

Risk-Taking Dynamics and Financial Stability

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- Crisis of 2008/09 has catapulted concerns about financial instability and financial policies to the top of the macroeconomic policy agenda
- Important strand of research: study factors behind booms and busts
 - e.g. sector-wide/aggregate distortions like exuberance, agency problems, externalities, ...

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 - This paper: focus instead on *composition* of the financial sector
 - booms & busts accompanied by growth & demise of high risk-takers
 - examples in 2000s: Countrywide, WaMu, AIG Fin Services, ...

Key Considerations

- **Heterogeneity:** market participants differ
- **Incomplete markets:** risk-sharing limited by agency problems
- **Distribution of wealth:** drives aggregate risk-taking dynamics

- **Pro-cyclicality** – Minsky's financial instability hypothesis
- **Novel compositional effects** of public policy interventions
- **Leaning against the wind** improves welfare (in both directions)

- **Booms and busts** Minsky (1986), Kiyotaki-Moore (1997), ...
- **Dynamics with heterogeneous agents**, e.g. beliefs (Blume and Easley (1992), Geanakoplos, 2009, Burnside et al., 2015), preferences (Borovicka, 2015), ...
- **Macro externalities** Lorenzoni (2008), Jeanne-Korinek (2010), Farhi-Werning (2016), Korinek-Simsek (2016), ...

Baseline setup

- $i = 1 \dots N$ types of bankers with log preferences

$$U_i = \sum \beta^t E [\log c_{it}]$$

with a unit mass of agents each and initial endowment k_{i0}
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- capital allocation in vector notation:

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Lemma (Exclusion of Inferior Strategies)

Bankers who earn geometric mean return $E[\log \tilde{R}_j] < \log \bar{R}$ will see

$$\lim_{T \rightarrow \infty} k_{jT} / K_T = 0 \quad \text{a.s.}$$

- **Definition 1: Increasing Risk:**

allocation k_t is riskier than k'_t for $K_t = K'_t$ if

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- **Definition 2: Volatility:**

n-period-ahead volatility $V_{t+n} = \frac{Std(K_{t+n})}{K_t}$

Proposition (Volatility and Procyclicality)

Volatility: *The more risky the wealth distribution of bankers, the greater the period-ahead volatility of aggregate wealth.*

Pro-cyclicality: *The more positive shocks the economy experiences,*

- *the greater the period-ahead volatility of aggregate wealth and*
- *the greater the loss from a negative shock.*

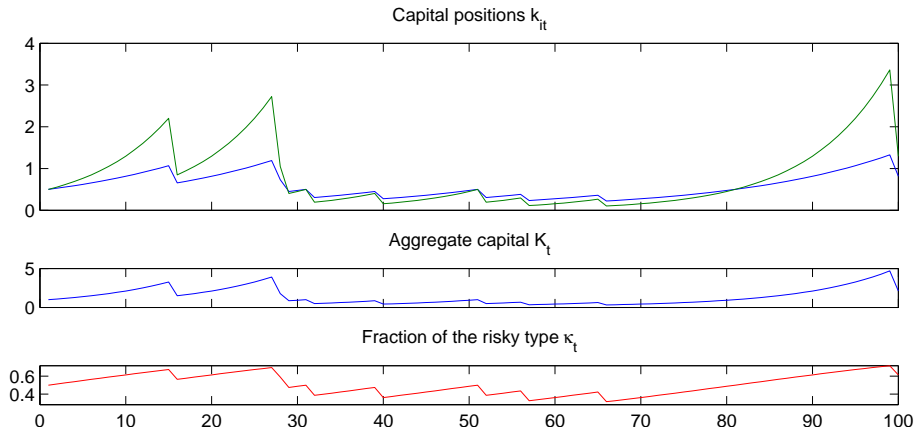
→ Minsky's financial instability hypothesis

Example with Two States of Nature

- two types $i = 1, 2$ and two states $\omega \in \{L, H\}$
- identical geometric mean return \bar{R} but different dispersion $s_1 < s_2$

$$\tilde{R}_i = \begin{cases} \bar{R}(1 + s_i)^{-\frac{1}{\rho}} & \text{in low state } L \text{ (with prob. } \rho) \\ \bar{R}(1 + s_i)^{\frac{1}{1-\rho}} & \text{in high state } H \text{ (with prob. } 1 - \rho) \end{cases}$$

Simulation 1: Volatility and Procyclicality



Optimal Capital Allocation

- under laissez faire, capital shares $\kappa_{it} = k_{it}/K_t$ fluctuate pro-cyclically
→ is this optimal?

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→ is this optimal?
- compare to capital allocation chosen by a social planner who
 - allocates capital among all technologies
 - allocates consumption to the agents of the economy

$$\max \sum_{i=1}^N \sum_{t=1}^{\infty} \lambda_i \beta^t E[\log c_{it}]$$

Proposition (Optimal Capital Allocation)

- 1 The planner allocates constant shares κ_i^* to each strategy to solve

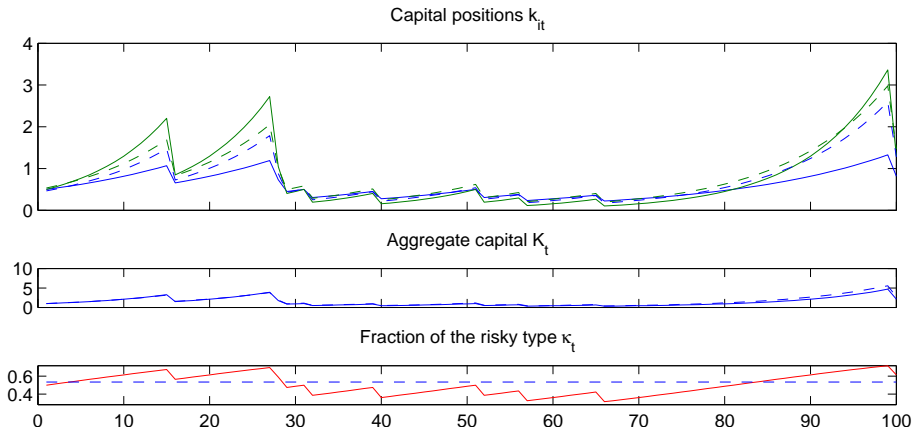
$$\max_{\kappa_i \in [0,1]} E \left[\log \sum_i \kappa_i \tilde{R}_i \right]$$

- 2 Economic growth is a.s. faster than in the decentralized economy.

Note:

- planner solves [static] portfolio allocation problem
- she overcomes the imperfections in risk markets

Simulation 1': Dynamics With Optimal Capital Allocation



How do we implement a more stable capital allocation?

Proposition (Effects of Financial Regulation)

Restricting risk-taking in a given period t leads to:

- *a **static effect** on period t volatility from restricting the choice set and*
- *a **dynamic effect** from changing the wealth composition in all future periods, which*
 - *reduces volatility if the period t shock is positive*
 - *increases volatility if the period t shock is negative*

→ the dynamic effect of regulation is inherently counter-cyclical

Model extension to capture spillovers to the real economy

- Unit mass of workers:
 - same (log) utility as bankers
 - supply one unit of labor
 - live hand-to-mouth so $c_t = w_t$,
- after shock realization, capital k'_{it} lent to real economy for production

$$y_{it} = Ak'_{it}{}^\alpha \ell_{it}^{1-\alpha}$$

- Equilibrium wage satisfies $w_t = (1 - \alpha)AK'_t$

Spillovers to the Real Economy

Observe: Workers care about stable supply of risk capital

Proposition (Spillovers)

- 1 *The results on (a) procyclicality and (b) on the optimal capital allocation continue to hold.*
- 2 *Aggregate bank capital creates spillovers to the real economy.*

Role of Financial Regulation:

- ensure stable supply of capital to the real economy
- desirable to stabilize capital shares of different investment technologies
- output and wages less volatile
- output and wages on average higher

Spillovers and bailouts

- $\exists \hat{K}$ such that workers benefit from providing one-way transfers to bankers
- assume transfers distributed in lump-sum fashion

Proposition (Bailouts and Natural Selection)

The introduction of bailouts

- *increases the fraction of capital controlled by high risk-types*
- *allows for long-run survival of inferior risk types (that would otherwise go extinct)*

Idiosyncratic Shocks

Big theme in literature on firm heterogeneity: technological churning

Three interpretations in our setup:

- 1 change in set of investment opportunities for given bankers, e.g.:
 - random change in technology
 - change in management/personnel
 - merger, take-over, firm entry and exit
- 2 reallocation of capital between bankers by outside investors
- 3 reallocation via public policy

→ for all three, implications captured by transition matrix $M = (m_{ij})$, where m_{ij} is fraction of capital of type i that is reallocated to type j

Law of motion under reallocation:

$$k_{t+1} = \beta M \tilde{R}_t k_t$$

- Optimal reallocation captured by matrix $M^* = \begin{pmatrix} \kappa_1^* & \cdots & \kappa_1^* \\ \cdots & \cdots & \cdots \\ \kappa_N^* & \cdots & \kappa_N^* \end{pmatrix}$
→ implements planner's constant capital shares
- Symmetric churning (two types): $M^{sym} = \begin{pmatrix} 1 - \mu & \mu \\ \mu & 1 - \mu \end{pmatrix}$

Proposition (Symmetric Churning, Two Risk Types)

- (i) *Introducing a small churning rate $\mu > 0$ is desirable iff $\kappa_t < \min \{\kappa^*, 1/2\}$ or $\kappa_t > \max \{\kappa^*, 1/2\}$*
- (ii) *The optimal churning rate is*

$$\mu^*(\kappa_t) = \min \{ \max \{ 0, \mu(\kappa_t) \}, 1 \} \quad \text{where} \quad \mu(\kappa_t) = \frac{\kappa_t - \kappa^*}{1 - 2\kappa_t}$$

Desirability of Random Churning

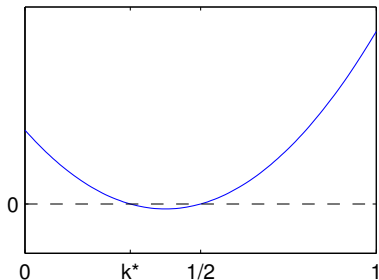
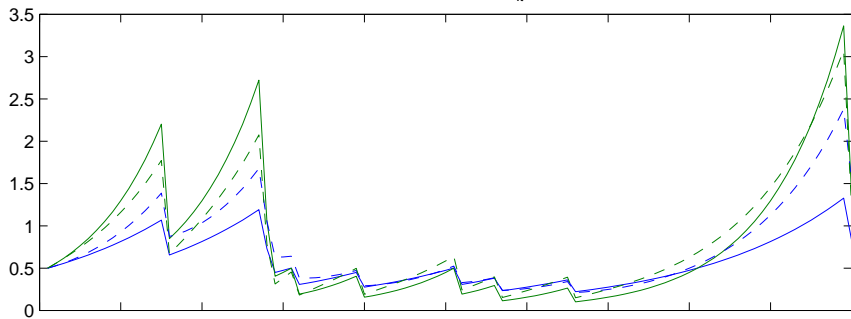


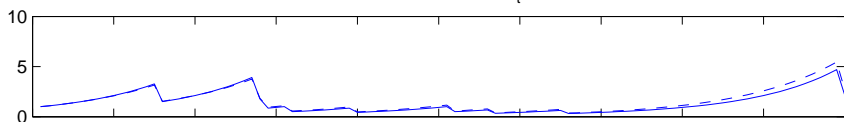
Figure: Desirability of introducing churning $\left. \frac{dE[\log K_{t+1}]}{d\mu} \right|_{\mu=0}$

Simulation 2: Random Churning

Capital positions k_{it}



Aggregate capital K_t



Fraction of the risky type κ_t



State-Dependent Idiosyncratic Shocks

- state-depnt reallocation matrix \tilde{M}_t
- law-of-motion $k_{t+1} = \tilde{M}_t \tilde{R}_t k_t$

Definition (Momentum and Contrarian Reallocations)

Momentum: M^L is upper-triangular and M^H is lower-triangular

Contrarian: vice versa

Proposition (Contrarian Reallocation and Volatility)

For $\kappa_t \in [\kappa^L, \kappa^H]$, contrarian reallocation reduces period-ahead volatility V_{t+1} whereas momentum-based reallocation increases period-ahead volatility.

Conclusions

- Heterogeneity in risk exposure creates novel channel that drives the riskiness of the aggregate economy
- Pro-cyclicality
- Policy interventions have dynamic selection effects
 - which financial institutions will benefit?
- Role for smoothing cycles – in both directions