# LOANRISK Predict ASSESSMENT denied?



# Who we are?

# /PayRangers.

- Data Science Team dari perusahaan layanan konsultasi data analytic and decisioning (DAD)
- Bermitra dengan Bank untuk memberikan solusi dalam masalah loan risk assessment

# **Our Team**











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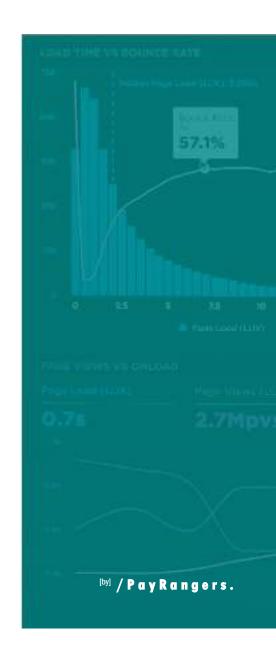
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# **Outline**

- 1. Business Understanding
- 2. What are the **problems** of bank?
- 3. What are the **goals** that the bank wants to achieve?
- 4. What **we do** to **solve the problem**?
- 5. Business Recommendation



# **Business Understanding**







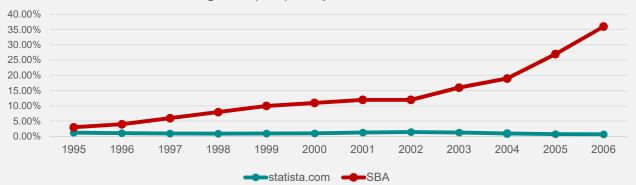


- The Small Business Administration (SBA) provides partial credit guarantees for Small Business in U.S
- The Bank as the lender will determine whether a loan application will be approved or rejected
- Since SBA loans only guarantee a portion of the entire loan balance, banks will incur some losses if a small business defaults on its SBA-guaranteed loan.

# Problem & Goals

# **Problem**





Dilihat dari data SBA setiap tahun angka *Non-Performing Loans (NPL)* terus meningkat.

Setiap tahun angkanya lebih buruk dibandingkan nila rata-rata NPL di U.S.

yang bersumber dari statista.com

The Federal Financial Institutions
Examination Council (FFIEC):

Portfolios with fewer than 6%

Non-Performing Loans

are deemed healthy.

Source : <a href="https://help.distressedpro.com/">https://help.distressedpro.com/</a>

Data SBA: CHGOFF 17% PIF 83%

Other problems (conventional business process):

- Manual form submission
- Manual review by bank analyst team
- Number of successful submissions reviewed



Meminimalisir resiko gagal bayar (charge ofi)

Target : Angka *Chargeoff* kurang dari **6%** 

**Business Metrics:** Default Rate





### **Identify The Most Important Factor**

Insight form historical data



### **Making Machine Learning Model**

Predict whether the loan be approved or denied



### **Business Recommendation**

Model impact and result to business based on data

# **01. Identify The Most Important Factor**

# **Dataset Overview**

Column: 27

Rows : 899.164

Target : MIS\_Status

■ CHGOFF (Charged off)

■ PIF (Paid in full)

Imbalance Target

No	Variable Name LoanNr_ChkDgt			
1				
2	Name			
3	City			
4	State			
5	Zip			
6	Bank BankState NAICS			
7				
8				
9	ApprovalFY			
10	NewExist			
11	FranchiseCode			
12	UrbanRural			
13	RevLineCr			
14	LowDoc			

No	Variable Name			
15	MIS_Status			
16	Term			
17	NoEmp			
18	CreateJob			
19	RetainedJob			
20	ApprovalDate			
21	ChgOffDate			
22	DisbursementDate			
23	DisbursementGros			
24	BalanceGross			
25	ChgOffPrinGr			
26	GrAppv			
27	SBA_Appv			

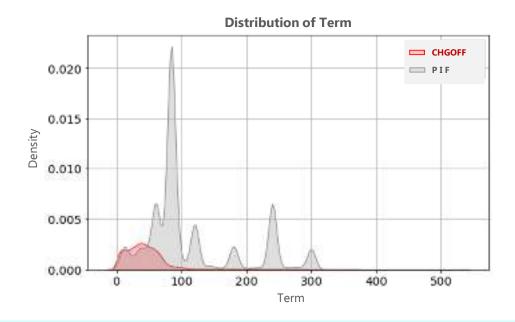


# **Data Pre-Processing**

1.	Handling Missing Value	<ul><li>Drop null value</li><li>Handling error values</li></ul>			
2.	Encoding	Label Encoding			
3.	Engineering	<ul><li>Feature Extraction</li><li>Feature Selection</li></ul>			
4.	Handling Outlier	<ul><li>Feature Transformation using BoxCox</li><li>Handling outlier using Z-score method</li></ul>			
5.	Split Data Set	80% Train & 20% Test			
6.	Scaling	Standardization			
7.	Handling Class Imbalance	Over sampling SMOTE			

# **Data Insight**

### 1. Pengaruh jangka waktu pinjaman (Term)



Semakin lama jangka waktu pinjaman, maka semakin kecil tingkat Chargeoff

### Why did it happen?

Cicilan perbulan lebih sedikit sehingga meringankan debitur

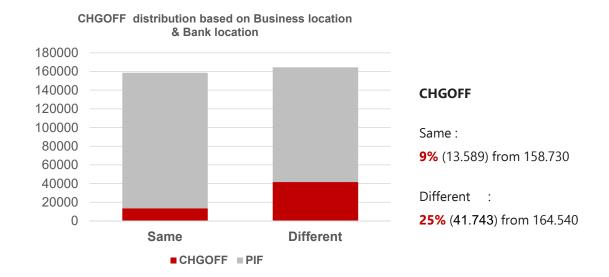
### **Recommendation:**

Memberikan perpanjagan jangka waktu pinjaman



## **Data Insight**

### 2. Pengaruh lokasi bisnis dengan lokasi Bank (StateSame)



Lokasi bisnis yang **sama** dengan lokasi Bank, angka *Chargeoff* lebih sedikit dibandingkan lokasi bisnis yang **berbeda** dengan lokasi bank.

### Why did it happen?

- ✓ Memudahkan Bank memantau bisnis debitur
- ✓ Informasi nasabah mudah diperoleh

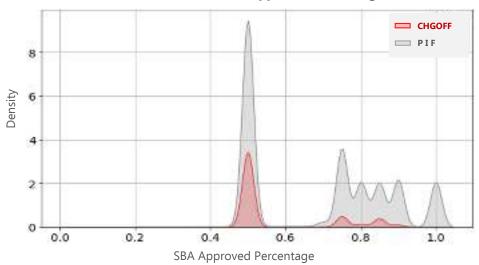
### **Recommendation:**

- Merekomendasikan peminjam untuk mengajukan pinjaman pada cabang Bank yang berlokasi sama dengan lokasi bisnis
- Marketing Campaign untuk memperluas nasabah dilokasi yang sama

# **Data Insight**

### 3. Pengaruh presentase jaminan yang diberikan oleh SBA





Program SBA Express (jaminan 50%) memiliki distribusi data pinjaman yang tinggi dengan tingat *Chargeoff* yang tinggi pula

Source: https://www.sba.gov/partners/lenders/7a-loan-program/terms-conditions-eligibility

### Why did it happen?

- ✓ Lebih mudah mendapatkan persetujuan
- ✓ Waktu *review* singkat

### **Recommendation:**

✓ Menggunakan *Machine Learning model* 

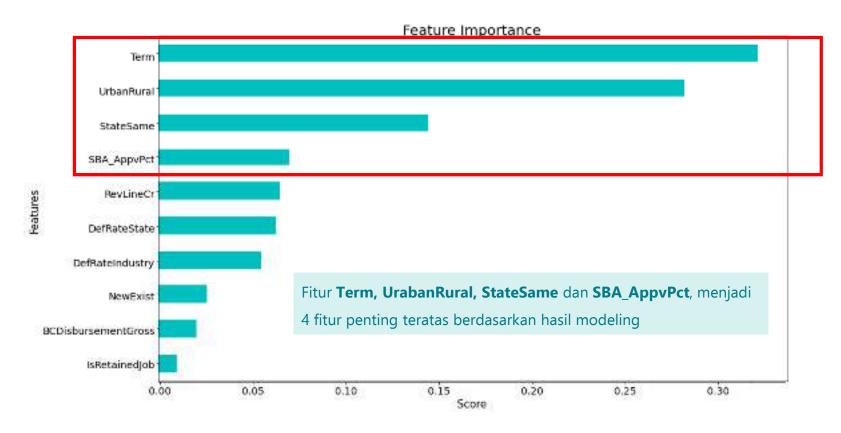
# **02. Making Machine Learning Model**

# **Model Comparison**

Evaluation Model Algorithm	Accuracy	Precision	Recall	AUC-test	AUC-train	Cross-Validation (AUC)
Decision Tree (after tuning)	0.93	0.77	0.84	0.96	0.98	92.76
Random Forest Classifier	0.95	0.85	0.83	0.97	1.00	96.24
Adaboost Classifier	0.91	0.70	0.84	0.95	0.96	95.31
XGBoost Classifier	0.93	0.79	0.83	0.97	0.98	96.43

- Matrics evaluasi yang digunakan adalah AUC
- XGBoost algoritma yang paling baik dengan gap antara AUC train dan test yang sangat kecil, dengan hasil Cross Validation yang paling tinggi dari yang lainnya

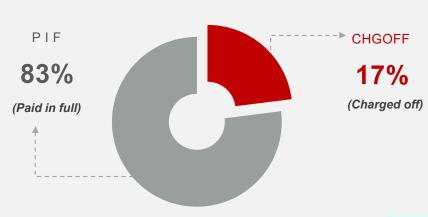
# **Feature Importance**



### **Impact of Machine Learning Models**

### **BEFORE**

### **Percentage of default loans**



Charge off before using ML Model 17% of 64.648 Creditor

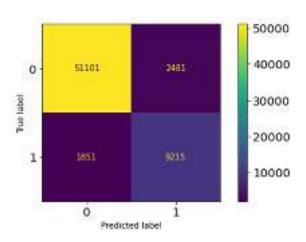
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### **AFTER**

### **Model prediction results**

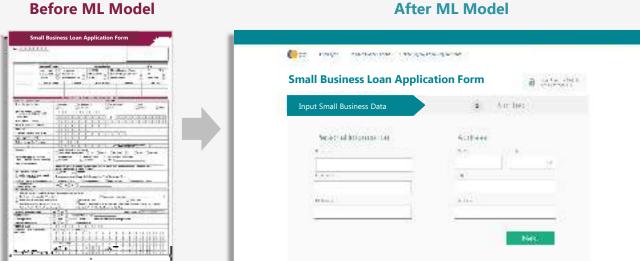


Charge off after using ML Model 4.63% of 53.582 Lost Creditor 3.45%

# Business Recommendation

### **Business Recommendation**

Menggunakan Machine Learning Model untuk melakukan review otomatis terhadap pengajuan pinjaman yang masuk, apakah harus di *approve* atau di *reject* berdasarkan *feature-feature* yang diinput

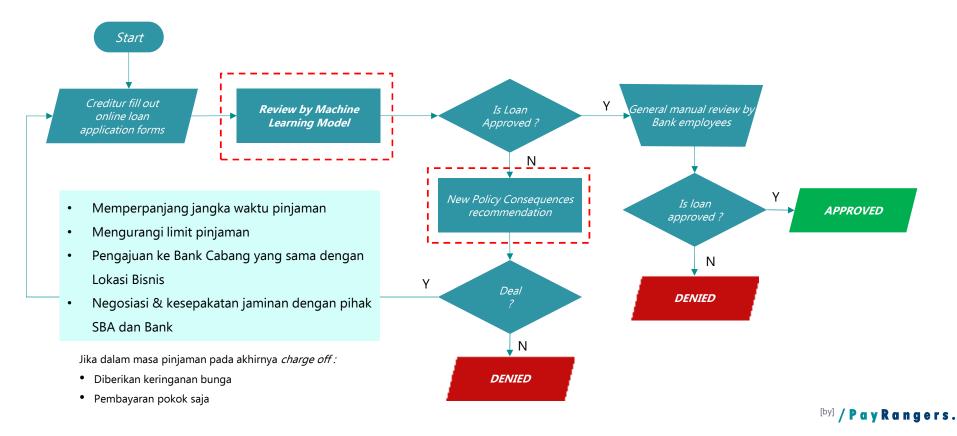


### **After ML Model**

- ✓ Meminimalisir angka resiko gagal bayar
- ✓ Mempercepat proses review
- ✓ Meminimalisir *human error*
- ✓ Reduce cost
- √ Kemudahan proses pengajuan pinjaman oleh debitur

### **Business Recommendation**

Predict whether the loan be approved or denied by Model:



# Thank You