

# Macro-financial determinants of default probability using copula

## A case study of Indonesian banks

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# Outline

- 1 Motivation
  - The Basic Motivation
  - Previous Works
- 2 Data/Methodology
  - Data
  - Methodology
- 3 Our Results
  - Main Results

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# Macro-financial linkages in financial stability

since aftermath GFC 2008.

- Financial and macroeconomic plays vital role in explaining business cycle.
- Shock in financial system is endogenous risk and potentially affect banks default probability.

# Macro-financial linkages in financial stability.

By understanding the macro-financial linkages on financial stability. . .

- May help predict if a bank will default on its portfolio
- Help regulators to understand the drivers of default probability
- Determine the appropriate policy to promote financial stability

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# Previous Works.

Method to assess a bank's probability of default . . .

- Supervisory approach relies on firm-specific information.
  - Different from market-based approach (such as Merton).
    - Central bank proprietary data.
- Based on a multivariate distribution
  - Copula approach can capture non-linear relationships between variables with complex data structures (see also Brechmann et al. [2013]; Pourkhanali et al. [2016]; Zhang [2014]).
  - Reflects the likelihood that losses come from the marginal distribution (tail risk) - latent factor.

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# Previous Works.

The drivers of default probability. . .

- Focus only a firm's intrinsic value or bank balance sheet or macroeconomic variables, but not using dependence modelling as copula does.[Weiß, Bostandzic, Neumann (2014), and Kleinow and Moreira (2016).]
- This paper analyze the interlinkage of bank-specific indicators and the macroeconomic variability to bank's default probability based on dependence modelling measurement (non-normal distribution).

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# Data(1)

We focus on Indonesian banks. . .

- Why Indonesia?
  - One of the large emerging economies in Asia.
  - Most suffered country among other peer countries in financial crises (Allen and Gale, 2000).
  - Serve as a benchmark due to the recent economic developments in emerging Asia
- Our data
  - 80 banks with 4800 observations from 2005–to 2019.
  - Coincides with the GFC 2008 and quantitative easing in 2013.
  - Included dummy periods of the pre-crisis, crisis and recovery periods of 2008
  - included dummy of bank owners: central state-owned, regional state-owned, and private commercial, also bank's region.

## Data(2)

The determinants are:

- Bank-specific variables: CAR, CET 1 Ratio (CAP), NPL, LLP, IEF, OPM, LR, DEPOSIT, and LOAN
- Macro-economic variables: HHI, GDP, CPI, Unemployment (UNE), POLICYRATE, and exchange rate (RER)

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# Default Probability

Following (Sklar et al., 1959) theorem, the bank's balance sheet was mapped to the copula function  $c(\cdot)$ :

$$E_t = P(t, T) \iiint_0^\infty G_2(A_{C_T}, A_{L_T}, B_{C_T}, B_{L_T}; T) \cdot c(F_{A_C}, F_{A_L}, F_{B_C}, F_{B_L}) f_{A_C} f_{A_L} f_{B_C} f_{B_L} dA_{C_T} dA_{L_T} dB_{C_T} dB_{L_T} \quad (1)$$

where  $c(\cdot)$  denotes the four-dimensional copula density function,  $F(\cdot)$  denotes the marginal cumulative distribution function, and  $f(\cdot)$  denotes the marginal probability density function.

Using a Monte Carlo simulation, the values of a bank's equity can be estimated as follows:

$$\tilde{E}_t = P(t, T) \frac{1}{N} \sum_{k=1}^N G_2(\tilde{A}_{C_{Tk}}, \tilde{A}_{L_{Tk}}, \tilde{B}_{C_{Tk}}, \tilde{B}_{L_{Tk}}; T) \quad (2)$$

where  $N$  is the number of simulations,  $E_t$ ,  $\tilde{A}_{C_{Tk}}$ ,  $\tilde{A}_{L_{Tk}}$ ,  $\tilde{B}_{C_{Tk}}$ , and  $\tilde{B}_{L_{Tk}}$  are the simulated values of equity, current, and long-term assets and [liabilities](#).

# GMM Panel Model

We follow two-step system GMM estimator (Arellano - Boyer, 1995; Blundell - Bond, 1998) using xtabond2 (Roodman,2009).

$$PD_{it} = \alpha + \delta \cdot PD_{i,t-1} + \beta \cdot V_{it} + \Sigma \varphi_i \cdot D_{it} + \varepsilon_{it} \quad (4)$$

where  $PD_{it}$  represents the probability of default of the bank  $i$  at year  $t$ ;  $PD_{i,t-1}$  denotes its lagged value,  $\delta$  measure the speed of mean reversion,  $\alpha$  is the constant term,  $V_{it}$  denotes the explanatory variables (banks-specific, structural, and macroeconomic variables).  $\beta$  is the vector of coefficient estimated, and  $\Sigma \varphi_i \cdot D_{it}$  represents the time dummies for the period 2005q1 – 2019q4. Finally,  $\varepsilon_{it}$  is the disturbance term.

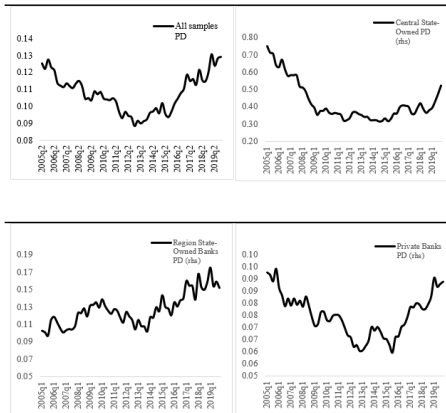


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# Evolution of Default Probability of Indonesian Banks

Figure 1. Evolution of probability of default of Indonesian banks



Source: Author calculation

# Basic Model

	Probability of Default-PD (1)	Probability of Default-PD (2)
Lagged dependent	0.873*** (0.049)	0.862*** (0.045)
Capital adequacy ratio	0 (0)	0 (0)
CET 1 ratio	-0.0002** (0.0001)	-0.0003*** (0.0001)
Non-performing loan ratio	0 (0)	0.001** (0.001)
Loan loss provision	0 (0)	0 (0)
Earnings ratio	0 (0.001)	0 (0)
Inefficiency ratio	-0.00008* (0)	-0.0002*** (0)
Loan ratio	0 (0)	0 (0)
Deposit ratio	0.00005** (0)	0.00005** (0)
Liquidity ratio	0.003* (0.002)	0 (0.003)
Concentration ratio/HHH(1)		0.001 (0.007)
Policy Rate(1)		-0.001*** (0.0002)
Real exchange rate(1)		-0.0003*** (0)
GDP growth(1)		-0.002*** (0.001)
Inflation rate(1)		0 (0.001)
Unemployment rate(1)		-0.049*** (0.019)
Constant	0.014 (0.011)	0.082 (0.054)
Time dummies	Yes	Yes
Observations	4720	4720
Sargan test (p-value)	609.05 (0.00)	651.01(0.00)
Hansen test (p-value)	14.17 (1.000)	73.41(0.132)
AB test AR (1) (p-value)	0.002	0.002
AB test AR (2) (p-value)	0.265	0.256

## Basic Model

Our basic model says: . . .

- Increase in bank solvability improves a bank's ability to absorb sudden losses and reduces the PD.
- Banks with lower efficiency are less likely to experience distress - skimping hypothesis.
- DEPOSIT and LR positively impact default probability proves a "moral hazard effect".
- Other macroeconomic variables aligned with expectation except unemployment rate positively impacted the default probability.

# Dummy crisis

	PD w/ Pre-crisis dummy	PD w/ Crisis dummy	PD w/ Post-crisis dummy
Lagged dependent	0.867*** (0.042)	0.861*** (0.045)	0.867*** (0.041)
Capital adequacy ratio	0 (0)	0 (0)	0 (0)
CET 1 ratio	-0.0004*** (0)	-0.0004*** (0)	-0.0004*** (0)
Non-performing loan ratio	0.001** (0.001)	0.001** (0.001)	0.001** (0.001)
Loan loss provision	0 (0)	0 (0)	0 (0)
Earnings ratio	0 (0)	0 (0)	0 (0)
Inefficiency ratio	-0.0002*** (0.0002)	-0.0002*** (0)	-0.0002*** (0)
Loan ratio	0 (0)	0 (0)	0 (0)
Deposit ratio	0.00005** (0)	0.00005** (0)	0.00005** (0)
Liquidity ratio	0 (0.003)	-0.001 (0.003)	0 (0.003)
Concentration ratio/HHH(1)	0.004 (0.008)	-0.001 (0.007)	0.003 (0.007)
Policy Rate(1)	-0.001*** (0)	-0.001*** (0)	-0.001*** (0)
Real exchange rate(1)	-0.0004*** (0)	-0.0004*** (0)	-0.0004*** (0)
GDP growth(1)	-0.002** (0.001)	-0.003*** (0.001)	-0.002** (0.001)
Inflation rate(1)	0 (0.001)	0.001 (0)	0.001 (0)
Unemployment rate(1)	-0.057*** (0.021)	-0.053*** (0.019)	-0.068*** (0.022)
Pre-crisis	-0.002 (0.002)		
Crisis		-0.002* (0.001)	
Post-crisis			0.004** (0.002)
constant	.066 (.051)	0.098* (0.055)	0.077 (0.052)
Observations	4720	4720	4720
Sargan test ( <i>p-value</i> )	654.97 (0.00)	648.91 (0.00)	651.99 (0.00)
Hansen test ( <i>p-value</i> )	74.11 (0.121)	73.45 (0.132)	74.69 (0.112)
AB test AR (1) ( <i>p-value</i> )	0.002	0.002	0.002
AB test AR (2) ( <i>p-value</i> )	0.258	0.257	0.262

# Dummy crisis

When we included dummy crisis. . .

- The positive value in the dummy crisis period indicates that Indonesian banks reduced their probability of default during the global financial crisis.
- A negative sign in recovery period demonstrates Indonesian banks increased their risk exposure , and as a result, PD was intensified.

# Dummy Bank Ownership and Region.

	PD w/bank's owner dummy	PD w/ regional dummy
Lagged dependent	0.849*** (0.048)	0.863*** (0.042)
Capital adequacy ratio	0 (0)	0 (0)
CET 1 ratio	-0.0003*** (0.0001)	-0.0003*** (0)
Non-performing loan ratio	0.001* (0.0003)	0.001** (0.001)
Loan loss provision	0 (0)	0 (0)
Earnings ratio	0 (0)	0 (0)
Inefficiency ratio	-0.0001*** (0)	-0.0002*** (0)
Loan ratio	0 (0)	-0.0002* (0)
Deposit ratio	0.00005** (0.00002)	0.00005** (0)
Liquidity ratio	-0.001 (0.003)	0.001 (0.002)
Concentration ratio/HHI(1)	0.004 (0.007)	-0.001 (0.007)
Policy Rate(1)	-0.001*** (0)	-0.001*** (0)
Real exchange rate(1)	-0.0003*** (0)	-0.0003*** (0)
GDP growth(1)	-0.002** (0.001)	-0.002** (0.001)
Inflation rate(1)	0 (0.001)	0 (0.001)
Unemployment rate(1)	-0.051** (0.02)	-0.047** (0.019)
Central State-Owned Banks	0.046*** (0.016)	
Regional state-owned banks	0.004 (0.006)	
Private Banks		
West region		0.012 (0.014)
Central region		-0.01** (0.005)
East region		-0.002 (0.007)
constant	0.056 (0.047)	0.099** (0.047)
Observations	4720	4720
Sargan test	628.13 (0.00)	649.69 (0.00)
Hansen test ( <i>p-value</i> )	74.50 (0.115)	71.55 (0.167)
AB test AR (1) ( <i>p-value</i> )	0.002	0.002
AB test AR (2) ( <i>p-value</i> )	0.256	0.258

# Dummy Bank Ownership and Region.

Impact of bank ownership and region. . .

- Central state-owned banks are at higher risk because their dominant and it affects the default probability as a whole.
- Regional state-owned banks in central region negatively associated with PD.



# Summary

- CET 1 ratio, inefficiency ratio, and deposit substantially **affect** a bank's default probability.
- Skimping hypothesis **linked** to a bank's inefficiency indicator
- NPL appeared to be **lagged** indicator when we included macroeconomic variables.
- Policy rate, real exchange rate, economic growth, and unemployment rate **reduce** default probability.
- Recommendations
  - Regulatory should focus on capital regulatory and deposit management policy.
  - The policy rate effectively anticipated the banks' default risk.

# QnA I

Questions?