

# GIMM Model Basics

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GIMM Hands-on Modeling Workshop

November 2022, Prague

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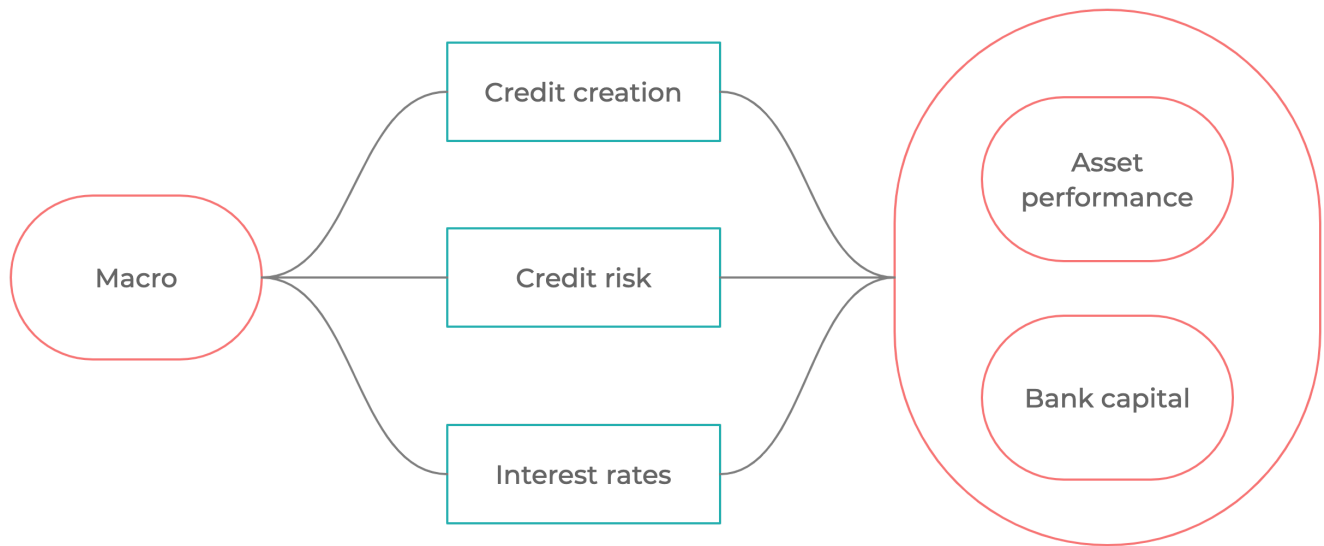
- Big-picture modeling of real-financial interactions with endogenous feedbacks
  - Macroprudential policy analysis
  - Scenario analysis with explicit assumptions
  - Synthesize a variety of insights and inputs, including expert judgment
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# What is it not for?

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- Forecasting
  - Statistical predictions
  - Deeply structural, research framework
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- Conceptual simulations, policy interventions
    - Gain insight into impact on the economy
  - Data-based projections, scenarios conditional on macroeconomic scenarios
    - Stress-testing exercises
  - Counterfactual simulations
  - Comparative static, analysis of equilibrium changes
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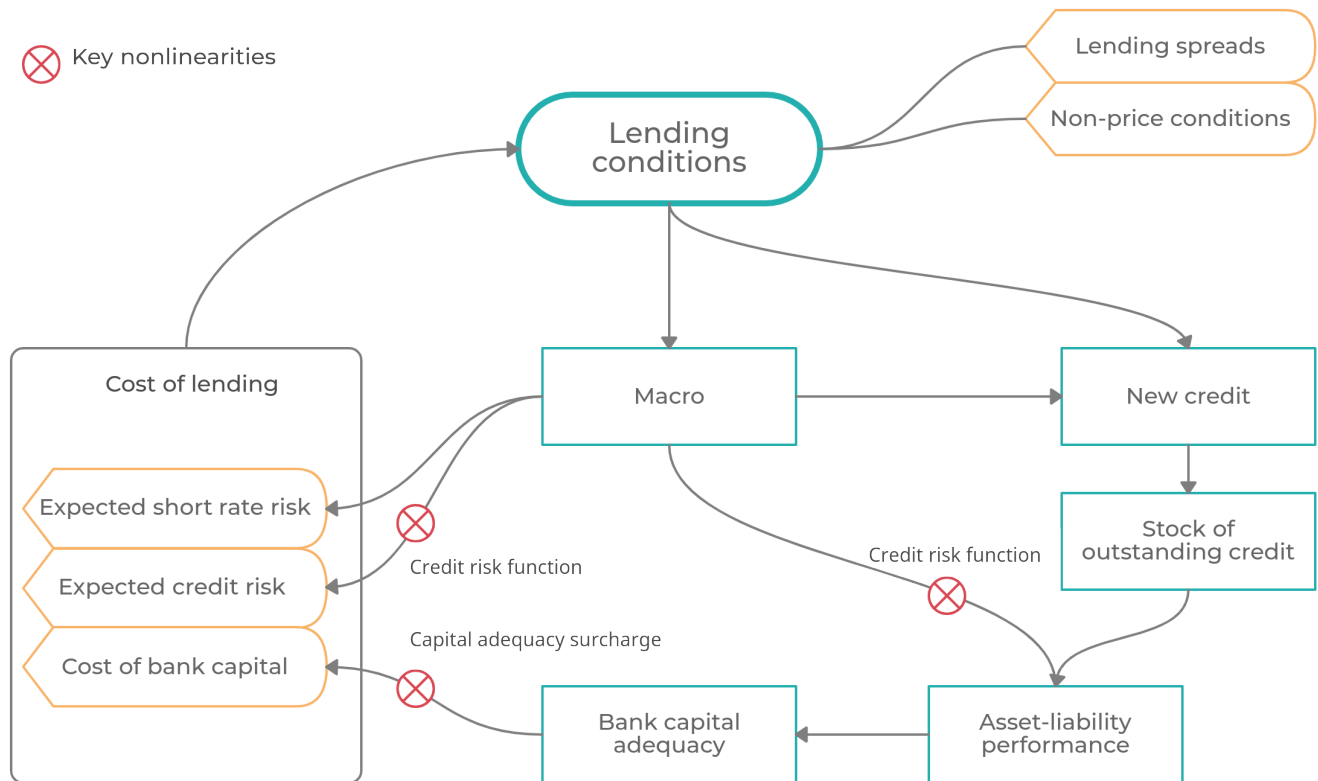


Error parsing Mermaid diagram!

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flowchart LR (Real

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# Bank balance sheet

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	<u>Assets</u>	<u>Liabilities</u>	
$le_t$	Net loans	Non-equity liabilities	$d_t$
$+ \sum l_t^k$	<i>Gross loans</i>	<i>Local currency</i>	$d_t^{\text{lc}} +$
$- \sum a_t^k$	<i>Allowances for credit losses</i>	<i>Foreign currency</i>	$d_t^{\text{fc}} +$
$ona_t$	Other net assets	Bank capital	$bk_t$

- All quantities explicitly tracked in LCU
  - Further decomposition possible
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# Riskless loan portfolio

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Assume loan portfolio consists of infinite amount of identical loans:

☐ ☐ ☐ ☐

⋮ ☐

	Origination	Time 1	Time 2	...
Closing book value	$l$	$(1 - \theta)l$	$(1 - \theta)^2 l$	...
Paydown		$\theta l$	$\theta (1 - \theta)l$	...
Interest		$rl_0 l$	$rl_1 (1 - \theta)l$	...
Total cash flow generated		$(\theta + rl_0)l$	$(\theta + rl_1) (1 - \theta)l_0$	...

Parameter  $\theta$  determines the average maturity of the loan portfolio, given as  $\frac{1}{\theta}$ .

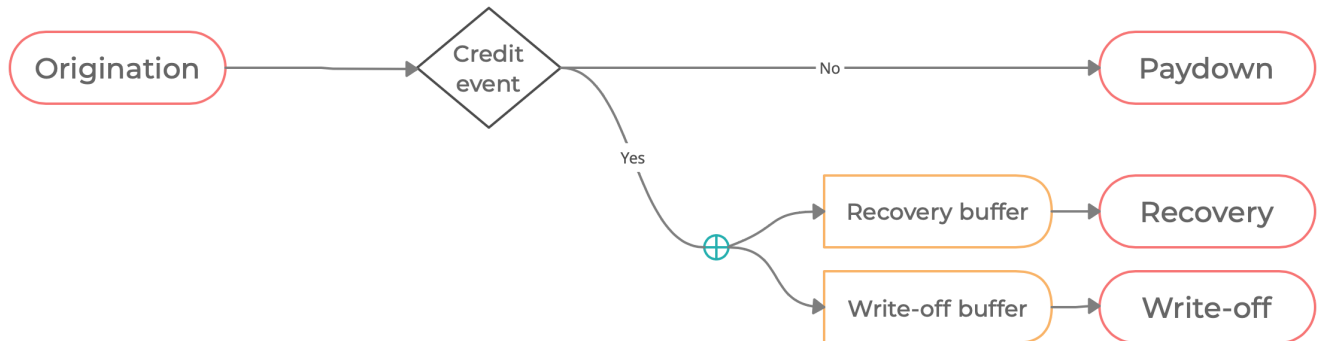
$$l_t = (1 - \theta)l_{t-1} + l_t^\Delta$$

where  $l_t^\Delta$  is the gross production of new loans.

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Each period, each loan faces a possibility of a credit event



Gross loans are classified either as performing or nonperforming:

$$l_t = lp_t + ln_t$$

NPLs break down into:

- $lnc_t$ : recovery buffer (to be gradually recovered, i.e. generating cashflows)
- $lnw_t$ : write-off buffer (to be gradually written-off),

$$ln_t = lnc_t + lnw_t$$

Split into the buffers takes place at the moment of the credit event, according to LGD parameter.

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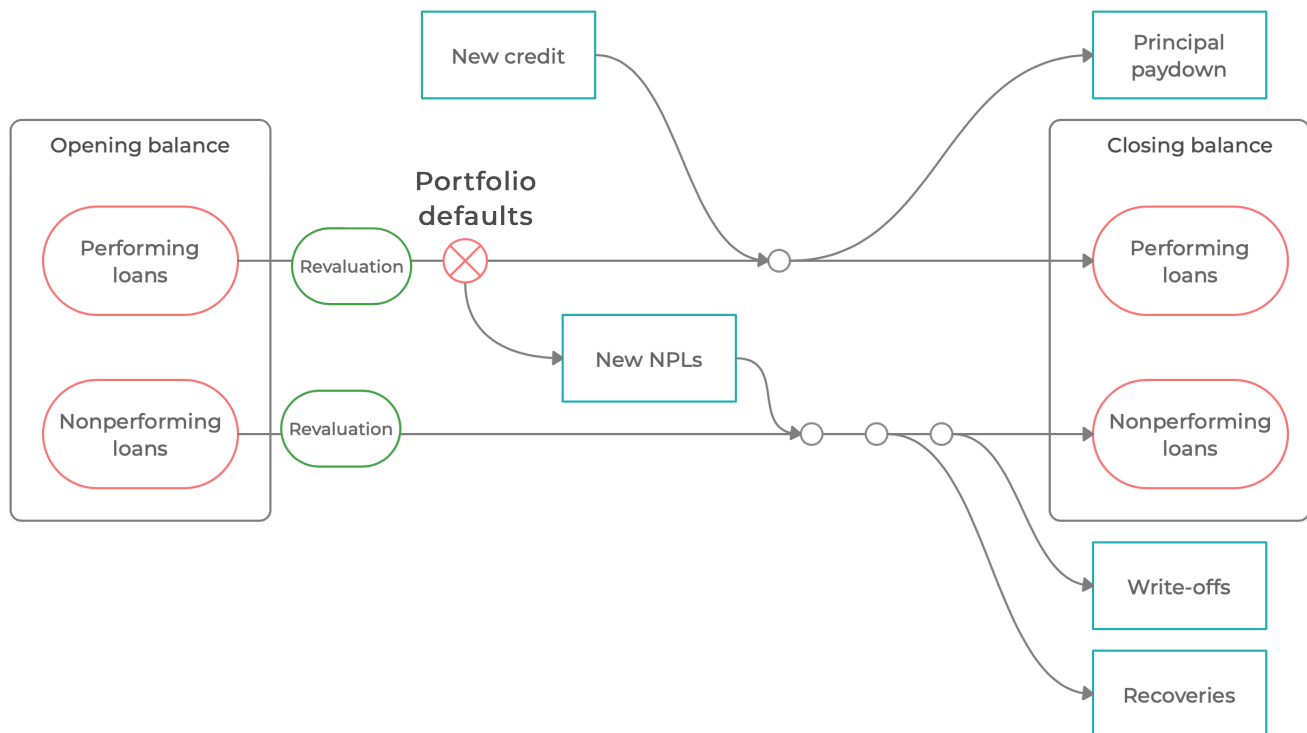
Portfolio division	Cash Flows	Credit event
Performing $lp_t$	Paydown, interest	(outflow) $ln_t^\Delta = q_t lp_t^0 \rightarrow$
Nonperforming recovery buffer $lnc_t$	Collections, repossession, resale	(inflow) $\rightarrow (1 - \lambda) ln_t^\Delta$
Nonperforming writeoff buffer $lnw_t$	None	(inflow) $\rightarrow \lambda ln_t^\Delta$

where:

- $ln_t^\Delta$  is the amount of loans that become non-performing in time  $t$
  - $q_t$  is the default rate
  - $\lambda$  is the LGD parameter
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# Stock-flow dynamics in loan portfolio

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Let's ignore revaluation (FX rate effects) for the moment.

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Assuming no exchange rate valuation.

## Performing loans

Closing balance rolled over from previous time  $t - 1$

$$lp_{t-1}$$

"Opening" balance at new time  $t$ : new information (defaults) arrives, adjust the balance for new information

$$lp_t^0 = lp_{t-1} - ln_t^\Delta$$

$$ln_t^\Delta = q_t lp_{t-1}$$

Throughout-period cash flows generated by the performing loans: paydown plus interest income

$$(\theta_{lp} + rl_{t-1}) lp_t^0$$

Closing balance after paydown and inclusive of new credit  $l_t^\Delta$

$$lp_t = (1 - \theta_{lp}) lp_t^0 + l_t^\Delta$$

## Nonperforming loans: Recovery buffer

Closing balance rolled over from previous time  $t - 1$

$$lnc_{t-1}$$

"Opening" balance at new time  $t$ : new information (defaults) arrives, adjust the balance for new information

$$lnc_t^0 = lnc_{t-1} + (1 - \lambda) ln_t^\Delta$$

Throughout-period cashflow generated by the recovery buffer

$$rec_t \, lnc_t^0$$

Closing balance after recovery cash flows

$$lnc_t = (1 - \theta_{lnc}) \, lnc_t^0$$

## Nonperforming loans: Write-off buffer

Closing balance rolled over from previous time  $t - 1$

$$lnw_{t-1}$$

"Opening" balance at new time  $t$ : new information arrives, adjust the balance for new information

$$lnw_t^0 = lnw_{t-1} + \lambda \, ln_t^\Delta$$

Throughout-period write-offs

$$w_t$$

Closing balance

$$lnw_t = lnw_t^0 - w_t$$

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Total loan portfolio has  $K$  segments (subportfolios, subclasses), each differing in its

- risk parameters
- responsiveness to macroeconomic conditions
- share of foreign exchange denomination

$$l_t = l_t^1 + \dots + l_t^K = \sum_{k=1}^K l_t^k$$

Each loan segment is tracked separately.

Equations presented above exist in  $K$  variants for each loan segment separately.

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- all balance sheet items tracked in LCY value
- FX valuation effects can be potentially very large
- we allow for FX revaluation effects where and if needed

Not presented here, not included in the simple model version.

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Let's have a look at file `loanPerformance.model` .

This part of the model is basically an Excel spreadsheet.

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- link between real economy and bank asset performance
  - expected credit risk gets (partially) priced in lending rates
  - expected credit risk reflected in allowances
  - banks hold excess capital to withstand unexpected increases in credit risk / worsening asset performance
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The credit risk function maps a macro conditions index (Basel II/III asymptotic single risk factor approach),  $z_t$ , into the actual portfolio default rate impact indicator,  $q_t$

$$q_t = f(z_t)$$

Sign and location conventions for  $z_t$ :

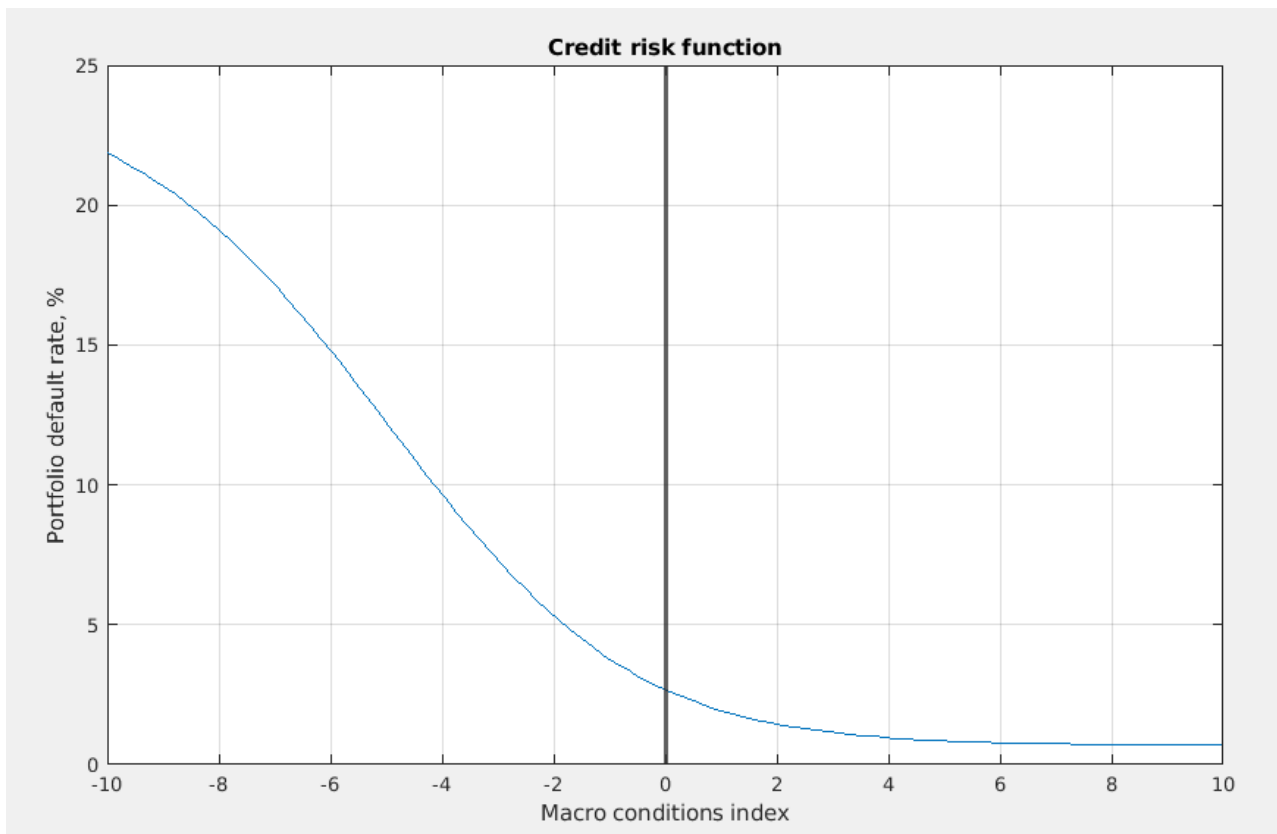
- $z_t = 0$  means normal (steady) macroeconomic and macrofinancial conditions
- $z_t > 0$  means good times
- $z_t < 0$  means bad times

The credit risk function determines

- the actual performance of the existing loan portfolio
  - the expected credit risk used in pricing new loans
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- key model feature
- very hard to estimate

$$f(z_t) = q_t = \underline{q} + (\bar{q} - \underline{q}) \left[ \frac{1}{1 + \exp - \frac{z - \mu}{\sigma}} \right]^{\exp \nu}$$



Combines:

- current macro performance: output gap
- borrower vulnerability: annualized credit (loans) to GDP ratio
- other possible factors

Constructed in deviations from the long-run sustainability trends

$$z_t^k = (\log y_t - \log \bar{y}_t) - c_1 \left( \left[ \frac{l^k}{4 py fwy} \right]_t - \overline{\left[ \frac{l^k}{4 py y} \right]}_t \right)$$

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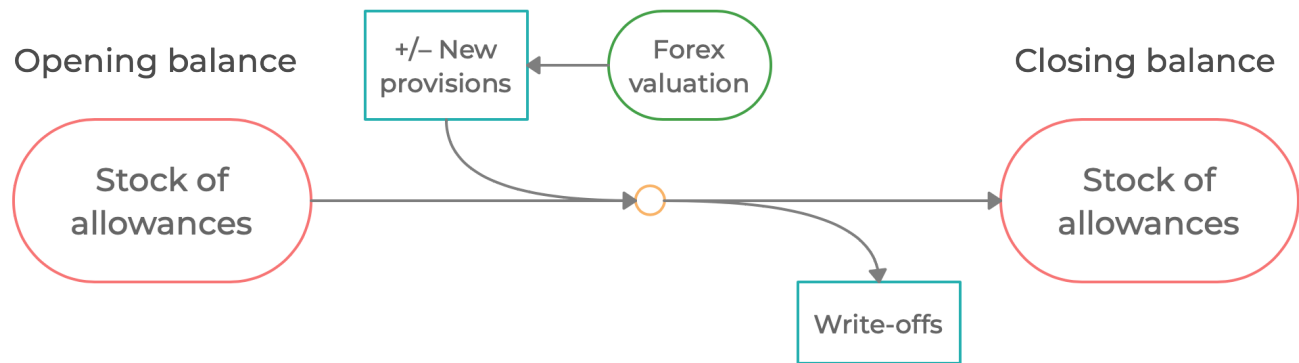
Let's have a look at all this in the model.

Open file `creditPerformanceShock.m`

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The allowances  $a_t$  (stock) are accumulated from provisions  $a_t^\Delta$  (flows).

$$a_t = a_{t-1} - w_t + a_t^\Delta$$



## Two types of allowances

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- incurred loss (IL) based (backward-looking) allowances,  $ab_t$ 
  - based on actual performance
- expected loss (EL) based (forward-looking) allowances,  $af_t$ 
  - based on expected performance

The math for EL-based allowances is ... long and complicated... Let's skip it for the moment and just note that:

- EL-allowances depend on expected credit risk over the lifetime of the loan
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# Let's revisit our simulation

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Again, please open file `creditPerformanceShock.m`

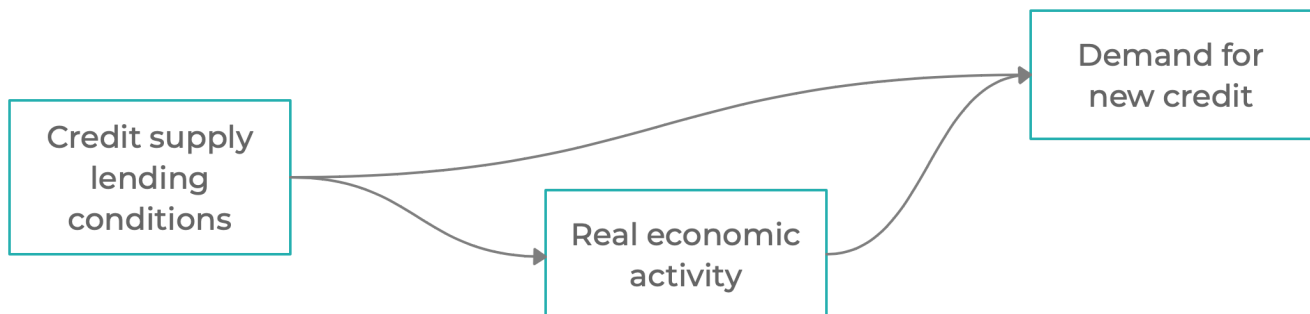
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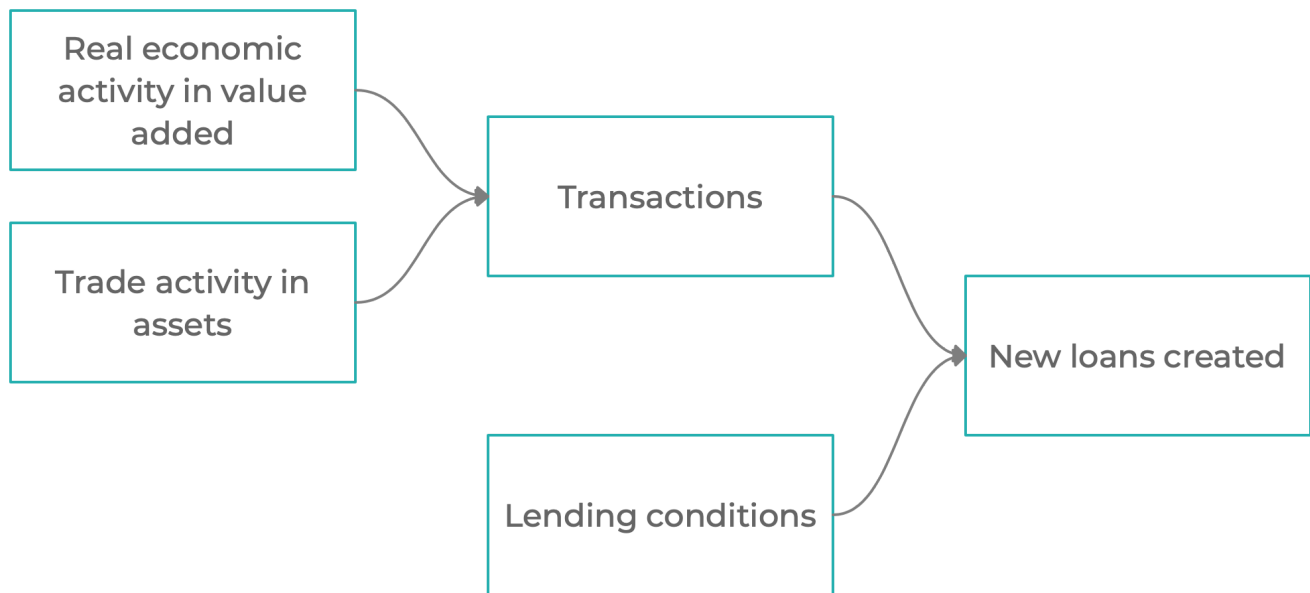


## Reduced-form approach



## Semistructural approach





Real-world lending conditions comprise

- price conditions: lending rates
  - non-price conditions: collateral requirements, insurance requirements, quantitative rationing
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New credit is demanded to finance current period's transactions

$$l_t^{\Delta k} = vel_t^k \cdot trn_t^k \cdot \exp \varepsilon_{l\Delta, t}^k$$

where

- $l_t^{\Delta k}$  is new credit in segment  $k$
  - $vel_t^k$  is the inverse velocity of new credit in segment  $k$
  - $trn_t^k$  is the value of all the transactions that need financing in segment  $k$
  - $\varepsilon_{l\Delta, t}^k$  is a shock to the new credit in segment  $k$
-

$$\begin{aligned}
 trn_t^k &= py_t y_t + c_1^k py_t y_t^{fws} \\
 vel_t^k &= c_0 \cdot vel_{t-1}^k + (1 - c_0) \cdot vel_{ss}^k \\
 &\quad - c_1 \cdot \hat{r}_t^{\Delta full} \\
 &\quad + c_2 \cdot \left( \left[ \frac{l^k}{ny} \right]_t - \overline{\left[ \frac{l^k}{ny} \right]_t} \right) \\
 &\quad + \varepsilon_{vel^k, t}
 \end{aligned}$$

where:

- $\hat{r}_t^{\Delta full}$  is a measure of lending conditions tightness
- $\left( \left[ \frac{l^k}{ny} \right]_t - \overline{\left[ \frac{l^k}{ny} \right]_t} \right)$  represents the current credit overhang over a sustainable level, expressed as a share of GDP; this concept is also known as the credit gap

Note that "credit trend"  $\overline{\left[ \frac{l^k}{ny} \right]_t}$  is exogenous.

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# Let's revisit our simulation

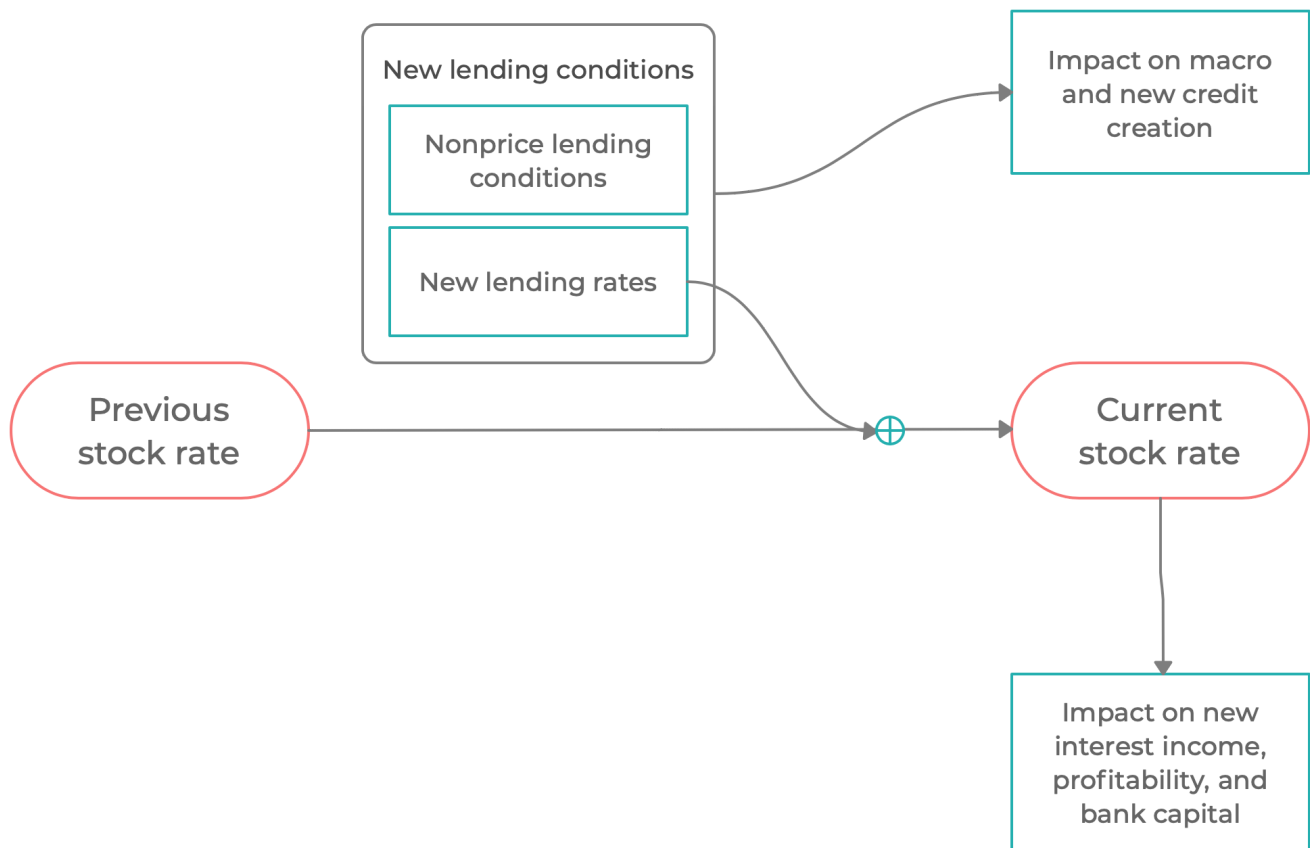
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Again, please open file `creditPerformanceShock.m`

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# Interest rates - stocks vs flows

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Each period, banks set lending rate on loans issued in that period

$$rl_t^\Delta$$

The average loan rate on the whole loan portfolio is

$$rl_t = rl_{t-1} + \Omega_t (rl_t^\Delta - rl_{t-1}) + \epsilon_{rl,t}$$

where

$$\Omega_t = \frac{\psi_{rl} (1 - \theta) lp_t^0 + l_t^\Delta}{lp_t}$$

Parameter  $\psi_{rl}$  determines the average duration of interest rate fix:

- $\psi_{rl} = 0$ : a fixed rate loan, no interest rate reset
  - $\psi_{rl} = 1$ : fully adjustable rate loan, interest rate resets every period
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# Let's revisit our simulation

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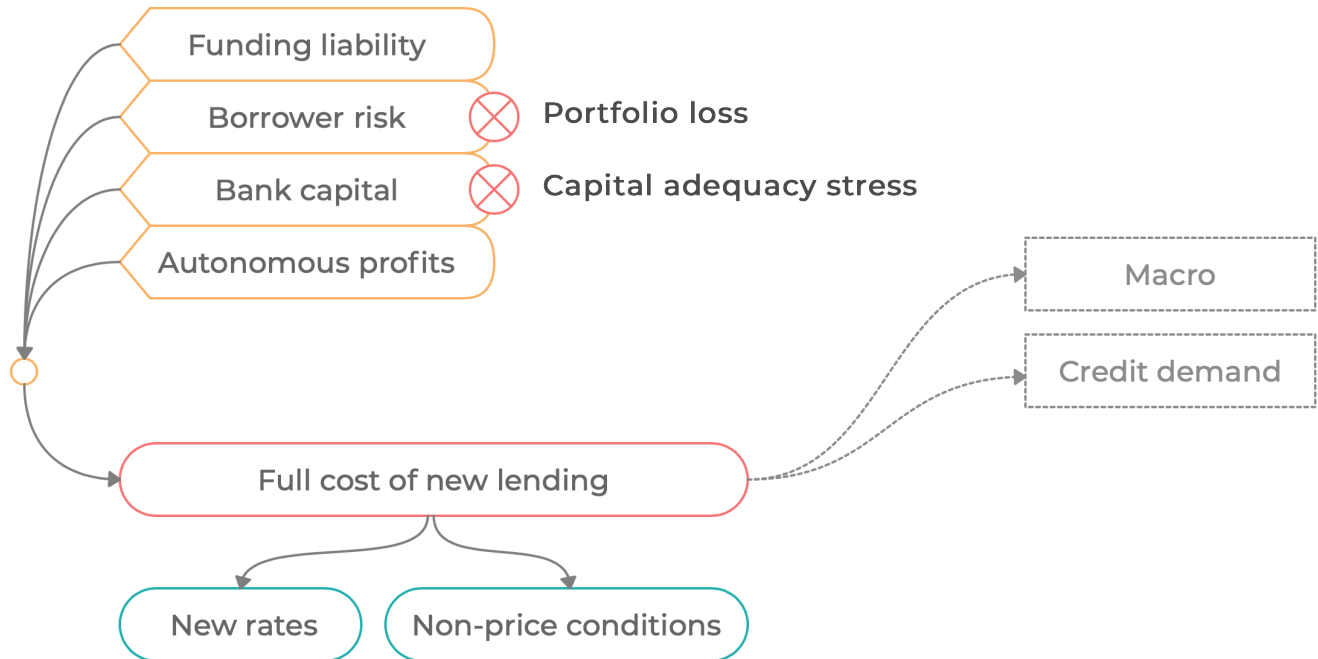
Again, please open file `creditPerformanceShock.m`

Note that banks cannot immediately reflect sudden increase in credit risk in higher interest rates for loans extended in previous periods.

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# Interest rate setting: overview

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## Short-term base rate

The short term base rate comprises

- the marginal cost of funding liabilities,  $rs_t$
- an autonomous profit margin,  $rl_t^{\Delta \text{apm}}$

$$rl_t^{\Delta \text{base}} = rs_t + rl_t^{\Delta \text{apm}}$$

## Forward-looking rate covering credit risk

The hypothetical lending rate covering the full credit is given by

$$rl_t^{\Delta \text{full},1} = (1 - \Psi_1) \left[ \frac{1 + rl_t^{\Delta \text{base}}}{1 - \lambda q_{t+1}} + \Psi_1 \frac{1 + rl_{t+1}^{\Delta \text{base}}}{1 - \lambda q_{t+2}} + \Psi_1^2 \frac{1 + rl_{t+2}^{\Delta \text{base}}}{1 - \lambda q_{t+3}} + \dots \right] + \epsilon$$

where

$$\Psi_1 = (1 - \psi_{rl,1}) (1 - \theta)$$

is the effective discount factor applied on each future base cost.

## Forward-looking rate covering cost of bank capital

The cost of bank capital (capital shortfall stress) is also reflected in loan pricing given by

$$rl_t^{\Delta \text{full},2} = (1 - \Psi_2) \left[ (1 + rx_t) + \Psi_2(1 + rx_t) + \Psi_2^2 (1 + rx_{t+2}) + \dots \right] + \epsilon_t^{rl\Delta \text{full},2}$$

## Overall lending rate

Overall hypothetical lending rate reflecting all costs is given by

$$1 + rl_t^{\Delta\text{full}} = (1 + rl_t^{\Delta\text{full},1}) \cdot (1 + rl_t^{\Delta\text{full},2})$$


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The hypothetical full-cost rate  $rl_t^{\Delta\text{full}}$  splits into

- a price component, i.e. the actually observed new lending rate;
- non-price conditions measured by an interest rate equivalent and passed on to borrowers)

$$rl_t^{\Delta} = rl_t^{\Delta\text{base}} + c_1 \left( rl_t^{\Delta\text{full}} - rl_t^{\Delta\text{base}} \right) + (1 - c_1) \left( rl_{ss}^{\Delta\text{full}} - rl_{ss}^{\Delta\text{base}} \right)$$

- hypothetical full-cost rate  $rl_t^{\Delta\text{full}}$  enters the aggregate demand and credit demand equations, as it represents the true cost of credit for borrowers
  - observed market lending rate  $rl_t^{\Delta}$  enters the bank profits calculations
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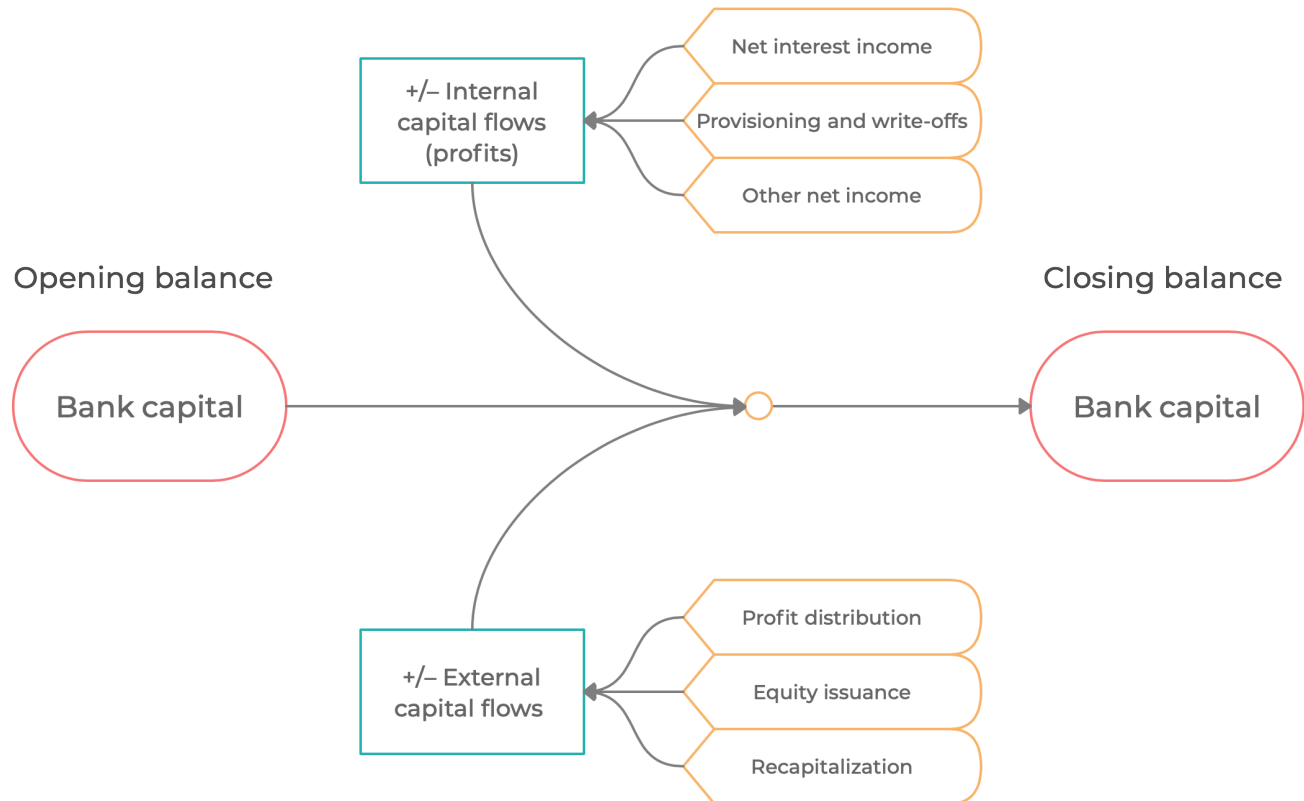
# Let's revisit our simulation

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Again, please open file `creditPerformanceShock.m`

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Key variable



$$bk_t = bk_{t-1} + prof_t + xcf_t$$

- $bk_t$  is bank capital (balance sheet capital)
- $prof_t$  is an internal flow of capital (profit or loss, PnL) recorded on the closing balance of the balance sheets at  $t - 1$  and credit events throughout  $t$
- $xcf_t$  is an external flow of capital throughout  $t$ : dividends paid out (–), new equity issuance (+), equity withdrawals by parents (–), recapitalization flows (+), etc.

Regulatory capital

$$regk_t = \left[ \frac{regk}{bk} \right]_t bk_t$$

Standard capital adequacy ratio

$$car_t = \frac{regk_t}{riskw_t l_t}$$

- $riskw_t$  is the effective average risk weight, an exogenous variable
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In the equilibrium (steady state), banks target a comfort level of CAR

$$car_t \longrightarrow car_t^{\text{tar}}$$

$$car_t^{\text{tar}} = car_t^{\text{min}} + car_t^{\text{exc}}$$

- $car_t^{\text{min}}$  is the regulatory minimum including regulatory buffers
- $car_t^{\text{exc}}$  is the excess capital target above the regulatory minimum to protect against unforeseen shocks

Why do banks not want to hold higher excess capital by themselves?

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Sums (almost) all developments on bank balance sheet:

- Interest income on loans (by segments)
  - Income on other assets
  - Interest expense on non-equity liabilities (by currency of denomination)
  - Other income (proxy for fees, commissions, etc)
  - Provisioning and write-offs
  - Exchange rate valuation
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# Let's revisit our simulation

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Again, please open file `creditPerformanceShock.m`

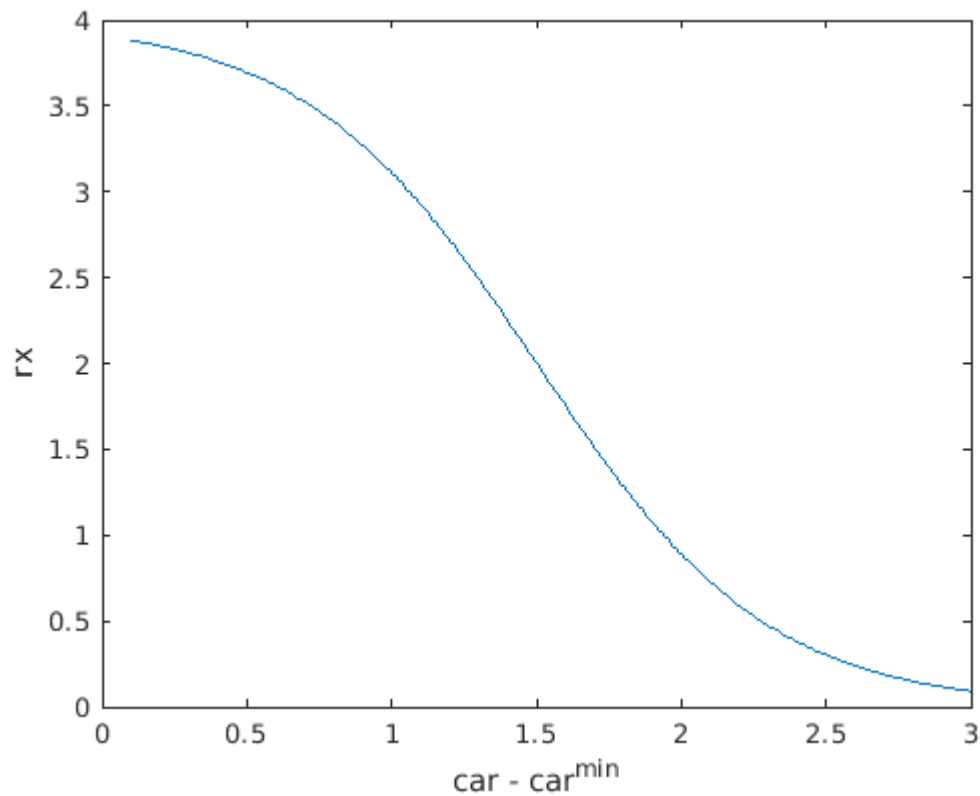
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We have not yet discussed the key issue:

What do the banks do, when their capital declines below "optimal" level?

- Change composition of assets
  - Reduce expenditures
  - Tighten lending conditions - the only macroeconomically relevant action, the only one we consider
-

When  $car_t$  approaches regulatory minimum  $car_t^{\min}$ , capital shortfall triggers increase in capital adequacy risk surcharge  $rx_t$ :



Another key nonlinearity.

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# Let's revisit our simulation

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Again, please open file `creditPerformanceShock.m`

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# Let's revisit our simulation

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Let me show you how to quickly set up a new simulation.

We will examine the impact of the shocks with and without the feedback mechanism.

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There is none in the model.

But we can represent a large number of policies:

- capital-based policies
- DTIs, LTVs, ...
- credit caps

We will see examples.

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Funding issues are not the current focus of the framework.

Deposits always available, liquidity not an issue.

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These are the key features of the model.

Let's examine it further - which scenario would be of interest to you?

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Estimation is very difficult.

We can employ various calibration strategies:

- Use other models - micro PD models, ...
- Match the model to data
  - Some parameters have directly observable real-world counterparts
  - Some parameters determine steady-state equilibrium
- Simulation exercises
  - Some parameters can be inferred by replicating interesting periods in data
  - Theoretical simulations, "smell-tests"
  - Comparison to other models

We have a calibration guide.

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Financial cycle can arise for multiple reasons.

Overly-optimistic expectations are one of them.

In structural model, we can work with expectations explicitly.

Let's have a look at file `simulateFinancialCycle.m`.

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