



EDA CASE STUDY

SUBMISSION

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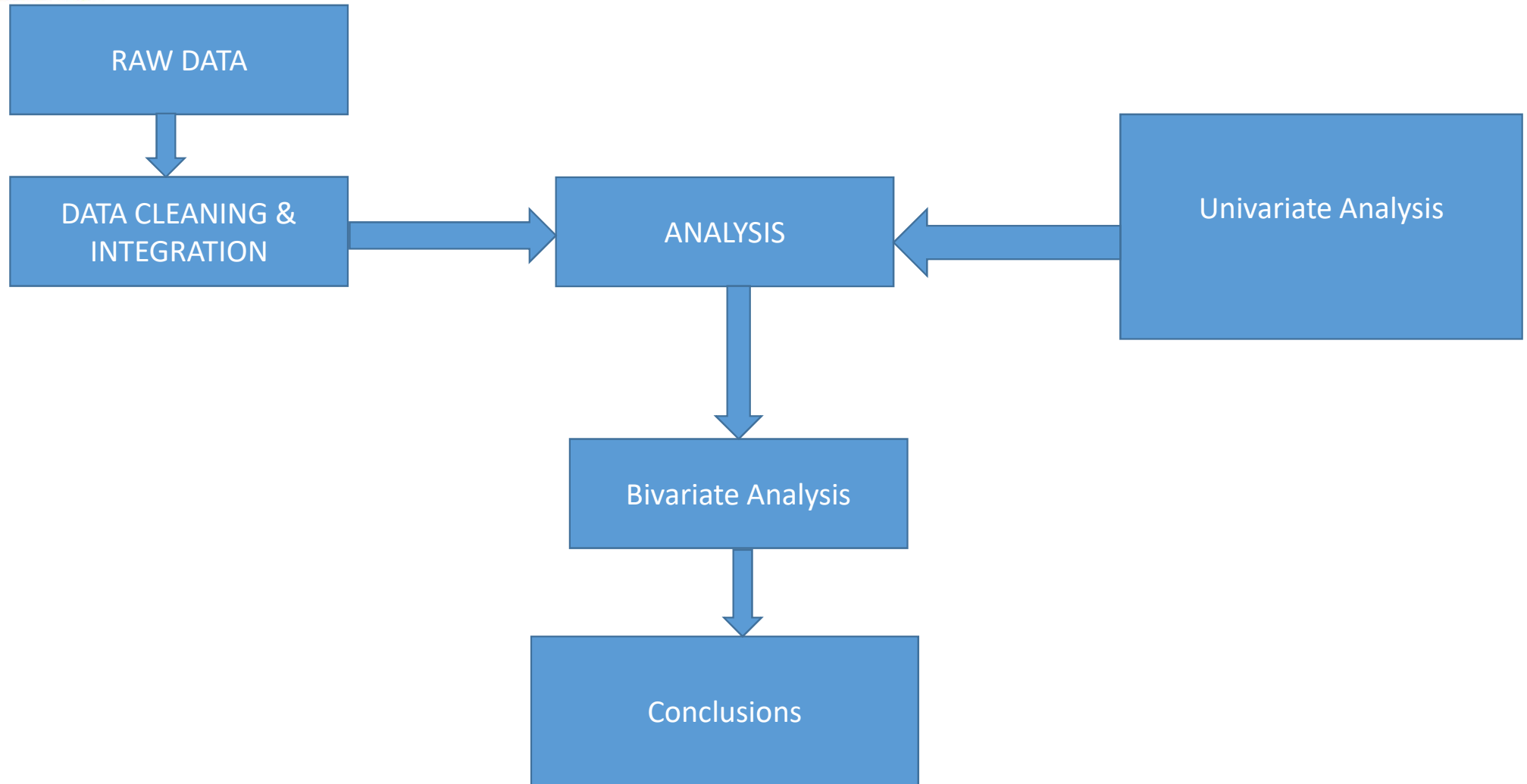


<Abstract>

Company wants to understand the driving factors (or driver variables) behind loan default, i.e. the variables which are strong indicators of default.

The company can utilise this knowledge for its portfolio and risk assessment.

If one is able to identify these risky loan applicants, then such loans can be reduced thereby cutting down the amount of credit loss. Identification of such applicants using EDA is the aim of this case study.

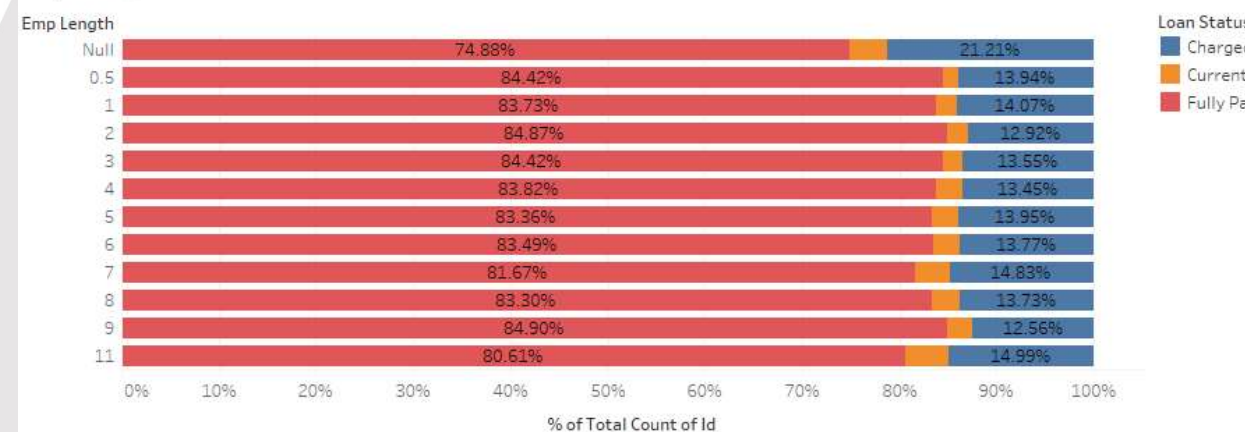




<Analysis>

- Univariate Analysis-I
- You see maximum defaulters where Employment length is not mentioned.

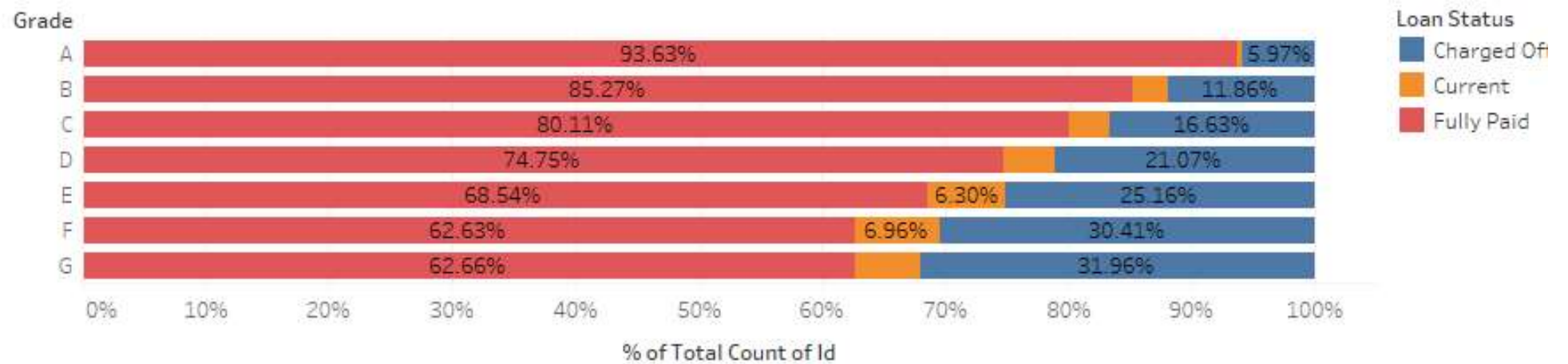
empl lengthwise





gradewise only

UpGrad

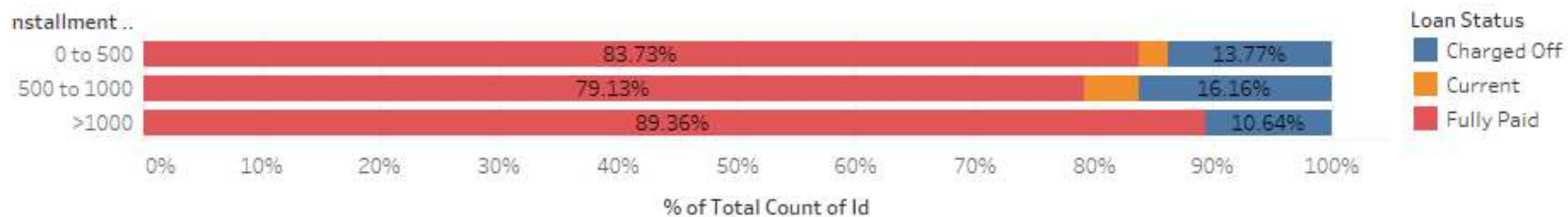


Univariate Analysis-II

You see as Grade decreases from A to G % of defaulters increases.

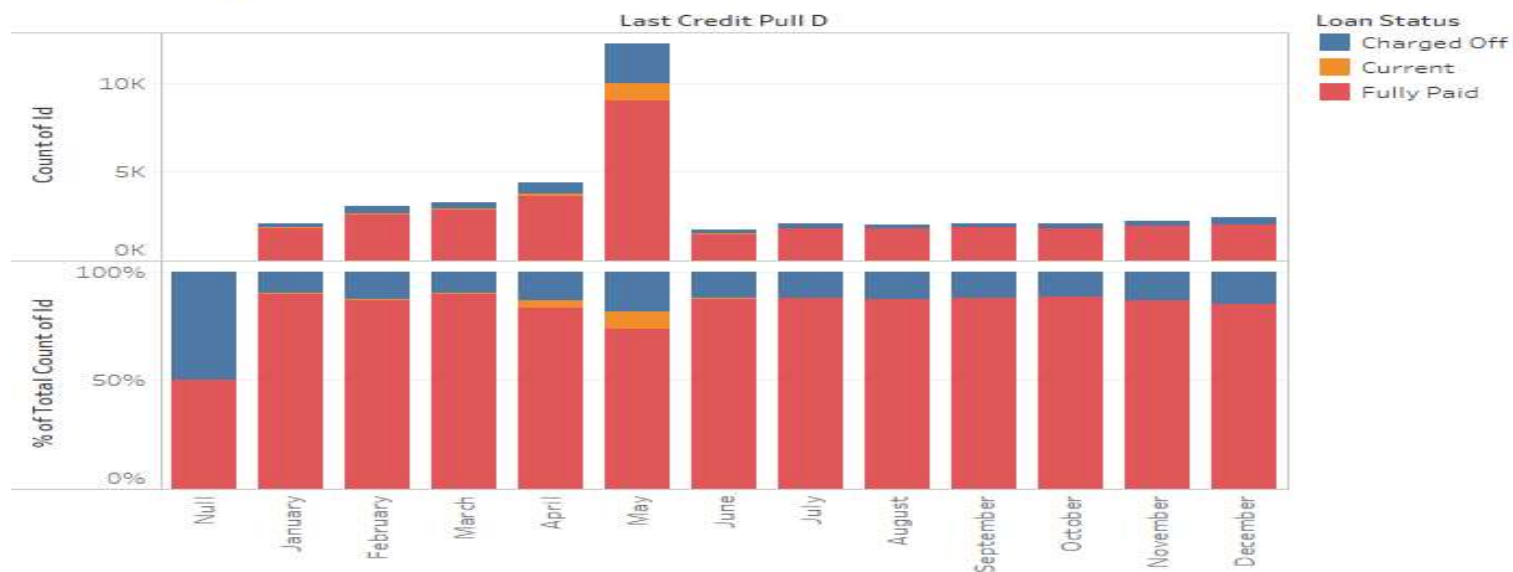


installment range only



Installment Range Analysis

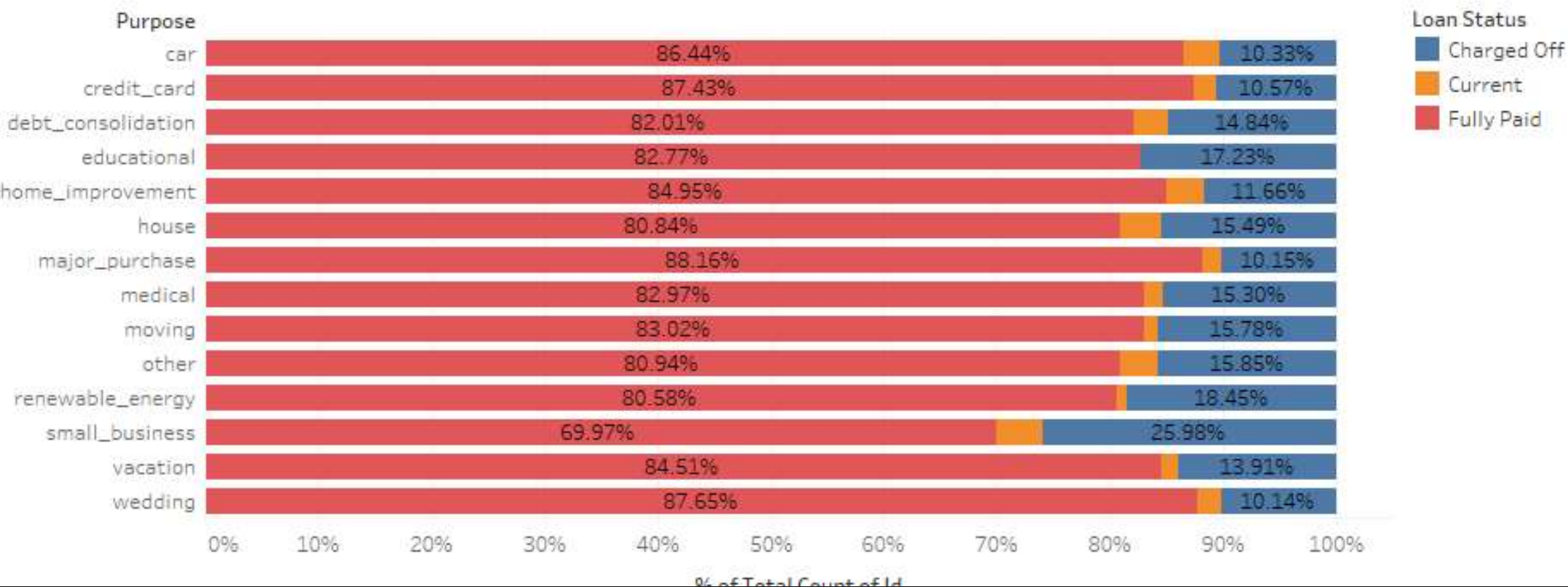
Last credit pull trend



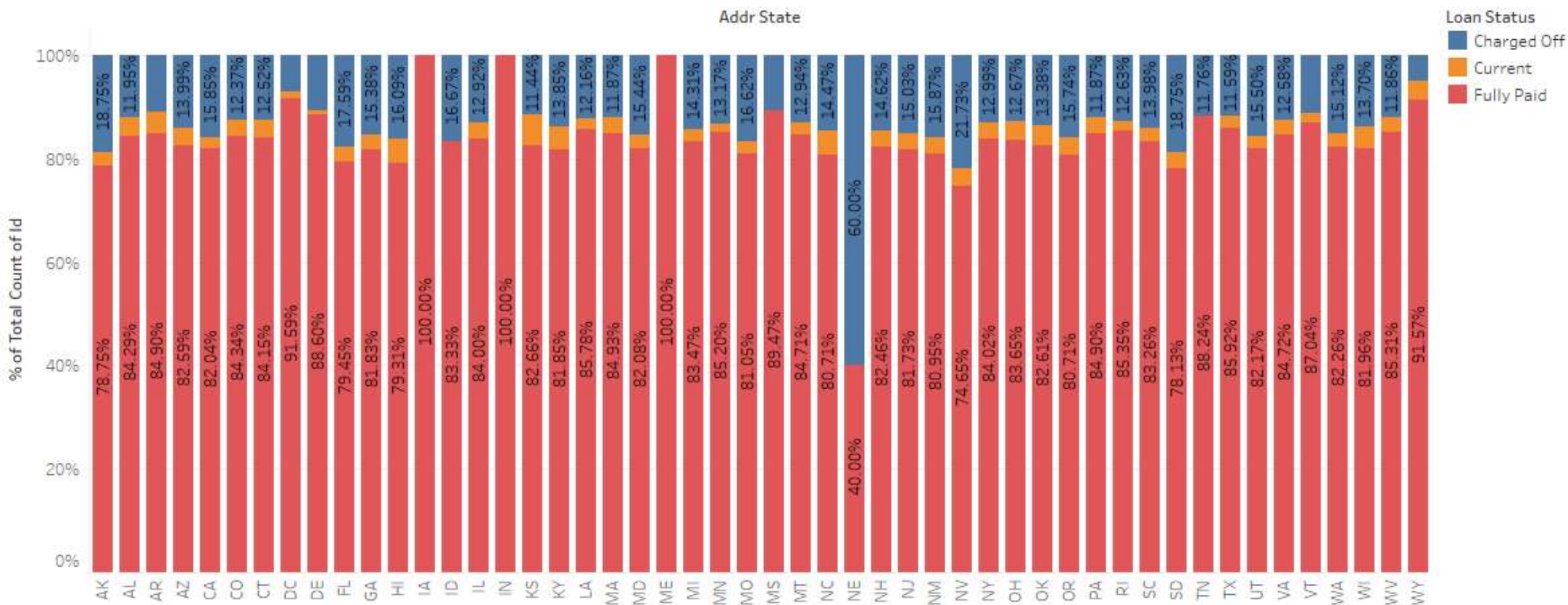
Count of Id and % of Total Count of Id for each Last Credit Pull D Month. Color shows details about Loan

Last Credit Pull Trend:- Unusual spike in number of loans during month of May

FIGURE 3.10: Loan Purpose



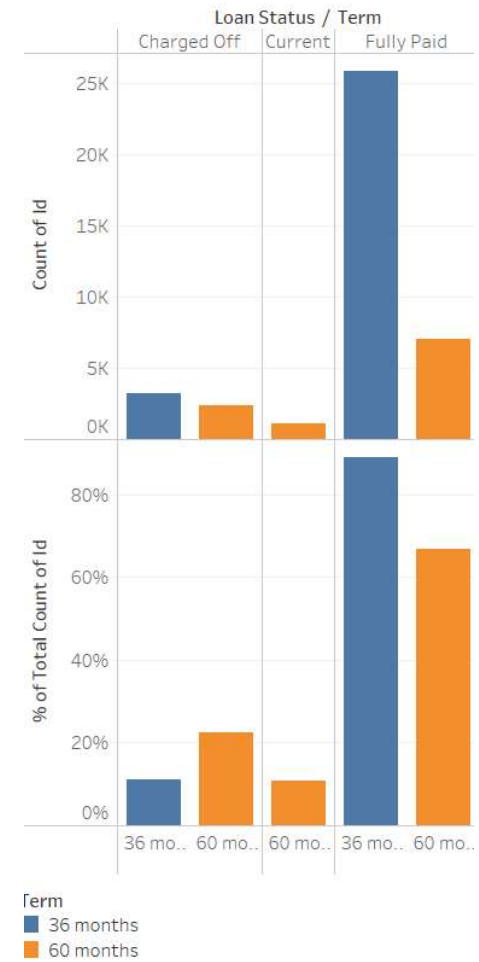
Looking at Purpose, you see maximum % of defaulters who took loan for business, banks may want to look closely business plan, model and revenue capacity.



Maximum % of defaulters from Nevada, may be people using money for gambling in Vegas.

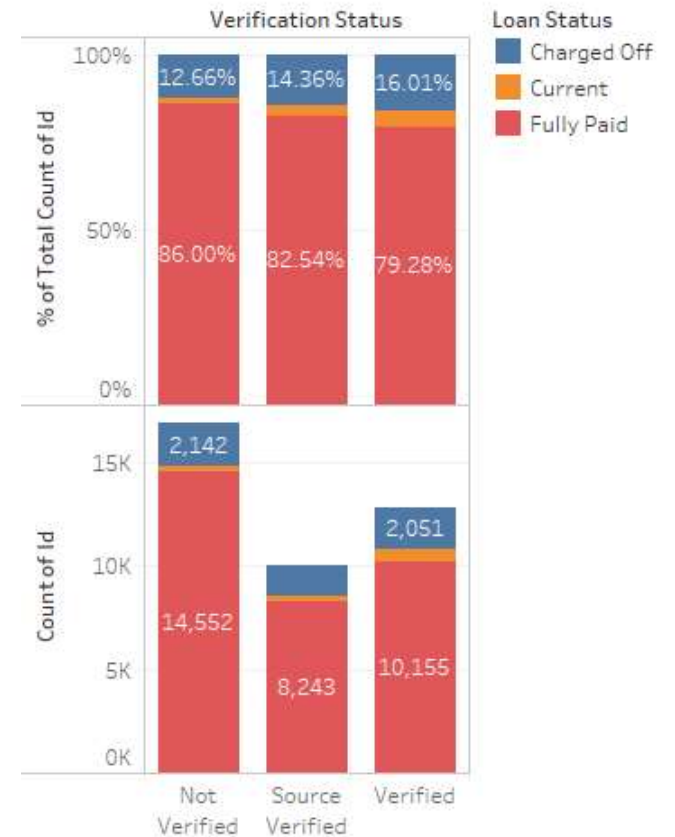
Term impact

Sheet 16

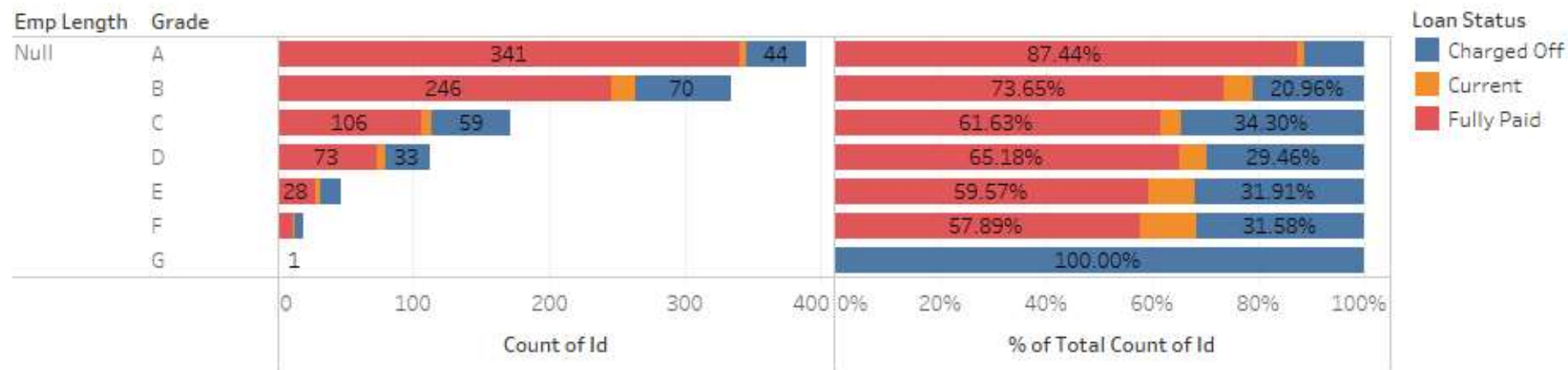


Verification

verification vs Request

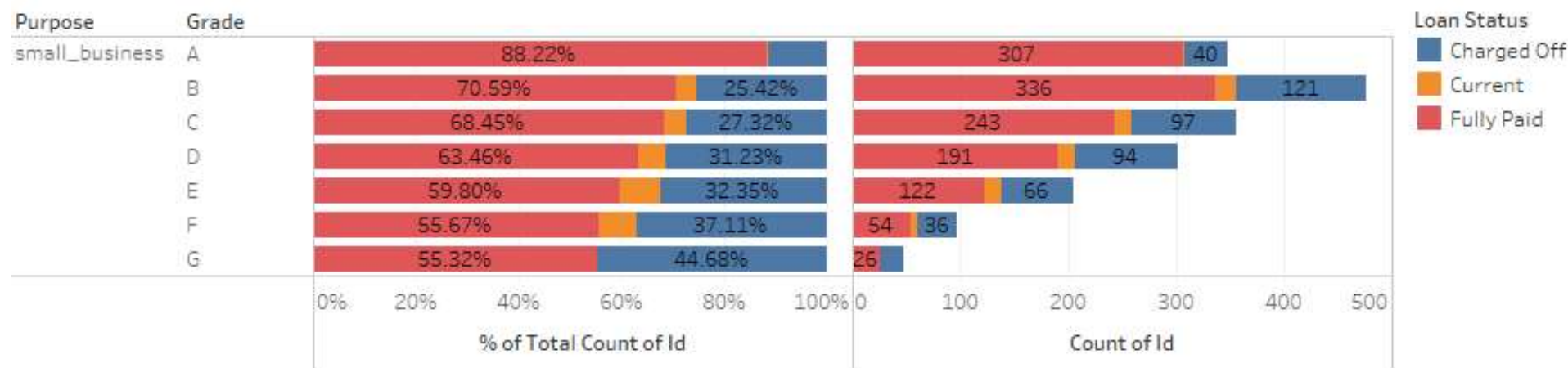


% of Total Count of Id and count of Id for



Count of Id and % of Total Count of Id for each Grade broken down by Emp Length. Color shows details about Loan Status. The view is filtered on Emp Length, which keeps Null.

Bivariate Analysis-I Employment length and grade



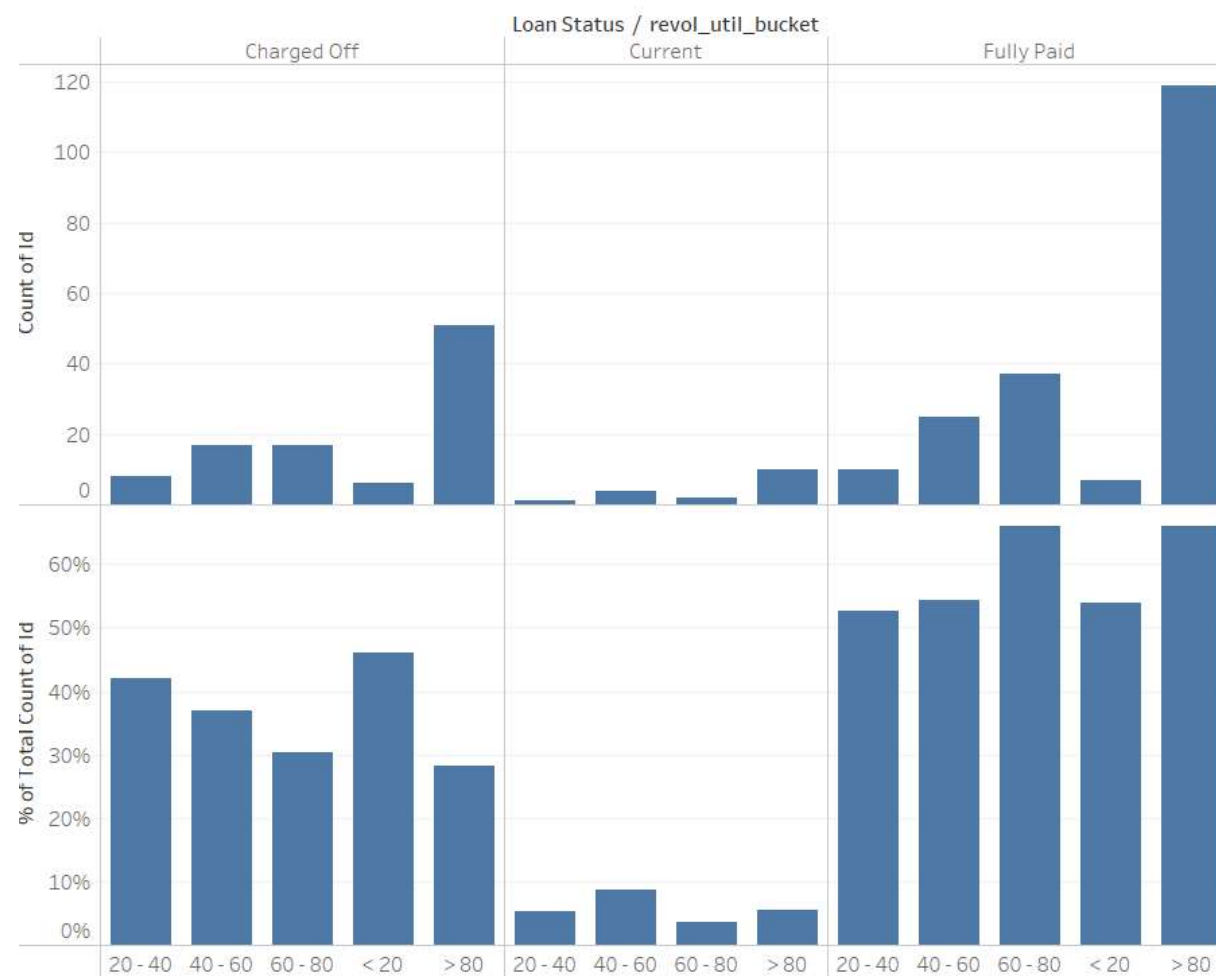
% of Total Count of Id and count of Id for each Grade broken down by Purpose. Color shows details about Loan Status. The view is filtered on Purpose, which keeps small_business.

Bivariate Analysis-II

Purpose and grade

Sheet 17 (2)

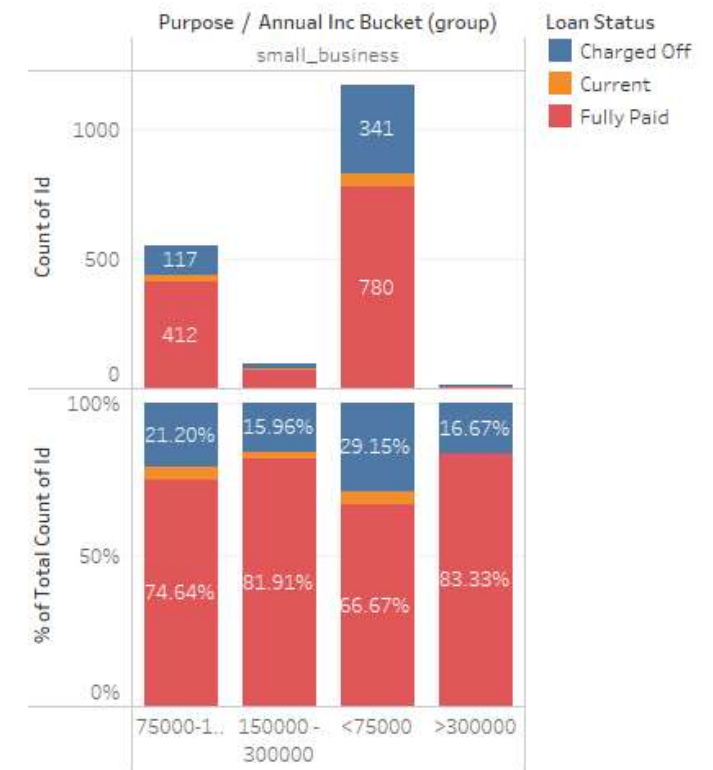
Bivariate Analysis-III
Loan Status vs
Revol bal
bucket and Int
Rate Bucket



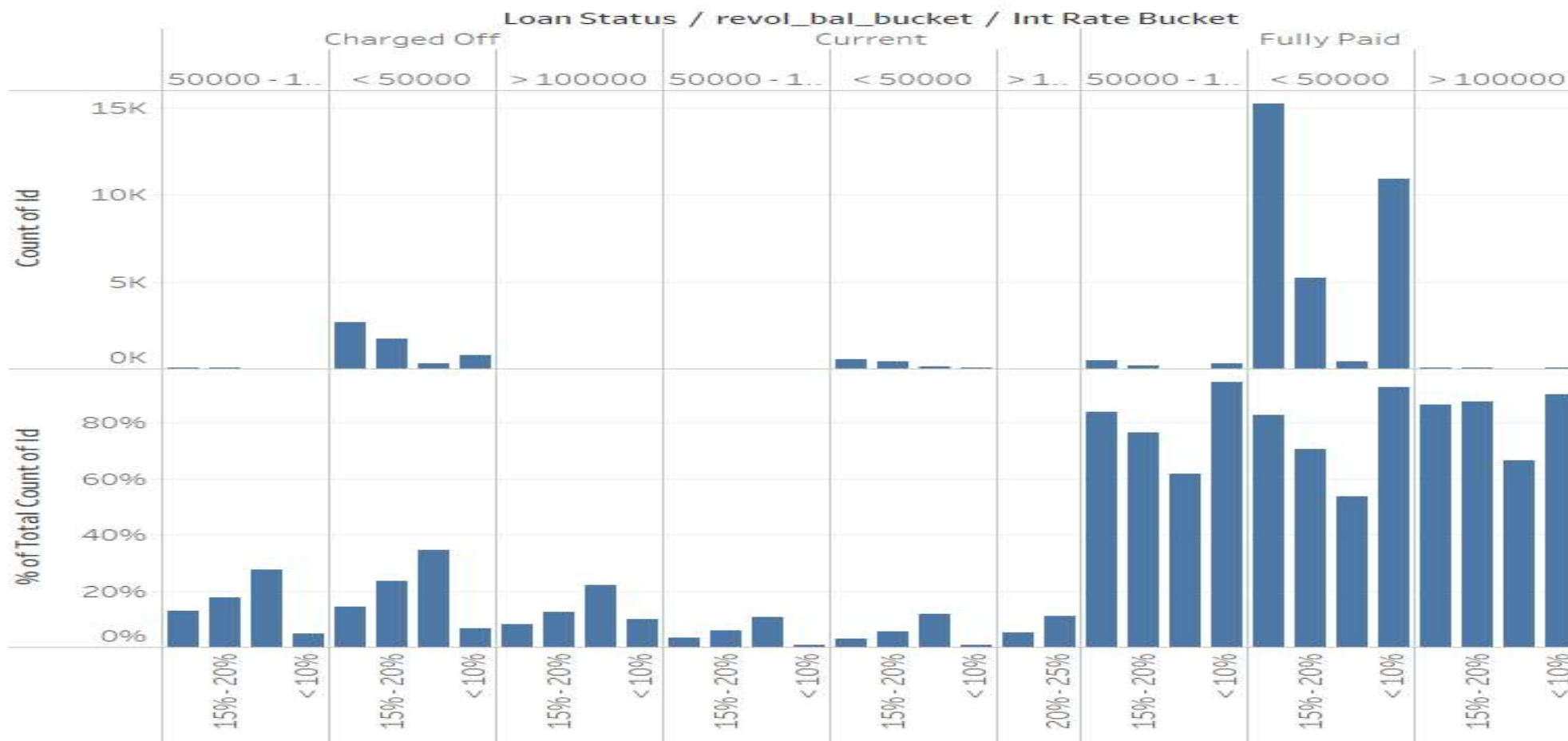
Count of Id and % of Total Count of Id for each revol_util_bucket broken down by Loan Status. The data is filtered on Exclusions (Int Rate Bucket, Loan Status, revol_util_bucket) and Grade. The Exclusions (Int Rate Bucket, Loan Status, revol_util_bucket) filter keeps 65 members. The Grade filter keeps 65 members.

Bivariate Analysis-IV Purpose along Annual Income bucket

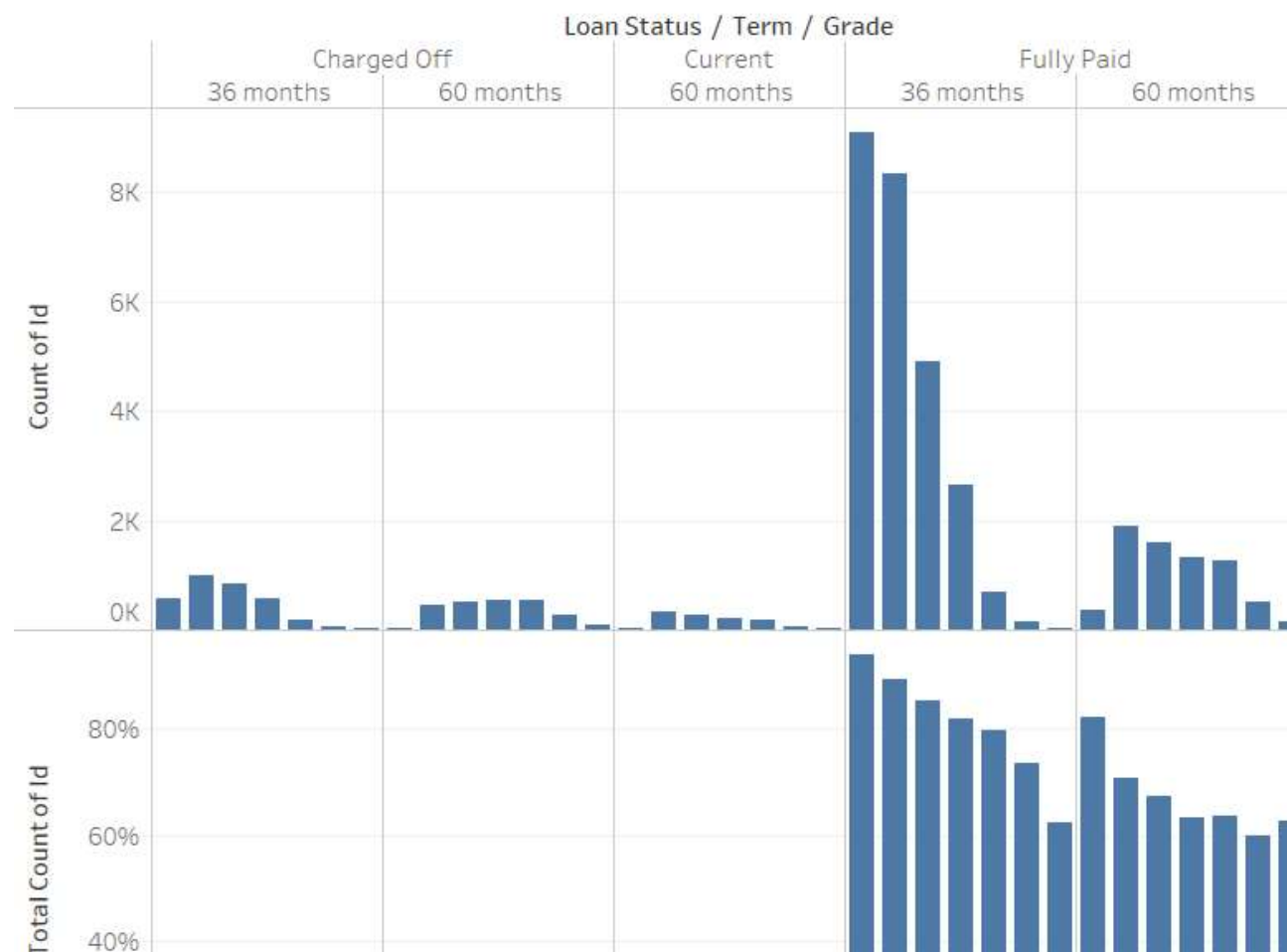
small business+annual inc
bucket



Count of Id and % of Total Count of Id for each Annual Income Bucket (group) broken down by Purpose. Color shows details about Loan Status. The data is filtered for small business.



Sheet 21





Conclusions

- As we have seen various plots, there are some very valuable insights we got, impact of different variables individually and also how other correlate with one another.
- Bankers can definitely use this insights to perform scrutiny on loan candidates, and avoid risk of loosing good customers and avoid defaulters.



Thanks