

Solve Hopenhayn (1992) Using MATLAB

Yijiang Zhou

Department of Economics
The Chinese University of Hong Kong

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Introduction and useful resources

- Hopenhayn (1992): workhorse model in firm and industry dynamics
- Lecture slides by Prof. Chris Edmond available at his [website](#).
 - Lecture 2 and 3 are essential for understanding the model and solution algorithm.
- Original MATLAB program by Prof. Alessandro Ruggieri available at his [GitHub page](#).
- Four MATLAB code files needed: `main.m`, `solve_vfi.m`, `compute_parameters.m` and `compute_statespace.m`.
- Detailed code notes on my [GitHub page](#).

Initialize model parameters

- Firm's production function (labor n is the only input):

$$y = zn^{\theta} \quad (1)$$

- Log of productivity z follows a $AR(1)$ process:

$$\ln z_t = a + \rho \ln z_{t-1} + \epsilon_t \quad (2)$$

$\sigma_{\ln z}^2$ calculated for later use given σ_{ϵ}^2 .

- Market demand is $D(p) \equiv D - p$ with D exogenously given.
- Discount rate β , fixed cost of operating c^f and entry cost c^e .

Initialize state space

- Discretize z : assume $\ln z$ follows a 21-state Markov chain on $\ln z_i$.
 - $\ln z_i$ are evenly spaced along the real line with $\ln z_1 < \ln z_2 < \dots < \ln z_{21}$ and $\ln z_1 = a - m\sigma_{\ln z}$, $\ln z_{21} = a + m\sigma_{\ln z}$.
- Calculate transition probability matrix for $\ln z_i$ using Tauchen's method: let $w = \ln z_i - \ln z_{i-1}$ and the transition probability π_{ij} is defined by:

$$\pi_{ij} = \begin{cases} \Phi\left(\frac{\ln z_1 + w/2 - \rho \ln z_i}{\sigma_\epsilon}\right) & \text{for } j = 1 \\ \Phi\left(\frac{\ln z_j + w/2 - \rho \ln z_i}{\sigma_\epsilon}\right) - \Phi\left(\frac{\ln z_{j-1} + w/2 - \rho \ln z_i}{\sigma_\epsilon}\right) & \text{for } 1 < j < 21 \\ 1 - \Phi\left(\frac{\ln z_1 - w/2 - \rho \ln z_i}{\sigma_\epsilon}\right) & \text{for } j = 21 \end{cases} \quad (3)$$

with Φ corresponding to the standard normal CDF.

Initialize state space

- Transition probability π defined as:

$$\pi_{ij} \equiv \text{Prob}(z_t = z_j | z_{t-1} = z_i) \quad (4)$$

- Initial productivity drawn from uniform distribution $g(z)$
- Discretize n : assume there are 251 grid points from 0 to 5000 for employment n .

Code for initializing z space

```

1 %% Grid for tech shock, z,
2 % by using Tauchen's method of finite state Markov approximation
3 m=3;
4 w=2*m*sigma/(Z-1);
5 lnz=(-m*sigma+a):w:(m*sigma+a);
6 z=exp(lnz);
7
8 %Markov transition matrix
9 p=zeros(Z);
10
11 %See formula on notes
12 p(:,1)=normcdf(((lnz(1)+w/2)*ones(Z,1)-ro*lnz')/stde,0,1);
13 p(:,Z)=ones(Z,1)-normcdf(((lnz(Z)-w/2)*ones(Z,1)-ro*lnz')/stde,0,1);
14 for j=2:(Z-1)
15     p(:,j)=normcdf(((lnz(j)+w/2)*ones(Z,1)-ro*lnz')/stde,0,1)-...
16         normcdf(((lnz(j)-w/2)*ones(Z,1)-ro*lnz')/stde,0,1);
17 end

```

Bisection on p

- Given price level p and employment choice n , we can write firm's profit f as:

$$f_i(p) = f(z_i, p, n) = z_i n^\theta p - n - c^f \quad (5)$$

where wage is normalized to 1.

- Bellman equation for incumbent firm is then:

$$v_i(p) = f_i(p) + \beta \max \left[0, \sum_{j=1}^{21} v_j(p) \pi_{ij} \right] \quad (6)$$

Bisection on p

- Free entry condition:

$$v^e(p) \equiv \sum_{i=1}^{21} v_i(p) g_i = c^e \quad (7)$$

- Easy to show $v^e(0) < 0$ and $v^e(p)$ is monotonically increasing in p .
- Given c^e , we can find the unique solution p^* using bisection.
 - Need to know $v^*(p)$ for each p . This is done by VFI.
 - Guess an initial p_0 between p_{min} and p_{max} , solve $v^*(p)$ with VFI, calculate $v^e(p_0)$ and compare to c^e .
 - Update new guess p_{j+1} as:

$$p_{j+1} = \begin{cases} \frac{p_j + p_{max}}{2} & \text{if } v^e(p_j) < c^e \\ \frac{p_{min} + p_j}{2} & \text{if } v^e(p_j) \geq c^e \end{cases} \quad (8)$$

- Stop when $\frac{|v^e(p) - c^e|}{c^e} \leq 10^{-6}$.

Code for bisection

```

1 %% Iterate over price of goods
2 d=1;
3 % Boundaries for price
4 pmin=0.01;
5 pmax=100;
6
7 while d>toler
8     % Guess prices
9     price=(pmin+pmax)/2;
10
11     % Solve firm value function iteration
12     [vinitial,dr,exit] = solve_vfi(price,z,Z,n,N,theta,beta,cf,p);
13
14     %Compute the decision rule for labor
15     decrule=zeros(1,Z);
16     for i=1:Z
17         decrule(i)=n(dr(i));
18     end
19
20     %Define expected value of entrant
21     value=inidis*vinitial';
22
23     % Update price till EV=ce
24     if value<ce
25         pmin=price;
26     else
27         pmax=price;
28     end
29
30     % Check convergence
31     d=abs(value-ce)/ce;
32 end

```

Value function iteration

- Write equation (6) into matrix form as:

$$\mathbf{v}^{k+1}(\mathbf{p})^\top = \mathbf{f}(\mathbf{p})^\top + \beta \max [\mathbf{0}, \mathbf{\Pi} \cdot \mathbf{v}^k(\mathbf{p})^\top] \equiv T((\mathbf{v}^k)^\top, \mathbf{p}) \quad (9)$$

k denotes number of iterations. Initial \mathbf{v}^0 is set to be $\mathbf{0}$.

- For every z and n , firm chooses optimal n^* that maximizes value function, i.e., elements of vector $\mathbf{v}^{k+1}(\mathbf{p})$ are obtained by solving:

$$v_i^{k+1}(\mathbf{p}) = \max_n \left[z_i n^\theta p - n - c^f + \beta \max \left(0, \sum_{j=1}^{21} v_j^k(\mathbf{p}) \pi_{ij} \right) \right] \quad (10)$$

Value function iteration

- Solving it yields $\mathbf{v}^{k+1}(p)$, policy $\mathbf{n}(p)$ and exit rule $\mathbf{x}(p)$ with elements (0 for exit and 1 for continue):

$$x_i(p) = 1 - \mathbb{1} \left(\sum_{j=1}^{21} v_j^k(p) \pi_{ij} < 0 \right) \quad (11)$$

- T in equation (9) is a contraction mapping. Iterating on T given initial guess \mathbf{v}^0 yields:

$$T((\mathbf{v}^k)^\top, p) = \mathbf{v}^{k+1}(p)^\top \rightarrow \mathbf{v}^*(p)^\top \text{ as } k \rightarrow \infty \quad (12)$$

- The program iterates on T until

$$\frac{\|\mathbf{v}^{k+1} - \mathbf{v}^k\|}{\|\mathbf{v}^k\|} \leq 10^{-8}$$

Code for VFI

```

1 function [vrevised,dr,exit] = solve_vfi(price,z,Z,n,N,theta,beta,cf,p)
2
3 %Value function iteration
4 d=1;
5 toler=1e-08;
6
7 % Store policy function
8 dr =zeros(1,Z); %record for policy function
9 exit=zeros(1,Z); %record for exit decision
10
11 % Guess value functions
12 vinitia=zeros(1,Z);
13 vrevised=zeros(1,Z);
14
15 % Fixed cost matrix
16 cost=cf*ones(N,1)';
17
18 while d>toler
19     for i=1:Z
20         fi = z(i)*n.^theta.*price - n - cost;
21         [vrevised(i),dr(i)]=max(fi+beta*max(p(i,:)*vinitia',0)*ones(N,1)');
22         exit(i)=1-1*(p(i,:)*vinitia'<0);
23     end
24     d=norm(vrevised-vinitia)/norm(vrevised);
25     vinitia=vrevised;
26 end
27
28 end

```

Distribution of incumbent firms

- In the previous section, we have solved p^* and $x(p^*)$.
- $\mu_{it} = \mu_t(z_i)$ denotes measure of firms with productivity z_i at t . It is also the element in μ_t , which evolves according to:

$$\mu_{t+1}^\top = \Psi(p^*) \mu_t^\top + m g^\top \equiv U(\mu_t^\top, p^*), \quad t = 0, 1, \dots \quad (13)$$

- μ is NOT a probability distribution vector.
- $\Psi(p^*)$ has elements

$$\psi_{ij}(p^*) = x_j(p^*) \pi_{ji}, \quad i, j = 1, \dots, n \quad (14)$$

- Easy to see $\psi_{ij}(p^*) = 0$ if firm exits with productivity draw z_j at t .
- Measure of firms at grid point z_i at $t + 1$ depends on transition probabilities and exit decisions of incumbents at t and flow of new entrants.

Distribution of incumbent firms

- Given m , we can calculate the stationary distribution μ as:

$$\mu^\top = m(I - \Psi(p^*))^{-1} \mathbf{g}^\top \equiv \mu(m, p^*)^\top \quad (15)$$

- Or: assume initial distribution $\mu^0 = \mathbf{g}$ and iterate on mapping U .
- Either way, μ is linear in m can be written into

$$\mu(m, p^*) = m \times \mu(1, p^*)$$

Mass of entry

- How do we know m ?
- Market clearing condition:

$$Y(m, p^*) = \sum_{i=1}^{21} y_i(p^*) [\mu_i(m, p^*) + mg_i] = D(p^*) \quad (16)$$

- Hence:

$$m^* = \frac{D(p^*)}{Y(1, p^*)} = \frac{D(p^*)}{\sum_{i=1}^{21} y_i(p^*) [\mu_i(1, p^*) + g_i]} \quad (17)$$

- Substituting m^* into equation (15) gives μ^* .

Exit threshold

- Using policy function $x(p)$ derived from the last section, we can calculate the exit threshold of productivity as:

$$z(p^*) = z_{i^*}, \quad i^* \equiv \min_i \left[\sum_{j=1}^n v_j(p^*) \pi_{ij} \geq 0 \right]$$

- Firms whose productivity draws are below z_{i^*} exit the market in the equilibrium.

Full stationary equilibrium

- Full stationary equilibrium: (p^*, z^*, μ^*, m^*)
 - p^* : price level for goods
 - z^* : exit threshold of productivity
 - μ^* : distribution of incumbent firms
 - m^* : mass of entrants
- They correspond to a steady-state of the dynamic system implied by the perfect foresight equilibrium $\{p_t, z_t, \mu_t, m_t\}_{t=0}^{\infty}$.

Code for calculating $\mu(1, p^*)$

```

1 %Given the value function and policy function, iterate on industry structure
2 %until it converges
3 d=1;
4 muinitial=inidis;
5 while d>toler
6     muexit=muinitial.*exit;           %exit decision
7     mustay=muexit*p;                 %update for the incumbents stay
8     mumentry=mustay+inidis;          %entry
9     murevised=mumentry./sum(mumentry);
10    d=norm(murevised-muinitial)/norm(murevised);
11    muinitial=murevised;
12 end

```

The author made a mistake...

- ...by scaling μ_{entry} , which is μ_{t+1} , using the sum of its elements.
- This makes no sense.
 - μ is not a probability distribution vector.
 - No need to scale it before calculating “distance” between μ_{t+1} and μ_t after each iteration.
- It has been tested, that the corrected code (next page) yields the same $\mu(1, p^*)$ as that produced by equation (15).

Corrected code for calculating $\mu(1, p^*)$

```

1 % mistake-free iteration
2 while d>toler
3     muexit=muinitial.*exit;                %exit decision
4     mustay=muexit*p;                      %update for the incumbents stay
5     mumentry=mustay+inidis;                %entry
6     murevised=mumentry;
7     d=norm(murevised-muinitial)/norm(murevised);
8     muinitial=murevised;
9 end

```

Corrected code for computing m^*

```

1 %% Calculating the entry mass M
2 % Using equilibrium condition in goods market
3 % mistake-free
4 murevised_sc = murevised ./ sum(murevised);
5
6 y=D-price;
7 Xstar=z(Z-sum(exit));
8 Pstar=price;
9 Size = (decrule)*murevised_sc';
10 Y=(decrule.^theta.*z)*murevised';
11 Mstar=y/[Y+(decrule.^theta.*z)*inidis'];
12 Exrate=sum(murevised_sc(1:Z-sum(exit)))*100;

```

Summary

Solve Hopenhayn model

- Step 1: use free entry condition (7) and monotonicity of $v^e(p)$ to construct a bisection on price p .
 - Rely on VFI to compute $v^*(p)$, $n(p)$ and $x(p)$ for each p .
- Step 2: given p^* and $x(p^*)$, use mapping U in equation (13) to compute $\mu(1, p^*)$, and equation (17) for m^* .
- Step 3: obtain z^* using exit decision $x(p^*)$.

Extras

- Changing initial distribution $g(z)$: only μ^* changes.
- Changing transition probability matrix to identity matrix: failed to solve μ^* .
- Increase entry cost c^e : price p^* increases and productivity threshold z^* decreases. Weaker election effect.
- Decrease fixed cost c^f : p^* decreases. How about z^* ?