

# Heterogeneous Agents, Projection, and Aggregation

## A Master Equation Approach

WORK IN PROGRESS

*Thomas Bourany*

THE UNIVERSITY OF CHICAGO

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## Solving Heterogeneous Agents models with aggregate risk is hard

- ▶ Huge literature since the classic papers of Krusell-Smith (KS) and Den Haan
- ▶ Main difficulty:
  - With rational expectations, GE, and aggregate shocks, agents need to forecast the dynamics of prices and aggregate variables
  - ⇒ The distribution of agents enters the household/firm decision problem
  - It gives rise to a “Master equation”, the value depends on an infinite-dimensional object  $g$

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  - It gives rise to a “Master equation”, the value depends on an infinite-dimensional object  $g$
- ▶ Today’s contribution:
  - Provide a method to solve this class of models with aggregate risk
  - Merge Krusell-Smith’s original idea with modern treatments of the Master Equation
  - Allow to benchmark “non-rational expectations” methods

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  - More recent methods bypassing this limitation – Second-order methods, *Bhandari, Bourany, Evans, Golosov (2023), Bayer, Luetticke, Weiss, Winkelmann (2025)*, Machine-Learning-based methods, *Fernandez-Villaverde, Hurtado, Nuno (2023), Huang (2023), Gu, Laurière, Merkel, Payne (2024)*, or others *Proehl (2019), Schaab (2021)* – may seem a little opaque or case-specific

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  - Mean Field Games w/ Common Noise, Master equation: *Cardaliaguet, Delarue, Lions, Lasry (2019)*
  - Also introduced in economics: *Schaab (2021), Bilal (2023), Gu, Laurière, Merkel, Payne (2024)*



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- ▶ Other applications (WIP)
  - Allows to benchmark “non-rational expectations” methods
    - Moll (2025): Should we depart from RE in HA Models? *Sargent/Sims “wilderness of non-rational expectations”*
  - Extends to general macro-finance models/portfolio choice

# Outline

1. Krusell Smith Model
2. Primer on the Master equation
3. Projection in Heterogeneous Agents Models
4. Numerical results for KS98
5. Testing bounded-rationality assumption in KS98.
6. “Macrofinance”: portfolio choice and Second Order Master Equation

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## Krusell-Smith (1998) recap

- Consumption-saving model,  $c, a$ , with
  - (i) idiosyncratic income risk  $z$ , (ii) incomplete market, (iii) credit constraints  $a \geq \underline{a}$
  - (iv) aggregate shock on aggregate TFP  $Z$ .
- Distribution of households  $g(a, z)$  over wealth and income

## Krusell-Smith (1998) recap

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  - (iv) aggregate shock on aggregate TFP  $Z$ .
- Distribution of households  $g(a, z)$  over wealth and income
- Firm side:

$$Y = ZK^\alpha \quad \Rightarrow \quad r = \alpha K^{\alpha-1} - \delta \quad w = (1-\alpha)K^\alpha$$

- Household problem (KS98)

$$V(a, z, g, Z) = \max_{c, a'} u(c) + \beta \mathbb{E}^{z', Z'} [V(a', z', g', Z') \mid z, Z]$$

$$s.t. \quad c + a' = zw + (1+r)a$$

$$g' = H(g, Z, Z')$$

- Equilibrium

$$K = \int_{a, z} a dg(a, z) \quad \forall Z$$

## General idea and KS98 global solution

- ▶ Difficulty: Value function  $V(a, z, g, Z)$  depends on the whole distribution  $g$
- ▶ Need to forecast the evolution of  $g \Rightarrow$  very difficult with aggregate risk
  - Need to follow distribution  $g_t$  on *every path* of  $\{Z_t\}_t$
  - Brute force: computationally intensive, c.f. Achdou, Bourany (2018)
- ▶ Krusell-Smith solution
  1. Assume the Household only care about aggregate capital / First-moment  $K = \int a dg(a, z)$
  2. Assume *Linear* forecasting rule for future capital

$$\log K' = a_1^Z \log K + a_2^Z$$

- Choose parameters  $(a_1^Z, a_2^Z)$  to match the *realized / simulated* path (Monte Carlo) of  $\{K_t\}_t$
- ▶ Proposal today:
  - remove assumption 2  $\Rightarrow$  bypass the linearity assumpt<sup>o</sup> (in that sense close to FVHN)
  - test robustness to 1 and 2, using methods based on the Master equation

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## Primer on Mean Field Games and the Master Equation

- ▶ Aiyagari model rewrites as PDEs: MFG system
  - States dynamics: saving and labor income shocks

$$da_t = [z_t w_t + r_t a_t - c_t] dt \quad z_j \sim \text{Markov jump process } \lambda_j$$



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### 1. Hamilton Jacobi Bellman Equation:

$$-\partial_t v(t, a, z) + \rho v(t, a, z) = \max_c u(c) + \mathcal{L}[v | c](t, a, z)$$

- Transport/Jump-Operator  $\mathcal{L}$ : *from agents' decision and shocks*

$$\mathcal{L}[v | c^*](t, a, z_j) = \underbrace{\partial_a v(t, a, z_j) [z_j \mathbf{w} + \mathbf{r} a - c^*]}_{\text{change in saving}} + \underbrace{\lambda_j (v(t, a, z_{-j}) - v(t, a, z_j))}_{\text{change in labor income}}$$

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### 2. Kolmogorov Forward Equation:

$$\partial_t g(t, a, z) = \mathcal{L}^*[g | c^*](t, a, z)$$

—  $\mathcal{L}^*$ : distribution dynamics comes from agents' decisions

- Equilibrium:

$$\iint_{z, a \geq \underline{a}} a dg(t, a, z_j) = K \quad r = \alpha K^{\alpha-1} - \delta$$

## Primer on the Master Equation

- The master equation combines in *one equation* both **HJB** and **KFE**
  - Case without aggregate risk *c.f. Cardaliaguet, Delarue, Lions, Lasry (2019), Schaab (2021), Bilal (2023)*

$$\begin{aligned}
 -\partial_t v(t, a, z, \mathbf{g}) + \rho v(t, a, z, \mathbf{g}) = & \overbrace{\max_c u(c) + \mathcal{L}[v | c^*](t, a, z)}^{\text{standard HJB continuation value}} + \\
 & \underbrace{\iint_{z, a} \frac{dv(t, a, z, \mathbf{g})}{dg} [(\tilde{a}, \tilde{z})] \mathcal{L}^*[\mathbf{g} | c^*](t, \tilde{a}, \tilde{z}) d\mathbf{g}(t, \tilde{a}, \tilde{z})}_{\text{change in } v \text{ due to the distribution dynamics}}
 \end{aligned}$$

- First part: **HJB**, how states  $(a, z)$  change agents' value  $v$
- **Novelty**: depends on how the distribution  $\mathbf{g}$  changes the value  $v$ 
  - Notice the forecast from agents  $(a, z)$  about all the other agents  $(\tilde{a}, \tilde{z})$

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  - Notice the forecast from agents  $(a, z)$  about all the other agents  $(\tilde{a}, \tilde{z})$
  - 1.  $\mathcal{L}^*[\mathbf{g} | c^*]$ : How the agents' decision change the distribution  $\mathbf{g}$
  - 2.  $dv/d\mathbf{g}$ : How the distribution changes the value
    - Needs to define derivatives in space of measures  $\frac{dv(\mathbf{g})[\tilde{\mathbf{x}}]}{d\mathbf{g}}$ : Lions' derivative

*Lions vs. Fréchet*

# Adding Aggregate Risk to the Master Equation

## ► Consider aggregate risk

- Agg. TFP follows a AR(1) – Ornstein-Uhlenbeck process

$$dZ_t = -\theta(Z_t - \bar{Z})dt + \hat{\sigma}dB_t^0$$

- The master equation doesn't change much: value  $v = v(t, a, z, g, Z)$

$$\begin{aligned}
 -\partial_t v + \rho v = & \underbrace{\max_c u(c) + \mathcal{L}[v|c](t, a, z)}_{\text{standard HJB continuation value}} \quad \underbrace{-\theta(Z - \bar{Z})v_Z + \frac{\hat{\sigma}^2}{2}v_{ZZ}}_{\text{direct effect of risk of } Z \text{ on } v} \\
 & + \underbrace{\iint_{z,a} \frac{dv(t, a, z, g, Z)}{dg} [(\tilde{a}, \tilde{z})] \mathcal{L}^*[g|c^*](t, \tilde{a}, \tilde{z}) dg}_{\text{change due to distribution dynamics}}
 \end{aligned}$$

- Why?
  - Aggregate shocks don't have *direct effects* on individual states!
  - As a result, distribution  $g$  is not affected/deformed directly by shocks  $dB_t^0$
  - If it were, it would become *second order*: much more complicated  $\Rightarrow$  Monster equation!
  - More later (portfolio problems) if time permits

## General (Second Order) Master Equation

- Include controlled drift, diffusion, jump on individual states  
+ mean-field interaction on drift, diffusion, and jump on aggregate states
- Encompasses most macro-finance models (e.g. portfolio choice)

$$\begin{aligned}
 \mathcal{H}(x, m, \mathcal{X}, V, D_x V, D_{xx} V) = & \max_c \mathcal{L}(x, m, \mathcal{X}, c) + b(x, m, \mathcal{X}, c) \cdot D_x V + \text{Tr}([\sigma \sigma' + \bar{\sigma} \bar{\sigma}'](x, m, \mathcal{X}, c) D_{xx} V) \\
 & \sum_{n=1}^{n_J^i} \lambda^n(x, m, \mathcal{X}, c) \left( V^n(x + \gamma(x, m, \mathcal{X}, c), x, m, \mathcal{X}) - V \right) \\
 -\partial_t V + \rho V = & \mathcal{H}(x, m, \mathcal{X}, V, D_x V, D_{xx} V, c^*) \\
 & + \mu(m, \mathcal{X}) \cdot D_{\mathcal{X}} V + \text{Tr}(\hat{\sigma} \hat{\sigma}' D_{\mathcal{X} \mathcal{X}} V) + \sum_{n=1}^{n_J^0} \hat{\lambda}^n(m, \mathcal{X}) \left( V \circ \hat{\gamma}^n(m, \mathcal{X}) - V \right) \\
 & + \int_{\mathbb{X}} D_m V(x, \cdot; y) \cdot D_p \mathcal{H}(y, \cdot) m(dy) + \int_{\mathbb{X}} \sum_{n=1}^{n_J^0} \lambda^n(y, \cdot) \Delta_m V(x, \cdot; y) \circ \gamma(y, \cdot) m(dy) \\
 & + \int_{\mathbb{X}} \text{Tr}[(\sigma \sigma' + \bar{\sigma} \bar{\sigma}')(y, \cdot) D_y (D_m V(x, m, \mathcal{X}; y))] (y, m, \mathcal{X}) m(dy) \\
 & + 2 \int_{\mathbb{X}} \text{Tr}(\bar{\sigma}(x, \cdot) \bar{\sigma}'(y, \cdot)' D_x D_m V(x, \cdot; y)) m(dy) + \int_{\mathbb{X}} \text{Tr}(\bar{\sigma}(y, \cdot) \hat{\sigma}(\mathcal{X}_t)' D_m D_{\mathcal{X}} V(x, m, \mathcal{X}; y)) m(dy) \\
 & + \int \int_{\mathbb{X}} \text{Tr}(\bar{\sigma}(y, \cdot) \bar{\sigma}'(y', \cdot)' D_{mm}^2 V)(x, \cdot; y, y') m(dy) m(dy')
 \end{aligned}$$

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## Projection assumption in KS98

► Back to KS98. What do households need for decisions?

- Require only changes in prices  $(r, w) \Rightarrow$  do not care of the distribution  $g$  *per se*
- Only need the change in one moment, *the mean* of  $g$ , to forecast future prices!

$$K = \iint_{a,z} a \, dg(a, z) \qquad r = \alpha K^{\alpha-1} - \delta$$

- Assumption (KS98 as well as here)

$$v(a, z, g, Z) \equiv \bar{v}(a, z, K^h, Z)$$

- **Novelty**: Nice property in Lions-derivative for the Master equation

$$\text{with } K^h = \int_x h(x) \, dg(x) \qquad \frac{d}{dg} v(x, g)[\tilde{x}] \equiv \frac{d}{dK^h} \bar{v}(x, K^h) h'(\tilde{x})$$

- First moment:  $h(x) = x$ , and  $\frac{dv(x, g)}{dg} \equiv \frac{d\bar{v}(x, K^h)}{dK^h}$



## Projection in the Master equation

- Rewrite Master Equation with projection on first-moment:  $v(a, z, \mathbf{g}, \mathbf{Z}) \equiv \bar{v}(a, z, \mathbf{K}, \mathbf{Z})$

$$\rho \bar{v} = \overbrace{\max_c u(c) + \mathcal{L}[\bar{v} | c]_{(a, z, \mathbf{K}, \mathbf{Z})}}^{\text{standard HJB continuation value}} \overbrace{-\theta(Z - \bar{Z})\bar{v}_Z + \frac{\hat{\sigma}^2}{2}\bar{v}_{ZZ}}^{\text{direct effect of risk of } \mathbf{Z} \text{ on } \bar{v}} + \bar{v}_K \underbrace{\int_{z, a} [r\tilde{a} + w\tilde{z} - c^*(\tilde{a}, \tilde{z}, \mathbf{K}, \mathbf{Z})] dg(\tilde{a}, \tilde{z})}_{\text{change in agents } (\tilde{a}, \tilde{z}) \text{ decisions}}$$

- Still dependence on  $\mathbf{g}$ ! How to “get rid of it”? (next slides!)

- Aggregation:

$$dK = \iint_{z, a} [r\tilde{a} + w\tilde{z} - c^*(\tilde{a}, \tilde{z}, \mathbf{K}, \mathbf{Z})] dg(\tilde{a}, \tilde{z})$$

$$d\mathbf{K} = r\mathbf{K} + w\bar{L} - \mathcal{C}(\mathbf{K}, \mathbf{Z} | \mathbf{g})$$

with aggregate consumption  $\mathcal{C}(\mathbf{K}, \mathbf{Z} | \mathbf{g}) = \iint_{z, a} c^*(\tilde{a}, \tilde{z}, \mathbf{K}, \mathbf{Z}) dg(\tilde{a}, \tilde{z})$

## The Master Equation becomes a fusion of two familiar equations

- The Master Equation becomes a “standard” HJB,  $v = v(a, z, \mathbf{g}, Z) \equiv \bar{v}(a, z, K, Z)$

$$\begin{aligned} \rho \bar{v} = \max_c u(c) + [\mathbf{w}z + \mathbf{r}a - c] \bar{v}_a + \lambda(\bar{v}(a, z', \cdot) - \bar{v}(a, z, \cdot)) \\ - \theta(Z - \bar{Z}) \bar{v}_Z + \frac{\hat{\sigma}^2}{2} \bar{v}_{ZZ} + \underbrace{[\mathbf{Z}K^\alpha - \delta K - \mathcal{C}(K, Z | \mathbf{g})]}_{=dK} \bar{v}_K \end{aligned}$$

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- Only issue:  $\mathcal{C}(\mathbf{K}, \mathbf{Z}|\mathbf{g})$  still depends on  $\mathbf{g}$  (next slide!)
- Looks exactly like the fusion of two standard models
  - RBC:  $v = v(K, Z)$

$$\rho v = \max_C u(C) + [\mathbf{Z}\mathbf{K}^\alpha - \delta\mathbf{K} - C]v_K - \theta(Z - \bar{Z})v_Z + \frac{\hat{\sigma}^2}{2}v_{ZZ}$$

- Aiyagari:  $v = v(a, z)$

$$\rho v = \max_c u(c) + [\mathbf{w}z + \mathbf{r}a - c]v_a + \lambda(v(a, z', \cdot) - v(a, z, \cdot))$$

## Agents' decision and global dynamical system

- With the Master equation and  $v = \bar{v}(a, z, K, Z)$  we obtain individual decisions:

$$c^*(\tilde{a}, \tilde{z}, K, Z) = u'^{-1}(\bar{v}_a(\tilde{a}, \tilde{z}, K, Z))$$

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⇒ get dynamical system for  $x = (a, z, K, Z)$

$$\begin{cases} da &= [z \overbrace{(1-\alpha)ZK^\alpha}^{=w} + \overbrace{(\alpha ZK^{\alpha-1} - \delta)}^{=r} a - c^*(a, z, K, Z)] dt \\ dz &= \gamma(z) dJ_t \quad \text{Markov, w/ intensity} \quad \lambda(z) \\ dK &= (ZK^\alpha - \delta K - \mathcal{C}(K, Z|g)) dt \\ dZ &= -\theta(Z - \bar{Z}) dt + \hat{\sigma} dB_t^0 \end{cases}$$

- For guess of  $g(a,z)$  and  $\mathcal{C}(K, Z|g) = \iint_{a,z} c^*(a, z, K, Z) dg(a,z) \Rightarrow$  complete characterization

⇒ Can get a Kolmogorov forward equation for system  $(a, z, K, Z)$  (!)

## “Master-” Kolmogorov Forward for the global system

- For a guess of  $g(a, z)$  and  $\mathcal{C}(K, Z|g) = \iint_{a,z} c^*(a, z, K, Z) dg(a, z)$ 
  - We can solve a “Master-KFE” for states  $x = (a, z, K, Z) \in \mathbb{X}$  to find the distribution  $\tilde{g}(x)$

$$0 = -\partial_a \left[ s(x, \bar{v}_a) \tilde{g}(x) \right] + \sum_n \lambda(z^n) \tilde{g}(x^n) - \lambda(z) \tilde{g}(x) \\ - \partial_K \left[ (ZK^\alpha - \delta K - \mathcal{C}(K, Z|g)) \tilde{g}(x) \right] - \partial_Z [-\theta(Z - \bar{Z}) \tilde{g}(x)] + \hat{\sigma} \partial_{ZZ}^2 \tilde{g}(x)$$

- Easy to get  $\tilde{g}$  from Master-HJB’s operator  $\mathcal{A}[\bar{v}]$  with finite-difference methods
- Consistency condition for rational-expectation equilibrium:

$$g(a, z)|_{K,Z} = \frac{\tilde{g}(a, z, K, Z)}{\tilde{g}(K, Z)} \quad \tilde{g}(K, Z) = \int_{\mathbb{X}} \delta_{\{\tilde{K}=K, \tilde{Z}=Z\}} d\tilde{g}(a, z, \tilde{K}, \tilde{Z})$$

- Using this  $g$ , we can obtain  $dK = (ZK^\alpha - \delta K - \mathcal{C}(K, Z|g))dt, \Rightarrow$  all we needed!!

## Summary and numerical methods

1. General Master equation: one equation for  $v(a, z, g, Z)$
2. Master Equation with “projection”:  $v = \bar{v}(a, z, K, Z)$ 
  - Start from guess  $g(a, z)$  and  $\mathcal{C}(K, Z|g)$
  - Solve Master-HJB: standard finite difference methods
  - Get individual decisions  $c^*(a, z, K, Z)$  and operator  $\mathcal{A}[\bar{v}]$  for  $x = (a, z, K, Z)$
3. “Master”-Kolmogorov forward for  $x = (a, z, K, Z)$ 
  - Solve for distribution  $\tilde{g}$  over all states  $(a, z, K, Z)$  for “free” with  $\mathcal{A}^*[\tilde{g}]$
  - Update  $g$  thanks to  $\tilde{g}$  and update  $\mathcal{C}(K, Z|g)$
  - Obtain aggregate dynamics: potentially very non-linear!!

$$dK = ZK^\alpha - \delta K - \mathcal{C}(K, Z|g)$$

### ► General procedure

- No need for deep-learning/splines/polynomials: use “**standard**” finite difference methods
- **RE**: Does **not** rely on bounded-rationality assumption of KS98, or forecasting rule

## Master-Equation with higher moments:

- HJB with 2nd-order moments:  $v(a, z, \mathbf{g}, Z) \equiv \bar{v}(a, z, \mathbf{K}, \mathbf{K}_2, L_2, KL, Z) = \bar{v}(a, z, \mathbf{K}, \mathbf{K}_2, \mathbf{Z})$
- $\mathbf{K}_2 = \mathbb{V}\text{ar}(a), L_2 = \mathbb{V}\text{ar}(z), KL = \mathbb{C}\text{ov}(a, z)$ .

$$\begin{aligned} \rho \bar{v} = \max_c & u(c) + (wz + ra - c) \bar{v}_a + \lambda(\bar{v}(a, z', \cdot) - \bar{v}(a, z, \cdot)) - \theta(Z - \bar{Z}) \bar{v}_Z + \frac{\hat{\sigma}^2}{2} \bar{v}_{ZZ} \\ & + \underbrace{(\mathbf{Z} \mathbf{K}^\alpha - \delta \mathbf{K} - \mathbb{E}^g[c^*])}_{=d\mathbf{K}} \bar{v}_K + \underbrace{\mathbb{C}\text{ov}^g(a, s^*)}_{d\mathbf{K}_2} \bar{v}_{K_2} \end{aligned}$$

- Similarly, solve dynamical system  $(a, z, \mathbf{K}, \mathbf{K}_2, \mathbf{Z})$ , the “master” KFE and then plug  $\mathbf{g}$  back into  $\mathbb{E}^g[c^*] = \iint c^* d\mathbf{g}$  and  $\mathbb{C}\text{ov}^g(a, s^*) = \iint (a - \bar{a})(ra + wz - c^*) d\mathbf{g}$



## Master-Equation with higher moments:

- HJB with 2nd-order moments:  $v(a, z, \mathbf{g}, Z) \equiv \bar{v}(a, z, \mathbf{K}, K_2, L_2, KL, Z) = \bar{v}(a, z, \mathbf{K}, K_2, \mathbf{Z})$
- $K_2 = \mathbb{V}\text{ar}(a), L_2 = \mathbb{V}\text{ar}(z), KL = \mathbb{C}\text{ov}(a, z)$ .

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- Theoretical insights:
  - In KS98, you don’t need all the moments!
    - $\bar{v}_{L_2}$  and  $\bar{v}_{KL}$  drops from HJB: No change in variance in labor income,  $dL_2 =$  and  $dKL = 0$
  - If  $\bar{v}_{K_2} < 0$  and  $\mathbb{C}\text{ov}^g(a, s^*) > 0$ , it reinforces the precautionary saving motive and lower value

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1. Krusell Smith Model
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## Numerical results – Summary

► Risk:

- Idiosyncratic risk: Two-state Markov process for labor income shocks  $z$
- Aggregate risk: Three-state Markov process for TFP  $Z$ , with  $\sigma(Z) = 12\%$

► I compare three economies:

1. RBC (Brock Mirman) model

⇒  $(v, g)$  value and distribution over aggregate capital and TFP  $(K, Z)$ .

2. Aiyagari model

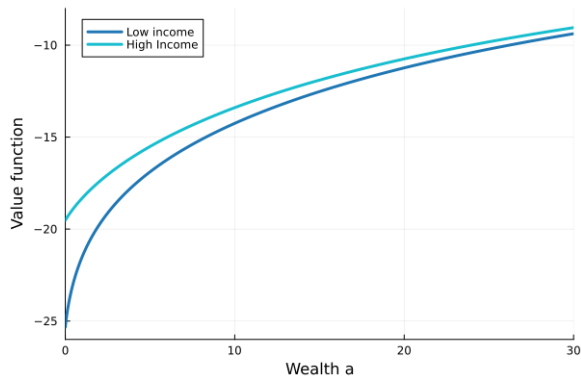
⇒  $(v, g)$  individual heterogeneity on  $(a, z)$ , for constant TFP  $Z = \bar{Z}$  and capital  $K = \bar{K}$ .

3. Krusell-Smith model

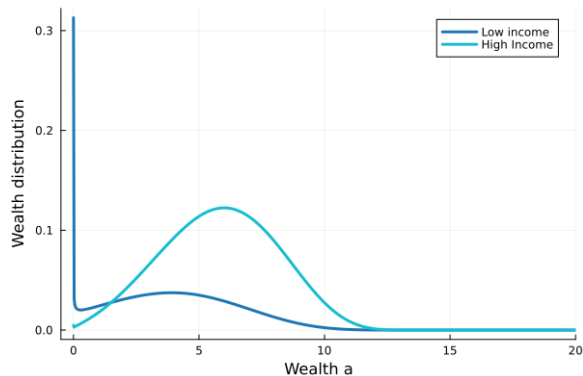
- Have both:  $(v, g)$  over  $(a, z, K, Z)$
- Iterate over  $dK = ZK^\alpha - \delta K - C(K, Z|g)$

## Recap – Aiyagari model

Value function  $v(a, z)$

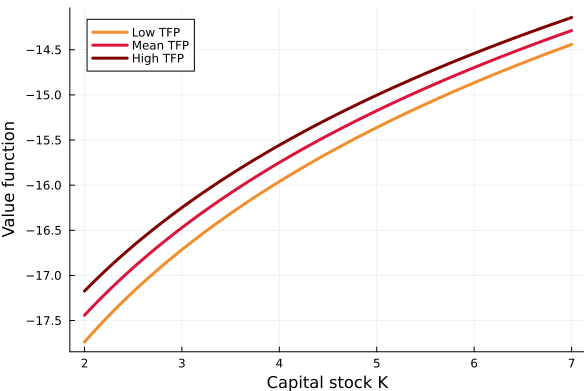


Distribution  $g(a, z)$

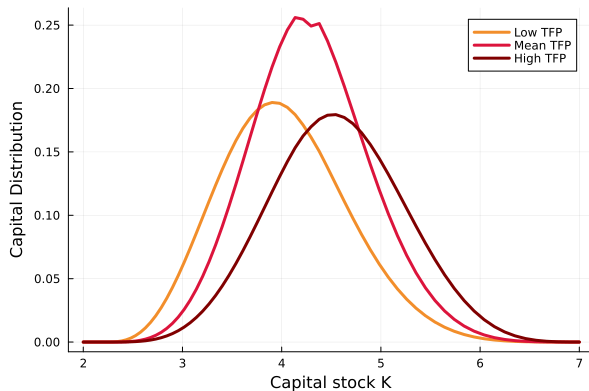


# Recap – Brock-Mirman / RBC

Value function  $v(K, Z)$

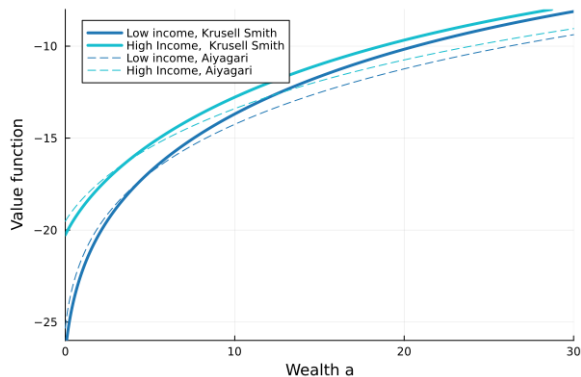


Distribution  $g(K, Z)$

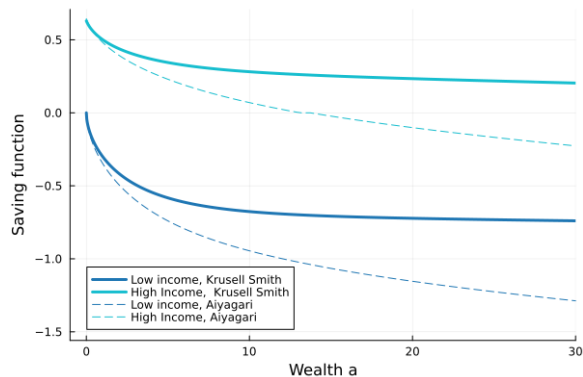


# Krusell-Smith: individual decisions, vs. Aiyagari

Value function  $v(a, z, \bar{K}, \bar{Z})$

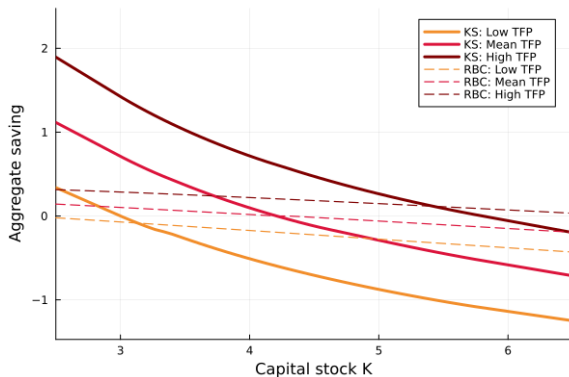


Saving  $s(a, z, \bar{K}, \bar{Z}) = wz + ra - c^*$



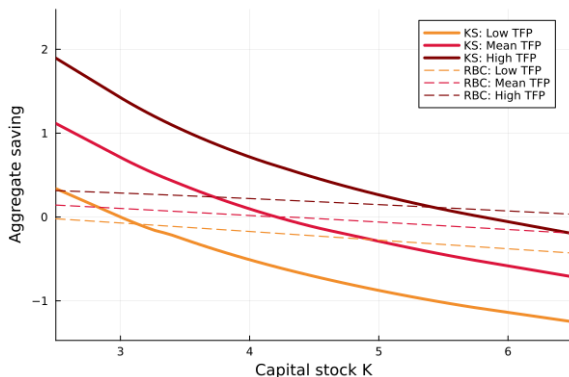
# Krusell-Smith: aggregate dynamics decisions, vs. RBC

Aggregate dynamics  $dK = ZK^\alpha - \delta K - C(K, Z|g)$

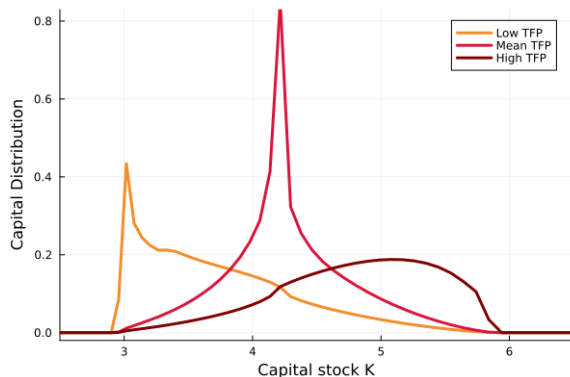


# Krusell-Smith: aggregate dynamics decisions, vs. RBC

Aggregate dynamics  $dK = \textcolor{blue}{Z}K^\alpha - \delta K - \mathcal{C}(K, \textcolor{blue}{Z}|g)$



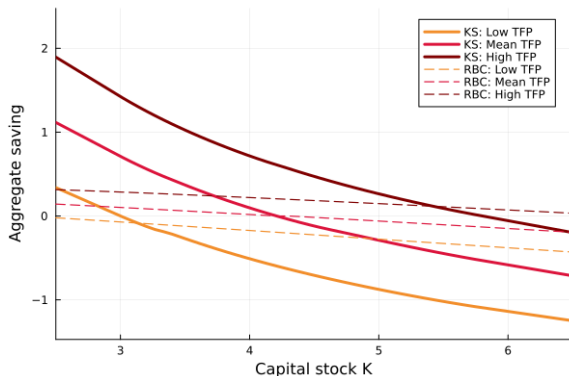
Distribution  $\tilde{g}(K, \textcolor{blue}{Z})$



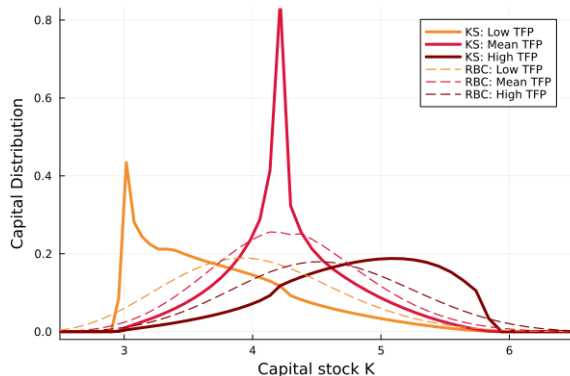


# Krusell-Smith: aggregate dynamics decisions, vs. RBC

Aggregate dynamics  $dK = \textcolor{blue}{Z}K^\alpha - \delta K - \mathcal{C}(K, \textcolor{blue}{Z}|g)$



Distribution  $\tilde{g}(K, \textcolor{green}{Z})$



## Effect of aggregate uncertainty and heterogeneity

### 1. Higher individual precautionary saving motives:

- Aggregate uncertainty affects the poor (reliant on labor income) more than the rich (hedging:  $r = MPK \uparrow$  when  $K \downarrow$ )
- Implies much more savings when rich: flatter savings function compared to Aiyagari

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2. Aggregate dynamics are more “reactive”
  - Individual heterogeneity implies a steeper aggregate saving function
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3. Change the distribution of economic activity
  - Output is fluctuating more in Krusell-Smith: 15.5% volatility compared to 14% for RBC
  - However, business cycles are less skewed – smaller right tail, more symmetric – and have less kurtosis – thinner tails of capital/output.

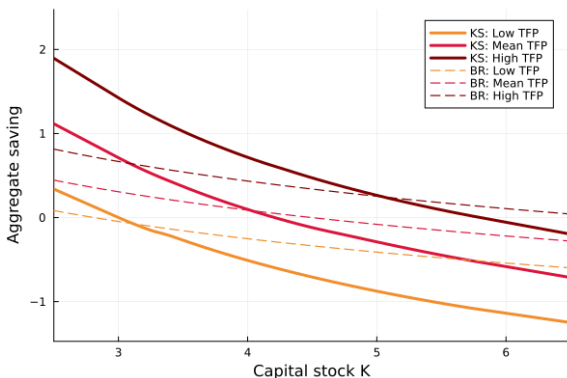
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## Bounded-rationality in Krusell-Smith

- Agents do not forecast the “true” capital dynamics  $dK$ 
  - Assume a log-linear rule:  $dK = \beta^Z \log(K) dt$
  - What do we “lose” from this assumption?

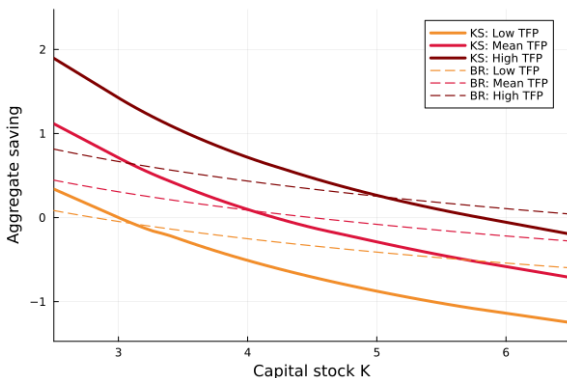
Aggregate dynamics  $dK = \beta^Z \log(K) dt$



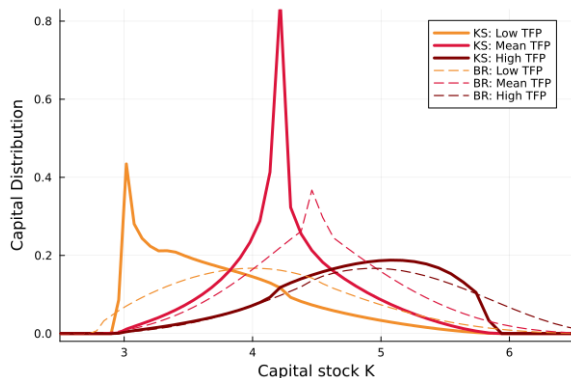
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## Aggregate Risk and Master Equation for portfolio problem

- Agg. risk: both TFP (OU)  $Z \sim \hat{\sigma} dB_t^0$  and *direct effect* on portfolio return (share  $\alpha$ )

$$da = (ra + zw - c)dt + \alpha a (R - r)dt + \alpha a \bar{\sigma} dB_t^0$$

- The master equation now becomes ***second order***! value  $v = v(a, z, g, Z)$  changes a lot!

$$\rho v = \underbrace{\max_{c, \alpha} u(c) + \mathcal{L}[v|c, \alpha](a, z, g, Z)}_{\text{standard HJB continuation value}} \underbrace{-\theta(Z - \bar{Z})v_Z + \frac{\hat{\sigma}^2}{2} v_{ZZ}}_{\text{direct effect of risk of } Z \text{ on } v} + \underbrace{\iint_{z, a} \frac{dv(a, z, g, Z)}{dg} [\tilde{a}, \tilde{z}] \mathcal{L}^*[g|c^*](\tilde{a}, \tilde{z}) dg(\tilde{a}, \tilde{z})}_{\text{deterministic evolution of the distribution}}$$

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# Conclusion

- ▶ I propose a new method for solving Heterogeneous Agent Models with aggregate risk
- ▶ Next steps:
  - Solving “more interesting” macro-finance models:  
Model with a meaningful distribution of portfolios, exposure, and impact of aggregate risk

## Primer on the Lions derivative

- Derivative in the space of distribution:  
how the value  $v(a, z, \mathbf{g})$  changes when the distribution of agents  $\mathbf{g}$  moves?

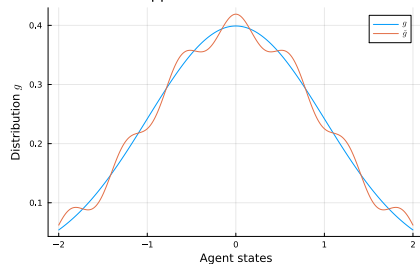
$$\begin{aligned} dv(a, z, \mathbf{g}) &\approx v(a, z, \tilde{\mathbf{g}}) - v(a, z, \mathbf{g}) \\ &\approx \iint_{\tilde{a}, \tilde{z}} \underbrace{\frac{\partial v(a, z, \mathbf{g})}{\partial \mathbf{g}}[\tilde{a}, \tilde{z}]}_{=\text{Fréchet}} (\tilde{\mathbf{g}}(\tilde{a}, \tilde{z}) - \mathbf{g}(\tilde{a}, \tilde{z})) \end{aligned}$$

$$\approx \iint_{\tilde{a}, \tilde{z}} \underbrace{\frac{d}{d\tilde{a}} \frac{\partial v(a, z, \mathbf{g})}{\partial \mathbf{g}}[\tilde{a}, \tilde{z}]}_{=\text{Lions}} \underbrace{d\tilde{a}}_{=\text{change in decision}} \mathbf{g}(\tilde{a}, \tilde{z})$$

- $\frac{\partial v(a, z, \mathbf{g})}{\partial \mathbf{g}}[\tilde{x}]$  Fréchet Derivative, for a change of  $\mathbf{g}$  in  $\tilde{x}$
- $\frac{dv(a, z, \mathbf{g})}{d\mathbf{g}}[\tilde{x}] = \frac{d}{dx} \frac{\partial v(a, z, \mathbf{g})}{\partial \mathbf{g}}[\tilde{x}]$  Lions Derivative, for a change of  $\tilde{x}$ , i.e. a *shift* in  $\mathbf{g}(\tilde{x})$

[back](#)

Point of approximation: Fréchet derivative



Point of approximation: Lions derivative

