

# CR - IRB - Retail x Corporates portfolio (Summary)

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Retail : PD, EAD, LGD

Corp : PD

## Outline:

### I PD

1. Popu. data collection
2. Seg. & sampling  
DD development
3. Model
  - i. Develop
    - Data treatment
    - Uni. ana.
    - Transform data
    - Multi. ana.
  - ii. validate
  - iii. Calibrate

long list  
↓  
short + list  
↓  
final shortlist

### II EAD

1. Data sampling
2. CCF modeling
3. EAD estimate
4. Backtest

### III LGD

1. Data + sampling
2. Recoveries modeling
3. Backtest

## Data Seg. DD. For Retail x Corp:

- Data : at least 1 business cycle (cover all states)  
Corp: 7-7y, Retail: > 5y (Base II)
- Train / test ~ 7/3 ; o-o-sample: 1mth after
- Segmentation: Corp: usually: industry size
- Default drivers:
 

Corp	Retail
<ul style="list-style-type: none"> <li>• Financial drivers: Profit, debt ratio...</li> <li>• Non-Fin :               <ul style="list-style-type: none"> <li>• level &amp; quality of mgmt</li> <li>• cic ingo</li> <li>• internal environment</li> <li>• Product ingo (length...)</li> <li>• credit history, pmt history</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>• Customer ingo (gender, age, Edu, marital Status...)</li> <li>• Product ingo</li> <li>• credit history, pmt</li> </ul>

## DETAIL I PD. 2

### 2. Seg. & sampling

+ Seg: segments of similar risk characteristic/profit based on:

- + Sta ana: clustering (K-mean, Agg. hic. clustering)
- decision tree
- + Expert perspective, business sense

+ Sample: + Stratified

+ Imbalance dataset → solution:

- oversampling / undersampling
- ensemble
- use Class weight = 'balanced'
- metric:  $F_1$ , Recall, Precision

### 3. Model

i. Develop

Data treat

- + Exclusion:
  - borrower exclusion, if missing > 20%
  - locality
  - Factor (variable)
- for Retail; 750% for Corp.

+ Missing value: (m.v)

- + Remove if:
  - m.v. Ran dom
  - Remove is not biased for model
- + Replace by mean / median for continuous variable
- + Group in 1 group: for discrete variable.

+ Outlier

- + true outlier:
  - $\bar{x} \pm 3\sigma$   $\bar{x} \pm 3\sigma$
  - ≠ value from business sense

+ data quality: treat as m.v issue



## a. Uni-variate ana.

predictive power  
stability  
business sense

by:

Statistical anal  
expert opinion

Sta ana. WOE: for group 0	Good
IV: 0.4 - 0.5	↑
Gini: - for uni. var. ana.	↑
=	5 - 10
For model	30 - 50%

test: seeing if G & B samples are statistically signifi.  $\neq$   
( $p < \alpha$ )

$$H_0: \bar{M}_X = \bar{M}_Y$$

PD: G & B sample  
( $\bar{M}$ ) of 1 variable  
EAD:  $\neq$  btw CCF  
LGD: CCF

+ Stepwise: mix both methods:

1 step fwd  
backward  
Repeat

+ Model selection based on

Sta test  
credit expert opinion

Sta ana:  $|p| < 0.7$   
VIF  $< 5$   
 $p < 0.05$   
gini: 30 - 50% or  $> 55\%$

## b. Transform

variables with  $\neq$  ranges  $\rightarrow$  bad Regression

How? num variable: normalization  $\rightarrow$  range (0,1)  
std  $X_{std} = \frac{X - \mu}{\sigma}$   
 $X_{new} \sim \text{Dist}(\mu=0; \sigma=1)$   
dis/categori: score adjustment

## c. Multi. ana

develop some models by variable combination  
select some optimized

How? + corr ana Pearson = corr(value) : continuous variable  
Spearman: corr(Rank) cont & dis

if  $|p| > 0.7$  or  $0.6 - 0.8$ :  
Keep 1: the one is better  
Keep both: but not together in 1 model  
disagree

+ multicollinearity:  $\exists$  linear rela btw  $(X_i)$

$$\text{check VIF} = \frac{1}{1 - R^2} \leftarrow R^2 \leftarrow X_1 = b_1 + b_2 X_2 + \dots$$

VIF  $> 5 \rightarrow$  remove  $X_1$

+ Stepwise Regression

+ backward: all variables at beginning  
Remove till all vari. meet requirement

+ Forward

add 1 vari. at a time

till no variable meet requirement

+ Stepwise

Condition to add/drop:  
most/least signifi. variable in the model



## Validate assess model performance

- Model stability : accuracy (if the imbalanced issue is treated already)
- Confusion matrix

- other:
- Gini : overall performance
  - KS peak — : 0.2 - 0.4
  - lift : relative — at a cutoff points
  - Hosmer Lemeshow test
  - Area under ROC : > 80%  
=  $(\text{Gini} + 1) / 2$

## Calibrate

add offset factor

when? if default rate in sample not  $\approx$  in population  
as time goes x more new data

## II EAD \$

$$\text{EAD} = \text{current outstanding} + \text{CCF} \left( \text{committed \$} - \text{current outstanding \$} \right)$$

estimate : additional borrowing up to default time

CCF: credit conversion factor  $\in [0, 1]$

## III LGD %

$$\% \text{LGD} = 1 - \frac{\sum \text{PV}(\text{Recoveries}) - \text{PV}(\text{cost})}{\text{EAD}}$$

## ⊗ Decision making PD.

Calculate PD for a customer → create score card → PD → set cutoffs

Some formulas

## WSE, IV

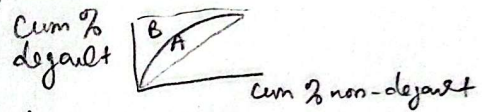
$$\text{WSE} = \ln \left( \frac{\%G}{\%B} \right)$$

$$\text{IV} = \sum [(\%G - \%B) \text{WSE}]$$

i.e.:

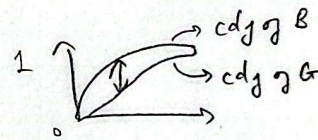
Category	#G	#B	%G	%B
Higher Edu				
no education				
$\Sigma$				

## Gini



$$\text{Gini} = \frac{A}{A+B}$$

## KS



KS = max distance btw cdf of G x B borrower  
 $\in [0, 1]$