Advanced Tools in Macroeconomics

Occasionally Binding Constraints

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 Previously we looked at models that could be written in the following way,

$$E_t[F(x_{t-1}, x_t, x_{t+1})] = 0$$

- Where x was a vector of endogenous and exogenous (possibly stochastic) variables
- Now we are going to look at models that are given by

$$E_t[F(x_{t-1}; x_t, z_t; x_{t+1}, z_{t+1})] = 0$$

- Mhere z_t is a discrete stochastic variable with some transition matrix T.
- The vector x_t can still contain other stochastic variables if you'd like, but wlog, it is assumed here that it doesn't)

- ▶ Suppose z_t can take on values in $Z = \{z^1, z^2, ..., z^J\}$.
- ► We will not linearize with respect to z but only with respect to x.
- ▶ That is, for each $z^i \in Z$ we will linearize the system around \bar{x} , such that

$$E_j F(\bar{x}; \bar{x}, z^i; \bar{x}, z^j) = D^i$$

▶ In fact, we could linearize around a different \bar{x} for each z^i if we would like to, but let's keep things simple.

- ► We indicate this period's state with superscript *i* and next-period's state with superscript *j*.
- ▶ The optimal choice of x_t will depend on z^i . Thus x_{t+1} will in turn depend on z^j (the exogenous state "tomorrow").
- Next period's state is not known, but it has a discrete distribution. So think of E_t as a sum and note that we have one realization of x_{t+1} for each j.

► Linearization of the system of equations gives

$$D^{i} + J_{x_{t-1}}^{i}(x_{t-1} - \bar{x}) + J_{x_{t}}^{i}(x_{t} - \bar{x}) + E_{j}[J_{x_{t+1}}^{j}(x_{t+1}(j) - \bar{x})|i] = 0,$$

- where $J^i_{x_{t-1}}$ is the Jacobian of $E_j[F(\bar{x},z^i;\bar{x},z^j;\bar{x})]$ with respect to the first argument, $J^i_{x_t}$ is the Jacobian with respect to the second argument, and $J^j_{x_{t+1}}$ is the Jacobian with respect to $x_{t+1}(j)$.
- ▶ Thus there are J "last Jacobians"; one for each j.

► We can again write this as

$$A^{i}u_{t-1}(i) + B^{i}u_{t}(i) + \sum_{i=1}^{J} C^{i}u_{t+1}(j) + D^{i} = 0, \quad \forall i$$

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Looks complicated? Let's make it more concrete.

Consumption/Savings problem with unemployment

The Euler equations for the employed and unemployed agent are

$$0 = -u'(a_{t-1}(1+r)+w-a_t)+\beta(1+r)[T_{e,e}u'(a_t(1+r)+w-a_{t+1}(e)) + T_{e,u}u'(a_t(1+r)-a_{t+1}(u))]$$

$$0 = -u'(a_{t-1}(1+r)-a_t) + \beta(1+r)[T_{u,e}u'(a_t(1+r)+w-a_{t+1}(e)) + T_{u,u}u'(a_t(1+r)-a_{t+1}(u))]$$

▶ Can take jacobian w.r.t a_{t-1} , a_t and $a_{t+1}(i)$, i = e, u, and evaluate around \bar{a}

Consumption/Savings problem with unemployment

The linearized regime switching system is given by

$$A^{e}u_{t-1}(e) + B^{e}u_{t}(e) + C^{e,e}u_{t+1}(e) + C^{e,u}u_{t+1}(u) + D^{e} = 0$$

$$A^{u}u_{t-1}(u) + B^{u}u_{t}(u) + C^{u,e}u_{t+1}(e) + C^{u,u}u_{t+1}(u) + D^{u} = 0$$

▶ We would look for solutions $u_t = E^i + F^i u_{t-1}$, i = e, u.

Let's go back to the general formulation:

$$A^{i}u_{t-1}(i) + B^{i}u_{t}(i) + \sum_{j=1}^{J} C^{j}u_{t+1}(j) + D^{i} = 0, \quad \forall i$$

We are looking for J policy functions of the type

$$u_t(i) = E^i + F^i u_{t-1}(i), \quad i = 1, 2, \dots, J$$

 \triangleright Time iteration means to find u_t as the solution to

$$A^{i}u_{t-1}(i)+B^{i}u_{t}(i)+\sum_{i=1}^{J}C^{j}(E_{n}^{j}+F_{n}^{j}u_{t}(i))+D^{i}=0, \quad \forall i$$

and update the coefficients E_{n+1}^i and F_{n+1}^i accordingly.

► Therefore we iterate on the equations

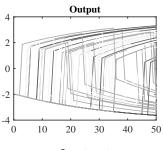
$$E_{n+1}^{i} = (B^{i} + \sum_{j=1}^{I} C^{j} F_{n}^{j})^{-1} (-(D + \sum_{j=1}^{I} C^{j} E_{n}^{j}))$$

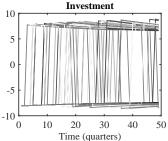
$$F_{n+1}^{i} = (B^{i} + \sum_{j=1}^{I} C^{j} F_{n}^{j})^{-1} (-A)$$

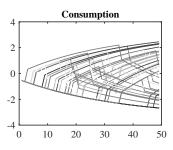
for
$$i = 1, 2, ..., J$$

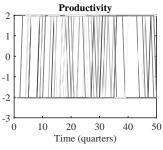
Until convergence

"Impulse Responses"



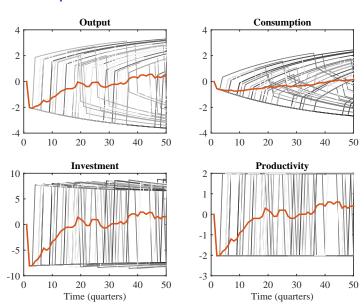






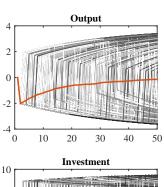
- Looks ok, but it's not pretty.
- ▶ Plot all possible sample paths? That would be 2^{50} . Or more generally if T is the length of the impulse response and N is the number of elements in Z, then there are N^T possible paths.
- Popular alternative: Plot the expected paths.
 - Quite good because this is what an econometrician would pick up if he had access to the data generated by the model.
 - Let's average over the samples plotted

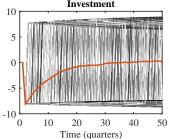
Impulse Responses

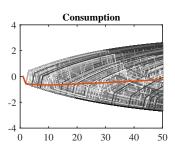


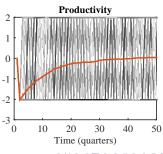
- ► That doesn't really look like an impulse response to me!
- Remedy: use more expected paths?
- ► The previous graph used 40 samples. Let's crank it up to 4,000!

Impulse Responses



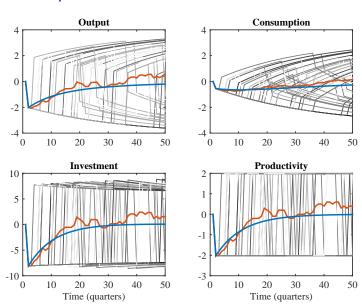






- ► Yep, much better!
- But it's time consuming.
- The calculation took several minutes on my desktop. And this is a simple model.
- It turns out that we can do this theoretically instead!

Impulse Responses



- Even better! All wiggles are gone.
- ► The theoretical paths are equal to the averaged sample paths as the sample goes to infinity.
- ► How is this done?

- ▶ Denote the expected value of u_{t+s} conditional on information available at time t, and conditional on being in state $z_{t+s} = z_j$ as $E_t[u_{t+s}|z_{t+s} = z_j]$.
- Because of the linearities of the policy functions, this can be written as

$$E_{t}[u_{t+s}|z_{t+s} = z_{j}] = \sum_{i=1}^{l} Pr(z_{t+s-1} = z_{i}|z_{t+s} = z_{j})$$

$$\times (E^{j} + F^{j}E_{t}[u_{t+s-1}|z_{t+s-1} = z_{i}])$$

$$E_{t}[u_{t+s}|z_{t+s}=z_{j}] = \sum_{i=1}^{l} Pr(z_{t+s-1}=z_{i}|z_{t+s}=z_{j}) \times (E^{j} + F^{j}E_{t}[u_{t+s-1}|z_{t+s-1}=z_{i}])$$

► Bayes' rule states that

$$P(A|B) = P(B|A)\frac{P(A)}{P(B)}$$

$$E_{t}[u_{t+s}|z_{t+s}=z_{j}] = \sum_{i=1}^{r} Pr(z_{t+s-1}=z_{i}|z_{t+s}=z_{j})$$
$$\times (E^{j} + F^{j}E_{t}[u_{t+s-1}|z_{t+s-1}=z_{i}])$$

Thus

$$Pr(z_{t+s-1} = z_i | z_{t+s} = z_j) = Pr(z_{t+s} = z_j | z_{t+s-1} = z_i)$$

$$\times \frac{Pr(z_{t+s-1} = z_i)}{Pr(z_{t+s} = z_i)}$$

If z follows transition matrix T, this can be written as

$$Pr(z_{t+s-1} = z_i | z_{t+s} = z_j) = Pr(z_{t+s} = z_j | z_{t+s-1} = z_i)$$

$$\times \frac{Pr(z_{t+s-1} = z_i)}{Pr(z_{t+s} = z_j)}$$

$$= T_{ij} \frac{v_{t+s-1,i}}{v_{t+s,j}}$$

▶ Where T_{ij} is the (i,j)th element of transition matrix T, and $v_{t+s,j}$ is the jth element of the vector

$$v_{t+s} = v_{t+s-1} \times T$$

for some initial v_t .

Thus our nasty equation

$$E_{t}[u_{t+s}|z_{t+s}=z_{j}] = \sum_{i=1}^{l} Pr(z_{t+s-1}=z_{i}|z_{t+s}=z_{j})$$

$$\times (E^{i} + F^{i}E_{t}[u_{t+s-1}|z_{t+s-1}=z_{i}])$$

turns into something more pleasant

$$E_{t}[u_{t+s}|z_{t+s} = z_{j}] = \sum_{i=1}^{l} T_{ij} \frac{v_{t+s-1,i}}{v_{t+s,j}} \times (E^{i} + F^{i}E_{t}[u_{t+s-1}|z_{t+s-1} = z_{i}])$$

And

$$E_t[u_{t+s}] = \sum_{i=1}^{r} v_{t+s,j} E_t[u_{t+s}|z_{t+s} = z_j]$$

- ➤ To implement this procedure, we still need to answer the following:
- ▶ What is the initial condition, u_{t-1} ?
- \blacktriangleright What is v_t ?

What is u_{t-1} ?

- This is somewhat arbitrary, but a good start is to assume that the economy is at it's long run expected value in period t; u_{ss}.
- ightharpoonup Given a long-run distribution v, this is given by

$$u_{ss} = \sum_{j=1}^{l} u_{j,ss} v_j$$

 \blacktriangleright Where $u_{j,ss}$ solves

$$u_{j,ss} = \sum_{i=1}^{I} T_{ij} \frac{v_i}{v_j} \times (E^i + F^i u_{i,ss}), \quad j = 1, \dots, I$$

We can either iterate to find $u_{j,ss}$, or to set it up as a linear system of equations.

► The nice thing about this starting value is that the expected value

$$E_t[u_{t+s}] = \sum_{j=1}^{l} v_{t+s,j} E_t[u_{t+s}|z_{t+s} = z_j]$$

will converge to u_{ss} as s goes to infinity.

► That is

$$\lim_{s\to\infty} E_t[u_{t+s}] = u_{ss}$$

What is v_t ?

- This is entirely up to you, and forms the basis of your impulse response.
- ▶ Setting $v_t = [0, 0, 1, 0, 0, ...]$ means that you know with certainty that you are in state 3 in period t.
- ▶ In the Ramsey, the starting condition was $v_t = [0, 1]$.