

Buera & Shin (2013) – Financial Friction and the Persistence of History: A Quantitative Exploration

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Introduction

Brief description of (incomplete) replication of Buera and Shin (2013). I denote the exogenous shock z , which they call e , and I also call γ what they call σ . I use μ as the agent distribution (while they had it as the distribution of just the exogenous process, they used G for the agent distribution).

The model is about solving a transition path in an infinite horizon model. Here I just write the steady-state problem.

Value Function Formulation

Buera and Shin (2013) present the household problem as a sequence problem. We rewrite it as a value function problem:

$$V(a, z) = \max_{c, a'} \left\{ \frac{c^{1-\gamma}}{1-\gamma} + \beta \mathbb{E}[V(a', z')|z] \right\} \quad (1)$$

subject to

$$c + a' \leq \max\{w, \pi(a, z|r, w)\} + (1+r)a$$

and

$$a' \geq 0$$

where

$$\pi(a, z|r, w) = \max_{l, k} \{f(z, k, l) - wl - (r + \delta)k\} \text{ subject to } k \leq \lambda a$$

and

$$f(z, k, l) = z(k^\alpha l^{1-\alpha})^{1-\nu}$$

This reduces to a one-state dynamic programming problem since the profit part can be solved analytically. Solving the VFI problem delivers the policy functions $a'(a, z)$, $e = e(a, z)$, $e \in \{0, 1\}$, where optimal occupation choice is

$$e(a, z) = \begin{cases} 1 & \text{if } \pi(a, z|r, w) > w \\ 0 & \text{otherwise.} \end{cases}$$

Analytical Solution

$$k_1 = \left[\left(\frac{1}{r + \delta} \alpha (1 - \nu) z \right)^{1 - (1 - \alpha)(1 - \nu)} \left(\frac{1}{w} (1 - \alpha)(1 - \nu) z \right)^{(1 - \alpha)(1 - \nu)} \right]^{1/\nu}$$

$$k^* = \min\{k_1, \lambda a\}$$

$$l^* = \left[\frac{1}{w} (1 - \alpha)(1 - \nu) z (k^*)^{\alpha(1 - \nu)} \right]^{\frac{1}{1 - (1 - \alpha)(1 - \nu)}}$$

$$\pi(a, z | r, w) = z (k^*)^\alpha (l^*)^{1 - \alpha} - w l^* - (r + \delta) k^*$$

General Equilibrium Conditions

There are two GE prices: interest rate r and wage rate w .

Labor market clearing:

$$\int l^*(a, z) \cdot \mathbb{I}_{\{e(a, z)=1\}} d\mu(a, z) = \int \mathbb{I}_{\{e(a, z)=0\}} d\mu(a, z)$$

Capital market clearing:

$$\int k^*(a, z) \cdot \mathbb{I}_{\{e(a, z)=1\}} d\mu(a, z) = \int a d\mu(a, z)$$

Implied by Walras law:

$$\int c(a, z) d\mu(a, z) + \int a d\mu(a, z) = \int [\mathbb{I}_{\{e(a, z)=0\}} \cdot w + \mathbb{I}_{\{e(a, z)=1\}} \cdot \pi(a, z)] d\mu(a, z) + (1 + r) \int a d\mu(a, z)$$

which simplifies to

$$\begin{aligned} C &= w \int \mathbb{I}_{\{e(a, z)=0\}} d\mu(a, z) + \int \mathbb{I}_{\{e(a, z)=1\}} \pi(a, z) d\mu(a, z) + rK \\ &= wL + Y - wL - (r + \delta)K + rK \\ &= Y - \delta K \end{aligned}$$

Calibration

External parameters. $\gamma = 1.5$, $\delta = 0.06$, $\alpha = 0.33$, $\lambda = \infty$

Internal calibration. Set v (span of control), η, ψ (dispersion and persistence of entrepreneurial ability) and β (discount rate) to match:

- Top 10% of employment (v, η)
- Top 5% of earnings (v, η)
- Entrepreneurs exit rate (ψ)

- Real interest rate (β)

Stochastic process. Entrepreneurial ability z follows a *truncated* Pareto distribution with density function

$$p(z) = \begin{cases} \eta z^{-(\eta+1)} & \text{if } z \geq 1 \\ 0 & \text{otherwise} \end{cases}$$

Each period, an individual keeps his old ability z with probability ψ . With complementary probability $1 - \psi$ he draws a new ability z' from $p(z')$ given above. Clearly, the parameter ψ governs the persistence of the stochastic process for ability, whereas η governs the dispersion. Given these assumption for the stochastic process of entrepreneurial ability, the continuation value in (1) can be written more explicitly as follows:

$$E[V(a', z')|z] = \psi V(a', z) + (1 - \psi) \int V(a', z') p(z') dz'.$$

In order to re-use the general routines for value function iteration and distribution iteration, it is convenient to *recast the stochastic process described above as a Markov chain*. First, we assume we have already discretized the Pareto distribution as $z \in \{z_1, z_2, \dots, z_n\}$ and $p(z) = [p_1, \dots, p_n]$ (a $n \times 1$ *column* vector such that $\sum_i p_i = 1$). Then, we observe that if the current ability shock is z_i , then next-period ability shock will be z_i with prob $\psi + (1 - \psi)p_i$ and z_j with prob $(1 - \psi)p_j$, for $j \neq i$. It is easy to check that $\sum \pi(z, z') = \psi + (1 - \psi)p_i + \sum_{j \neq i} (1 - \psi)p_j = 1$, i.e. each row i of the Markov chain sums to one, for all $i = 1, \dots, n$. Therefore the $n \times n$ Markov chain for the shock (z, z') can be written as

$$\Pi_{n \times n} = \psi \mathbf{I}_n + (1 - \psi) \mathbf{1}_n p^T$$

or, more in detail, as

$$\Pi_{n \times n} = \psi \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 1 \end{bmatrix} + (1 - \psi) \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} [p_1 \quad p_2 \quad \dots \quad p_n]$$

where \mathbf{I}_n is the $n \times n$ identity matrix and $\mathbf{1}_n$ is an $n \times 1$ *column* vector of 1s.

Note: For the replication I used the files `support.dat` and `dist.dat` kindly provided by the authors. If I replicate the discretization as described in their paper, I get values for z and $p(z)$ which are quite different from the ones used in the paper.

Appendix: Profit Maximization Derivations

$$\pi(a, z|r, w) = \max_{l, k} \{z(k^\alpha l^{1-\alpha})^{1-\nu} - wl - (r + \delta)k\}, \quad \text{s.t. } k \leq \lambda a$$

First Order Conditions (Unconstrained):

$$\frac{\partial \pi}{\partial l} = (1 - \alpha)(1 - \nu)zk^{\alpha(1-\nu)}l^{(1-\alpha)(1-\nu)-1} - w = 0$$

$$\frac{\partial \pi}{\partial k} = \alpha(1 - \nu)zk^{\alpha(1-\nu)-1}l^{(1-\alpha)(1-\nu)} - (r + \delta) = 0$$

Solving gives:

$$l^* = \left[\frac{1}{w}(1 - \alpha)(1 - \nu)zk^{\alpha(1-\nu)} \right]^{\frac{1}{1-(1-\alpha)(1-\nu)}}$$

$$k_1 = \left[\left(\frac{1}{r + \delta} \alpha(1 - \nu)z \right)^{1-(1-\alpha)(1-\nu)} \left(\frac{1}{w}(1 - \alpha)(1 - \nu)z \right)^{(1-\alpha)(1-\nu)} \right]^{1/\nu}$$

$$k^* = \min\{k_1, \lambda a\}$$