

# Time-Series Cross-Section Analysis

## Enders, Chapter 1: Difference Equations

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# Outline

- 1 Difference Equations and Time Series Analysis
- 2 Solving Difference Equations (First-Order)
- 3 Solving Difference Equations (General)

# Stochastic Difference Equations

“A difference equation expresses the value of a variable as a function of its own lagged values, time, and other variables...The reason for introducing...[these] equations is to make the point that time-series econometrics is concerned with the estimation of difference equations containing stochastic components” (Enders, p.3).

- 1<sup>st</sup>, 2<sup>nd</sup>, and  $n^{\text{th}}$  difference  $\Delta$ ,  $\Delta^2$ , and  $\Delta^n$
- 1<sup>st</sup>, 2<sup>nd</sup>, and  $n^{\text{th}}$  order linear difference equations

$$y_t = a_0 + a_1 y_{t-1} + x_t$$

$$y_t = a_0 + a_1 y_{t-1} + a_2 y_{t-2} + x_t$$

$$y_t = a_0 + \sum_{i=1}^n a_i y_{t-i} + x_t$$

# Theory Evaluation

Many economic theories generate model specifications in the form of difference equations

- The Random Walk Hypothesis
- Reduced-form and Structural Equations
- Error-Correction: Forward and Spot Prices
- Non-linear dynamics

Paper stones is an example of political theory that generates a model specification in the form of difference equations

## Solving by Iteration

“A solution to a difference equation expresses the value of  $y_t$  as a function of the elements of the  $x_t$  sequence and  $t \dots$  and possibly initial conditions” (Enders, p.9).

$$y_t = a_0 + a_1 y_{t-1} + \varepsilon_t$$

$$y_1 = a_0 + a_1 y_0 + \varepsilon_1$$

$$y_2 = a_0 + a_1 [a_0 + a_1 y_0 + \varepsilon_1] + \varepsilon_2$$

$$y_2 = a_0 + a_1 a_0 + (a_1)^2 y_0 + a_1 \varepsilon_1 + \varepsilon_2$$

$$\vdots$$

$$y_t = a_0 \sum_{i=0}^{t-1} (a_1)^i + (a_1)^t y_0 + \sum_{i=0}^{t-1} (a_1)^i \varepsilon_{t-i}$$

# The Dynamics of First-Order Difference Equations

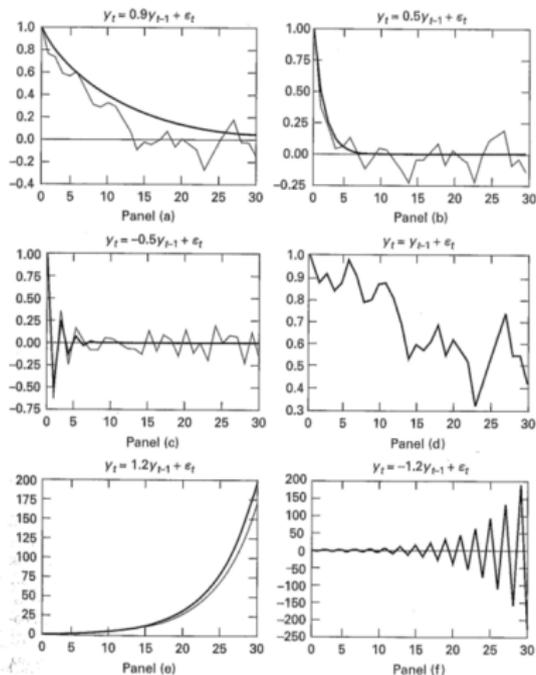


FIGURE 1.2 Convergent and Nonconvergent Sequences

## Solving $n^{\text{th}}$ Order Difference Equations

The complete  $n^{\text{th}}$  order difference equation is

$$y_t = a_0 + \sum_{i=1}^n a_i y_{t-i} + x_t$$

The homogeneous portion of the  $n^{\text{th}}$  order difference equation is

$$y_t = \sum_{i=1}^n a_i y_{t-i}$$

- A homogeneous solution to an  $n^{\text{th}}$  order difference equation is a solution to the homogeneous portion of the difference equation. There should be  $n$  solutions.
- A particular solution is a solution to the original complete difference equation.
- A general solution to an  $n^{\text{th}}$  order difference equation is a particular solution plus all homogeneous solutions.

# The Solution Methodology

- 1 Form the homogeneous equation and find all  $n$  homogeneous solutions;
- 2 Find a particular solution;
- 3 Obtain the general solution as the sum of the particular solution and a linear combination of all homogeneous solutions;
- 4 Eliminate the arbitrary constant(s) by imposing the initial condition(s) on the general solution.

# Solving Homogeneous Difference Equations

Ex. 1 (First-order):  $y_t = .9y_{t-1}$

- The homogeneous solution will take the form  $y_t^h = A\alpha^t$
- The goal is to solve for  $A$  and  $\alpha$
- Substitute for  $y_t$

$$A\alpha^t - .9A\alpha^{t-1} = 0$$

- Divide by  $A\alpha^{t-1}$

$$\alpha - .9 = 0$$

- So, now we have

$$y_t^h = A(.9)^t$$

# Solving Homogeneous Difference Equations

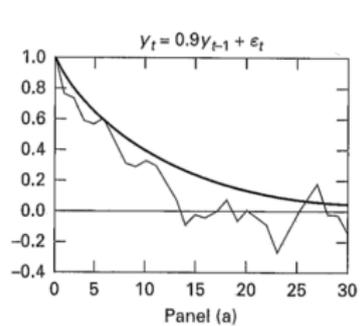
Ex. 1 (First-order):  $y_t = .9y_{t-1}$

- We can eliminate the arbitrary constant if we know the outcome in the initial period  $y_0$

$$y_0 = A(.9)^0$$

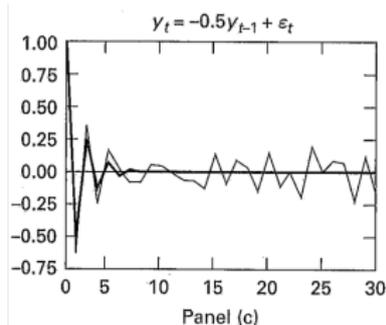
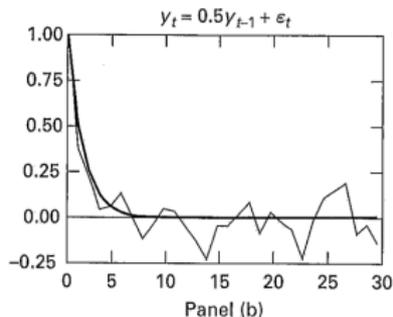
- If we set  $y_0 = 1$ , we have our final solution

$$y_t^h = (.9)^t$$



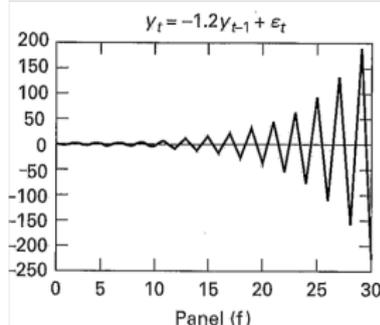
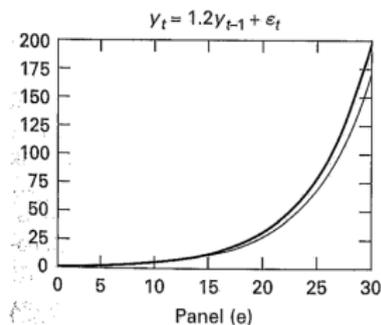
# Stability Conditions for First-Order Solutions

If  $|\alpha| < 1$ , then  $\alpha^t$  converges to zero as  $t$  goes to infinity.  
Convergence is direct if  $0 < \alpha < 1$  and oscillatory if  $-1 < \alpha < 0$ .



# Stability Conditions for First-Order Solutions

If  $|\alpha| > 1$ , the solution is not stable. If  $\alpha > 1$ , then  $\alpha^t$  converges to infinity as  $t$  goes to infinity. If  $\alpha < 1$  and the solution oscillates explosively.



# Solving Homogeneous Difference Equations

Ex. 2 (Second-order):  $y_t = 3 + .9y_{t-1} - .2y_{t-2}$

- Again, the homogeneous solutions will take the form  $y_t^h = A\alpha^t$
- The goal is to solve for  $A_1$ ,  $A_2$ ,  $\alpha_1$ , and  $\alpha_2$
- Substitute for  $y_t$

$$A\alpha^t - .9A\alpha^{t-1} + .2A\alpha^{t-2} = 0$$

- Divide by  $A\alpha^{t-2}$

$$\alpha^2 - .9\alpha + .2 = 0$$

- There are two solutions. We solve for  $\alpha_1$  and  $\alpha_2$  using the quadratic formula

$$\alpha_1, \alpha_2 = \frac{.9 \pm \sqrt{.81 - 4(.2)}}{2} = .5, .4$$

## Solving Homogeneous Difference Equations

Ex. 2 (Second-order):  $y_t = 3 + .9y_{t-1} - .2y_{t-2}$

- So, now we have

$$y_t = A_1(.5)^t + A_2(.4)^t$$

- We can eliminate the arbitrary constants if we know the outcome in the initial periods  $y_0$  and  $y_1$
- If we set  $y_0 = 13$  and  $y_1 = 11.3$ , for instance, we two equations and two unknowns. To these equations, we need to add the particular (steady-state) solution ( $c = 3/(1 - .9 + .2)$ ).

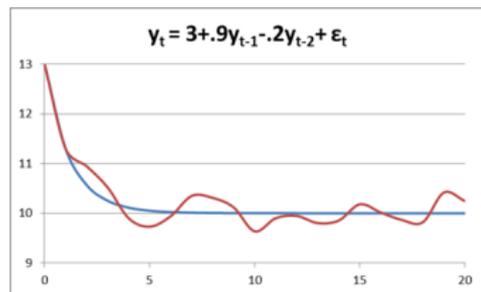
$$\begin{aligned} 13 &= 10 + A_1 + A_2 \\ 11.3 &= 10 + A_1(.5) + A_2(.4) \end{aligned}$$

## Solving Homogeneous Difference Equations

Ex. 2 (Second-order):  $y_t = 3 + .9y_{t-1} - .2y_{t-2}$

- Solving gives us  $A_1 = 1$  and  $A_2 = 2$ , and our final solution is

$$y_t = (.5)^t + 2(.4)^t$$

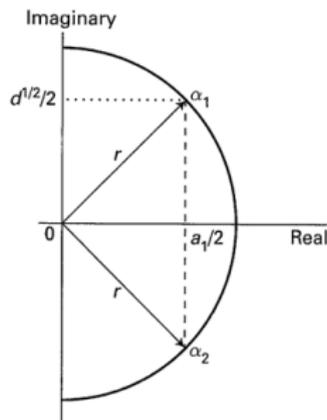


$$x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

$$\alpha = \frac{-a_1 \pm \sqrt{a_1^2 - 4a_2}}{2}$$

## Stability Conditions for Second-Order Solutions

*stability requires as all characteristic roots lie within the unit circle (Enders, p. 29).*



$$r < 1$$

$$\alpha = \frac{-a_1 \pm \sqrt{a_1^2 - 4a_2}}{2}$$

**FIGURE 1.6** Characteristic Roots and the Unit Circle

## Stability Conditions for Higher-Order Systems

Higher-Order Systems:  $y_t - \sum_{i=1}^n a_i y_{t-i} = 0$

Oftentimes, we do not need to solve for the characteristic roots of higher-order systems.

- 1 A necessary condition for stability is  $\sum_{i=1}^n a_i < 1$
- 2 A sufficient condition for stability is  $\sum_{i=1}^n |a_i| < 1$
- 3 The process contains a unit root if  $\sum_{i=1}^n a_i = 1$

# Particular Solutions for Deterministic Processes

If  $\mathbf{x}_t = \mathbf{0}$ , the difference equation becomes

$$y_t = a_0 + a_1 y_{t-1} + a_2 y_{t-2} + \dots + a_n y_{t-n},$$

which is solved when  $\Delta y_t = 0$  or  $y_t = y_{t-1} = y_{t-2} = y_{t-n} = c$ .

- Substituting for  $y_t$  gives

$$c = a_0 + a_1 c + a_2 c + \dots + a_n c$$

- Solving for  $c$  gives

$$c = a_0 / (1 - a_1 - a_2 - \dots - a_n)$$

- Thus, a particular solution is

$$y_t = a_0 / (1 - a_1 - a_2 - \dots - a_n)$$

# Particular Solutions for Stochastic Processes

## The Method of Undetermined Coefficients

- 1 Since linear equations have linear solutions, we know the form of the solution.
- 2 Posit a linear *challenge solution* that includes all the terms thought to appear in the solution.
- 3 Solve for the undetermined coefficients.

# The Method of Undetermined Coefficients

Ex. 3 (First-Order):  $y_t = 3 + .9y_{t-1} + \varepsilon_t$

- Posit a linear *challenge solution* for the stochastic portion of the particular solution

$$y_t = \sum_{i=0}^{\infty} \alpha_i \varepsilon_{t-i}$$

- Substitute the challenge solution into the difference equation

$$\alpha_0 \varepsilon_t + \alpha_1 \varepsilon_{t-1} + \alpha_2 \varepsilon_{t-2} + \dots = .9[\alpha_0 \varepsilon_{t-1} + \alpha_1 \varepsilon_{t-2} + \alpha_2 \varepsilon_{t-3} + \dots] + \varepsilon_t$$

# The Method of Undetermined Coefficients

Ex. 3 (First-Order):  $y_t = 3 + .9y_{t-1} + \varepsilon_t$

- Collect like terms

$$(\alpha_0 - 1)\varepsilon_t + (\alpha_1 - .9\alpha_0)\varepsilon_{t-1} + (\alpha_2 - .9\alpha_1)\varepsilon_{t-2} + \dots = 0$$

- Verify that there are coefficient values that make the challenge solution a solution for the difference equation.

$$\begin{aligned}(\alpha_0 - 1) &= 0 \\(\alpha_1 - .9\alpha_0) &= 0 \\(\alpha_2 - .9\alpha_1) &= 0 \\&\vdots\end{aligned}$$

- Solving for  $\alpha_j$ , we have  $\alpha_j = (.9)^j$

## Putting it all together

Ex. 3 (First-Order):  $y_t = 3 + .9y_{t-1} + \varepsilon_t$

- This gives the general solution

$$y_t = 30 + A(.9)^t + \sum_{i=0}^{\infty} (.9)^i \varepsilon_{t-i}$$

- We can eliminate the arbitrary constant if we have an initial value for  $y_0$ .

$$y_0 = 30 + A + \sum_{i=0}^{\infty} (.9)^i \varepsilon_{-i}$$

## Putting it all together

Ex. 3 (First-Order):  $y_t = 3 + .9y_{t-1} + \varepsilon_t$

- Substituting  $A$  into the general solution gives

$$y_t = 30 + \left[ y_0 - 30 - \sum_{i=0}^{\infty} (.9)^i \varepsilon_{-i} \right] (.9)^t + \sum_{i=0}^{\infty} (.9)^i \varepsilon_{t-i}$$

- Collecting like terms, we have

$$y_t = 30 + (.9)^t [y_0 - 30] + \sum_{i=0}^{\infty} (.9)^i \varepsilon_{t-i} - (.9)^t \sum_{i=0}^{\infty} (.9)^i \varepsilon_{-i}$$

## Putting it all together

Ex. 3 (First-Order):  $y_t = 3 + .9y_{t-1} + \varepsilon_t$

- The stochastic portion of this solution can be simplified. To see this, write out the case for  $t = 1$ .

$$\varepsilon_1 + (.9)\varepsilon_0 + (.9)^2\varepsilon_{-1} + \dots - (.9)[\varepsilon_0 + (.9)\varepsilon_{-1} + \dots]$$

- Thus, we have

$$y_t = 30 + (.9)^t [y_0 - 30] + \sum_{i=0}^{t-1} (.9)^i \varepsilon_{t-i}$$

# Lag Operators and their Properties

$$L^i y_t \equiv y_{t-i}$$

- 1 The lag of a constant is a constant:  $Lc = c$ .
- 2 The **distributive law** holds:  
 $(L^i + L^j)y_t = L^i y_t + L^j y_t = y_{t-i} + y_{t-j}$ .
- 3 The **associative law** holds:  
 $L^i L^j y_t = L^i (L^j y_t) = L^i y_{t-j} = y_{t-i-j}$ .
- 4  $L$  raised to a negative number is the lead operator:  
 $L^{-i} y_t = y_{t+i}$ .
- 5 For  $|a| < 1$  the infinite sum  $(1 + aL + a^2 L^2 + a^3 L^3 + \dots)y_t$  converges to  $y_t / (1 - aL)$ .

# Lag Operators and their Properties

$$L^i y_t \equiv y_{t-i}$$

- Lag Operators allow us to write high-order difference equations,

$$(1 - a_1L - a_2L^2 + \dots a_pL^p)y_t = a_0 + (1 + b_1L - b_2L^2 + \dots b_qL^q)\varepsilon_t$$

$$A(L)y_t = a_0 + B(L)\varepsilon_t$$

as well as their particular solutions, compactly:

$$y_t = a_0/A(L) + B(L)\varepsilon_t/A(L).$$