

Computational Methods for Buera & Shin (2013): Financial Frictions and the Persistence of History

Replication Notes

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Abstract

This document provides detailed computational notes for replicating the stationary equilibrium and Figure 2 of [1]. We describe the model setup, the individual optimization problem, value function iteration, computation of the stationary distribution, and the general equilibrium algorithm. Implementation details for Python with Numba acceleration are provided.

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1 Model Environment

1.1 Preferences

There is a continuum of infinitely-lived agents with measure one. Agents have CRRA preferences over consumption:

$$U = \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t u(c_t), \quad u(c) = \frac{c^{1-\sigma} - 1}{1 - \sigma} \quad (1)$$

where $\beta \in (0, 1)$ is the discount factor and $\sigma > 0$ is the coefficient of relative risk aversion. For $\sigma = 1$, we have $u(c) = \log(c)$.

1.2 Technology

Each agent is endowed with entrepreneurial ability $e \geq 1$. An entrepreneur with ability e operates a production technology:

$$y = f(e, k, \ell) = e \cdot (k^\alpha \ell^{1-\alpha})^{1-\nu} \quad (2)$$

where:

- k is capital input
- ℓ is labor input
- $\alpha \in (0, 1)$ is the capital share in variable factors
- $\nu \in (0, 1)$ is the entrepreneur's share (span-of-control parameter)

The span-of-control parameter ν implies decreasing returns to scale in (k, ℓ) , which bounds firm size even with perfect credit markets.

1.3 Entrepreneurial Ability Process

Ability e follows a Markov process. With probability ψ , ability remains unchanged. With probability $1 - \psi$, a new ability is drawn from the stationary Pareto distribution:

$$\Pr(e'|e) = \psi \cdot \mathbf{1}_{e'=e} + (1 - \psi) \cdot g(e') \quad (3)$$

where the stationary distribution has Pareto density:

$$g(e) = \eta \cdot e^{-(\eta+1)}, \quad e \geq 1 \quad (4)$$

with CDF $G(e) = 1 - e^{-\eta}$ and tail parameter $\eta > 0$.

1.4 Financial Frictions

Agents face a collateral constraint on capital:

$$k \leq \lambda \cdot a \quad (5)$$

where a is the agent's asset holdings and $\lambda \geq 1$ is the financial friction parameter:

- $\lambda = 1$: Financial autarky (no external borrowing)
- $\lambda = \infty$: Perfect credit markets
- $\lambda \in (1, \infty)$: Intermediate financial development

1.5 Markets

Markets are competitive:

- Labor market: wage w
- Capital rental market: rental rate $r + \delta$ where r is the interest rate and δ is depreciation
- Asset market: agents can save at rate r (no borrowing beyond collateral)

2 Individual Optimization Problem

2.1 Occupational Choice

Each period, an agent with state (a, e) chooses to be either:

1. **Worker**: Supplies 1 unit of labor, earns wage w
2. **Entrepreneur**: Operates a firm, earns profit $\pi(a, e; w, r, \lambda)$

The agent chooses the occupation that maximizes income:

$$\text{Occupation} = \begin{cases} \text{Entrepreneur} & \text{if } \pi(a, e; w, r, \lambda) > w \\ \text{Worker} & \text{otherwise} \end{cases} \quad (6)$$

2.2 Entrepreneur's Static Problem

An entrepreneur with assets a and ability e solves:

$$\pi(a, e; w, r, \lambda) = \max_{k, \ell \geq 0} \{e \cdot (k^\alpha \ell^{1-\alpha})^{1-\nu} - w\ell - (r + \delta)k\} \quad (7)$$

subject to:

$$k \leq \lambda \cdot a \quad (8)$$

2.2.1 First-Order Conditions

The FOCs for an interior solution are:

$$\frac{\partial f}{\partial k} = (r + \delta) \Rightarrow \alpha(1 - \nu) \cdot e \cdot (k^\alpha \ell^{1-\alpha})^{-\nu} \cdot k^{\alpha-1} \ell^{1-\alpha} = r + \delta \quad (9)$$

$$\frac{\partial f}{\partial \ell} = w \Rightarrow (1 - \alpha)(1 - \nu) \cdot e \cdot (k^\alpha \ell^{1-\alpha})^{-\nu} \cdot k^\alpha \ell^{-\alpha} = w \quad (10)$$

2.2.2 Unconstrained Optimal Capital

Combining the FOCs, the unconstrained optimal capital k^* satisfies:

$$k^* = \left[\frac{\alpha(1-\nu)}{r+\delta} \right]^{\frac{1-(1-\alpha)(1-\nu)}{\nu}} \cdot \left[\frac{(1-\alpha)(1-\nu)}{w} \right]^{\frac{(1-\alpha)(1-\nu)}{\nu}} \cdot e^{1/\nu} \quad (11)$$

Derivation: Define the span $s \equiv 1 - \nu$. From the FOCs:

$$\frac{k}{\ell} = \frac{\alpha}{1-\alpha} \cdot \frac{w}{r+\delta} \quad (12)$$

$$\ell = \left[\frac{(1-\alpha)s \cdot e \cdot k^{\alpha s}}{w} \right]^{\frac{1}{1-(1-\alpha)s}} \quad (13)$$

Substituting into the capital FOC and solving yields the expression above.

2.2.3 Constrained Solution

The optimal capital is:

$$\hat{k}(a, e) = \min\{k^*, \lambda a\} \quad (14)$$

Given \hat{k} , the optimal labor is:

$$\hat{\ell}(a, e) = \left[\frac{(1-\alpha)(1-\nu) \cdot e \cdot \hat{k}^{\alpha(1-\nu)}}{w} \right]^{\frac{1}{1-(1-\alpha)(1-\nu)}} \quad (15)$$

And profit is:

$$\pi(a, e) = e \cdot (\hat{k}^\alpha \hat{\ell}^{1-\alpha})^{1-\nu} - w\hat{\ell} - (r+\delta)\hat{k} \quad (16)$$

2.3 Dynamic Problem: Bellman Equation

The agent's dynamic optimization problem is:

$$V(a, e) = \max_{a' \geq 0} \{ u(c) + \beta \mathbb{E}[V(a', e')|e] \} \quad (17)$$

subject to the budget constraint:

$$c + a' = y(a, e) + (1+r)a \quad (18)$$

where:

$$y(a, e) = \max\{w, \pi(a, e)\} \quad (19)$$

2.3.1 Expected Continuation Value

Given the ability transition process:

$$\mathbb{E}[V(a', e')|e] = \psi \cdot V(a', e) + (1-\psi) \sum_{e'} g(e')V(a', e') \quad (20)$$

3 Value Function Iteration

3.1 Discretization

3.1.1 Ability Grid

Following the paper (page 235), we discretize ability into $n_e = 40$ grid points $\{e_1, \dots, e_{40}\}$:

1. Points $j = 1, \dots, 38$: Equidistant in CDF space

$$G(e_j) = 0.633 + \frac{j-1}{37}(0.998 - 0.633), \quad j = 1, \dots, 38 \quad (21)$$

2. Points $j = 39, 40$: Capture the tail

$$G(e_{39}) = 0.999, \quad G(e_{40}) = 0.9995 \quad (22)$$

Grid points are computed via the inverse CDF:

$$e_j = (1 - G(e_j))^{-1/\eta} \quad (23)$$

Probability masses:

$$\Pr(e_1) = \frac{G(e_1)}{G(e_{40})}, \quad \Pr(e_j) = \frac{G(e_j) - G(e_{j-1})}{G(e_{40})}, \quad j = 2, \dots, 40 \quad (24)$$

3.1.2 Asset Grid

We use n_a grid points with curvature scaling to concentrate points near zero:

$$a_i = a_{\min} + (a_{\max} - a_{\min}) \left(\frac{i-1}{n_a - 1} \right)^{\gamma}, \quad i = 1, \dots, n_a \quad (25)$$

where $\gamma = 2$ provides quadratic spacing. Typical values: $n_a = 501$, $a_{\min} = 10^{-6}$, $a_{\max} = 4000$.

3.2 VFI Algorithm

Algorithm 1: Value Function Iteration

```

INPUT: Grids (a_i), (e_j), probabilities Pr(e_j), prices (w,r), parameters
INITIALIZE: V^0(a_i, e_j) = u(w + r*a_i)/(1-beta) for all i,j
SET: tolerance eps = 1e-5, iteration n = 0

REPEAT:
  n = n + 1
  FOR j = 1, ..., n_e: [Can parallelize over j]
    Compute expected value:
    EV_j(a') = psi*V^{n-1}(a', e_j) + (1-psi)*SUM_j' Pr(e_j')*V^{n-1}(a', e_j')
    FOR i = 1, ..., n_a:

```

```

Solve entrepreneur: (pi_ij, k_ij, l_ij) = SolveEntrepreneur(a_i, e_j, w, r, lambda)
Occupation: y_ij = max{w, pi_ij}
Income: I_ij = y_ij + (1+r)*a_i
Optimal savings: a'_ij = argmax_{a' < I_ij} {u(I_ij - a') + beta*EV_j(a')}
Update: V^n(a_i, e_j) = u(I_ij - a'_ij) + beta*EV_j(a'_ij)
Compute: diff = max_{i,j} |V^n(a_i, e_j) - V^{n-1}(a_i, e_j)|
UNTIL: diff < eps

OUTPUT: Value function V, policy functions a'(a,e), occ(a,e)

```

3.3 Solving the Savings Problem

For each state (a_i, e_j) with income I , we find:

$$a'_* = \arg \max_{k: a_k < I} \{u(I - a_k) + \beta \cdot EV(a_k)\} \quad (26)$$

Grid search: Since u is strictly concave and the grid is finite, we can use:

1. Linear search: Check all feasible a_k (simple but $O(n_a)$)
2. Binary search: Exploit concavity to find the peak ($O(\log n_a)$)

In practice, linear search with early stopping (break when $c \leq 0$) works well since the asset grid is ordered.

4 Stationary Distribution

4.1 Transition Matrix

The state space is $\mathcal{S} = \{1, \dots, n_a\} \times \{1, \dots, n_e\}$ with $|\mathcal{S}| = n_a \times n_e$ states.

We index states as $s = (i-1) \cdot n_e + j$ for (a_i, e_j) .

The transition matrix $Q \in \mathbb{R}^{|\mathcal{S}| \times |\mathcal{S}|}$ has entries:

$$Q_{s',s} = \Pr(\text{state } s' | \text{state } s) = \Pr(a' = a_{i'}, e' = e_{j'} | a = a_i, e = e_j) \quad (27)$$

Given the policy function $a'(a_i, e_j) = a_{i'(i,j)}$:

$$Q_{s',s} = \begin{cases} \psi + (1-\psi) \cdot \Pr(e_{j'}) & \text{if } i' = i'(i, j) \text{ and } j' = j \\ (1-\psi) \cdot \Pr(e_{j'}) & \text{if } i' = i'(i, j) \text{ and } j' \neq j \\ 0 & \text{otherwise} \end{cases} \quad (28)$$

4.2 Computing the Stationary Distribution

The stationary distribution $\mu^* \in \mathbb{R}^{|\mathcal{S}|}$ satisfies:

$$\mu^* = Q\mu^*, \quad \sum_s \mu_s^* = 1 \quad (29)$$

4.2.1 Method 1: Power Iteration

Algorithm 2: Power Iteration for Stationary Distribution

```

INITIALIZE: mu^0 = (1/|S|) * 1 (uniform distribution)

REPEAT:
    mu^{n+1} = Q * mu^n
    mu^{n+1} = mu^{n+1} / ||mu^{n+1}||_1 (normalize)
UNTIL: ||mu^{n+1} - mu^n||_inf < eps

OUTPUT: mu* = mu^{n+1}

```

4.2.2 Method 2: Eigenvalue Method

Find the eigenvector corresponding to eigenvalue 1:

$$Qv = v \Rightarrow \mu^* = \frac{v}{\|v\|_1} \quad (30)$$

In practice, use sparse matrix methods since Q has at most n_e non-zero entries per column.

4.3 Sparse Matrix Implementation

The transition matrix Q is highly sparse. For each state $s = (i, j)$:

- There is exactly one destination asset index $i' = i'(i, j)$
- But n_e possible destination ability indices $j' \in \{1, \dots, n_e\}$

Total non-zero entries: $n_a \times n_e \times n_e$.

Store in Compressed Sparse Row (CSR) format for efficient matrix-vector multiplication.

5 Aggregate Variables

Given the stationary distribution $\mu(a, e)$ (reshaped to $n_a \times n_e$), we compute:

5.1 Aggregate Capital

$$K = \sum_{i,j} \mu(a_i, e_j) \cdot \hat{k}(a_i, e_j) \cdot \mathbf{1}_{\text{entrepreneur}}(a_i, e_j) \quad (31)$$

5.2 Aggregate Labor Demand

$$L^d = \sum_{i,j} \mu(a_i, e_j) \cdot \hat{\ell}(a_i, e_j) \cdot \mathbf{1}_{\text{entrepreneur}}(a_i, e_j) \quad (32)$$

5.3 Aggregate Output

$$Y = \sum_{i,j} \mu(a_i, e_j) \cdot f(e_j, \hat{k}(a_i, e_j), \hat{\ell}(a_i, e_j)) \cdot \mathbf{1}_{\text{entrepreneur}}(a_i, e_j) \quad (33)$$

5.4 Aggregate Assets

$$A = \sum_{i,j} \mu(a_i, e_j) \cdot a_i \quad (34)$$

5.5 Share of Entrepreneurs

$$\text{share}_e = \sum_{i,j} \mu(a_i, e_j) \cdot \mathbf{1}_{\text{entrepreneur}}(a_i, e_j) \quad (35)$$

5.6 External Finance

$$\text{ExtFin} = \sum_{i,j} \mu(a_i, e_j) \cdot \max\{0, \hat{k}(a_i, e_j) - a_i\} \cdot \mathbf{1}_{\text{entrepreneur}}(a_i, e_j) \quad (36)$$

5.7 Total Factor Productivity

Using standard capital share of 1/3:

$$\text{TFP} = \frac{Y}{K^{1/3} L^{2/3}} \quad (37)$$

6 General Equilibrium

6.1 Market Clearing Conditions

6.1.1 Labor Market

Labor supply equals labor demand:

$$L^s = 1 - \text{share}_e = L^d \quad (38)$$

Workers supply 1 unit each, so total labor supply is the fraction of workers.

6.1.2 Capital Market

Capital demand equals total assets:

$$K = A \quad (39)$$

In equilibrium, entrepreneurs' capital is funded by all agents' savings.

6.2 Equilibrium Algorithm

Algorithm 3: General Equilibrium Computation

```
INPUT: Initial guesses ( $w^0, r^0$ ), tolerance  $\tau$ , max iterations  $N$ 
INITIALIZE: best_error = inf
```

```
FOR n = 1, ..., N:
    Solve VFI with ( $w^{n-1}, r^{n-1}$ ) -> policies ( $a'(\cdot), \text{occ}(\cdot)$ )
    Compute stationary distribution  $\mu$ 
    Compute aggregates:  $K, L^d, A, \text{share}_e$ 
```

```
Excess demands:
     $\text{ExcL} = L^d - (1 - \text{share}_e)$ 
     $\text{ExcK} = K - A$ 
```

```
IF  $|\text{ExcL}| < \tau$  AND  $|\text{ExcK}| < \tau$ :
    CONVERGED
```

```
Update prices:
     $w_{\text{new}} = w^{n-1} * (1 + \gamma_w * \text{ExcL})$ 
     $r_{\text{new}} = r^{n-1} + \gamma_r * \text{ExcK}$ 
```

```
Damped update:
     $w^n = \theta * w^{n-1} + (1-\theta) * w_{\text{new}}$ 
     $r^n = \theta * r^{n-1} + (1-\theta) * r_{\text{new}}$ 
```

```
Apply bounds:  $w^n$  in [0.01, 2.0],  $r^n$  in [-0.06, 0.12]
```

```
OUTPUT: Equilibrium ( $w^*, r^*$ ) and aggregates
```

Tuning parameters:

- Step sizes: $\gamma_w = 0.3, \gamma_r = 0.01$
- Damping: $\theta = 0.5$
- Tolerance: $\tau = 10^{-3}$

6.3 Intuition for Price Adjustments

- **Excess labor demand ($L^d > L^s$):** Raise wage w to reduce labor demand and encourage more agents to become workers
- **Excess capital demand ($K > A$):** Raise interest rate r to reduce capital demand (higher rental cost) and encourage more savings

7 Implementation Details

7.1 Numba JIT Compilation

Key functions are compiled with `@njit(cache=True)` for C-level performance:

- `solve_entrepreneur`: Static profit maximization
- `utility`: CRRA utility evaluation
- `bellman_operator`: Main VFI iteration
- `compute_aggregates_on_grid`: Aggregate computation

7.2 Parallelization

The Bellman operator is parallelized over ability states using `@njit(parallel=True)` with `prange`:

```
@njit(cache=True, parallel=True)
def bellman_operator(...):
    for i_z in prange(n_z): # Parallel loop
        for i_a in range(n_a):
            # Solve individual problem
```

This provides near-linear speedup with the number of CPU cores.

7.3 Memory Efficiency

- Use sparse matrices for the transition matrix Q
- Pre-compute expected values $EV(a', e)$ before the inner loops
- Store policy functions as integer indices rather than values

7.4 Numerical Stability

- Bound rental rate: $r + \delta > 10^{-8}$
- Bound wage: $w > 10^{-8}$
- Bound consumption: $c > 10^{-10}$ before computing utility
- Use -10^{10} as utility for infeasible consumption

8 Calibration

9 Computational Performance

10 Results Summary

Key findings:

Parameter	Value	Target/Source
σ	1.5	Standard value
β	0.904	Interest rate $\approx 4.5\%$
α	0.33	Capital share
ν	0.21	Span-of-control
δ	0.06	Depreciation rate
η	4.15	Top 10% employment share
ψ	0.894	Exit rate of establishments

Table 1: Baseline calibration from [1]

Component	Time per λ	Notes
VFI (per iteration)	$\sim 2\text{s}$	501×40 states
VFI convergence	$\sim 50 - 200$ iter	Depends on initial guess
Stationary distribution	$\sim 0.5\text{s}$	Power iteration
GE iteration	$\sim 10 - 50$ iter	Price adjustment
Total per λ	$\sim 20 - 100\text{s}$	Varies with convergence

Table 2: Approximate computation times (single core, 3.5GHz)

- Financial autarky ($\lambda = 1$) reduces GDP by $\sim 32\%$ relative to perfect credit
- TFP loss from misallocation: $\sim 22\%$
- Interest rates decline with tighter frictions (excess savings due to precautionary motives)

References

- [1] Buera, Francisco J., and Yongseok Shin. *Financial frictions and the persistence of history: A quantitative exploration*. Journal of Political Economy 121.2 (2013): 221-272.

λ	ExtFin/GDP	GDP (rel.)	TFP (rel.)	r
∞	1.69	1.00	1.00	4.6%
2.00	1.26	0.83	0.87	-2.0%
1.75	1.06	0.81	0.86	-3.7%
1.50	0.75	0.78	0.84	-4.0%
1.25	0.44	0.73	0.81	-4.5%
1.00	0.00	0.68	0.78	-6.0%

Table 3: Replication of Figure 2: Long-run effect of financial frictions