

**PROJECT - BASED VIRTUAL INTERNSHIP** 

# CREDIT RISK MODELING & BUSINESS RECOMMENDATION - HOME CREDIT INDONESIA

ROY FIRMAN SIHOMBING

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I am a Mathematics graduate with a strong interest in data science. Currently, I am enrolled in the Data Science Bootcamp at Dibimbing.id to sharpen my skills in Python, SQL, data analysis, and visualization. I have hands-on experience in warehouse data management, sales tracking, and spreadsheet automation. Known for analytical thinking, problem-solving skills, and clear strong communication, I thrive both in team collaborations and independent work.







## **Company Profile**

Home Credit Indonesia is a consumer finance company that has been operating since 2013, providing installment-based lending services without the need for credit cards. The company focuses on expanding financial inclusion by offering accessible, fast, and transparent financing solutions for customers across Indonesia. With millions of customers served nationwide, Home Credit Indonesia enables individuals to purchase consumer goods, electronics, and other necessities through flexible loan products. Guided by its mission to make credit more inclusive and responsible, the company continues to data-driven approaches and advanced risk modeling to improve decision-making and ensure sustainable business growth.





#### **Overview**

- 1 Lending services across diverse customer segments.
- 2 Challenge: approve good customers while avoiding risky ones.

#### **Problem Statement**

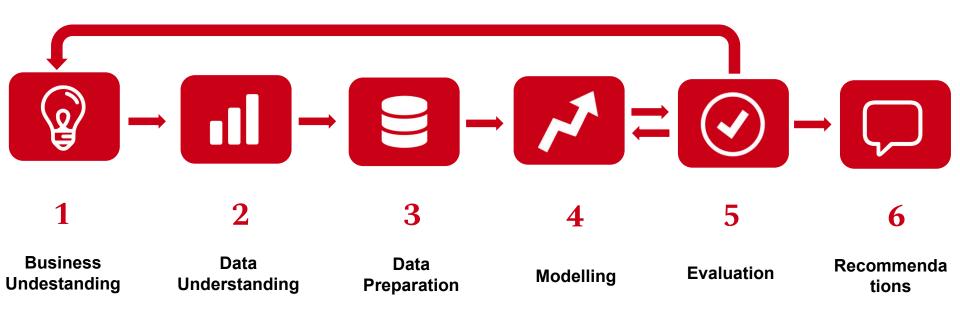
One of the main challenges faced by the company is that good customers are being rejected. This issue not only leads to lost revenue opportunities but also limits financial inclusion, as eligible individuals are unfairly excluded from accessing credit

### **Objectives**

- Reduce false rejections by ensuring that eligible customers are correctly approved.
- Support financial inclusion by widening access to credit for good customers.
- Maintain business growth by minimizing revenue loss from rejected eligible applicants.



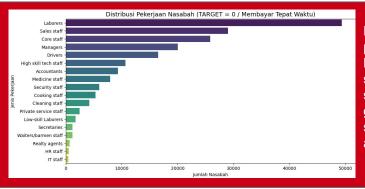
## **Workflow Process**



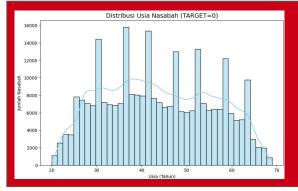


## **Initial Insights**

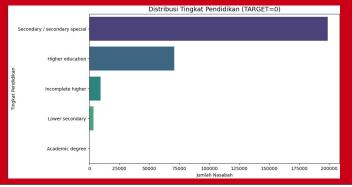




Most on-time payers are laborers, sales staff, and core staff, while groups like IT staff and realty agents are fewer.



Customers aged 30–45 dominate on-time repayment, with fewer very young or older borrowers.



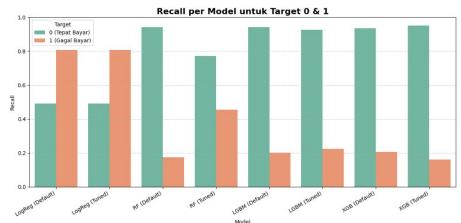
Majority of on-time payers have **secondary education**, followed by higher education.



Around 91.9% customers pay on time, while 8.1% default.



### **Model Performance**



Since our business goal is to avoid rejecting eligible customers, recall is the most important metric. By maximizing recall for Target 0, we ensure that as many good customers as possible are correctly approved, minimizing false rejections

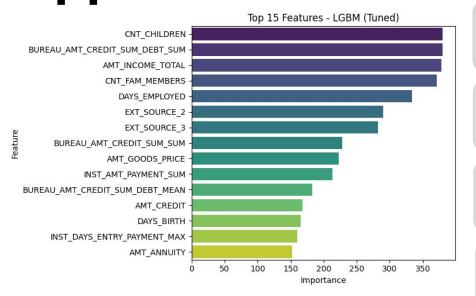
|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0            | 0.9314    | 0.9287 | 0.9301   | 11308   |
| 1            | 0.2144    | 0.2216 | 0.2179   | 993     |
| accuracy     |           |        | 0.8716   | 12301   |
| macro avg    | 0.5729    | 0.5751 | 0.5740   | 12301   |
| weighted avg | 0.8736    | 0.8716 | 0.8726   | 12301   |
| weighted avg | 0.8736    | 0.8716 | 0.8726   | 123     |

| 7 7     |        | précision | recall | f1-score | support |
|---------|--------|-----------|--------|----------|---------|
|         | 0      | 0.9315    | 0.9265 | 0.9290   | 11308   |
|         | 1      | 0.2108    | 0.2236 | 0.2170   | 993     |
| aco     | uracy  |           |        | 0.8698   | 12301   |
| macr    | o avg  | 0.5711    | 0.5750 | 0.5730   | 12301   |
| weighte | ed avg | 0.8733    | 0.8698 | 0.8715   | 12301   |

Validation and Test results are consistent, showing good generalization and reliability

# **Key Drivers of Credit Approval**







Family factors : CNT\_CHILDREN, CNT\_FAM\_MEMBERS



Financial status: AMT\_INCOME\_TOTAL, AMT\_CREDIT, AMT\_GOODS\_PRICE



Employment & age : DAYS\_EMPLOYED, DAYS\_BIRTH



Risk indicators: EXT\_SOURCE\_2, EXT\_SOURCE\_3



Credit history & payment behavior : BUREAU\_\*, INST\_\*

# **Business Recommendations**





Strengthen credit scoring by using family and income factors to better identify eligible customers.



Prioritize recall for Target 0 to reduce false rejections of good applicants.



Regularly retrain the model with updated credit bureau and payment behavior data to maintain accuracy and fairness.

#### Note:

Due to limited computational resources, the current modeling results are not yet fully optimized. For stronger insights and improved performance, future iterations should include more comprehensive hyperparameter tuning and extended computation.



### **THANK YOU!**

Data allows us to see beyond numbers, unlocking fair and inclusive access to credit for everyone.