R406: Applied Economic Modelling with Python

Systems of Difference Equations

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Lecture Contents

Introduction

Introduction

- So far we considered single-variable equations
- Also, we stuck to the autonomous case; we shall do the same in the current topic
- Suppose that instead of having just one equation in one variable, we have the following set of two equations in two variables:

$$y_t = ax_{t-1} + by_{t-1}$$
$$x_t = cx_{t-1} + dy_{t-1}$$

- Apparently, this is already a system of difference equations
- Such systems turn out to be very useful in describing relationships among economic variables which are not confined to a single direction of causality

A note on classification

- **Autonomous** vs. **non-autonomous** systems were already introduced on-the-fly
- **Non-homogeneous** systems are those in which at least one equation is non-homogeneous
- Likewise, **non-linear** systems of equations are the ones in which there is at least one non-linear equation
- As with the single-equation case, we will consider only linear systems

Analysis of linear systems of equations

Matrix form of systems

- The simple homogeneous system example given earlier can be written as follows:
$$\begin{bmatrix} x_t \\ y_t \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x_{t-1} \\ y_{t-1} \end{bmatrix} \quad (*)$$

- Using matrix shorthand, this is the same as:

$$\mathbf{u}_t = \mathbf{A}\mathbf{u}_{t-1}$$

where $\mathbf{u}_t = [x_t, y_t]'$, and $\mathbf{A}$ is the coefficient matrix

- By analogy, the non-homogeneous counterpart to this system would look as follows:

$$\mathbf{u}_t = \mathbf{A}\mathbf{u}_{t-1} + \mathbf{b}$$

where $\mathbf{b}$ is a vector of conformable size (can be time-dependent but let's forget about this for the time being)

Equilibrium

- Take $(*)$ are the reference point
- Equilibrium implies that $x_t = x^*$ and $y_t = y^*$, $\forall t$
- Therefore, in equilibrium:

$$\begin{bmatrix} x^* \\ y^* \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x^* \\ y^* \end{bmatrix},$$

or:

$$\mathbf{u}^* = \mathbf{A}\mathbf{u}^*$$

- Move everything to the LHS of the equation:

$$\mathbf{u}^* - \mathbf{A}\mathbf{u}^* = \mathbf{0} \Leftrightarrow (\mathbf{I} - \mathbf{A})\mathbf{u}^* = \mathbf{0}$$

Equilibrium (2)

- If $I - A$ is a non-singular matrix, the equilibrium corresponds to $u^* = 0$
- Now, return to the non-homogeneous system; equilibrium again requires that $u_t = u^*$, $\forall t$, i. e.:

$$u^* = Au^* + b$$

- Move everything except b to the LHS:

$$u^* - Au^* = b$$

Equilibrium (3)

- This is the same as:

$$(\mathbf{I} - \mathbf{A})\mathbf{u}^* = \mathbf{b}$$

- If $\mathbf{I} - \mathbf{A}$ is a non-singular matrix, then the equilibrium of the system is found where:

$$\mathbf{u}^* = (\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}$$

- This looks very much like the result for the single-equation case which shouldn't be surprising

Stability

- As in the single-equation case, in order to analyse stability, we need to solve the system
- The procedure that is followed is analogical
- Start with the homogeneous case:

$$\mathbf{u}_t = \mathbf{A}\mathbf{u}_{t-1} = \mathbf{A}(\mathbf{A}\mathbf{u}_{t-2}) = \mathbf{A}^2\mathbf{u}_{t-2} = \ldots = \mathbf{A}^t\mathbf{u}_0$$

- In the non-homogeneous case the solution is:

$$\mathbf{u}_t = \mathbf{A}^t\mathbf{u}_0 + (\mathbf{I} + \mathbf{A} + \mathbf{A}^2 + \ldots + \mathbf{A}^{t-1})\mathbf{b}$$

Stability (2)

- Note that a non-homogeneous system can be reduced to a homogeneous one by expressing it in deviations from equilibrium
- Therefore, after subtracting $\mathbf{u}^*$ from both sides of the matrix equation, we have:
$$\mathbf{u}_t - \mathbf{u}^* = \mathbf{A}(\mathbf{u}_{t-1} - \mathbf{u}^*)$$
- Denote $\mathbf{z}_t = \mathbf{u}_t - \mathbf{u}^*$; then:
$$\mathbf{z}_t = \mathbf{A}\mathbf{z}_{t-1}$$
- This is a first-order homogeneous system
- Therefore, from now on the discussion can proceed with homogeneous systems without losing generality with respect to the non-homogeneous case

Stability (3)

Theorem 1

If the matrix $\mathbf{A}$ has two distinct eigenvalues r and s, then there exists a matrix $\mathbf{V} = [\mathbf{v}^r \quad \mathbf{v}^s]$ composed of the eigenvectors corresponding to r and s such that:

$$\mathbf{D} = \begin{bmatrix} r & 0 \\ 0 & s \end{bmatrix} = \mathbf{V}^{-1}\mathbf{A}\mathbf{V}$$

- If we pre-multiply both sides of the Theorem's result by $\mathbf{V}$ and post-multiply them by $\mathbf{V}^{-1}$, we get:

$$\mathbf{V}\mathbf{D}\mathbf{V}^{-1} = \mathbf{V}\mathbf{V}^{-1}\mathbf{A}\mathbf{V}\mathbf{V}^{-1} = \mathbf{A}$$

- It is also easy to see that $\mathbf{A}^t = \mathbf{V}\mathbf{D}^t\mathbf{V}^{-1}$

Stability (3)

- Using this, we can write:

$$\mathbf{u}_t = \mathbf{A}^t \mathbf{u}_0 = \mathbf{V}\mathbf{D}^t\mathbf{V}^{-1}\mathbf{u}_0$$

- But this can be stated explicitly as follows:

$$\mathbf{u}_t = \mathbf{V}\begin{bmatrix} r^t & 0 \\ 0 & s^t \end{bmatrix}\mathbf{V}^{-1}\mathbf{u}_0$$

Theorems

Theorem 2

A necessary and sufficient condition for the system of difference equations to be globally asymptotically stable is that all eigenvalues of the matrix $\mathbf{A}$ have moduli strictly less than 1.

Theorem 3

If all eigenvalues of $\mathbf{A}$ have moduli strictly less than 1, the difference equation is globally asymptotically stable and every solution converges to the constant vector $(\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}$.

Transformation to autonomous form

- Consider the non-autonomous equation

$$y_{t+1} = f(t, y_t), \quad t = 0, 1, 2, \ldots$$

- We can introduce a variable x_t and consider the modified system

$$y_{t+1} = f(x_t, y_t)$$
$$x_{t+1} = x_t + 1$$

together with the initial condition $x_0 = 0$

- The resulting system is autonomous, which has been achieved at the expense of an increase in the dimension of the original problem

- Sometimes such a transformation can facilitate the analysis of the system

Order reduction

- Consider the equation

$$y_{t+p} = f(y_t, y_{t+1}, \ldots, y_{t+p-1}), \quad t = 0, 1, 2, \ldots \tag{1}$$

- Introduce the following variables

$$
\begin{array}{rcl}
x_{1,t} & = & y_t \\
x_{2,t} & = & y_{t+1} \\
x_{3,t} & = & y_{t+2} \\
& \ldots & \\
x_{p,t} & = & y_{t+p-1}
\end{array}
$$

- Then we can write the system

$$
\begin{array}{rcl}
x_{1,t+1} & = & x_{2,t} \\
x_{2,t+1} & = & x_{3,t} \\
& \ldots & \\
x_{p-1,t+1} & = & x_{p,t} \\
x_{p,t+1} & = & f(x_{1,t}, x_{2,t}, \ldots, x_{p,t})
\end{array}
\tag{2}
$$

Order reduction (2)

- This approach trades the p-th order equation (1) for the first-order system of p equations (2)
- The representation of a problem as a system of equations can be more convenient to work with in certain cases