

# **Simulating Models with Heterogeneous Agents**

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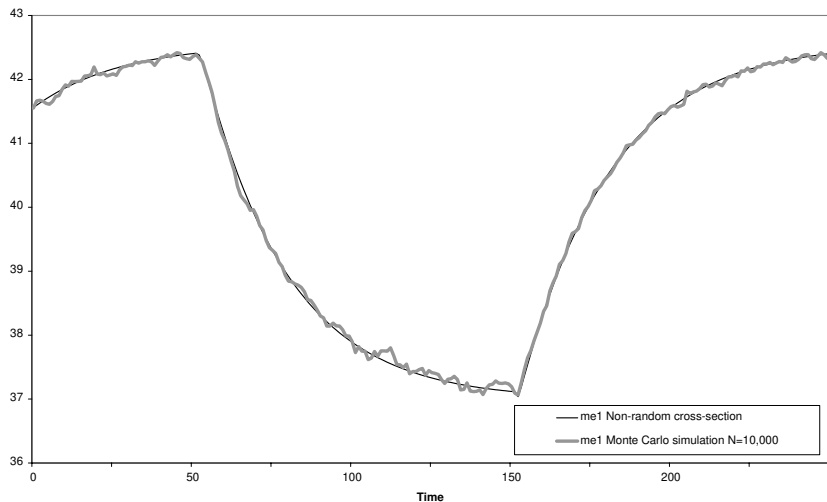
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# Two different ways to go

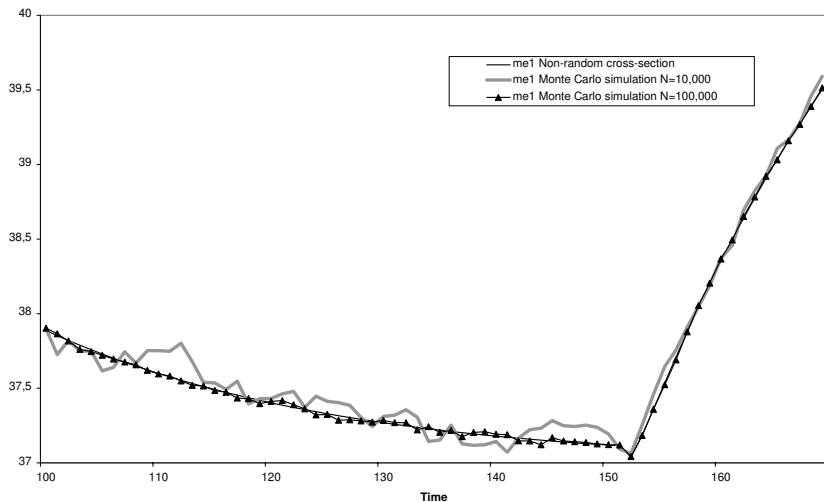
- Simulate a panel with a large number of agents
  - Monte Carlo integration to calculate cross-sectional moments  
⇒ cross-sectional sampling variation
- Grid method of Young (2010)
  - faster
  - avoids cross-sectional sampling variation  
⇒ much more accurate

# MOTIVATIONAL PICS TO USE GRID METHOD

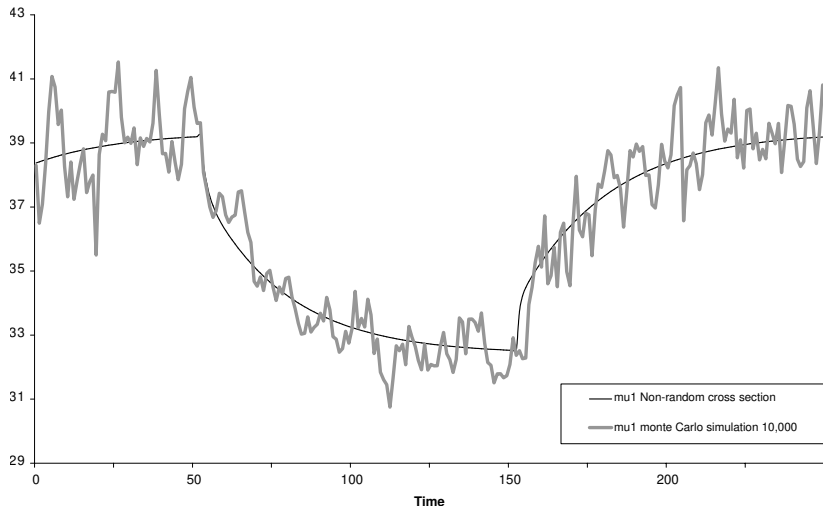
# Average capital stock employed agents



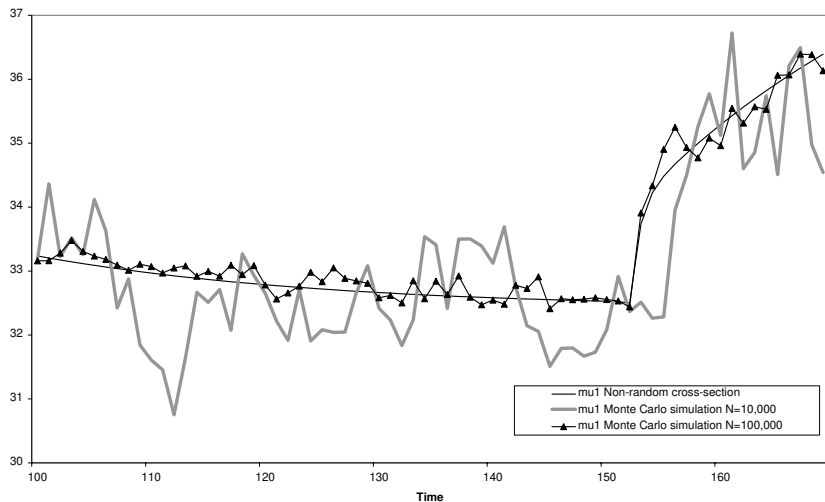
# Average capital stock employed agents



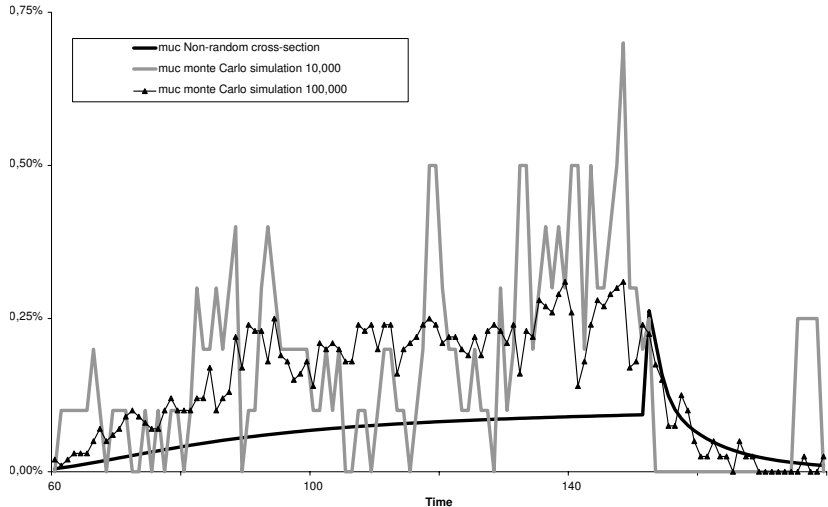
# Average capital stock unemployed agents



# Average capital stock unemployed agents



# Fraction of agents at constraint





# THE METHODS

# What is given?

- A policy function for individual choice  $k'(k_{i,t}, \varepsilon_{i,t}, S_t)$ 
  - $S_t$ : the aggregate state variables
- initial distribution for  $t = 1$ 
  - characterizes the density of capital holdings of the employed and unemployed.
- Approximate laws of motion for aggregate variables (e.g., for  $K_t$  or any other moment) are NOT used. If needed, for example to calculate  $r_t$  or  $w_t$ , then actual cross-sectional moments are used.

# Simulation method

- Cross-section consists of  $I$  agents.
- For each  $t$ 
  - Draw  $\varepsilon_{i,t}$  for each  $i$ 
    - Draw  $\varepsilon_{i,t}$ s such that unemployment rate *and* transition probabilities are exactly equal to theoretical value.  
(e.g., if  $u_t = 0.04$  and  $I = 10,000$ , then 400  $\varepsilon_{i,t}$ s should be equal to 0.)
    - This does NOT eliminate sampling noise, since it doesn't ensure these are imposed across capital holdings
  - Calculate  $k_{i,t+1} = k'(k_{i,t}, \varepsilon_{i,t}, S_t)$  for each  $i$ .
- Cross-sectional moments follow directly from cross-section.  
E.g.,  $K_t = \sum_i^I k_{i,t}$ .

# Grid method

- Fine grid with nodes:  $\kappa_i, i = 0, 1, \dots, I$
- Only mass  $AT$  grid points
  - $f_{i,t}^\varepsilon$  : mass of agents with  $k_t^\varepsilon = \kappa_i, i = 0, 1, \dots, I$
  - $\varepsilon$  : employment status
  - no mass in between grid points
- If  $k'_i \geq 0$  is binding  $\implies f_{0,t}^\varepsilon > 0$

# Grid method

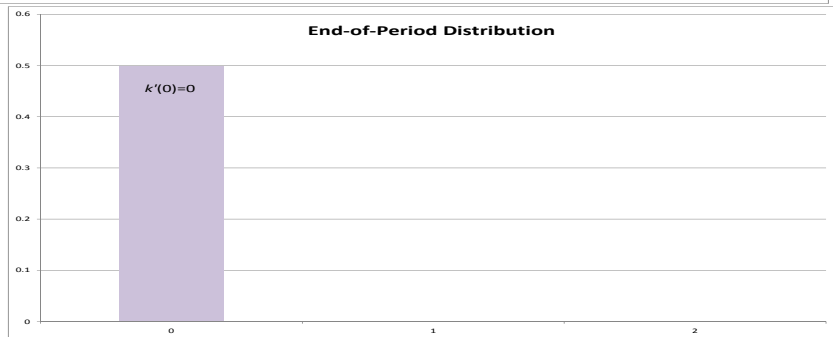
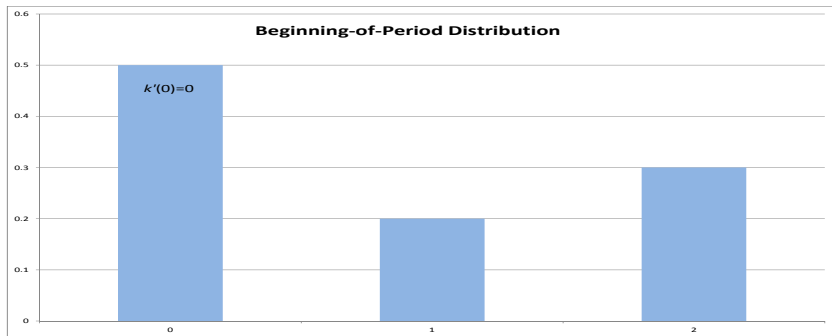
- Fix employment status (for now) to illustrate procedure
  - remain within the period  $t$  (for now), that is, go from beginning-of-period to end-of-period distribution

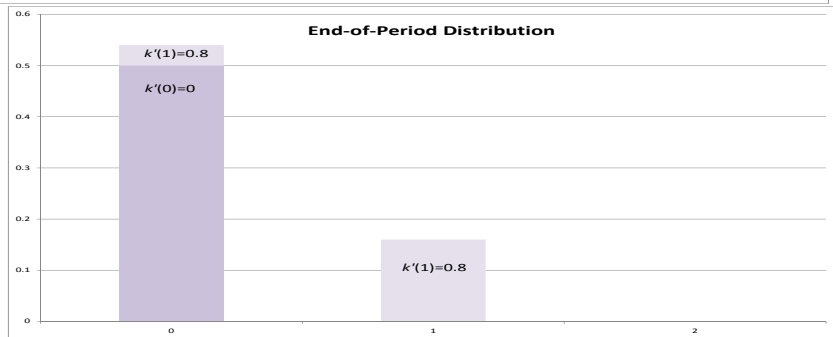
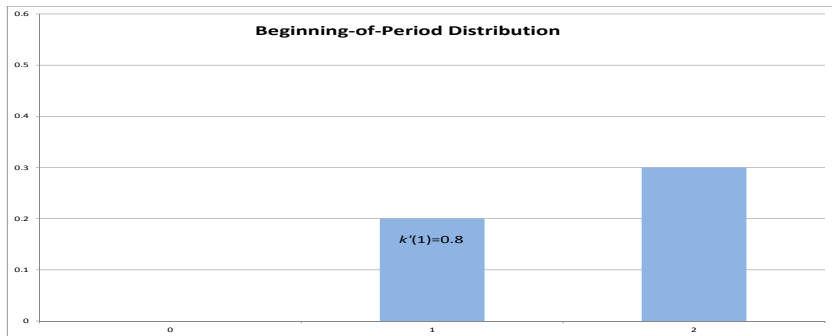
# Grid method

- focus on node  $j$  with mass  $f_t^{\varepsilon,j}$  and capital value  $\kappa_j$
- find  $i$  such that  $k'(\kappa_j, \varepsilon, \cdot)$  satisfies

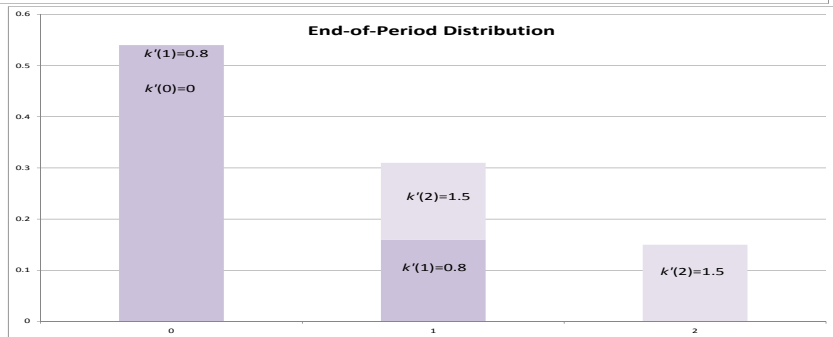
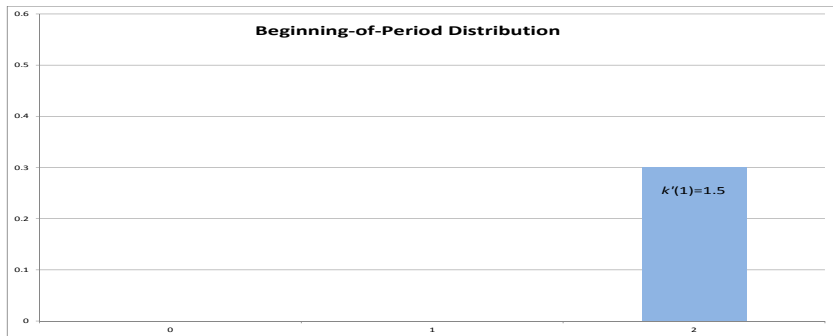
$$\kappa_{i-1} < k'(\kappa_j, \varepsilon, \cdot) \leq \kappa_i$$

- if  $k'(\kappa_j, \varepsilon, \cdot) > \kappa_I$ ,  $i = I$









## From beginning to end-of-period distribution

- Set all end-of-period fractions equal to zero:

$$p_t^{\varepsilon,i} = 0 \quad \forall i$$

- For each  $j$  allocate beginning-of-period  $f_t^{\varepsilon,j}$  to relevant end-of-period  $p_t^{\varepsilon,i}$ s:
- if  $k'(\kappa_j, \varepsilon, \cdot) < \kappa_I$  then

$$\omega_t^{i,j} = \frac{k'(\kappa_j, \varepsilon, \cdot) - \kappa_{i-1}}{\kappa_i - \kappa_{i-1}}$$

$$p_t^{\varepsilon,i-1} = p_t^{\varepsilon,i-1} + f_t^{\varepsilon,j} \left(1 - \omega_t^{i,j}\right)$$

$$p_t^{\varepsilon,i} = p_t^{\varepsilon,i} + f_t^{\varepsilon,j} \omega_t^{i,j}$$

- if  $k'(\kappa_j, \varepsilon, \cdot) \geq \kappa_I$  then  $p_t^{\varepsilon,I} = p_t^{\varepsilon,I} + f_t^{\varepsilon,j}$

## Next period's beginning-of-period distribution

- Use transition laws to go from end-of-period  $t$  distribution to beginning-of-period  $t + 1$  distribution
- $\phi_{00z_t z_{t+1}}$  : population transition probability of remaining unemployed for current and next period's value of  $z$
- $\phi_{11z_t z_{t+1}}$  : population transition probability of remaining employed for current and next period's value of  $z$

## Next period's beginning-of-period distribution

Use **population** transition probabilities; these are exogenous and should be exactly identical at each node.

$$\varepsilon_{t+1} = 0 : f_{t+1}^{0,i} = \phi_{00z_t z_{t+1}} p_t^{0,i} + (1 - \phi_{11z_t z_{t+1}}) p_t^{1,i}$$

$$\varepsilon_{t+1} = 1 : f_{t+1}^{1,i} = (1 - \phi_{00z_t z_{t+1}}) p_t^{0,i} + \phi_{11z_t z_{t+1}} p_t^{1,i}$$

In contrast to simulate method with  $I$  agents, the grid method imposes that transition probabilities are exactly identical across the whole distribution of capital holdings (and equal to correct theoretical values).

# Praise for grid method

- Simulating using a histogram with  $I$  nodes is roughly as expensive as simulating with  $I$  agents.
- But with say 1,000 nodes you get much more accurate answer than with say 100,000 agents.
- **Why?** Because you avoid sampling noise

**Note:** Before Young (2010), the literature used a more complicated grid method. Instead of going through each beginning-of-period node and allocating the associated mass,  $f^{\varepsilon,j}$ , to  $p^{\varepsilon,i}$ s, it considered each end-of-period node and then *using the inverse of the policy function* determined which elements of  $f^{\varepsilon,j}$  would get to that node. More complicated because you had to calculate the inverse.

# References

- Young, E. R., 2010, Solving the incomplete markets model with aggregate uncertainty using the Krusell-Smith algorithm and non-stochastic simulations, Journal of Economic Dynamics and Control
  - This paper introduces the grid method explained in these slides