

Solving Models with Heterogeneous Agents Xpa algorithm

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Individual agent

- Subject to employment shocks
 - $\varepsilon_{i,t} \in \{0, 1\}$ or $\varepsilon_{i,t} \in \{u, e\}$
- Incomplete markets
 - only way to save is through holding capital

Alternatives to inequality constraint

- Penalty function in the utility function

$$u(c_{i,t}, k_{i,t}) = \ln(c_{i,t}) - P(k_{i,t+1})$$

- Assumptions about $P(k_{i,t})$

standard inequality: $k_{i,t+1} \geq 0$ differentiable alternative

$$P(k_{i,t+1}) = \begin{cases} \infty & \text{if } k_{i,t+1} < 0 \\ 0 & \text{if } k_{i,t+1} \geq 0 \end{cases} \quad p(k_{i,t+1}) = \frac{\partial P(k_{i,t+1})}{\partial k_{i,t+1}} \leq 0$$

Alternatives to inequality constraint

Alternative to $P(\cdot)$ in utility function

- Individual interest rate depends on amount invested

$$r_{i,t} = r_t - P(k_{i,t+1}) \text{ with } \frac{\partial P(k_{i,t+1})}{\partial k_{i,t+1}} > 0$$

- Advantage: nicer economic story
- Disadvantage:
 - you have to take a stand on what happens with the resources (thrown away? intermediary?)
 - both Euler and budget constraint are affected

Laws of motion

- aggregate productivity, a_t , can take on two values
- employment status, ε_t , can take on two values
- probability of being (un)employed depends on a_t
- transition probabilities are such that unemployment rate only depends on current a_t
- $\implies u_t = u(a_t)$ with $u_b = u(1 - \Delta_a) > u_g - u(1 + \Delta_a)$.

Individual agent

$$\max_{\{c_{i,t}, k_{i,t+1}\}_{t=0}^{\infty}} E \sum_{t=0}^{\infty} \beta^t (\ln(c_{i,t}) - P(k_{i,t+1}))$$

s.t.

$$c_{i,t} + k_{i,t+1} = r_t k_{i,t} + (1 - \tau_t) w_t \bar{l} \varepsilon_{i,t} + \mu w_t (1 - \varepsilon_{i,t}) + (1 - \delta) k_{i,t}$$

for **given** processes of r_t and w_t , this is a relatively simple problem

Individual agent: Euler equation

$$\frac{1}{c_{i,t}} = p(k_{i,t+1}) + \mathbf{E}_t \left[\frac{\beta(r_{t+1} + 1 - \delta)}{c_{i,t+1}} \right]$$

cost of $k_{i,t+1}$ \uparrow = reduction in penalty + usual term

Firm problem

$$r_t = a_t \alpha \left(\frac{K_t}{\bar{l}(1 - u(a_t))} \right)^{\alpha - 1}$$

$$w_t = a_t (1 - \alpha) \left(\frac{K_t}{\bar{l}(1 - u(a_t))} \right)^{\alpha}$$

Government

$$\tau_t w_t \bar{l} (1 - u(a_t)) = \mu w_t u(a_t)$$

$$\tau_t = \frac{\mu u(a_t)}{\bar{l}(1 - u(a_t))}$$

Aggregate variables agents care about

- r_t and w_t
- They only depend on aggregate capital stock and a_t
- !!! This is not true in general for equilibrium prices
- Agents are interested in all information that forecasts K_t
- In principle that is the complete cross-sectional distribution of employment status and capital levels

Equilibrium - first part

- Individual policy functions solving the agent's max problem
- A wage and a rental rate given by equations above.

Equilibrium - second part

- A transition law for the cross-sectional distribution of capital, consistent with the individual policy function.
 - f_t = beginning-of-period cross-sectional distribution of capital and the employment status *after* the employment status has been realized.

$$f_{t+1} = Y(a_{t+1}, a_t, f_t)$$

- a_{t+1} does not affect the period t cross-sectional distribution of capital
- a_{t+1} does affect the *joint* cross-sectional distribution of capital and employment status

Overview

- ① What if individual policy rules are linear in *levels*?
- ② What if individual policy rules are polynomials in *levels*?
- ③ What if individual policy rules are *not* polynomials in the levels?
- ④ Topics:
 - What if there are non-differentiabilities
 - Economy with bonds
 - the price of a bond—unlike rental rate—is not a simple function of K_t

Linear policy rule for individual problem

- Suppose policy rules are given and equal to:

$$\text{if } \varepsilon = u : k'_u = \Psi_{u,0}(s) + \Psi_{u,1}(s)k$$

$$\text{if } \varepsilon = e : k'_e = \Psi_{e,0}(s) + \Psi_{e,1}(s)k$$

- linear in k , completely general in all other dimensions

Linear policy rule for individual problem

- s is the set of aggregate state variables and consist for sure of a, K_u, K_e
- Xpa determines "endogenously" whether other elements should be added to this list

Linear policy rule for individual problem

- Policy rules:

$$\text{if } \varepsilon = u : k'_u = \Psi_{u,0}(s) + \Psi_{u,1}(s)k$$

$$\text{if } \varepsilon = e : k'_e = \Psi_{e,0}(s) + \Psi_{e,1}(s)k$$

- Can we calculate K' from this?
- If the answer is yes, then we can calculate r' (given a')

Notation

End of this period

- \widehat{K}_u : *end-of-period* aggregate capital stock of unemployed
- \widehat{K}_e : *end-of-period* aggregate capital stock of employed

Beginning of the next period

- K'_u and K'_e : corresponding *beginning-of-period* equivalents

Transition laws

From end of this period to beginning of next period

- $K'_u \neq \widehat{K}_u$ and $K'_e \neq \widehat{K}_e$ because employment status changes
- Apply transition laws to go from $\widehat{K}_e, \widehat{K}_u$ to K'_e, K'_u

More notation

- $g_{\varepsilon\varepsilon'aa'}$: mass of agents with employment status ε now and ε' next period for given values of a and a'

$$g_{uuaa'} + g_{euaa'} + g_{eeaa'} + g_{ueaa'} = 1$$

From end to beginning-of-period moments

$$K'_u = \frac{g_{uuaa'}\widehat{K}_u + g_{euaa'}\widehat{K}_e}{g_{uuaa'} + g_{euaa'}}$$

$$K'_e = \frac{g_{ueaa'}\widehat{K}_u + g_{eeaa'}\widehat{K}_e}{g_{ueaa'} + g_{eeaa'}}$$

$$K' = u(a')K'_u + (1 - u(a'))K'_e$$

Given the formulas for \widehat{K}_u and \widehat{K}_e we can calculate these.

Simpler case (to learn method)

Suppose individual policy function are as follows:

$$\text{if } \varepsilon = u : k'_u = \Psi_{u,0} + \Psi_{u,k}k + \Psi_{u,a}a + \Psi_{u,K_u}K_u + \Psi_{u,K_e}K_e$$

$$\text{if } \varepsilon = e : k'_e = \Psi_{e,0} + \Psi_{e,k}k + \Psi_{e,a}a + \Psi_{e,K_u}K_u + \Psi_{e,K_e}K_e$$

This immediately gives

$$\begin{aligned} \widehat{K}_u &= \widehat{M}_u(1) = \int k'_u(\cdot) dF_u(k) \\ &= (\Psi_{u,0} + \Psi_{u,a}) + (\Psi_{u,k} + \Psi_{u,K_u}) K_u + \Psi_{u,K_e} K_e, \\ \widehat{K}_e &= \widehat{M}_e(1) = \int k'_e(\cdot) dF_e(k) \\ &= (\Psi_{e,0} + \Psi_{e,a}) + (\Psi_{e,k} + \Psi_{e,K_e}) K_e + \Psi_{e,K_u} K_u, \end{aligned}$$

- Applying transition laws gives K'_u , K'_e , and K'

What about the state variables?

- Above, we assumed that s consists of $k, \varepsilon, a, K_u, K_e$ and possibly other characteristics of the distribution. But with linear policy rules, no other moments are needed.
- Intuition?

Simple way to solve for coefficients

- 1 Assume linear law of motion for K'_u and K'_e
- 2 Find linear approximation for k'_u and k'_e
- 3 Get new linear law of motion for K'_u and K'_e by *explicit aggregation* (and transition laws)
- 4 Iterate until convergence

More general linear policy rules

Policy rules:

$$\text{if } \varepsilon = u : k'_u = \Psi_{u,0}(s) + \Psi_{u,1}(s)k$$

$$\text{if } \varepsilon = e : k'_e = \Psi_{e,0}(s) + \Psi_{e,1}(s)k$$

This immediately gives

$$\widehat{K}_u = \widehat{M}_u(1) = \int k'_u(k, s_{-k}) dF_u(k) = \Psi_{u,0}(s) + \Psi_{u,1}(s)M_u(1),$$

$$\widehat{K}_e = \widehat{M}_e(1) = \int k'_e(k, s_{-k}) dF_e(k) = \Psi_{e,0}(s) + \Psi_{e,1}(s)M_e(1),$$

- This law of motion could be non-linear in K_u and K_e !!!!
- Applying transition laws gives K'_u , K'_e , and K'

Progress so far

- Given linear individual policy rule:
 - we can get law of motion for aggregate variables
 - determine what the set of aggregate state variables are
- Haven't said anything yet on how to find individual policy rules
 - We'll answer that for somewhat more general case with quadratic individual policy rules

Second-order policy rule

Policy rules:

$$k'_u = \Psi_{u,0}(s) + \sum_{i=1}^2 \Psi_{u,i}(s)k^i \text{ and}$$

$$k'_e = \Psi_{e,0}(s) + \sum_{i=1}^2 \Psi_{e,i}(s)k^i,$$

- quadratic in k , completely general in all other dimensions

Second-order policy rule

- s is the set of state variables and consist for sure of a, K_u, K_e
- Xpa determines "endogenously" whether other elements should be added to this list

Aggregate second-order policy

- Aggregation gives

$$\begin{aligned}\widehat{K}_u &= \widehat{M}_u(1) = \Psi_{u,0}(s) + \sum_{i=1}^2 \Psi_{u,i}(s)M_u(i), \\ \widehat{K}_e &= \widehat{M}_e(1) = \Psi_{e,0}(s) + \sum_{i=1}^2 \Psi_{e,i}(s)M_e(i),\end{aligned}$$

- What are the state variables?
 - Not only $M_u(1)$ and $M_e(1)$, but also $M_u(2)$ and $M_e(2)$

Aggregate second-order policy

- \implies we need laws of motion for $\widehat{M}_u(2)$ and $\widehat{M}_e(2)$;
 - these together with transition laws would give laws of motion for $M'_u(2)$ and $M'_e(2)$

Laws of motion for second-order terms

- Moments are

$$\widehat{M}_\varepsilon(2) = \int (k'_\varepsilon(k, s_{-k}))^2 dF_\varepsilon(k)$$

- Do I have a law of motion for $(k'_\varepsilon)^2$? Yes, of course, namely

$$\begin{aligned} (k'_\varepsilon)^2 = & \Psi_{\varepsilon,0}^2(s) + 2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,1}(s)k + \\ & (2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,2} + (\Psi_{\varepsilon,1}(s))^2)k^2 \\ & + 2\Psi_{\varepsilon,1}(s)\Psi_{\varepsilon,2}(s)k^3 + (\Psi_{\varepsilon,2}(s))^2k^4. \end{aligned}$$

What is the problem?

Laws of motion for second-order terms

- Aggregating

$$(k'_\varepsilon)^2 = \Psi_{\varepsilon,0}^2(s) + 2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,1}(s)k + (2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,2} + (\Psi_{\varepsilon,1}(s))^2)k^2 + 2\Psi_{\varepsilon,1}(s)\Psi_{\varepsilon,2}(s)k^3 + (\Psi_{\varepsilon,2}(s))^2k^4.$$

gives

$$\widehat{M}_\varepsilon(2) = (\Psi_{\varepsilon,0}(s))^2 + 2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,1}(s)\widehat{M}_\varepsilon(1) + (2\Psi_{\varepsilon,0}(s)\Psi_{\varepsilon,2} + (\Psi_{\varepsilon,1}(s))^2)\widehat{M}_\varepsilon(2) + 2\Psi_{\varepsilon,1}(s)\Psi_{\varepsilon,2}(s)\widehat{M}_\varepsilon(3) + (\Psi_{\varepsilon,2}(s))^2\widehat{M}_\varepsilon(4).$$

Avoiding infinite-regress problem

- Define

$$y'_\varepsilon = (k'_\varepsilon)^2$$

Come up with a *separate* 2nd-order approximation for y'_ε :

$$y'_\varepsilon = (k'_\varepsilon)^2 \approx \Psi_{\varepsilon,(k')^2,0}(s) + \Psi_{\varepsilon,(k')^2,1}(s)k + \Psi_{\varepsilon,(k')^2,2}(s)k^2$$

- $\Psi_{\varepsilon,j}$ coeffs have no direct relationship to $\Psi_{\varepsilon,(k')^2,j}$ coeffs

Aggregation in second-order case

$$k'_\varepsilon = \Psi_{\varepsilon,0}(s) + \sum_{i=1}^2 \Psi_{\varepsilon,i}(s)k^i \text{ gives}$$

$$\widehat{K}_\varepsilon = \widehat{M}_\varepsilon(1) = \Psi_{\varepsilon,0}(s) + \sum_{i=1}^2 \Psi_{\varepsilon,i}(s)M_\varepsilon(i),$$

and

$$(k'_\varepsilon)^2 = \Psi_{\varepsilon,(k')^2,0}(s) + \sum_{i=1}^2 \Psi_{\varepsilon,(k')^2,i}(s)k^i \text{ gives}$$

$$\widehat{M}_\varepsilon(2) = \Psi_{\varepsilon,(k')^2,0}(s) + \sum_{i=1}^2 \Psi_{\varepsilon,(k')^2,i}(s)M_\varepsilon(i),$$

Basic formulation

- Policy rules:

$$k'_u = \Psi_{u,0}(s) + \sum_{i=1}^I \Psi_{u,i}(s)k^i \text{ and } k'_e = \Psi_{e,0}(s) + \sum_{i=1}^I \Psi_{e,i}(s)k^i,$$

- Aggregation gives

$$\begin{aligned}\widehat{K}_u &= \widehat{M}_u(1) = \Psi_{u,0}(s) + \sum_{i=1}^I \Psi_{u,i}(s)M_u(i), \\ \widehat{K}_e &= \widehat{M}_e(1) = \Psi_{e,0}(s) + \sum_{i=1}^I \Psi_{e,i}(s)M_e(i),\end{aligned}$$

- Get additional *separate* policy rules for each polynomial term in policy function

Full program

How to find the Ψ coefficients?

- Iterative perturbation procedures
- general projection method

Iterative perturbation solutions

- ➊ Guess a law of motion for aggregate law of motion
- ➋ Conditional on this solve for individual law of motion
- ➌ Explicitly aggregate and update aggregate law of motions

Perturbation solution with discrete support

- Write law of motion for a_t and ε_t as

$$\begin{aligned}a_t &= \bar{a} + \rho_a a_{t-1} + e_{a,t} && \text{with } E[e_{a,t}] = \sigma_a^2 \\ \varepsilon_{i,t} &= \bar{\varepsilon} + \rho_\varepsilon \varepsilon_{t-1} + e_{\varepsilon,t} && \text{with } E[e_{\varepsilon,t}] = \sigma_\varepsilon^2\end{aligned}$$

- $\rho_a, \rho_\varepsilon, \sigma_a^2, \sigma_\varepsilon^2$ are such that autocovariances and variances correspond to original process

General projection procedure

- Suppose a second-order solution is used:

$$s_{i,t} = [\varepsilon_{i,t}, k_{i,t}, a_t, M_{u,t}(1), M_{e,t}(1), M_{u,t}(2), M_{e,t}(2)]$$

- Basic idea:
 - set up a grid
 - define error term at each grid point
 - define loss function
 - use minimization routine to find Ψ coefficients

General projection procedure

- Solve for $\psi_{k'}(s; \Psi)$ by making model equations "fit" on grid
- Notation: we have included $\varepsilon_{i,t}$ in s and $\psi_{k'}(s; \Psi)$ describes behavior of both employed and unemployed agent

Individual policy rules & projection methods

- $\{s_\kappa\}_{\kappa=1}^\chi$ the set of state variables with χ nodes
- $M_\kappa = \{a_\kappa, M_{u,\kappa}(1), M_{e,\kappa}(1), M_{u,\kappa}(2), M_{e,\kappa}(2)\}$
- κ indicates a grid point not a period

First-order condition

$$\begin{aligned}
 & \left(\begin{array}{c} (r(a_\kappa, K_\kappa) + 1 - \delta)k_\kappa \\ + (1 - \tau(a_\kappa))w(a_\kappa, K_\kappa)\bar{l}\varepsilon_\kappa + \mu w(a_\kappa, K_\kappa)(1 - \varepsilon_\kappa) \\ - \psi_{k'}(s_\kappa; \Psi) \end{array} \right)^{-\nu} \\
 = & p(\psi_{k'}(s_\kappa; \Psi)) + \mathbf{E} \left[\begin{array}{c} \beta(r(a', K') + (1 - \delta)) \times \\ \left(\begin{array}{c} (r(a', K') + 1 - \delta)\psi_{k'}(s_\kappa; \Psi) \\ + (1 - \tau(a'))w(z', K')\bar{l}\varepsilon' \\ + \mu w(a', K')(1 - \varepsilon') \\ - \psi_{k'}(s'; \Psi) \end{array} \right)^{-\nu} \end{array} \right]
 \end{aligned}$$

Individual problem errors (usual part):

$$\begin{aligned}
 u_{\kappa} = & - \left(\begin{array}{c} (r(a_{\kappa}, K_{\kappa}) + 1 - \delta)k_{\kappa} \\ + (1 - \tau(a_{\kappa}))w(a_{\kappa}, K_{\kappa})\bar{l}\varepsilon_{\kappa} + \mu w(a_{\kappa}, K_{\kappa})(1 - \varepsilon_{\kappa}) \\ - \psi_{k'}(s_{\kappa}; \Psi) \end{array} \right)^{-\nu} \\
 & + p(\psi_{k'}(s_{\kappa}; \Psi)) + \sum_{a'} \sum_{\varepsilon'} \left[\begin{array}{c} \beta(r(a', K') + (1 - \delta)) \times \\ \left(\begin{array}{c} (r(a', K') + 1 - \delta)\psi_{k'}(s_{\kappa}; \Psi) \\ + (1 - \tau(a'))w(z', K')\bar{l}\varepsilon' \\ + \mu w(a', K')(1 - \varepsilon') \\ - \psi_{k'}(s'; \Psi) \end{array} \right)^{-\nu} \\ \pi(a', \varepsilon' | a_{\kappa}, \varepsilon_{\kappa}) \end{array} \right] \times
 \end{aligned}$$

Individual problem errors (new part):

- Error for $y' = (k')^2$:

$$u_{\kappa}^* = -\psi_{(k')^2}(s_{\kappa}; \Psi) + (\psi_{k'}(s_{\kappa}; \Psi))^2$$

Errors only depend only on known things and Ψ

$$\begin{aligned}
 r(a_\kappa, K_\kappa) &= \alpha a_\kappa (K_\kappa / (\bar{l}(1 - u(a_\kappa))))^{\alpha-1} \\
 w(a_\kappa, K_\kappa) &= (1 - \alpha) a_\kappa^\alpha (K_\kappa / (\bar{l}(1 - u(a_\kappa))))^\alpha \\
 r(a', K') &= \alpha a' (K' / (\bar{l}(1 - u(a'))))^{\alpha-1} \\
 w(a', K') &= (1 - \alpha) a' (K' / L(z'))^\alpha \\
 \tau(a) &= \frac{\mu u(a)}{\bar{l}(1 - u(a))} \quad \text{and} \quad \tau(a') = \frac{\mu u(a')}{\bar{l}(1 - u(a'))} \\
 s' &= \{k', \varepsilon', a', M'_u(1), M'_e(1), M'_u(2), M'_e(2)\} \\
 &= \left\{ \begin{array}{l} \psi_{k'}(s_\kappa; \Psi), \varepsilon', a', \\ M'_u(1), M'_e(1), M'_u(2), M'_e(2) \end{array} \right\}
 \end{aligned}$$

Errors only depend only on known things and Ψ

How to get $K', M'_u(1), M'_e(1), M'_u(2), M'_e(2)$ in terms of current state variables?

- 1 $K' = u(a')M'_u(1) + (1 - u(a'))M'_e(1)$
- 2 Express $M'_u(j)$ and $M'_e(j)$ in terms of a', a , and $\widehat{M}_u(j)$ and $\widehat{M}_e(j)$
- 3 Explicitly aggregate to get expressions for $\widehat{M}_u(j)$ and $\widehat{M}_e(j)$

Topics

- non-polynomial basis functions
 - procedure
 - bias correction
- Bond economy

If individual policy rules are not polynomials in levels

- Suppose

$$\begin{aligned} b(k'_u) &= \Psi_{u,0}(s) + \Psi_{u,1}(s) b(k_u) \text{ if } \varepsilon = u \text{ and} \\ b(k'_e) &= \Psi_{e,0}(s) + \Psi_{e,1}(s) b(k_e) \text{ if } \varepsilon = e, \end{aligned}$$

where $b(k)$ is some function, e.g. $\ln(k)$

If individual policy rules are not polynomials in levels

- Since you need to know K' you need a law of motion for k' .
- Thus, *in addition* to policy rule for $b(k')$ also obtain policy rules for k' .
- How to get policy rule for k' ?
 - Use linear approximation to $b(k')$, or
 - Solve two individual policy rules
 - one to solve for aggregate law of motion and
 - one to describe individual behavior

linear approximation for linear spline

- Taking linear approximation when using linear spline is trivial:
 - Get aggregate law of motion for \hat{K}_u simply by evaluating policy rule of unemployed at $k_u = K_u$
 - Get aggregate law of motion for \hat{K}_e simply by evaluating policy rule of unemployed at $k_e = K_u$

Bias correction

- If policy rules are *not* polynomials \implies inconsistency between
 - individual policy rules and aggregate law of motion
- Estimate of the mean bias can be found from model without aggregate uncertainty

Bias correction

- \tilde{K}_ε and \hat{K}_ε very accurate solution for beginning and end-of-period capital stocks in model without aggregate uncertainty
- Xpa aggregate law of motion without bias correction:

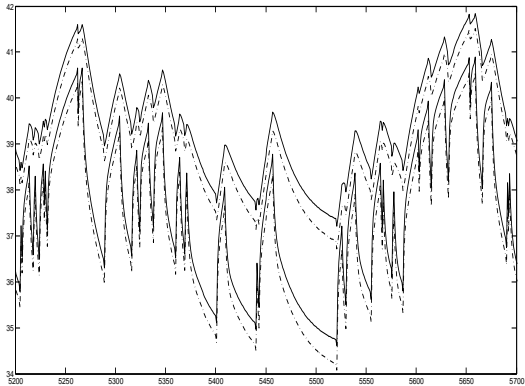
$$\hat{K}_\varepsilon = \Psi_{\varepsilon,0}(M) + \Psi_{\varepsilon,1}(M)K_\varepsilon.$$

- Bias correction ζ_ε :

$$\zeta_\varepsilon = \hat{K}_\varepsilon - \Psi_{\varepsilon,0}(\tilde{M}) - \Psi_{\varepsilon,1}(\tilde{M})\tilde{K}_\varepsilon.$$

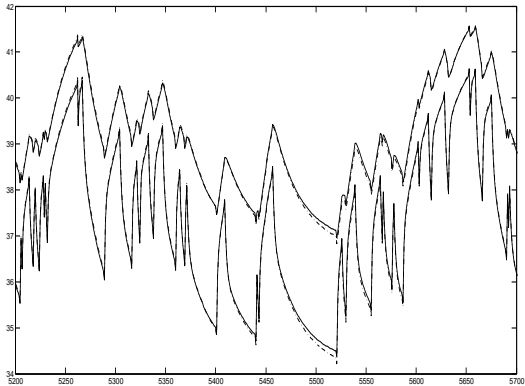
No bias correction

Figure: Simulated values of K_u and K_e without bias correction



With bias correction

Figure: Simulated values of K_u and K_e with bias correction



Would the R2 pick this up?

No, the $R^2 > 0.99997$ in both cases

Imposing equilibrium

- KS economy: at $r_t = a(K_t/L_t)^{\alpha-1}$ equilibrium is ensured for any law of motion for K_t
- For other types of assets this is not that easy
- But *exactly* imposing equilibrium is important
 - errors are unlikely to be exactly average on zero
 - \implies errors accumulate and at some point simulation is meaningless

Bond economy

- endowment economy
- borrowing and lending in one and two-period riskless bonds
- penalty functions instead of inequality constraints

Bond economy - equations

$$\frac{q_t^1}{c_{i,t}} = \beta \mathbb{E}_t \frac{1}{c_{i,t+1}} + p(b_{i,t+1}^1)$$

$$\frac{q_t^2}{c_{i,t}} = \beta \mathbb{E}_t \frac{q_{t+1}^1}{c_{i,t+1}} + p(b_{i,t+1}^2)$$

$$c_{i,t} + q_t^1 b_{i,t+1}^1 + q_t^2 b_{i,t+1}^2 = y_{i,t} + b_t^1 + q_t^1 b_t^2,$$

Bond economy - (too) simple approach

- guess law of motion for q_t^1 and q_t^2
- solve individual problem
- problem with simulation: equilibrium is not imposed
- problem with Xpa: equilibrium not imposed off grid points and during simulation

Bond economy - imposing equilibrium

- Instead of solving for $b^j(s_{i,t}, m_t)$ solve for $b^j(q_t^j, s_{i,t}, m_t)$
- Solve for q_t^j from

$$\int b^j(q_t^j, s_{i,t}, m_t) di = 0$$

- How do I get these $b^j(q_t^j, s_{i,t}, m_t)$?

Bond economy - imposing equilibrium

- Use model equations:

$$\frac{q_t^1}{c_{i,t}} = \beta E_t \frac{1}{c_{i,t+1}} + p(b_{i,t+1}^1)$$

$$\frac{q_t^2}{c_{i,t}} = \beta E_t \frac{q_{t+1}^1}{c_{i,t+1}} + p(b_{i,t+1}^2)$$

$$c_{i,t} + q_t^1 b_{i,t+1}^1 + q_t^2 b_{i,t+1}^2 = y_{i,t} + b_t^1 + q_t^1 b_t^2$$

- and add the following two equations that define d_{t+1}^1 and d_{t+1}^2

$$b_{i,t+1}^1 + q_t^1 = d_{t+1}^1 \text{ and}$$

$$b_{i,t+1}^2 + q_t^2 = d_{t+1}^2$$

Bond economy - imposing equilibrium

- This gives the following solutions
 - $q(m_t)$
 - $b^j(s_{i,t}, m_t)$
 - $d^j(s_{i,t}, m_t)$
- Take **none** of these literally. Except use
 - $b_{t+1}^j = b^j(q_t^j, s_{i,t}, m_t) = d^j(s_{i,t}, m_t) - q_t^j$

Bond economy - imposing equilibrium

- Imposing equilibrium

$$\int b^j(q_t, s_{i,t}, m_t) di = 0$$

gives

$$q_t^j = \int d^j(s_{i,t}, m_t) di$$

Bond economy - imposing equilibrium

- Alternative definitions for $d^j(\cdot)$ are possible
- !!!! But one does need that

$$\frac{\partial b^j(q_t, s_{i,t}, m_t)}{\partial q_t} < 0$$

that is, you have a demand equation.

Something practical

- The following slides work out a particular case
- The individual policy rule is of a simpler form than used in discussing the general case above
- Given this simpler form the slides go through the steps of XPA

Something practical

- Suppose that

$$a_t = \bar{a} + \rho_a a_{t-1} + e_{a,t} \quad \text{with } E[e_{a,t}] = \sigma_a^2$$

$$\varepsilon_{i,t} = \bar{\varepsilon}(1 - \rho_\varepsilon) + \rho_\varepsilon \varepsilon_{i,t-1} + e_{\varepsilon,t} \quad \text{with } E[e_{\varepsilon,t}] = \sigma_\varepsilon^2$$

- Note that

$$E[\varepsilon_{i,t}] = \bar{\varepsilon}$$

Something practical

- Individual policy rule is higher-order

$$k' = \Psi_0^* + \Psi_k k + \Psi_\varepsilon \varepsilon + \Psi_{k\varepsilon} k\varepsilon + \Psi_{u,a} a + \Psi_{u,K} K$$

- This can be written as

$$k' = \Psi_0 + \Psi_k k + \Psi_\varepsilon \varepsilon + \Psi_{k\varepsilon} \rho k \varepsilon_{-1} + \Psi_{k\varepsilon} e_\varepsilon + \Psi_{u,a} a + \Psi_{u,K} K$$

Something practical

Aggregation of

$$k' = \Psi_0 + \Psi_k k + \Psi_\varepsilon \varepsilon + \Psi_{k\varepsilon} \rho k \varepsilon_{-1} + \Psi_{k\varepsilon} e_\varepsilon + \Psi_a a + \Psi_K K$$

gives

$$K' = (\Psi_0 + \Psi_\varepsilon \bar{\varepsilon}) + (\Psi_k + \Psi_{u,K}) K + \Psi_{k\varepsilon} \rho M_{k\varepsilon} + \Psi_{u,a} a$$

where

$$M_{k\varepsilon} = \int k \varepsilon_{-1} dF(k, \varepsilon_{-1})$$

Something practical

- Aggregate state variables: $a, K, M_{k\varepsilon-1}$
- Why not K_u and K_e separately?
- Why don't we have to use transition laws?
- How we get law of motion for $M'_{k\varepsilon}$?
- **Answer:** define

$$y' = k'\varepsilon \approx \tilde{\Psi}_0 + \tilde{\Psi}_k k + \tilde{\Psi}_\varepsilon \varepsilon + \tilde{\Psi}_{k\varepsilon} \rho k \varepsilon_{-1} + \tilde{\Psi}_{k\varepsilon} e_\varepsilon + \tilde{\Psi}_a a + \tilde{\Psi}_K K$$

Something practical

Aggregation of

$$y' = \tilde{\Psi}_0 + \tilde{\Psi}_k k + \tilde{\Psi}_\varepsilon \varepsilon + \tilde{\Psi}_{k\varepsilon} \rho k \varepsilon_{-1} + \tilde{\Psi}_{k\varepsilon} e_\varepsilon + \tilde{\Psi}_a a + \tilde{\Psi}_K K$$

gives

$$M'_{k\varepsilon} = \left(\tilde{\Psi}_0 + \tilde{\Psi}_\varepsilon \bar{\varepsilon} \right) + \left(\tilde{\Psi}_k + \tilde{\Psi}_{u,K} \right) K + \tilde{\Psi}_{k\varepsilon} \rho M_{k\varepsilon} + \tilde{\Psi}_{u,a} a$$

References

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 - survey article available online, also describes link between Xpa and the perturbation method of Preston & Roca
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