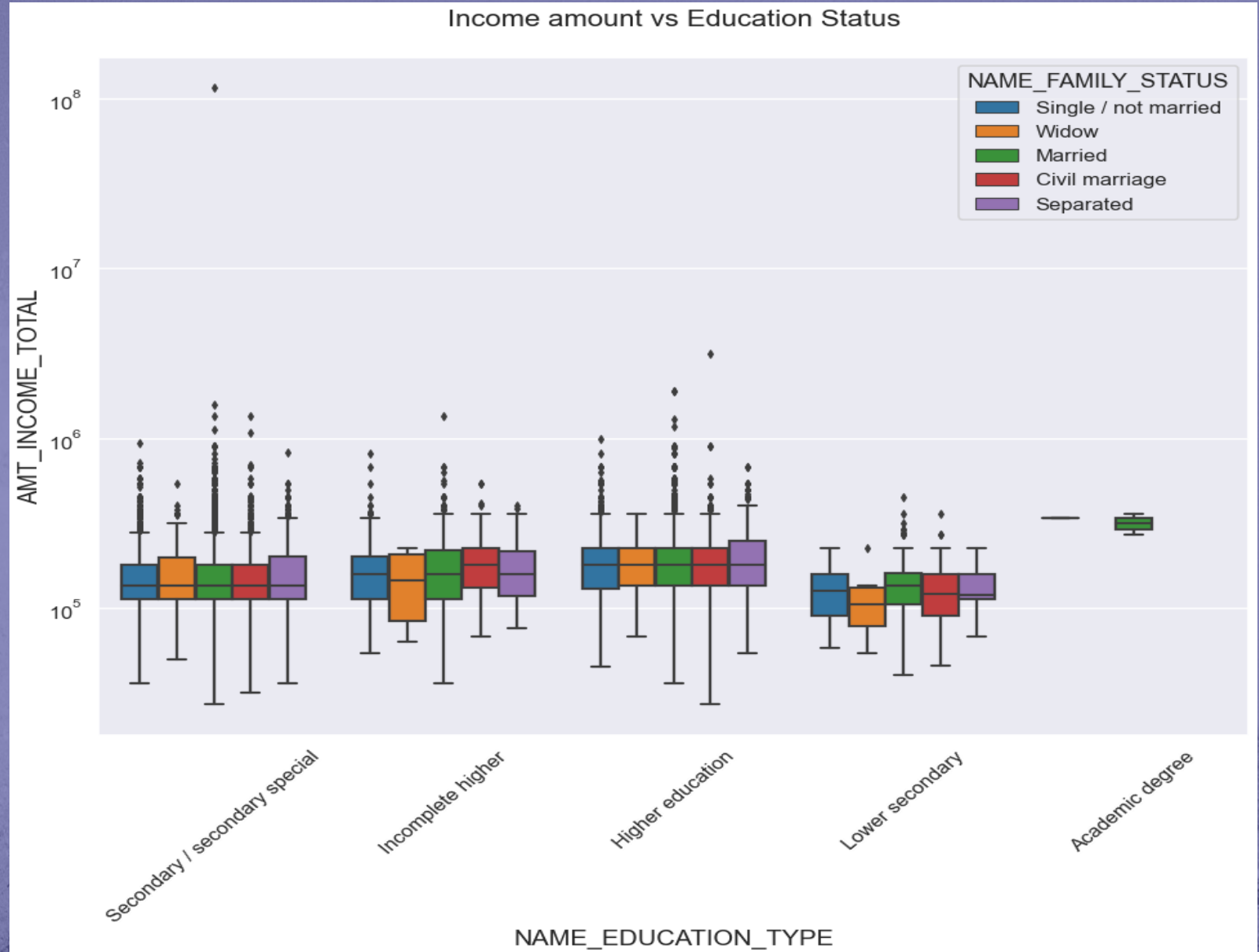




# INCOME AMOUNT VS. EDUCATION STATUS

Observation from the graph:

- Income amount is mostly equal with family status.
- Less outlier are having for Academic degree but there income amount is little higher than Higher education.
- Lower secondary are have less income amount than others.





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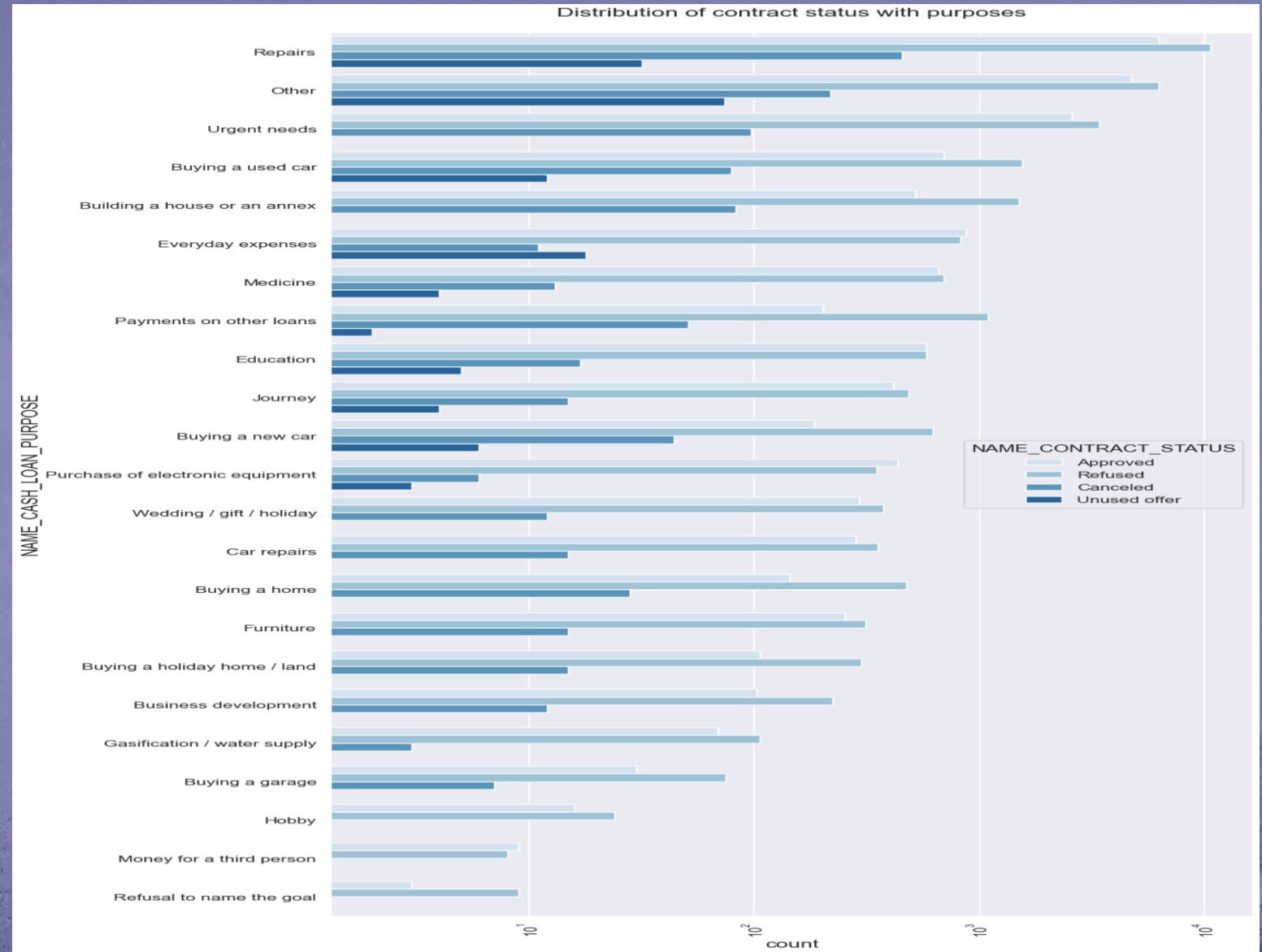
# UNIVARIATE ANALYSIS AFTER MERGING PREVIOUS DATA



# DISTRIBUTION OF CONTRACT STATUS WITH PURPOSES

Observation from the graph:

- Most rejection of loans came from purpose 'repairs'.
- For education purposes we have equal number of approvals and rejection
- 'Payment on other loans' and 'buying a new car' are having significant higher rejection than approvals

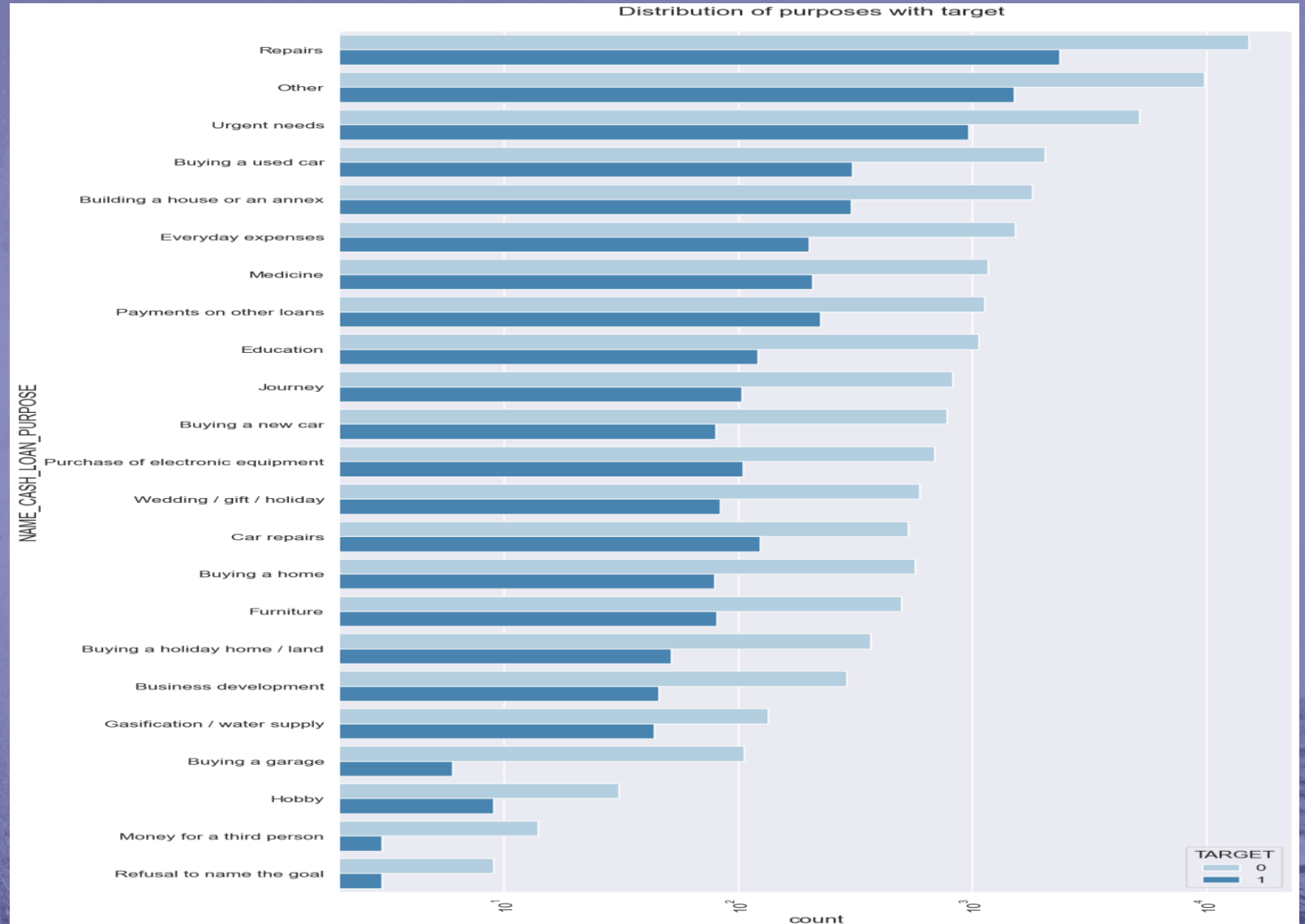




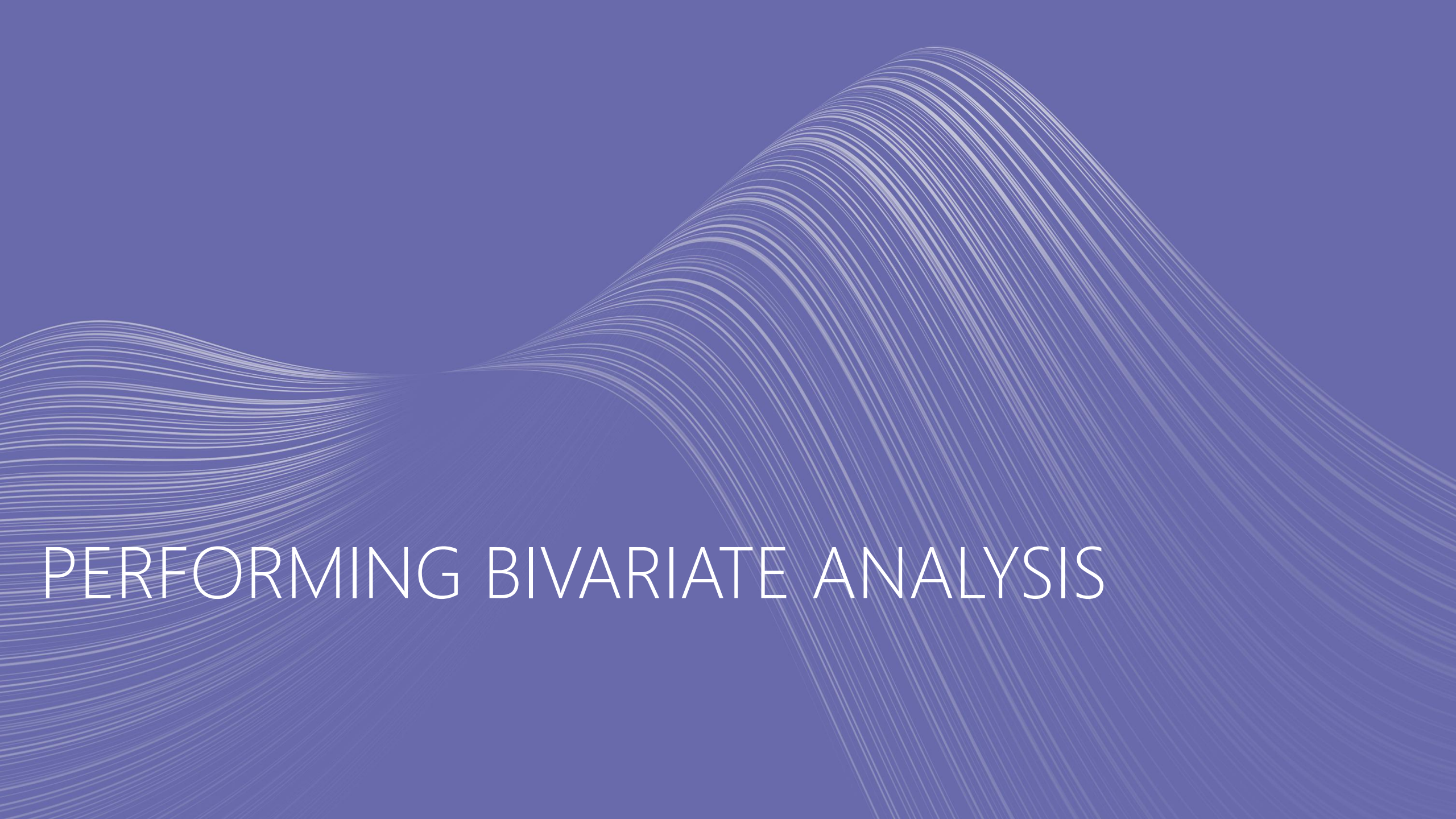
# DISTRIBUTION OF PURPOSES WITH TARGET

Observation from the graph:

- Most rejection of loans came from purpose 'repairs'.
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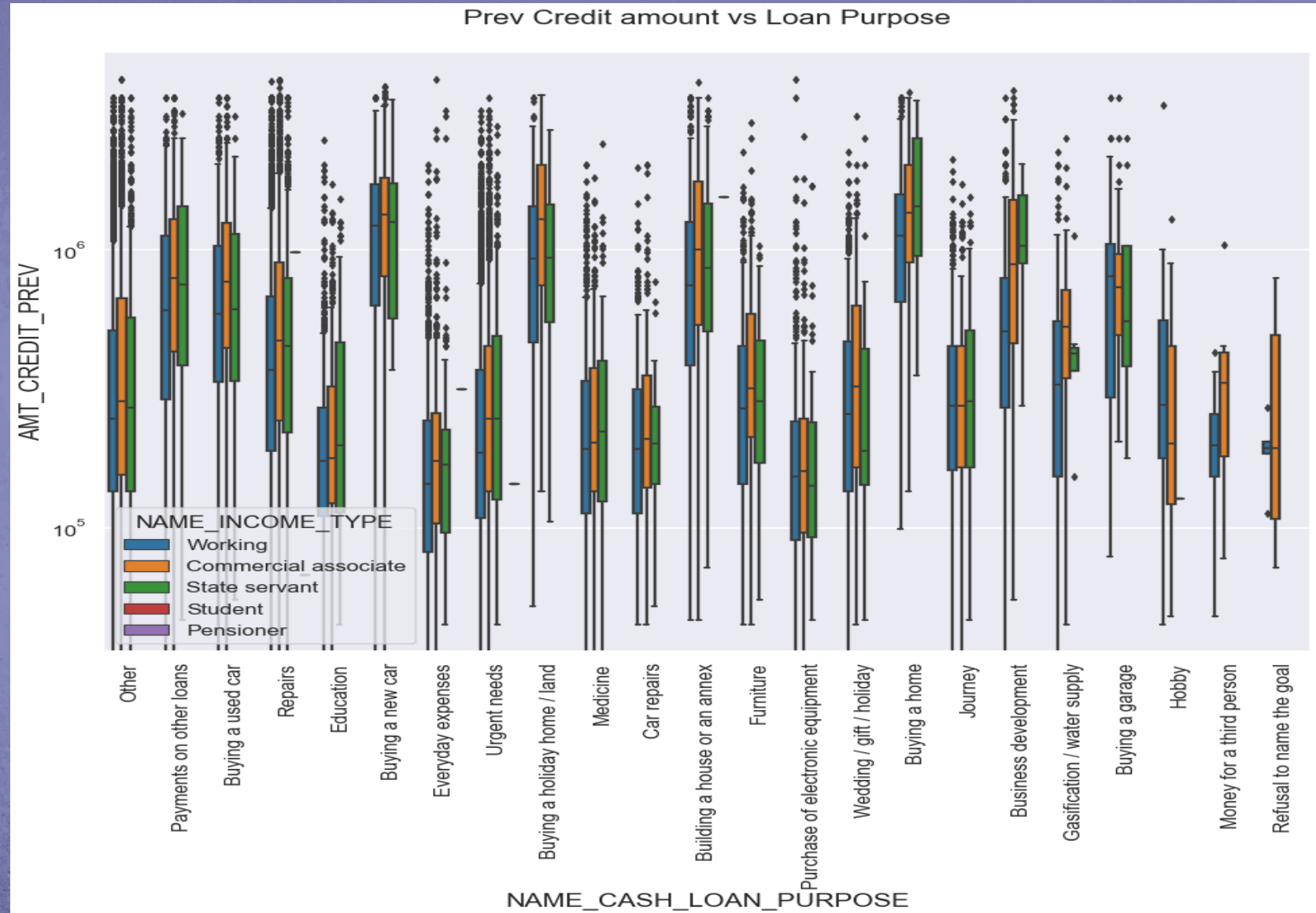
# PERFORMING BIVARIATE ANALYSIS



# PREVIOUS CREDIT AMOUNT VS. LOAN PURPOSE

Observation from the graph:

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

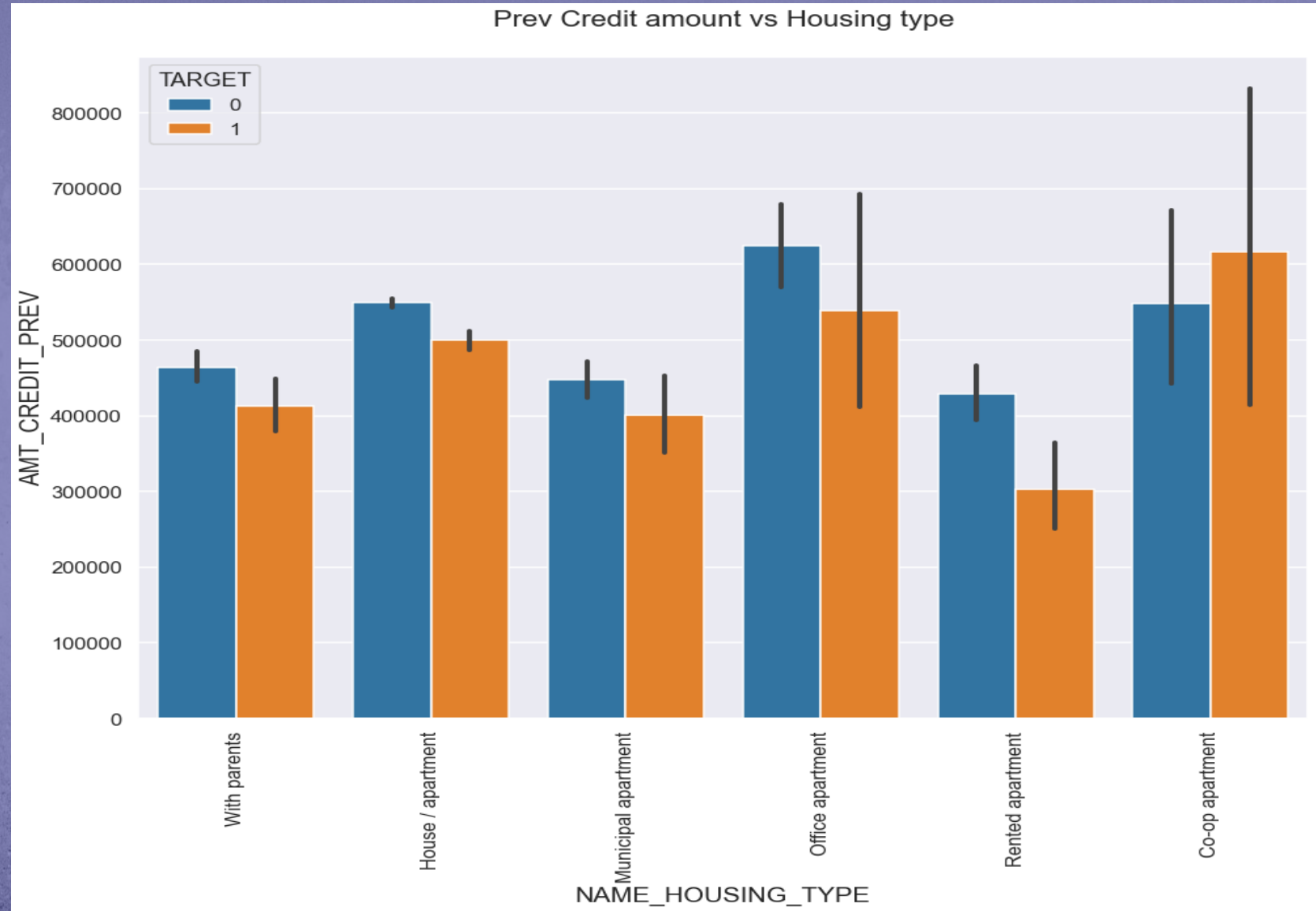




# PREVIOUS CREDIT AMOUNT VS. HOUSING TYPE

Observation from the graph:

- 'House / apartment', 'Office apartment' are having higher credit in target 0
- 'Co-op apartment' is having higher credit in target 1.
- Based on the above analysis, 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', 'House / apartment' or 'Municipal apartment' for successful payments.





# CONCLUSION

- High Focus Area for Banks:
  - Banks should focus more on contract type 'Student', 'pensioner' and 'Businessman' with housing type other than 'Co-op apartment' for successful payments.
  - Banks should focus on clients from housing type 'With parents' as they are having least number of unsuccessful payments.
- Less Focus Area for Banks:
  - Banks should focus less on income type 'Working' as they are having most number of unsuccessful payments.
  - Loan purpose 'Repair' is having higher number of unsuccessful payments on time.







THANK YOU