

LendingClub

Credit Risk Modeling Project

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Lending Club Business Model

Lending Club (www.LendingClub.com) is a San Francisco based online provider of loans for debt consolidation, small business expenses, moving costs, home improvements, medical expenses or other select major one time purchase or expenditures.

Lending Club's website advertises consolidating your variable, high interest debt with a single, low, fixed rate loan with APRs ranging from 5.99% to 35.96% (depending on credit score, loan amount, loan terms, and credit usage and history).



Business Problem Statement and Objective

Statement of Problem

Improve on Lending Club's existing risk management and risk pricing strategy in order to maximize economic value to the owners of the organization.

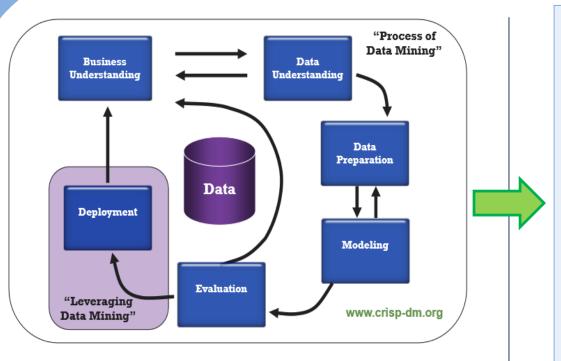
Objectives

- Utilize historical loan default data to credit risk segmentation used in credit scoring so proper risk pricing component can be utilized to ensure returns are achieved across entire credit spectrum during pricing process.
- This will be accomplished if the team can build a credit scoring model that will utilize the existing independent variables that will have higher predictive power over the legacy model, and result in a lower default rate, which will increase operating profits (and shareholder value).





CRISP-DM Approach



- **Exploratory Data Analysis**
- Created CAR Customer **Analytics Record**
- **Excluded Rejects**
- Classification / Regression Model using CAR
- Evaluated the Model based on Probability Scores
- Built Model 2 Weighted Regression
- Appended the Probability Score to the CAR
- **Created Supervised** Clustering Model based on CAR





	Class and Years										
Class	Year	Loan Count	Method								
Α	2007-2011	42,535	FICO								
В	2012-2013	188,123	FICO & Variable Measures								
С	2014-2015	656,724	FICO & Revised Variable Measures								

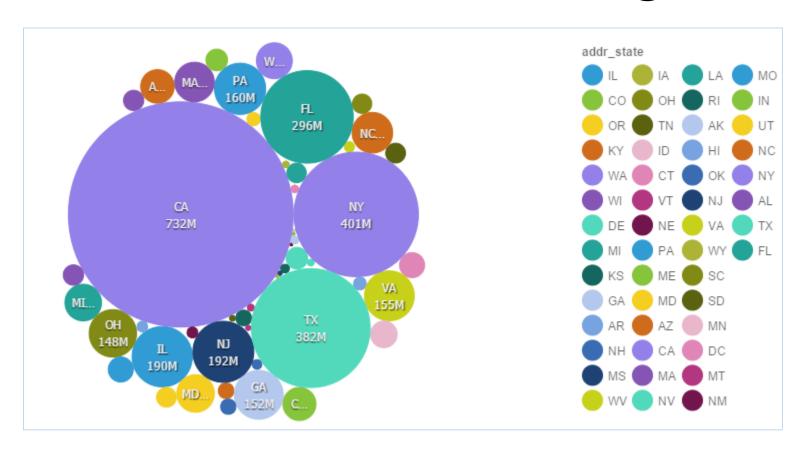
	Variable Measures											
Borrower Assessmen 🔻	Loan Characteristic 🔻	Borrower Assessment 💌	Credit History	Borrower Indebtedness								
Grade	Loan Purpose	Annual Income	Credit History Length	Loan Amount to Annual Income								
Sub- Grade	Loan Amount	Housing Situation	Delinquency 2 years	Annual Installment to Income								
Interest Rate		Emp Length	Inquires Last 6 mo	Debt to Income								
			Public Records									
			Revolving Utilization									
			Open Accounts									
			Months Since Last Delinq									

	Lending Club Interesting Facts												
Variable	Best	ROI	Worst	Loss									
Grade:	E	8.99%	G	18.65%									
Purpose:	Wedding	8.07%	Small Business	13.01%									
Year:	2013	8.29%	2007	14.80%									
FICO:	>674 <= 679	8.33%	<=664	8.54%									
Annual Income:	>150K	8.74%	<= 28K	8.10%									
Inquires:	<=0	8.08%	>3	10.94%									

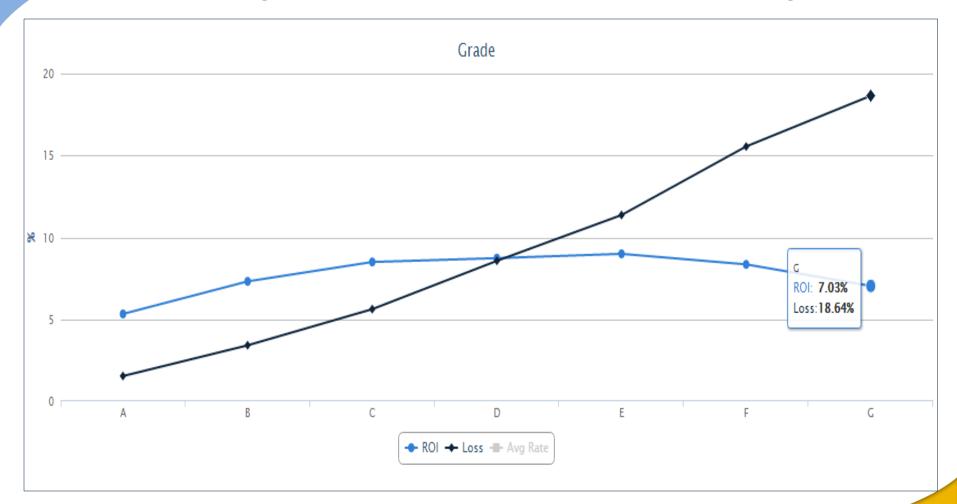




Data Understanding

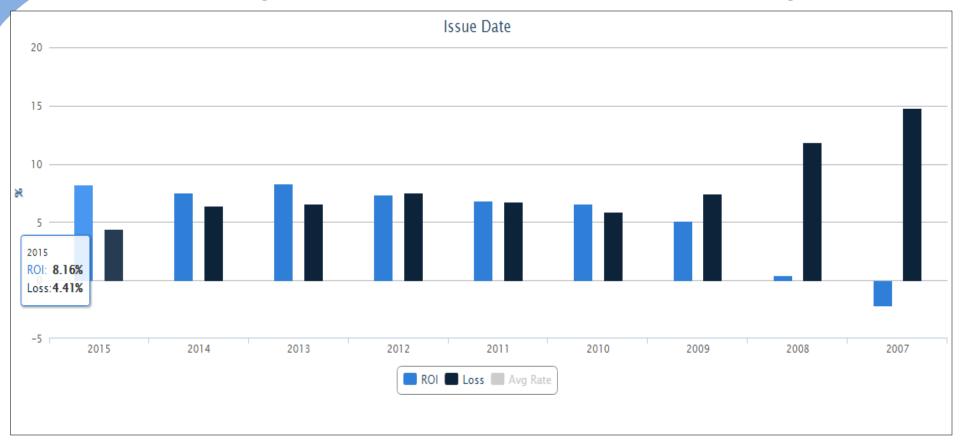








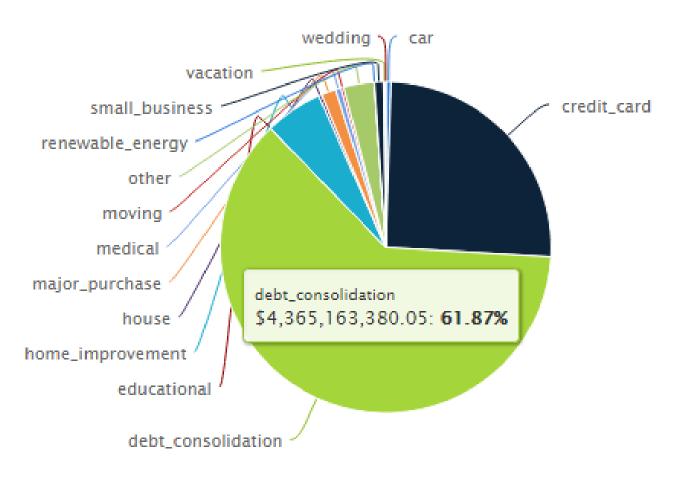






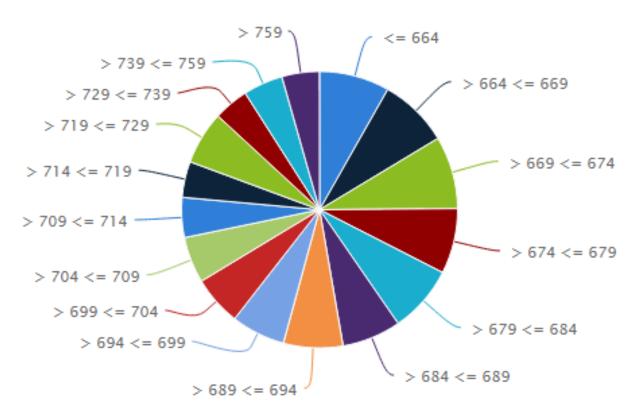


Purpose





Fico Range High







Classification Model – Executive Summary

- The credit risk modeling team utilized SAP Predictive Analytics to perform data discovery on Lending Club's historical loan portfolio and then create a classification model using the statistically relevant independent variables to improve the prediction of a borrower's probability of default.
- This modeling strategy yielded an improved identification and segmentation of the independent variables and resulted in a measurable improvement to the existing probability of default rate of 49.89%

Classification Model 1 - Summary

Eligible Population

Number of records: 333,162

•Description of the population on which the score is computed

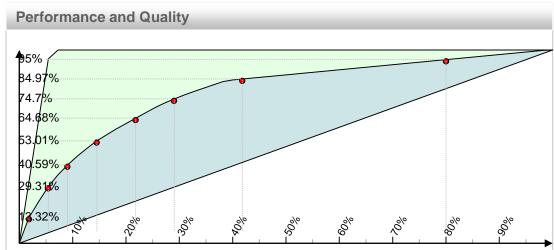
Target Population

Observed Target Frequency: 5.73%

Description of the Modeled Phenomena

Predictive Power (KI) = 0.601 Prediction Confidence (KR) = 0.995





By targeting 5% of the clients we detect 28% of the entire population of interest, or 5.4 times better than without a predictive model

Variable contribution

Less contributive	Variable Name	Most contributive
{c}	loan_type	{a}
{163;168;169;171;172;174;	mo_sin_old_rev_tl_op	KxMissing
{A1}	sub_grade	{KxOther}
{10;11;3;4}	num_bc_sats	KxMissing
]13 ; 211]	mo_sin_rcnt_tl	KxMissing
{0;KxOther}	num_tl_30dpd	KxMissing
[0;37]	num_actv_rev_tl	KxMissing
]96000 ; 7.4464e+006]	annual_inc	[1896 ; 26700] & KxMissing
[52900 ; 560800]	total_bc_limit	KxMissing
{0}	num_tl_op_past_12m	KxMissing





Classification Model 2 Summary

Eligible Population

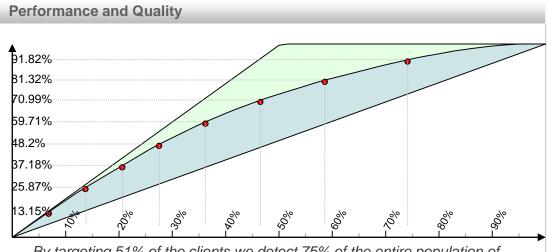
Number of records: 38,136

•Description of the population on which the score is computed

Target Population

Observed Target Frequency: 49.89%

•Description of the Modeled Phenomena



By targeting 51% of the clients we detect 75% of the entire population of interest, or 1.5 times better than without a predictive model

Variable contribution

Less contributive	Variable Name	Most contributive
{a}	loan_type	{c}
KxMissing	num_tl_120dpd_2m	{KxOther;0}
KxMissing	num_actv_rev_tl	[0; 29]
{E4;F2;F3;F4;F5}	sub_grade	{A1;A2}
[1896 ; 22440] & KxMissing	annual_inc]94000 ; 136500]
{f}	met	{t}
KxMissing	mths_since_recent_bc]62 ; 538]
KxMissing	tot_cur_bal]273322 ; 2.54717e+006]
KxMissing	acc_open_past_24mths	[0; 2[
KxMissing	num_actv_bc_tl	[1;18]

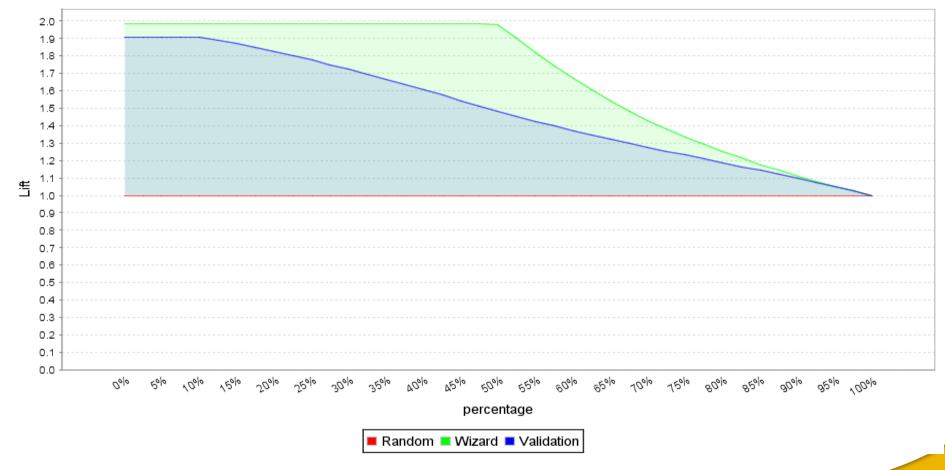
Predictive Power (KI) = 0.652

Prediction Confidence (KR) = 0.989

★★★★

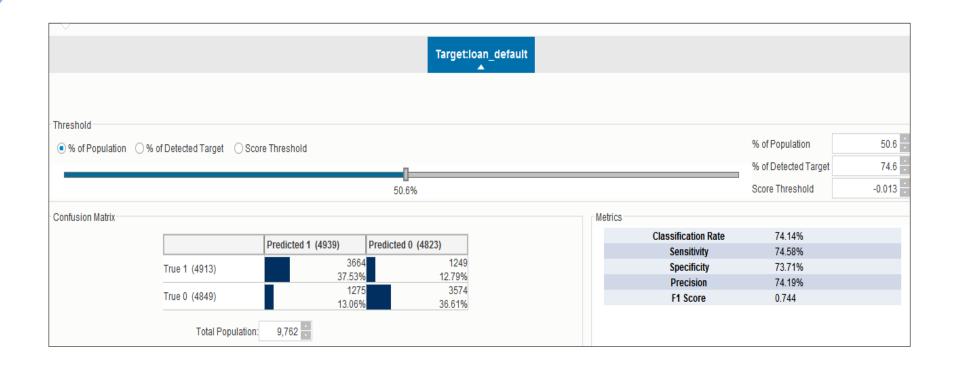
Lift Curve— Model 2

Performance





Confusion Matrix







Model Evaluation

Probability Scores - Model 1 (KI 0.601)	Probability Scores - Model 2 (KI 0.652)	TRUE	False Positives	Total records	Loan Amount gain (\$)	Loan Amount Loss (\$)	Net Profit in \$
> 90%		179	10	189	1,630,650	142,875	1,487,775
> 80%		268	20	288	2,259,050	215,475	2,043,575
> 70%		379	56	435	3,143,900	554,100	2,589,800
> 60%		418	78	496	3,413,775	842,100	2,571,675
<10%		9338	275576	284914	134,529,650	3,993,112,3 50	(-3,858,582,700)
	> 95%	120	61169	61289			
	< 10%	2453	732	3185	24,085,175	9,946,550	14,138,625
	<20%	5689	10609	16298	66,726,925	1,44,479,40 0	-77752475

Defaults identified by Model 2: 2453

Total Defaults(Baseline): 19068

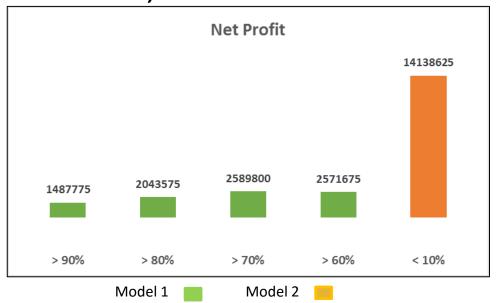


Recommendations

 Classification Model 2 can be used to predict the Loan Defaults in future. Borrowers with Probability Score < 10% can be Rejected.

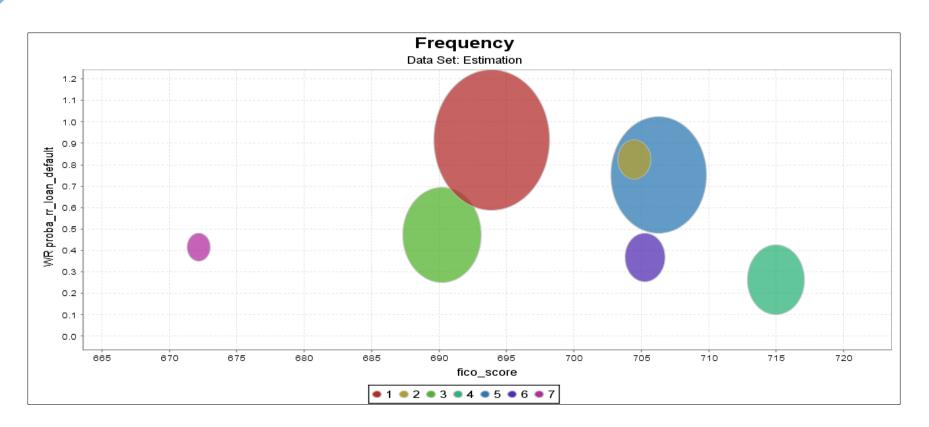
Based on the Evaluation, the Profit with Model 2 is

\$ **14,138,625**



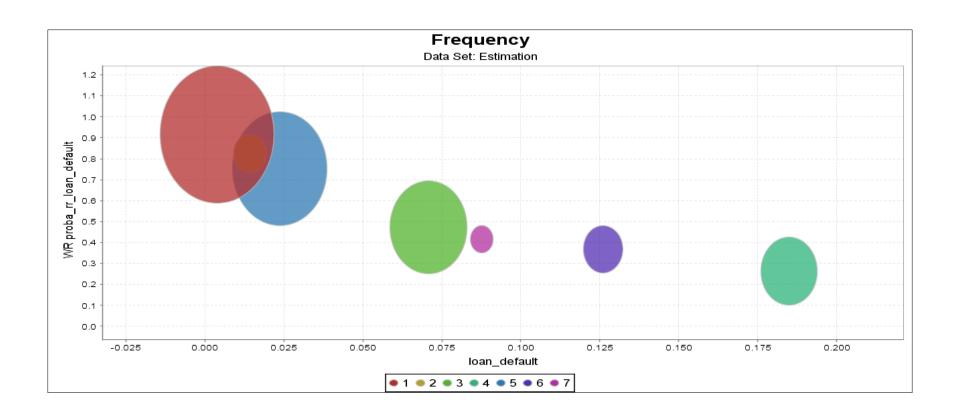


Clustering Model – 2 (fico_score)



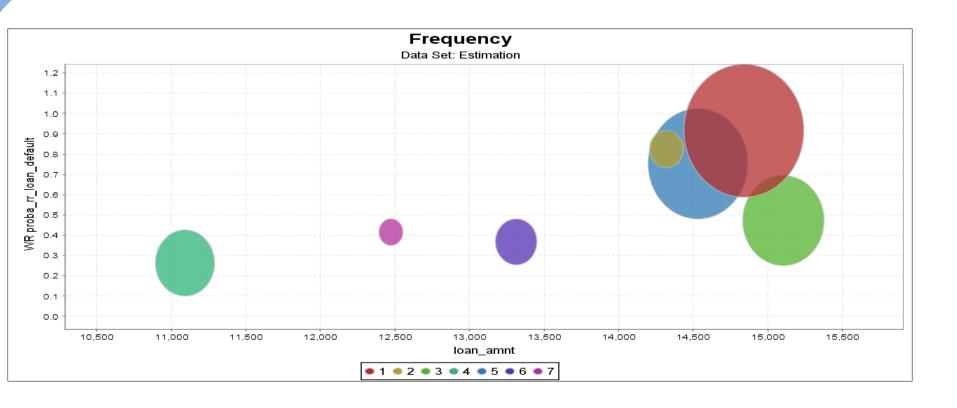


Loan Default



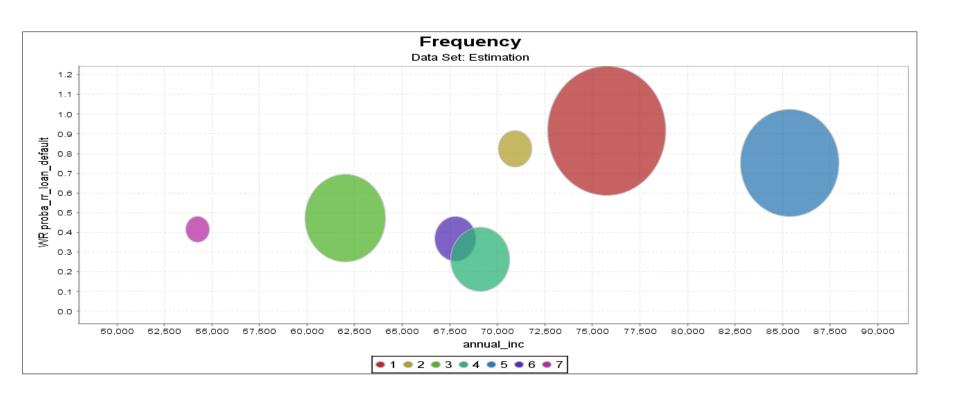


Loan Amount



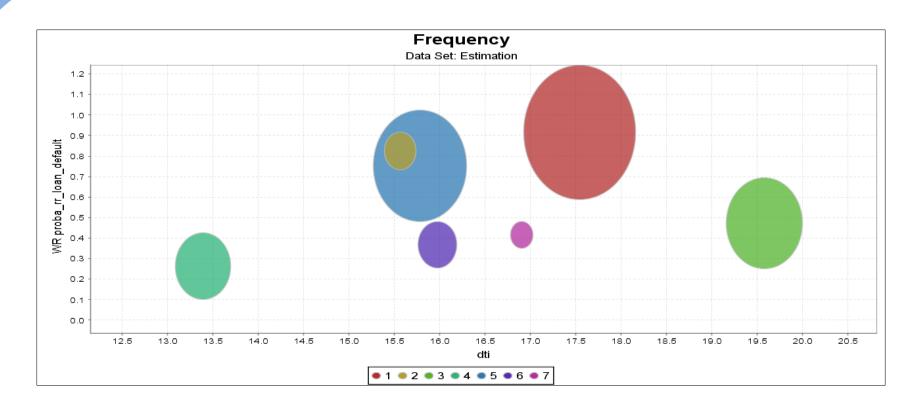


Annual Income



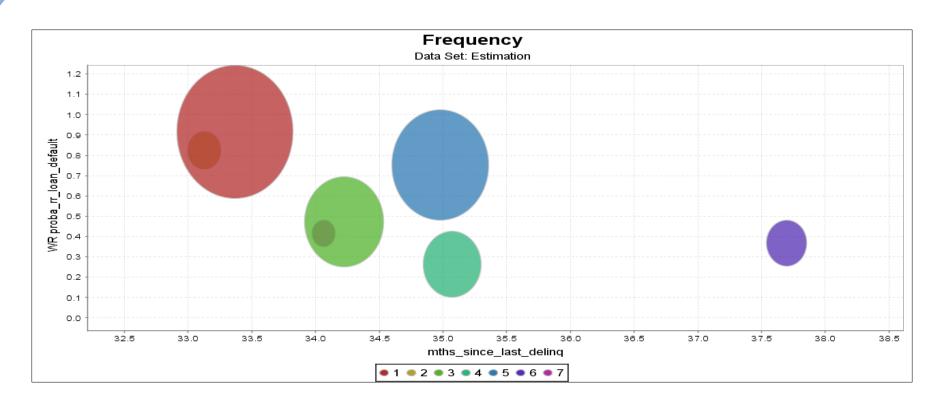


Debt to Income Ratio



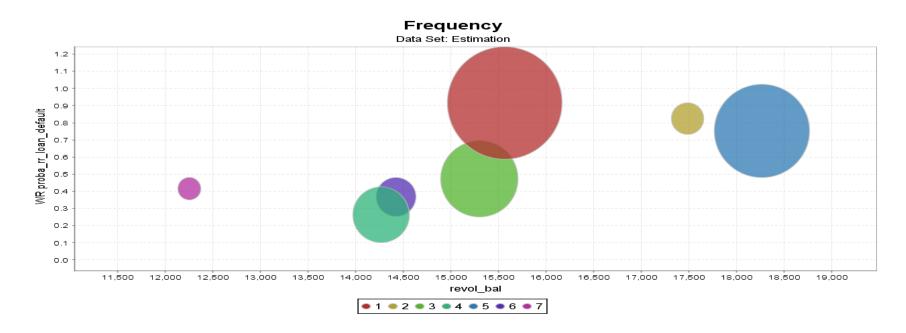


Months since last delinquency





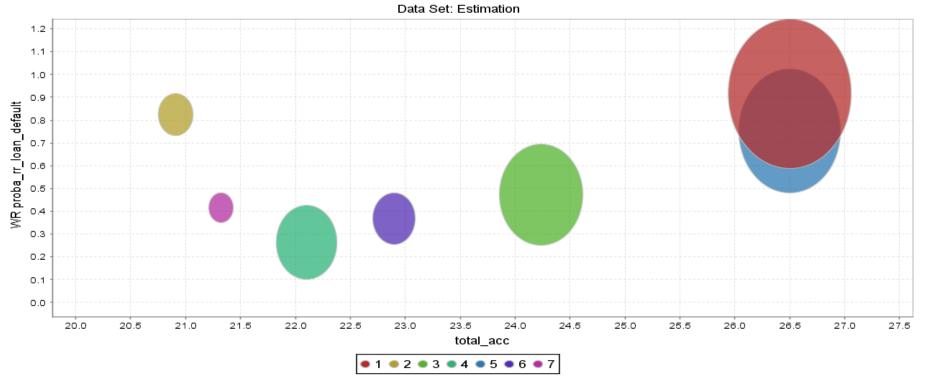
Total credit revolving balance





Total Accounts

Frequency





Appendix



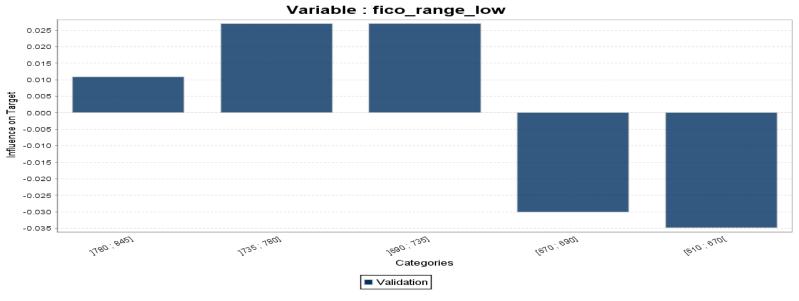
CAR - Sample

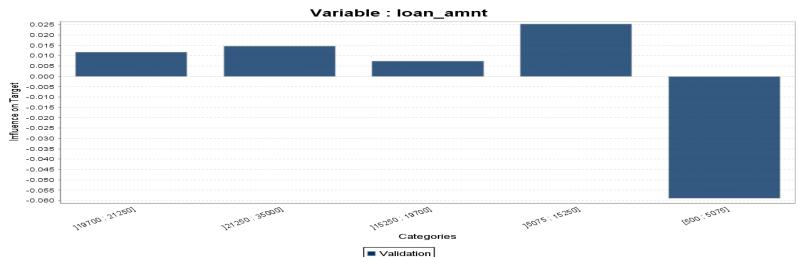
											addr_sta		sub_gra		acc_now
id	member_id	loan_default	fico_range_low	fico_range_high	fico_score	loan_amnt	annual_inc	dti	home_ownership	zip_code	te	grade	de	is_inc_v	_delinq
54734	80364	1	735	739	737	25000	85000	19.48	RENT	941xx	CA	В	B4	Verified	0
														Not	
55521	107577	0	640	644	642	1000	30000	23.84	RENT	600xx	IL	F	F2	Verified	0
														Not	
55742	114426	1	705	709	707	7000	65000	14.29	RENT	112xx	NY	В	B5	Verified	0
56413	129814	0	645	649	647	7000	189500	22.47	MORTGAGE	956xx	CA	F	F2	Verified	0

												mo_sin						
											mo_sin	rcnt_r	total_re	num_re				
mths_since_last_maj		num_tl_op_	mo_sin	tot_hi_c	tot_cur	avg_cur	num_bc	num_ac	num_bc							tot_coll	loan_ty	
or_derog	num_sats	past_12m	_rcnt_tl	red_lim	_bal	_bal	_tl	tv_bc_tl	_sats	vr_dlq	v_tl_op	р	m	_gt_0	_rev_tl	_amt	ре	met
																	a	t
																	a	f



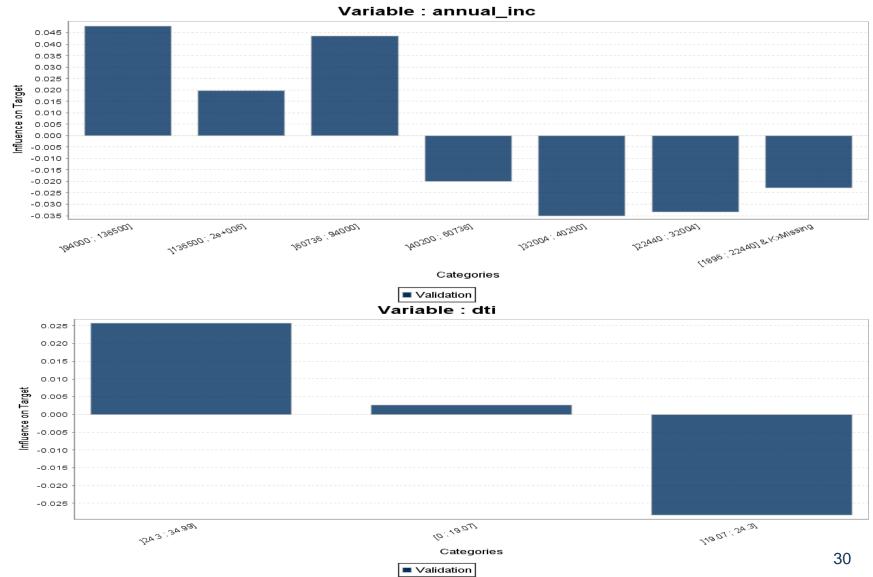
Variable Contribution



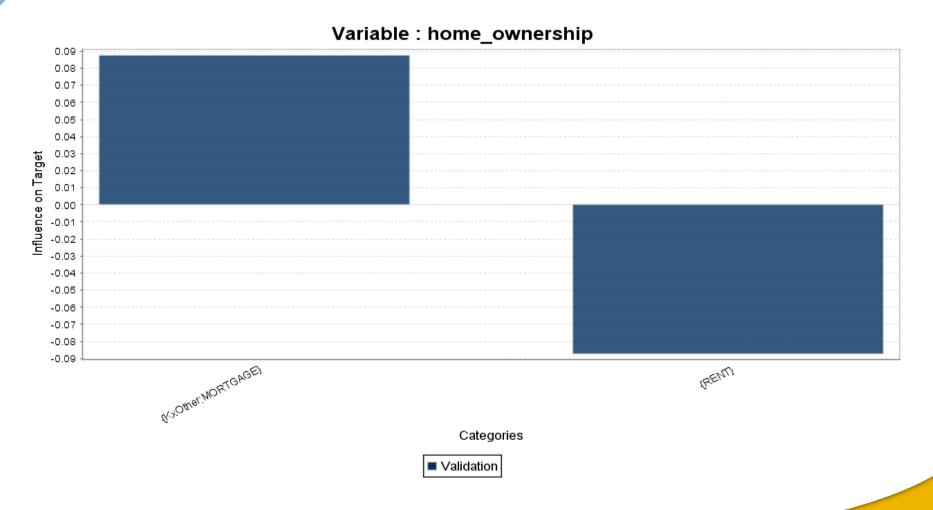




Variable Contributions - Contd

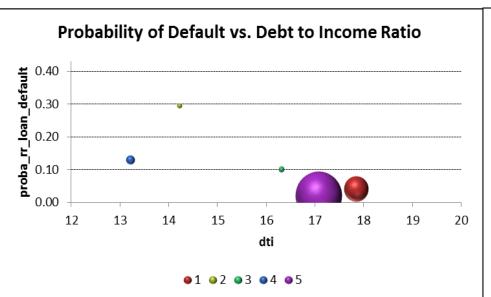


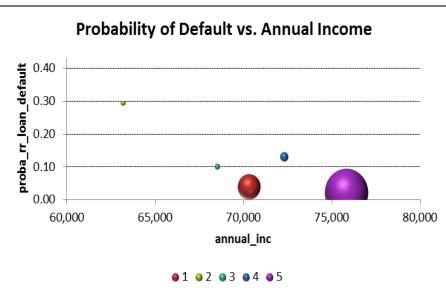
Variable Contribution - Contd





Data Discovery - Cluster Analysis Model 1



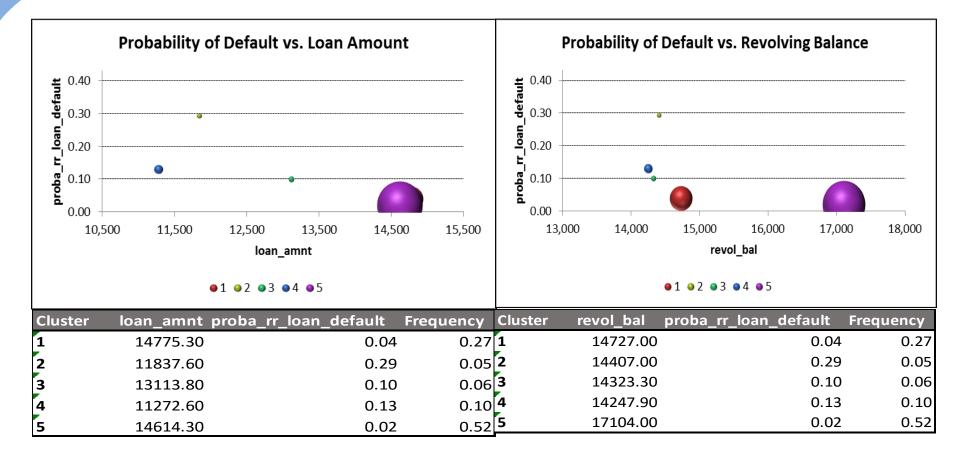


Cluster	dti	proba_rr_loan_default	Frequency	Cluster	annual_inc proba_r	r_loan_default	Frequency
1	17.84	0.04	0.27	1	70312.90	0.04	0.27
2	14.22	0.29	0.05	2	63172.30	0.29	0.05
3	16.31	0.10	0.06	3	68506.90	0.10	0.06
4	13.21	0.13	0.10	4	72297.70	0.13	0.10
5	17.08	0.02	0.52	5	75830.20	0.02	0.52





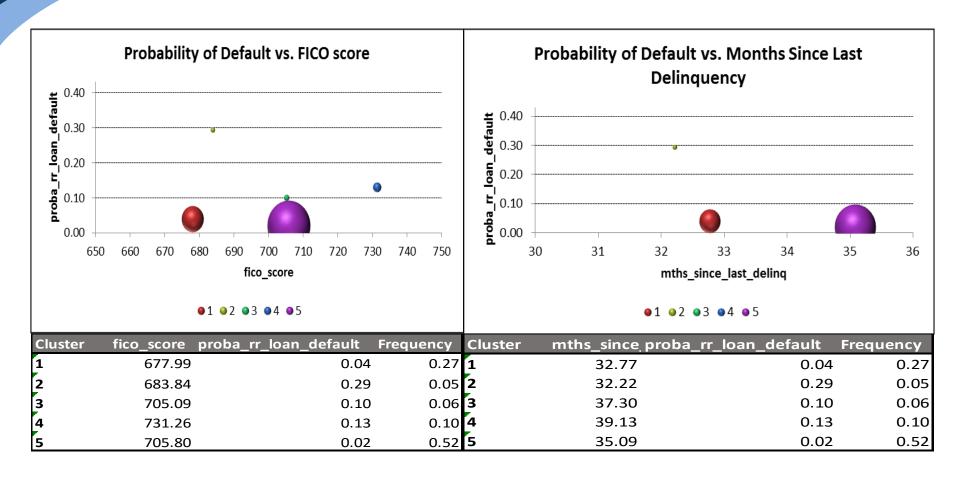
Data Discovery - Cluster Analysis Model 1..







Data Discovery - Cluster Analysis Model 1..









LendingClub

Questions / Discussion