



# Lending Club Case Study

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# Lending Club Case Study

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Operates a "peer-to-peer" lending website for personal loans. The company assesses applicants' risk and lets investors lend directly to individuals or spread their money across a number of loans.

They want to assess the risk by analysing a dataset with previous loan related data

The purpose is to find out some variables which might help them to predict the occurrence of default in order to lower their risks and increase profits

# Cleaning the Dataset

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# Data Cleaning Summary

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Initial number of rows : 39717

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loan.shape
```

Initial number of columns : 111

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(39717, 111)
```

No Summary rows at the top or bottom of the table found

Number of columns with all “N/A” values = 54

Number of columns found to be not required for analysis = 35

Removed all rows with loan status as “Current”, as current loans will not help us to determine if the loan will end up being default or not

Removed outliers

Cleaned numeric data by removing %, <, >, string values

After Cleaning

Number of rows: 36970

Number of Columns: 23

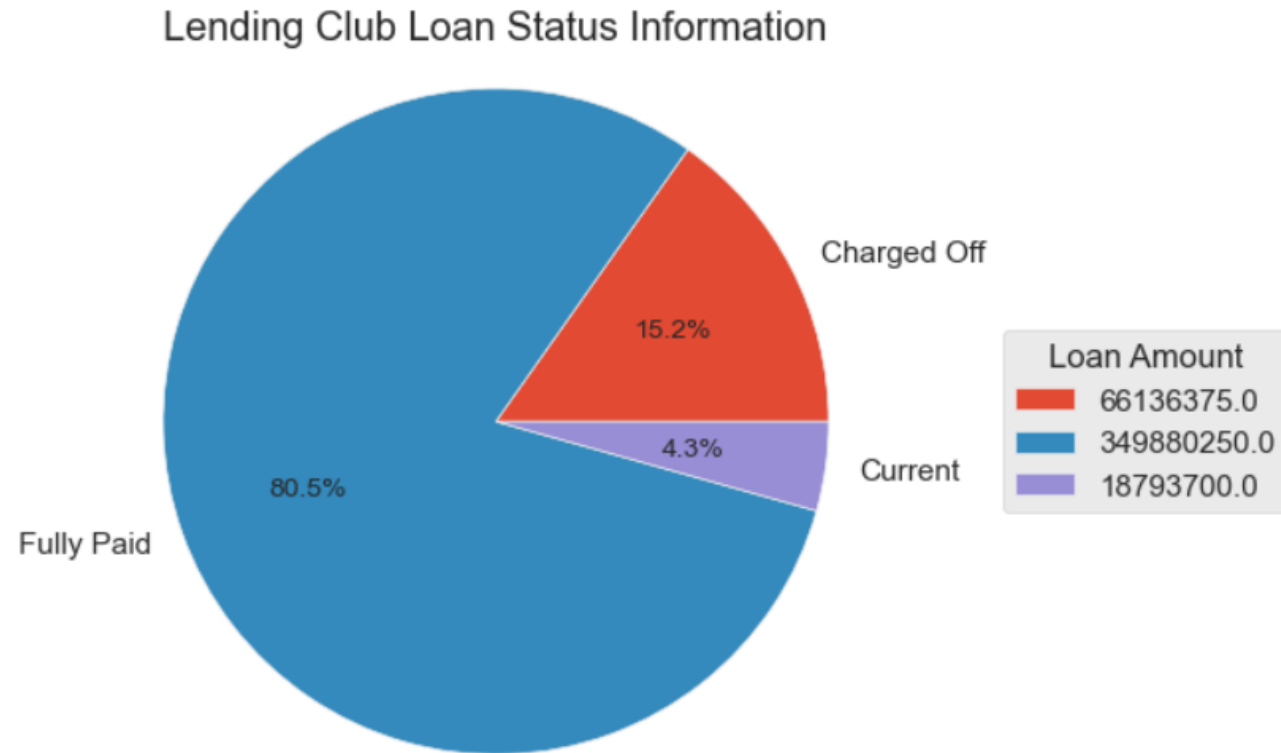
# Loan Status Info

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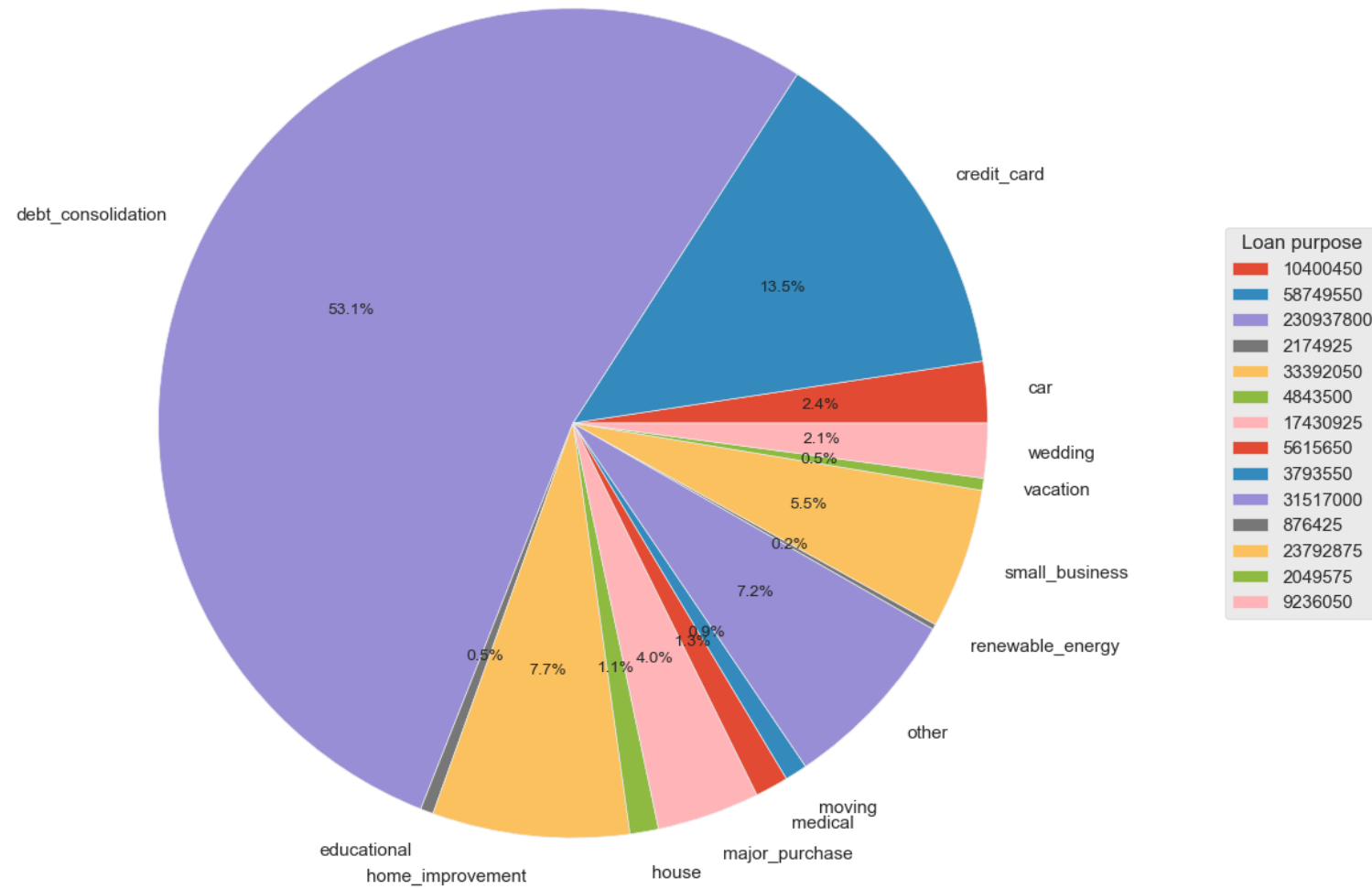
“Current” Loans would not help in determining if the loan would default.

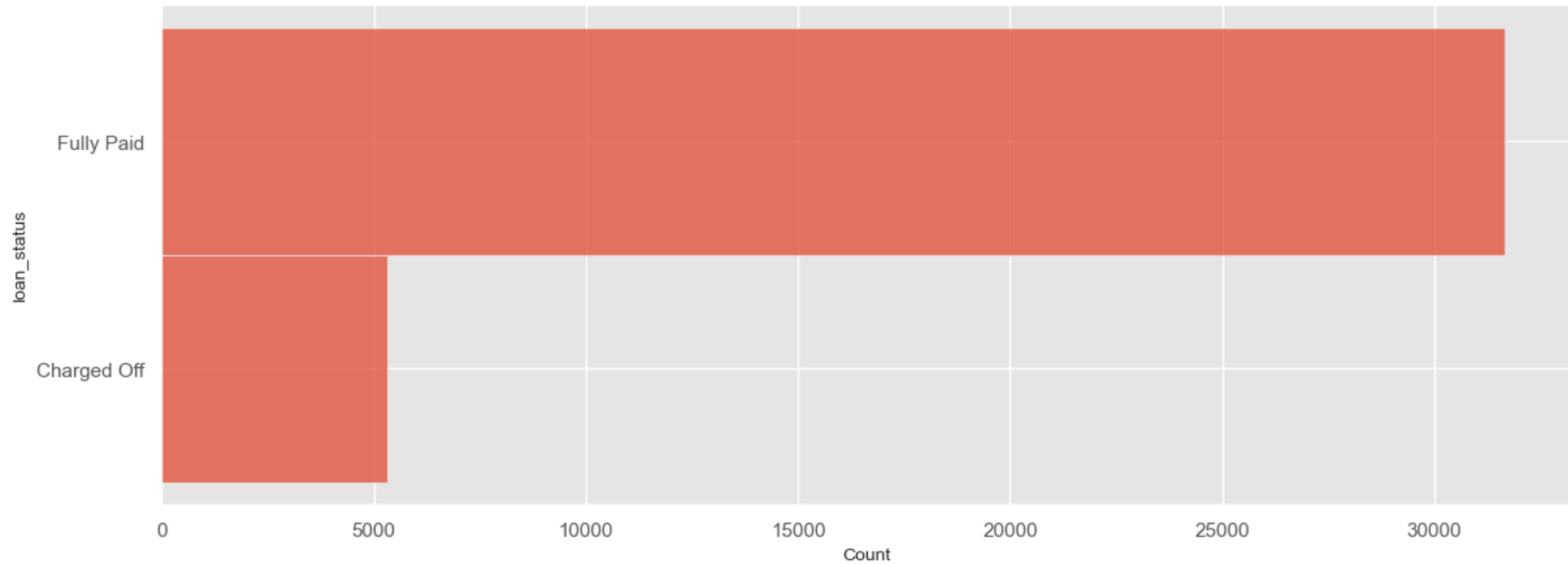
The Pie chart shows the loan status.

We can see that 4.3 % of the loans are current



Lending Club Loan purpose Information





# Loan Status post cleanup

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# Univariate Analysis

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# Purpose and approach

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The purpose of the Univariate analysis is to identify the variables which have an impact on the loan default.

This will help Lending Club to make an informed decision when approving the loan

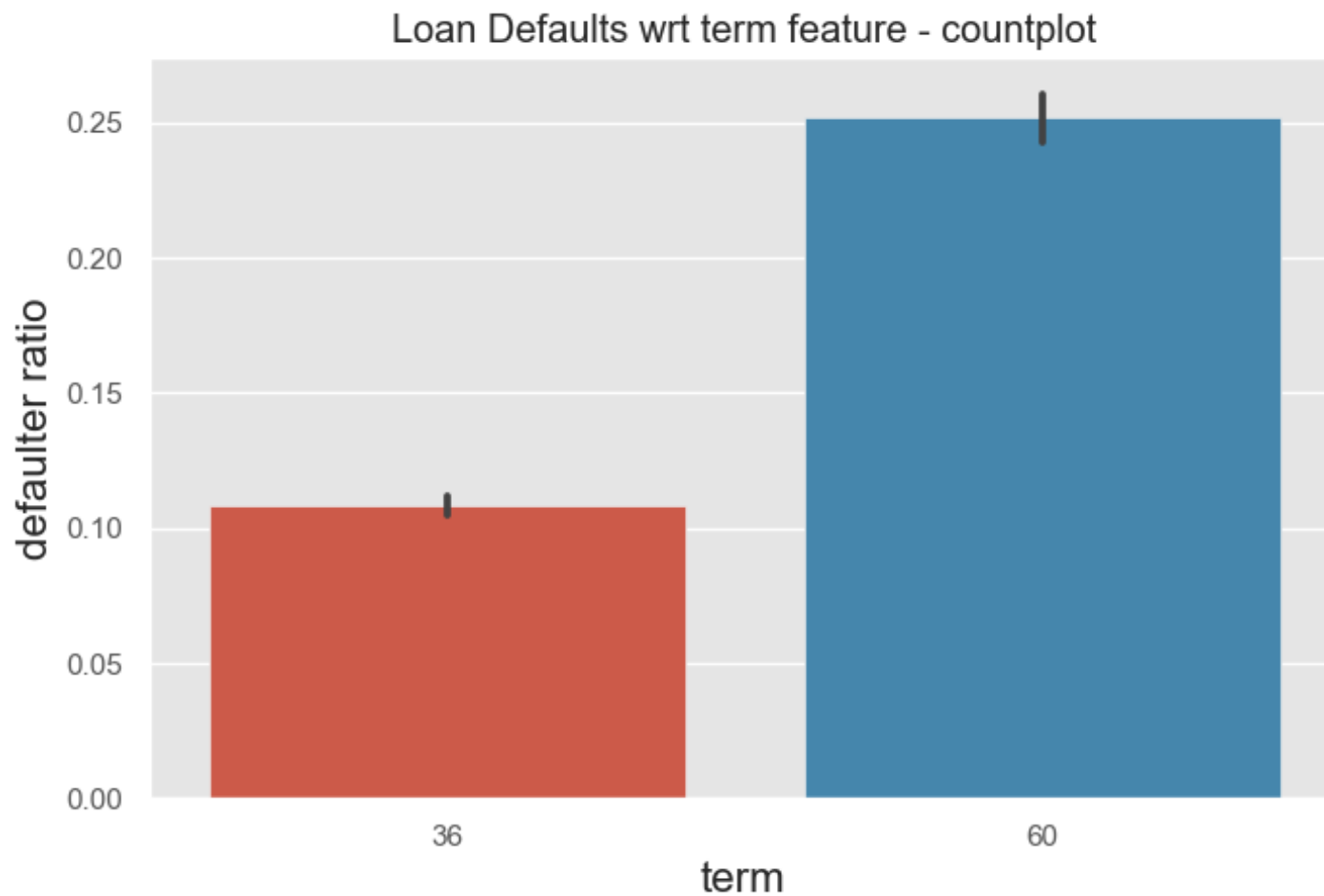
A function has been written which taken the feature as input. The feature can be any of the columns

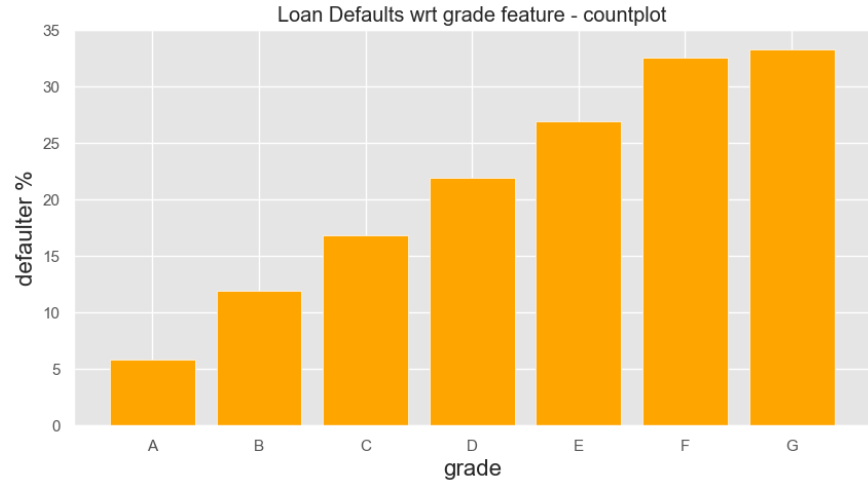
The function plots a countplot with the data and can be used to infer the impact

# Default wrt term

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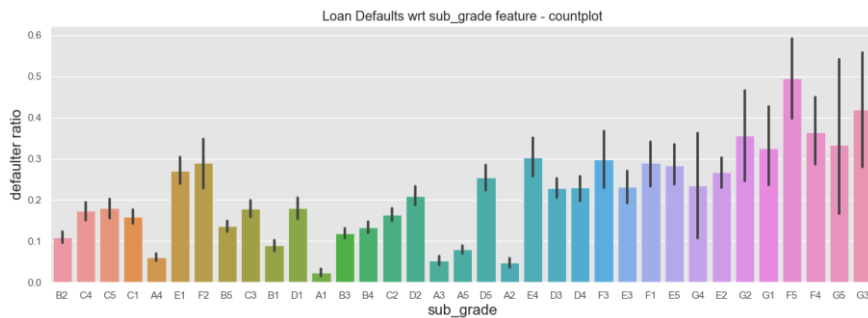
Defaulter rate is higher for loan term of 60 months





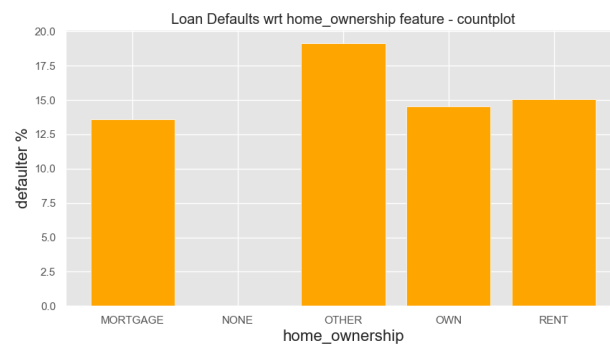
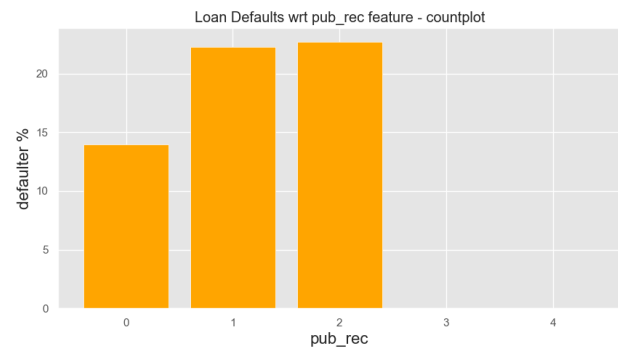
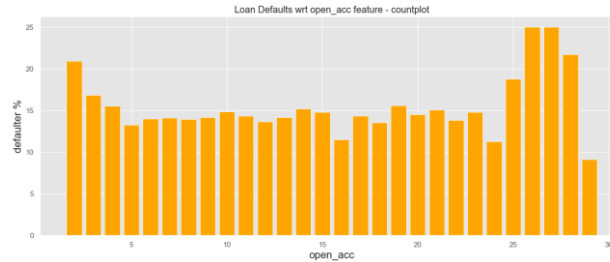
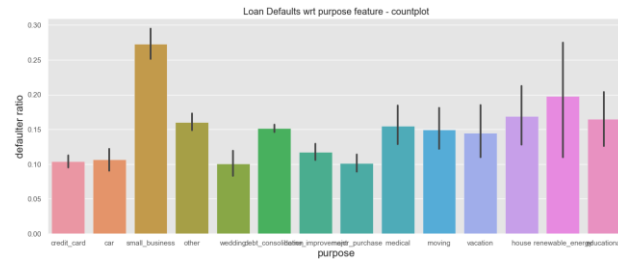
# Default wrt Grade and Subgrade

It can be inferred from the plots that both grade and sub\_grade have a relation with default rate



# Default wrt purpose, open\_acc, pub\_rec, home\_ownership

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We can infer the following from the plots:

Purpose – Big spike noticed for “small\_business”. Thus purpose is relevant and useful.

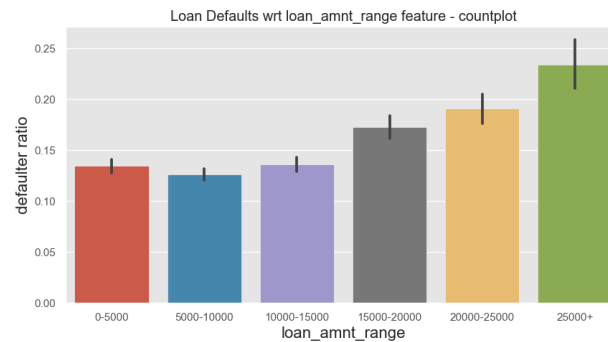
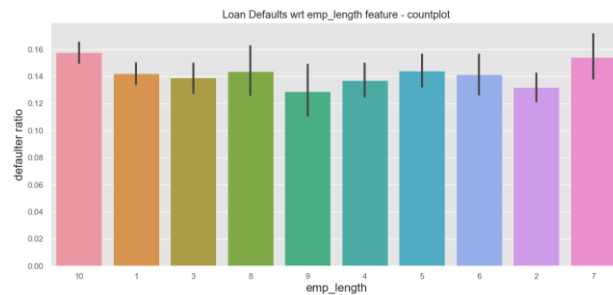
Open\_acc – Almost constant. Its variable is not useful.

Pub\_rec – rate of defaulters rises by > 5% for pub\_rec > 0. We will consider this as useful

Home\_ownership – Only some increase is seen for “Other”, but as we don’t have data on what “Other” comprises of, we will not consider this variable.

# Default wrt emp\_length, loan\_amt\_range, year, earliest\_cr\_line

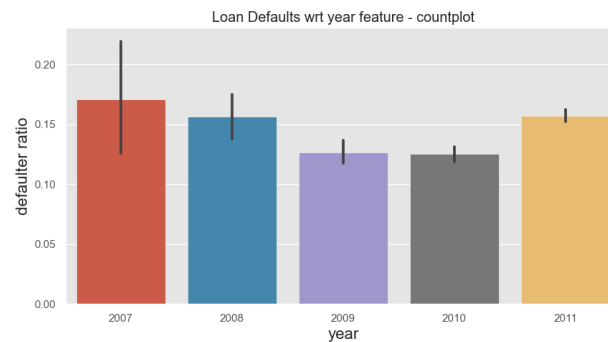
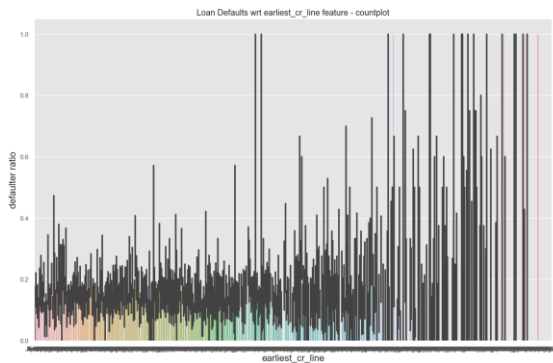
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The following can be inferred from the charts:

Emp\_length – defaulter ratio is almost constant, thus this variable is not useful.

Loan\_amnt\_range – Ratio is increasing, thus this variable is useful.

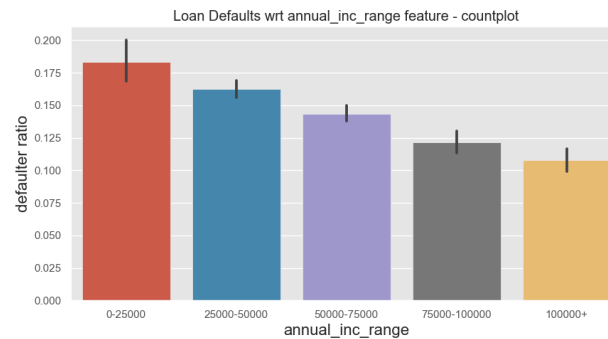
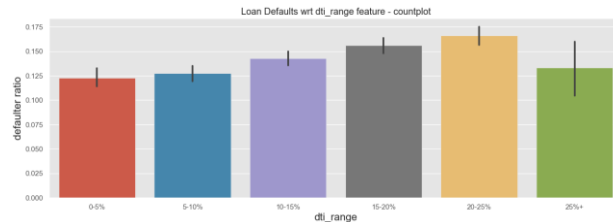


Earliest\_cr\_line – Ratio is almost constant, thus the variable is not useful.

Year – Ratio is almost constant over the years, thus this variable also cannot be useful.

# Default wrt dti\_range, annual\_inc\_range, installment, int\_rate\_range

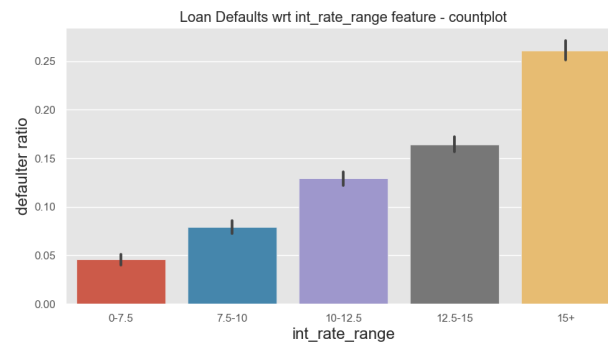
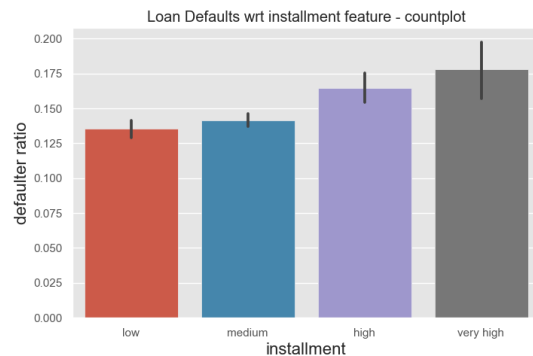
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Following can be inferred:

Dti\_range – Defaulter rate increases with increase of dti. Thus this variable is also useful.

Annual\_inc\_range – Defaulter rate reduces with increase in annual income. Thus this variable is useful.



Installment – Defaulter ratio is increasing with increase in installment, thus this variable is useful.

Int\_rate\_change – Defaulter ratio is increasing with increase in int\_rate. This variable is useful.

# Univariate Results Summary

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The following are the important features we have deduced using univariate analysis:-

term, grade, sub\_grade, purpose, pub\_rec, funded\_amnt, int\_rate, annual\_inc, dti, installment

# Bivariate Analysis

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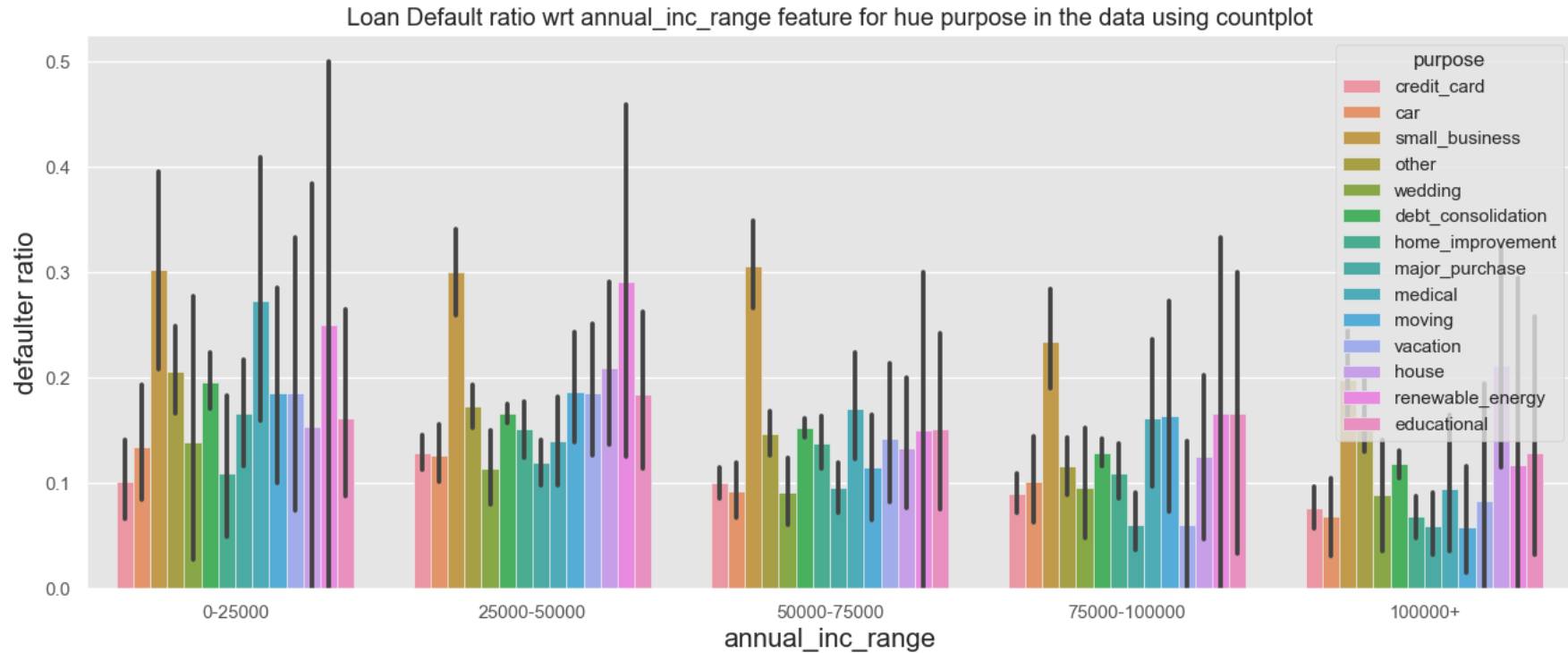
# Purpose and Approach

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The purpose of the Bivariate analysis is to identify the relation between 2 variables and identify if they have an impact on the loan default.

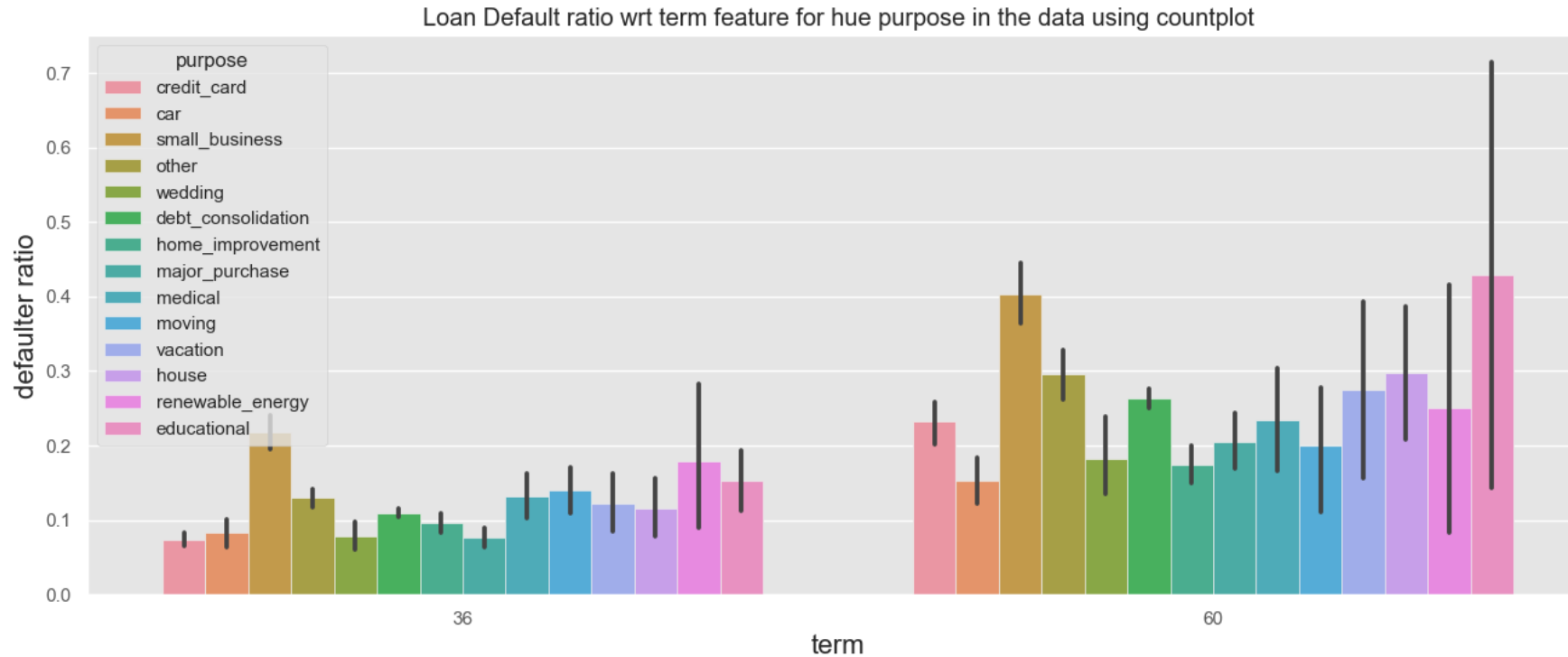
This will help Lending Club to make an informed decision when approving the loan

A function has been written which takes 2 features as input. The features can be any of the columns



# Annual\_inc\_range and purpose

From the above, we can infer that there is no correlation between the annual\_income\_range and purpose wrt defaulter ratio



# Term and purpose

As we can see straight lines on the plot, default ratio increases for every purpose wrt term

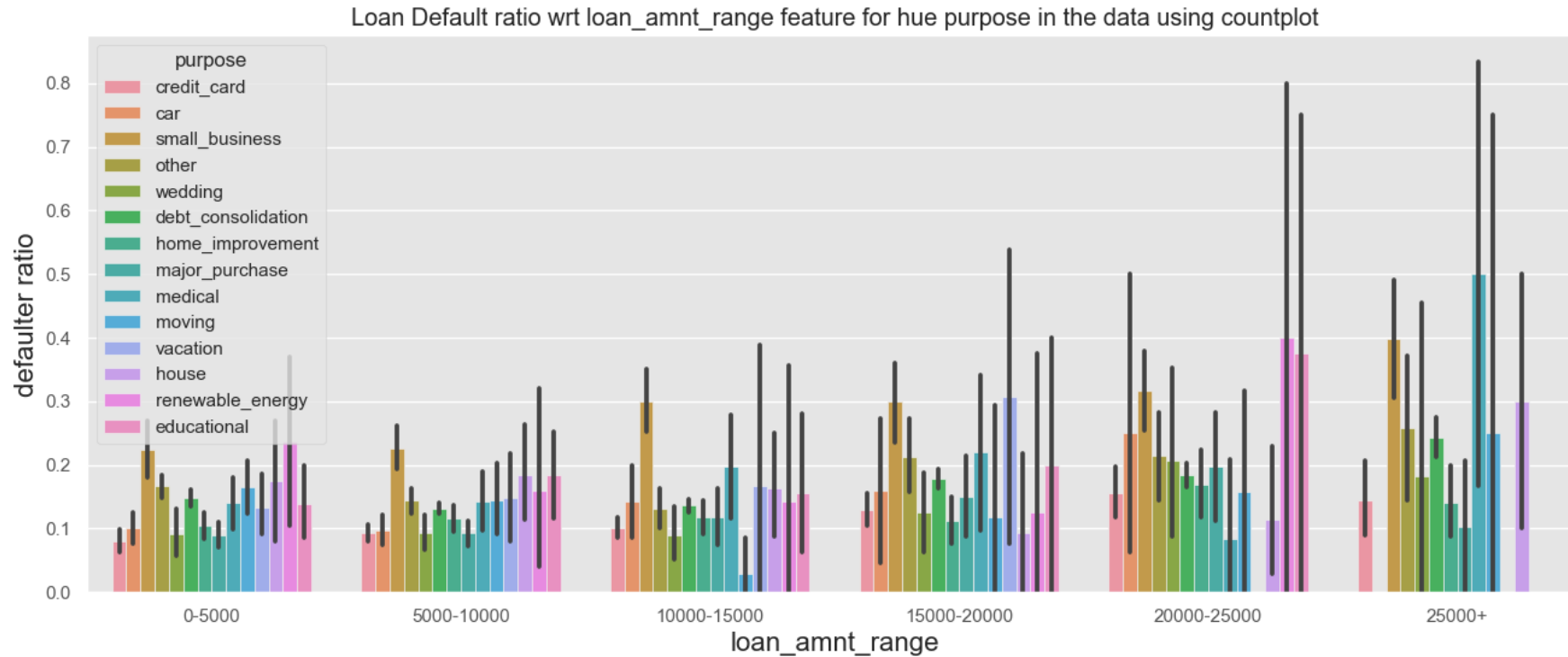
There is a relation



# Grade and prpose

As we can see straight lines on the plot, default ratio increases for every purpose wrt grade

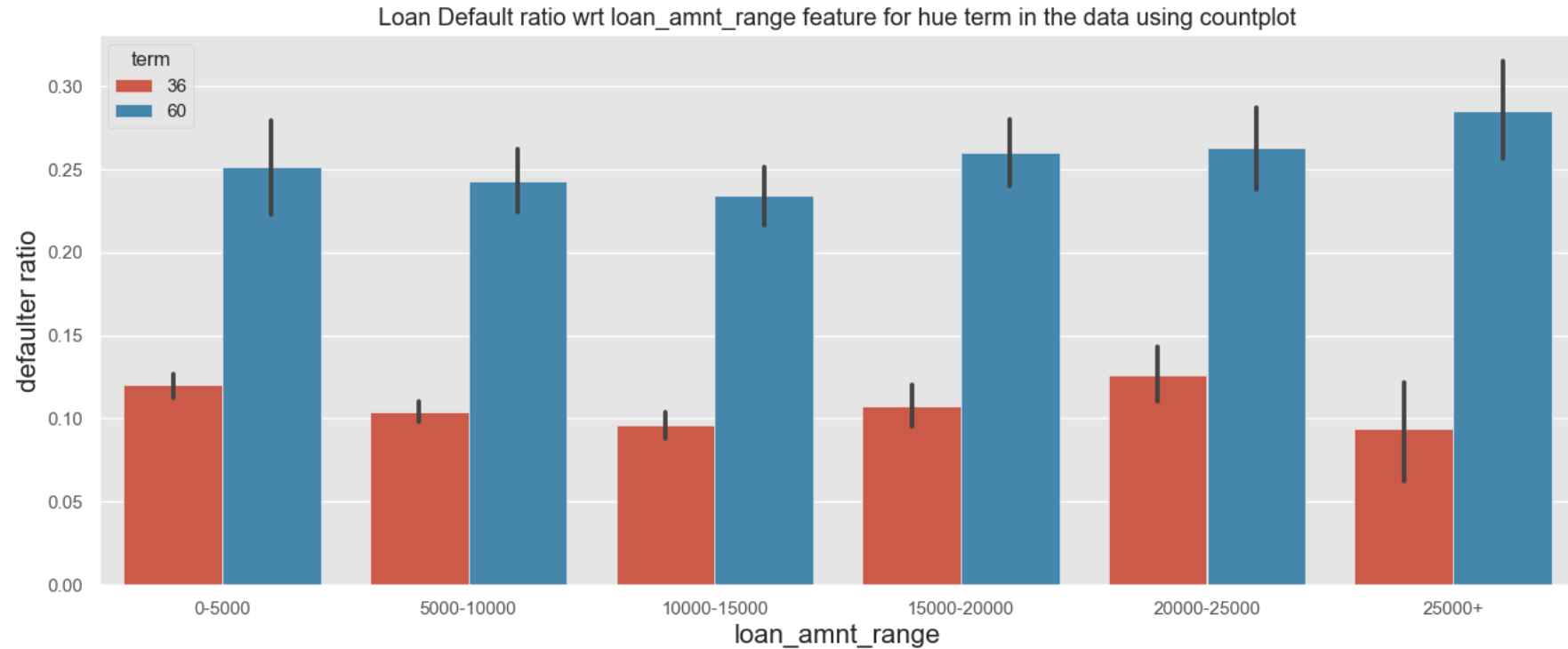
There is a relation



# Loan\_amnt\_range and purpose

As we can see straight lines on the plot, default ratio increases for every purpose wrt loan\_amnt\_range

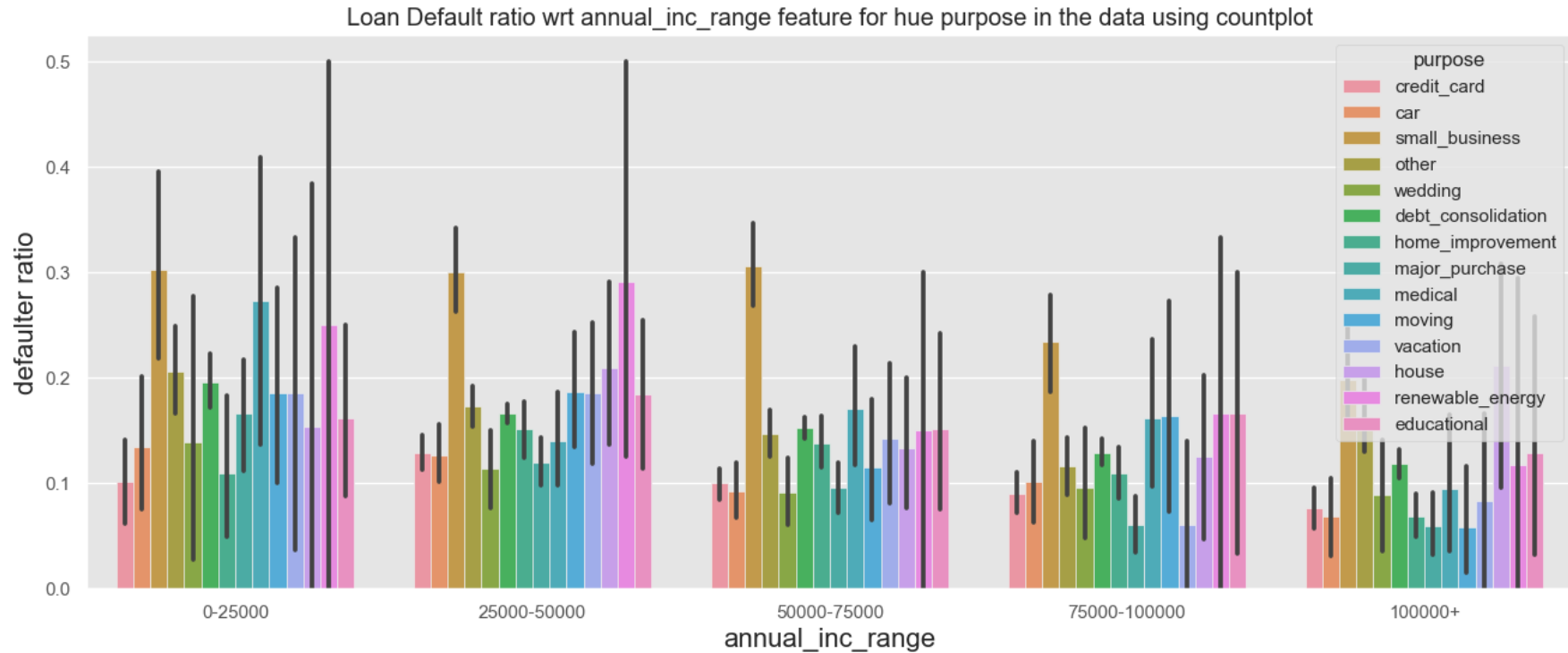
There is a relation



# Loan\_amnt\_range and Term

As we can see straight lines on the plot, default ratio increases for every term wrt loan\_amnt\_range

There is a relation



# Annual\_inc\_range and purpose

As we can see straight lines on the plot, default ratio increases for every purpose wrt annual\_inc\_range

There is a relation

# Final Findings

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After analysing all the related features available in the dataset, we have come to an end, deducing the main driving features for the Lending Club Loan Default analysis:

The best driving features for the Loan default analysis are: term, grade, purpose, pub\_rec, int\_rate, installment, annual\_inc, funded\_amnt