



GRAMENER CASE STUDY SUBMISSION

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Abstract

Summary

"Consumer Finance Company" specialize in lending loans to urban customers. They are interested in understanding the **driving** factors (or driver variables) behind loan default, which can help in identifying such loans applications and improving there portfolio and risk assessment.

Design

The analysis is designed aligned to exploring data set, identifying variables aligned to domain. Some important variables which we consider from domain are as below

Charged (Off: Loan accour	nts which a	re defaul	ted, will	l be a primary	variable in rela	tion to whi	ch other va	ariable are	analysed.

- ☐ Fully Paid: Loan accounts which are fully paid, will serve as an important variable for ratio and comparative analysis.
- ☐ Open accounts: Number of open credit lines which can help us to understand a credit scoring for an application
- ☐ Interest rate: Interest rate at which loan is granted

Assumption

As we can't determine the status of Current accounts whether they will be defaulted or not we are not going to consider them for analysis. The ratio analysis will be based on Charged-Off and Fully Paid accounts.





We try explore the complete dataset *holistically* by establishing a correlation among them as to understand how various variables are related.

Assumption: To establish a correlation for some of the categorical variables we use label encoding to represent categorical bins to number.

Note: In next consecutive exploratory analysis we are going to detail out specific variables

																					_	
loan_amnt	1	0.98	0.94	0.3	0.93	0.27	0.061	-0.034	0.013	0.17	-0.046	0.31	0.26	0.88	0.85	0.85	0.73	0.047	0.14	0.076	0.47	-0.032
funded_amnt	0.98	1	0.96	0.3	0.96	0.26	0.061	-0.034	0.013	0.17	-0.047	0.31	0.25	0.9	0.86	0.86	0.74	0.049	0.14	0.078	0.48	-0.033
funded_amnt_inv	0.94	0.96	-1	0.3	0.9	0.25	0.07	-0.04	-0.0028	0.16	-0.048	0.29	0.24	0.87	0.91	0.84	0.73	0.029	0.13	0.064	0.47	-0.038
int_rate	0.3	0.3	0.3	1	0.27	0.046	0.11	0.16	0.13	0.0076	0.099	0.095	-0.045	0.29	0.28	0.17	0.52	0.095	0.13	0.07	0.17	0.083
installment	0.93	0.96	0.9	0.27	1	0.26	0.051	-0.022	0.011	0.17	-0.042	0.31	0.23	0.86	0.82	0.85	0.64	0.058	0.12	0.077	0.41	-0.029
annual_inc	0.27	0.26	0.25	0.046	0.26	1	-0.12	0.021	0.036	0.15	-0.014	0.28	0.23	0.25	0.24	0.25	0.18	0.0062	0.022	0.016	0.14	-0.012
dti	0.061	0.061	0.07	0.11	0.051	-0.12	1	-0.034	0.00045	0.29	-0.0047	0.23	0.23	0.059	0.066	0.037	0.1	-0.012	0.025	0.011	0.0093	0.0067
delinq_2yrs	-0.034	-0.034	-0.04	0.16	-0.022	0.021	-0.034	1	0.0083	0.01	0.01	-0.054	0.066	-0.025	-0.031	-0.04	0.022	0.029	0.012	0.014	-0.013	0.0057
inq_last_6mths	0.013	0.013	-0.0028	0.13	0.011	0.036	0.00045	0.0083	1	0.094	0.023	-0.021	0.11	-0.0084	-0.019	-0.021	0.026	0.031	0.018	0.012	0.028	0.014
open_acc	0.17	0.17	0.16	0.0076	0.17	0.15	0.29	0.01	0.094	1	0.0027	0.29	0.69	0.16	0.15	0.16	0.12	-0.019	0.017	0.006	0.08	0.0092
pub_rec	-0.046	-0.047	-0.048	0.099	-0.042	-0.014	-0.0047	0.01	0.023	0.0027	1	-0.059	-0.02	-0.05	-0.05	-0.061	-0.0016	-0.0034	-0.0063	-0.0071	-0.033	0.84
revol_bal	0.31	0.31	0.29	0.095	0.31	0.28	0.23	-0.054	-0.021	0.29	-0.059	1	0.31	0.29	0.27	0.28	0.24	0.0057	0.043	0.022	0.13	-0.046
total_acc	0.26	0.25	0.24	-0.045	0.23	0.23	0.23	0.066	0.11	0.69	-0.02	0.31	1	0.22	0.22	0.23	0.15	-0.023	0.024	0.011	0.17	-0.0066
total_pymnt	0.88	0.9	0.87	0.29	0.86	0.25	0.059	-0.025	-0.0084	0.16	-0.05	0.29	0.22	1	0.97	0.97	0.82	0.014	0.03	0.028	0.51	-0.039
total_pymnt_inv	0.85	0.86	0.91	0.28	0.82	0.24	0.066	-0.031	-0.019	0.15	-0.05	0.27	0.22	0.97	1	0.94	0.81	0.00047	0.023	0.018	0.5	-0.043
total_rec_prncp	0.85	0.86	0.84	0.17	0.85	0.25	0.037	-0.04	-0.021	0.16	-0.061	0.28	0.23	0.97	0.94	1	0.68	-0.02	-0.095	-0.059	0.57	-0.049
total_rec_int	0.73	0.74	0.73	0.52	0.64	0.18	0.1	0.022	0.026	0.12	-0.0016	0.24	0.15	0.82	0.81	0.68	1	0.078	0.092	0.042	0.24	0.0017
total_rec_late_fee	0.047	0.049	0.029	0.095	0.058	0.0062	-0.012	0.029	0.031	-0.019	-0.0034	0.0057	-0.023	0.014 -	0.00047	-0.02	0.078	1	0.1	0.095	-0.063	-0.0053
recoveries	0.14	0.14	0.13	0.13	0.12	0.022	0.025	0.012	0.018	0.017	-0.0063	0.043	0.024	0.03	0.023	-0.095	0.092	0.1	1	0.8	-0.072	-0.0055
ollection_recovery_fee	0.076	0.078	0.064	0.07	0.077	0.016	0.011	0.014	0.012	0.006	-0.0071	0.022	0.011	0.028	0.018	-0.059	0.042	0.095	0.8	1	-0.043	-0.0071
last_pymnt_amnt	0.47	0.48	0.47	0.17	0.41	0.14	0.0093	-0.013	0.028	0.08	-0.033	0.13	0.17	0.51	0.5	0.57	0.24	-0.063	-0.072	-0.043	- 1	-0.023
pub_rec_bankruptcies	-0.032	-0.033	-0.038	0.083	-0.029	-0.012	0.0067	0.0057	0.014	0.0092	0.84	-0.046	-0.0066	-0.039	-0.043	-0.049	0.0017	-0.0053	-0.0055	-0.0071	-0.023	- 1
	ban_amnt	funded_amnt	funded_amnt_inv	nt_rate	nstallment	annual_inc	€	deling_2yrs	nq_last_6mths	open_acc	par_duq	revol_bal	btal_acc	btal_pymnt	total_pymnt_inv	btal_rec_prncp	total_rec_int	total_rec_late_fee	recoveries	ction_recovery_fee	last_pymnt_amnt	rec_bankruptcies

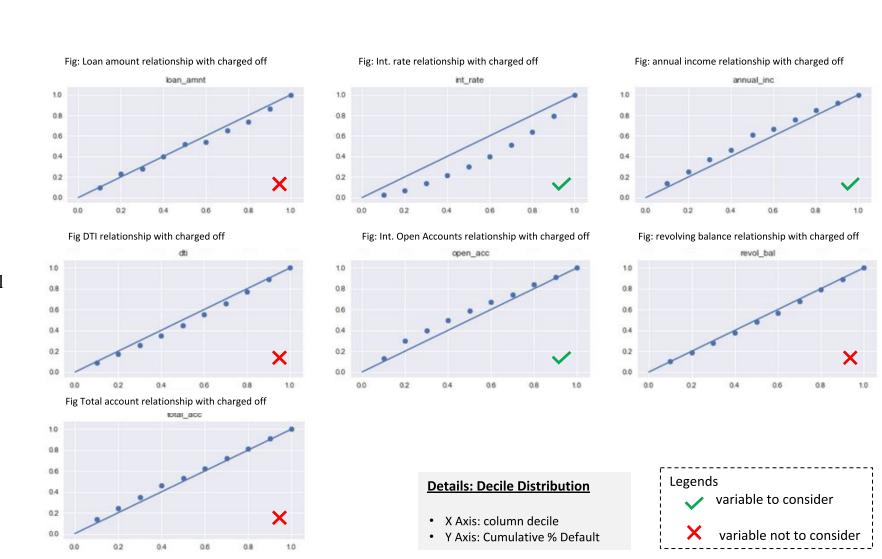




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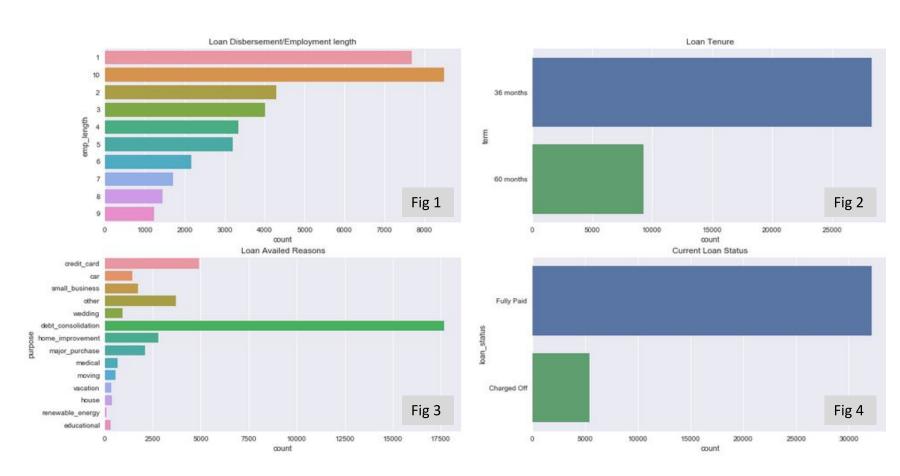




Univariate Exploration

We explore data set available to understand how the loan accounts are distributed around various variables

- 1. Fig 1: Top 3 employment length got maximum loans are 10,2 and 1
- 2. Fig 2: All loans tenure are either 36 month (short term) and 60 months (long term)
- 3. Fig 3: Loan distribution across various purpose for which load been disbursed
- 4. Fig 4: Loan application proportion of Fully Paid over Charged Off



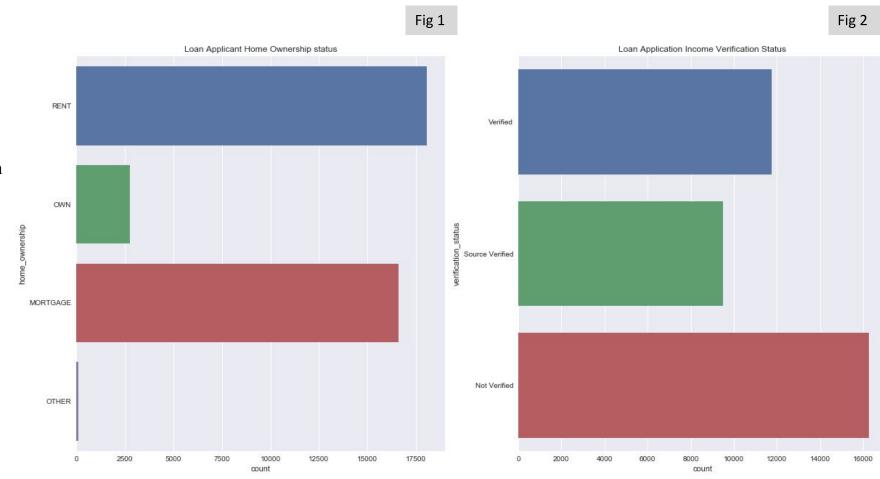




Univariate Exploration

We explore data set available to understand how the loan accounts are distributed around various variables

- 1. Fig 1: Home ownership distribution of loan applicants
- 2. Fig 2: Verification status of loan applications



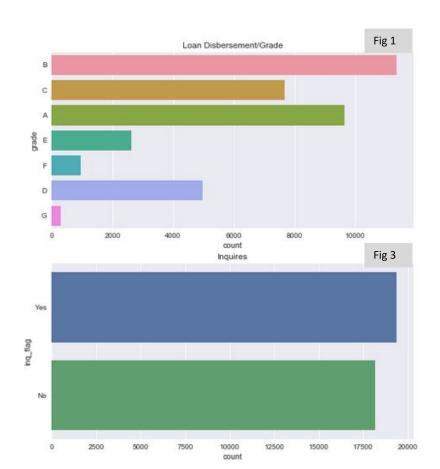


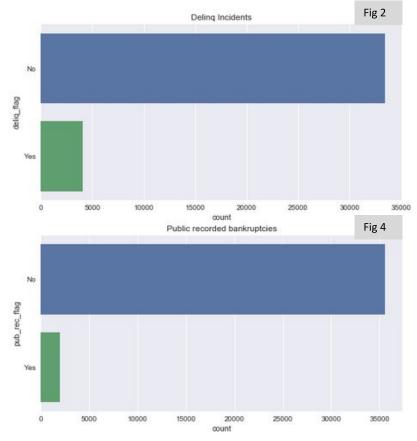


Univariate Exploration

We explore data set available to understand how the loan accounts are distributed around various variables

- 1. Fig 1: Loan accounts distribution over Grade
- 2. Fig 2: Delinquency incidents reported for loan accounts
- 3. Fig 3: Inquires distribution
- 4. Fig 4: Public recorded bankruptcies









Bivariate Exploration

We explore data set bivariable to understand how the loan accounts and home ownership are related

Identification:

1. It indicates that maximum *charged-off* occurred for loan accounts where home ownership status are either *rent* or *mortgage*

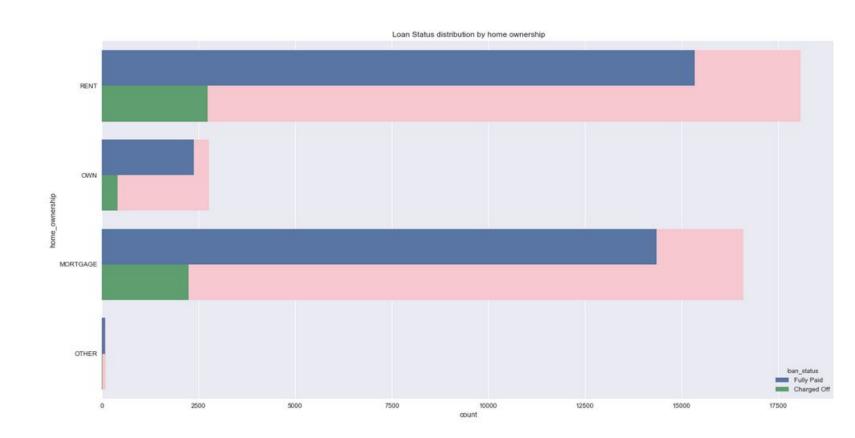






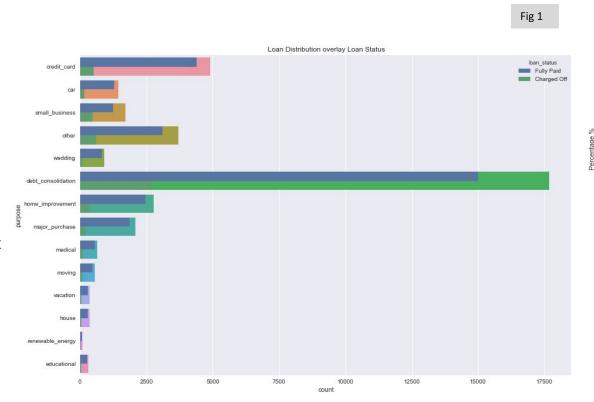
Fig 2

Exploring Data Sets

Bivariate Exploration

We explore data set bivariable to understand how the loan accounts and loan purpose are distributed

- 1. Fig 1: It indicates that maximum charged-off occurred for loan accounts where purpose of loan is debt consolidation, small business, other and credit cards
- 2. Fig 2: It shows that loan charged-off % is very evenly distributed across employment length.









Pre-analysis Summary

After exploring various variables of data set in relationship with charged off or Fully Paid accounts we concluded and identified a list of variables listed in below table.

All variables identified below are attributed by if they hold direct/indirect relationship with charged off, do they have a conjunct variable which forms the basis of bivariate analysis and if they need to be considered at there value or need to be ranged.

Variables	Relationship with charge off	Conjunct variable	Range variable
int_rate	Direct	-	False
open_acc	Indirect	-	True
emp_length	Indirect	annual_inc	False
annual_inc	Direct	emp_length	True
purpose	Direct	-	False
term	Direct	-	False





Charged Off Analysis

As we have observed in data exploration there is not much **strong one to one** relationship between variable and 'Charged-Off'%'. We are going to perform analysis based on bivariable

Bivariate Analysis

We explore data set bivariable how various variables are in groups are contributing to default rated i.e. 'Charged-Off'%'

Identification:

We able to identify some of the variables which we will be looking for our analysis due to there relationship to the *'Charged -Off'* accounts.

	Category_1	Category_2	Level_1	Level_2	Def%	Def%_LC1	Def%_LC2	Pop%	Pop%_LC1	Pop%_LC2	base_Percent
307	term	home_ownership	60 months	RENT	27.9	25.1	15.1	9.9	24.7	48.2	14.4
343	term	inq_flag	60 months	Yes	27.3	25.1	16.6	13.6	24.7	51.6	14.4
4	grade	deliq_flag	E	No	27.2	26.7	14.2	5.8	7.0	89.1	14.4
381	term	purpose	60 months	debt_consolidation	26.2	25.1	15.2	12.9	24.7	47.1	14.4
139	inq_flag	grade	Yes	D	23.7	16.6	21.8	7.7	51.6	13.3	14.4
3	grade	deliq_flag	D	No	22.5	21.8	14.2	10.9	13.3	89.1	14.4
43	inq_flag	deliq_flag	Yes	Yes	18.4	16.6	16.2	5.8	51.6	10.9	14.4
115	inq_flag	grade	Yes	С	18.3	16.6	16.8	13.0	51.6	20.4	14.4
337	purpose	inq_flag	other	Yes	18.2	16.1	16.6	5.1	9.9	51.6	14.4
301	purpose	home_ownership	other	RENT	18.1	16.1	15.1	5.4	9.9	48.2	14.4

Index

- Def%: default rate aggregated
- Def% LC1: default rate of Level 1 & Category 1
- Def% LC2: default rate of Level 2 & Category 2
- Pop%: overall population %
- Pop%_LC1: population % of Level 1 & Category 1
- Pop%_LC2: population % of Level 2 & Category 2





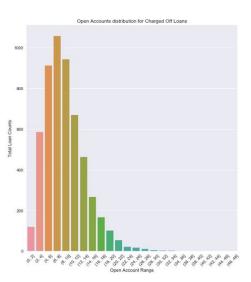
Charged Off Analysis

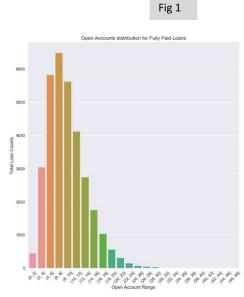
Bivariate Analysis

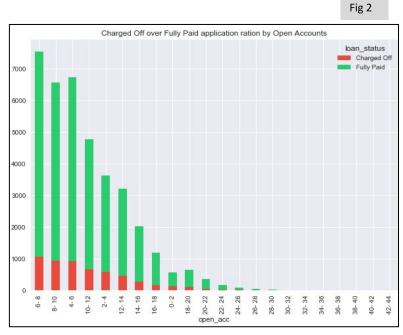
We grouped open accounts with range index of 2 and explored loan account distribution of charged off and fully paid accounts

Observation:

Fig 1: The distribution of charged off accounts and fully paid accounts on open-accounts range is a uniform normal distribution







Identification:

Fig 2: The account having open accounts in range of 2-10 will be contributing to highest default rate as there <u>credit lines are low</u>.





Fig 2

Charged Off Analysis

Bivariate Analysis

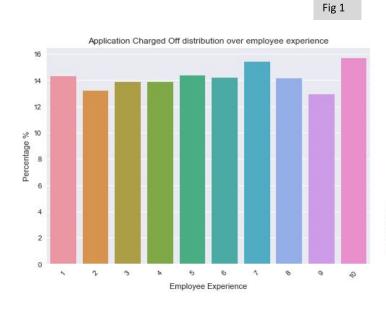
As we observed in employment length alone was unable to provide an insight grouping same with annual income provide an interesting insight

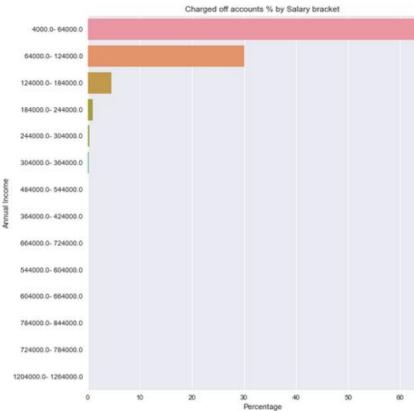
Observation:

Fig 1: uniform normal distribution of charged off accounts over employment length

Identification:

Fig 2: Loan account with employee experience of 1,2 and 10 years if in annual income bracket of 4,000-1,24,000 will hold high probability of defaulting loans.









Summary

As per analysis made below is the summary, which can help in filtering loan application which can account to default.

S.No.	Variables	Conjunct variable	Analysis
1	emp_len		A loan give to an applicant bearing employment experience of 1,2 & 10 year hold high chances of default. The possible reason can be either job instability for an experience employee, a less experience employee may be pursuing higher education
2	Int_rate	term	A high interest rate reflects a relationship with charged off and in conjunction if the loan term is of long term i.e. 60 months it is likely that a person may default the loan.
3	term	home_ownership, purpose,int_rate,emp_len	A long term loan i.e. 60 months if taken for debt_consolidation or for small business with home_ownership status of Rent or mortgage it's very likely to default.
4	open_acc		An open_acc between range of 0-10 reflect a very instable/low credit line and hold very high probability of defaulting a loan account.
5	annual_inc	emp_len,term,int_rate,open_acc	An annual income range of 4000-124000/annum if in conjunction with employment experience of $1/2/10$ offered a loan of long term with high interest rate & low credit lines holds very high probability to default