



Credit Scorecard Model
Using Logistic Regression
Study Case: Home Credit

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HOME CREDIT

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Business Understanding





Background

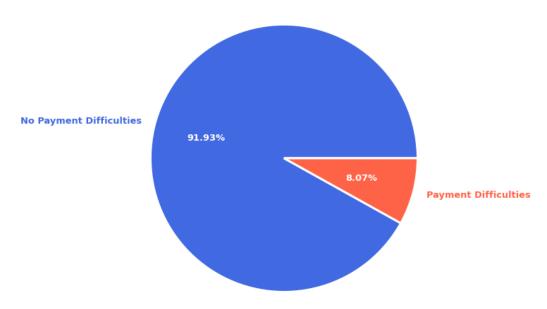
Founded in 1997, Home Credit is an international consumer finance provider with operations in eight countries. Home Credit focus on responsible lending model empowers underserved customers with little or no credit history to access financing, enabling customers to borrow easily and safely, both online and offline.



Dataset Overview

- There are about 91% (282,686
 applicants) loans which indicates
 that client did not had any
 problems in repaying the loan in
 given time.
- While 9% of the total loans (24,825 applicants) involved the clients having problems in repaying the loan.

Distribution of Client Repayment Abilities



HOME CREDIT

Problem Statement

A major challenge for banks and other finance lending agencies is to decide for which candidates to approve loans. In order to make sure this underserved population has a positive loan experience, Home Credit makes use of a variety of alternative data to predict their clients' repayment abilities.

Business Objectives

Create a credit scoring system where the inputs are various features describing the financial and behavioral history of the loan applicants, in order to automatically predict whether the loan will be repaid or defaulted.

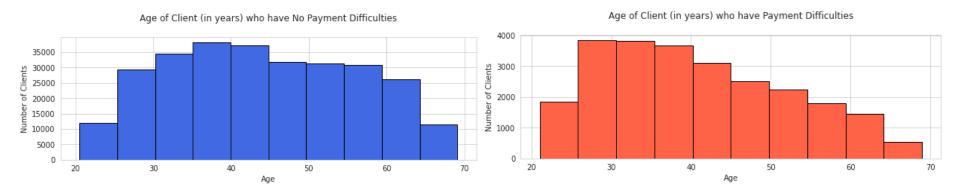






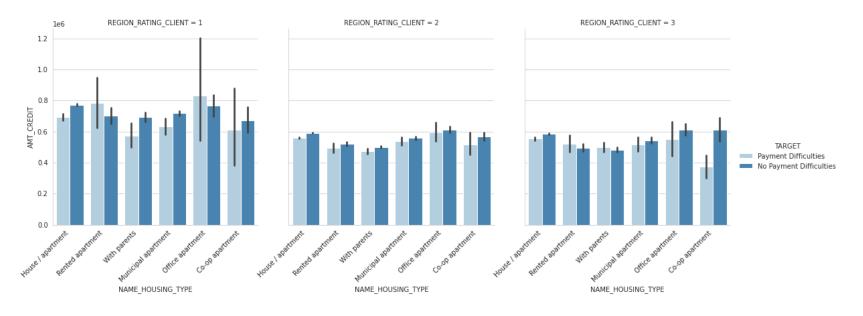
Business Insights





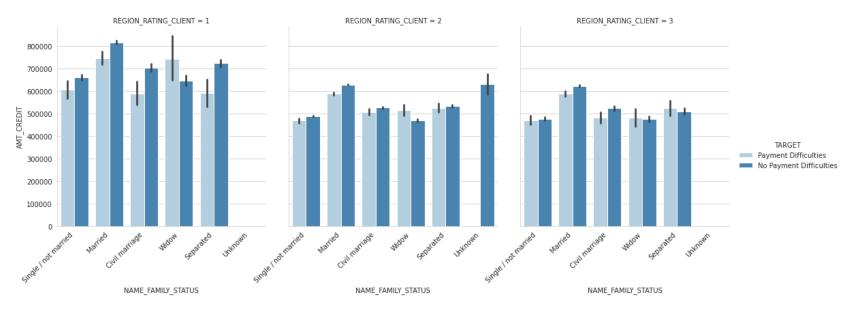
Clients who have no payment difficulties are client in the range of 35-45 years. While clients who have payment difficulties are client in the range of 25-35 years.





Clients who lives in rented apartment and office apartment and their region have a rating of 1, have a problem repaying the loans compared to client in region with rating of 2 or 3.





Clients who have a family status as widow, whether they live in a region with a rating of 1, 2, or 3, have problems repaying loans for moderate to higher amounts credit of the loan. For clients who have a family status as separated and live in region with a rating of 3, also have problems repaying loans for a moderate amount credit of the loan.





Data

Processing



HOME CREDIT



Data Cleansing

Check Data Duplicates
Check Missing Values
Drop Feature
Simple Imputer Median.



Feature Selection

Data Splitting (80:20)
Categorical Data
(Chi-Square)
Numerical Data
(ANOVA)



Feature Engineering

Information Value (IV)
Weight of Evidence Binning
Weight of Evidence Transform



Modeling and Evaluation

Logistic Regression

Class Weight = Balanced

AUROC

PR Curve

KS Statistics

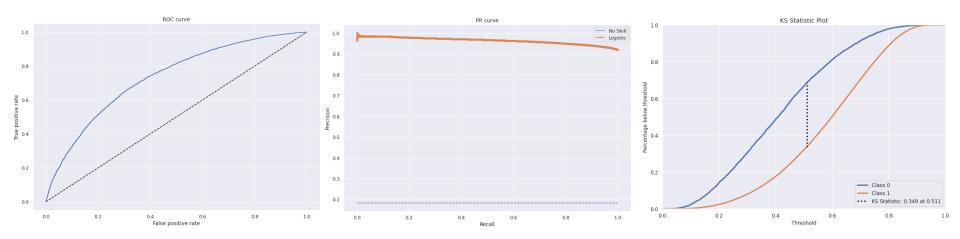




<u>Data</u> Modeling







Model Result

Models	MEAN AUROC	GINI
Random Forest	0.6578	0.3155
Decision Tree	0.5351	0.0702
Logistic Regression	0.7310	0.4620

Based on AUROC score 0.7310, PR AUC score 0.9642, and, KS score 0.349, the Logistic Regression model is considered as a good performance model.





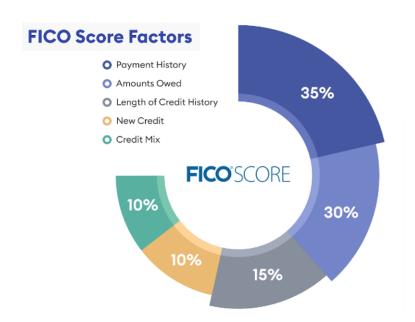
<u>Credit</u> Scorecard





Using FICO Score

A FICO score is a credit score created by the Fair Isaac Corporation (FICO). Lenders use borrowers' FICO scores along with other details on borrowers' credit reports to assess credit risk and determine whether to extend credit. Most credit scores have a 300-850 score range. The higher the score, the lower the risk to lenders.







Threshold = 0.5

Accept Score	N Approved	N Rejected	Approval Rate	Rejection Rate
560.0	39840	21663	0.647773	0.352227

Best Threshold

Accept Score	N Approved	N Rejected	Approval Rate	Rejection Rate
512.0	55663	5840	0.905045	0.094955

We will choose the best threshold with accept score 512 to avoid higher rejection rate with threshold 0.5 but will monitoring model's performance in production.

Setting Loan Approval Cut-offs

Base (Intercept) = 560

Min Score = 300

Max Score = 850

Threshold = 0.5

Best Threshold = 0.29682

Accept Threshold = 512

Application Test = 61503 Applicants

Model = Logistic Regression







Feature	Specific Feature	Score
CODE GENDER	Male	-10
CODE_GENDER	Female or XNA	10
	Academic degree	60
	Higher education	3
NAME_EDUCATION_TYPE	Incomplete higher	-9
	Lower secondary	-31
	Secondary / secondary special	-21
	Single or Unknown	-2
	Civil marriage	-4
NAME_FAMILY_STATUS	Married	7
	Separated	-3
	Widow	3
	Businessman or Commercial Associate	8
	Pensioner or Maternity leave	7
NAME_INCOME_TYPE	Student or Unemployed	-40
	State servant	22
	Working	3
REG_CITY_NOT_LIVE_CITY	0	7
	1	-6
FLAG DOCUMENT 3	0	9
FLAG_DUCUMENT_3	1	-8

REGION_RATING_CLIENT_W_CITY	0	0
	1	18
	2	9
	<0.0147	2
	0.0147-0.0292	1
REGION_POPULATION_RELATIVE	0.0292-0.0436	-2
	0.0436-0.0581	1
	>0.0581	-1
	<2	-7
	2-4	3
VEAR LAST BURNE SHANCE	4-6	2
YEAR_LAST_PHONE_CHANGE	6-8	5
	8-10	5
	>10	0
	<30	-2
	30-40	-10
YEAR_BIRTH	40-50	-2
	50-60	6
	>60	8
	<846000	-4
	846000-1647000	0
AMT_CREDIT	1647000-2448000	9
	2448000-3249000	3
	>3249000	-7

	<4	-11
	4-8	-6
YEAR_ID_PUBLISH	8-12	-4
	12-16	4
	>16	17
	<17	-3
	17-34	0
YEAR_REGISTRATION	34-51	-3
	>51	6
	<0.0855	-53
	0.0855-0.171	-35
	0.171-0.256	-24
	0.256-0.342	-13
EVE COURCE A	0.342-0.427	-5
EXT_SOURCE_2	0.427-0.513	3
	0.513-0.598	10
	0.598-0.684	21
	0.684-0.769	38
	>0.769	59
	<0.0901	-68
	0.0901-0.18	-49
	0.18-0.269	-33
	0.269-0.359	-14
THE COURSE A	0.359-0.448	-1
EXT_SOURCE_3	0.448-0.538	5
	0.538-0.627	26
	0.627-0.717	36
	0.717-0.806	48
	>0.806	51



Name: Novr	rizal Roynanda		Score Calculation
Base Score			560
	CODE_GENDER	Male	-10
	NAME_EDUCATION_TYPE	Academic Degree	60
	NAME_FAMILY_STATUS	Single	-2
	NAME_INCOME_TYPE	Working	3
	REG_CITY_NOT_LIVE_CITY	0	7
	FLAG_DOCUMENT_3	0	9
	REGION_RATING_CLIENT_W_CITY	2	9
	REGION_POPULATION_RELATIVE	0.0600	-1
	YEAR_LAST_PHONE_CHANGE	7	5
	YEAR_BIRTH	26	-2
	AMT_CREDIT	800000	-4
	YEAR_ID_PUBLISH	5	-6
	YEAR_REGISTRATION	5	-3
	EXT_SOURCE_2	0.420	-5
	EXT_SOURCE_3	0.420	-1
	Total		619

Simulation

- If Total Score > Accept Threshold:
 Approve
- If Total Score < Accept Threshold:
 Reject
- Novrizal's Score (619) is higher than Accept Threshold (512), then his loan would be Approved







Visit My GitHub!

You can see the entire project documentation here from my github

@novrizalrnd