#### Advanced Macroeconomics I Lecture 3 (2)

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## SUFE Spring 2011



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- The accumulation path will spend infinite time arbitrarily close to the steady state
- In a very small region a continuous function can be arbitrarily well approximated by a linear function, using the first-order Taylor expansion of the function
  - we will be able to approximate the speed of convergence arbitrarily well as time passes

 If the starting point is far from the steady state, we will make mistakes that might be large initially, but these mistakes will become smaller and smaller and eventually become unimportant

#### Local and global convergence

- The global convergence theorem applies only for the one-sector growth model
- Local convergence applies to a general dynamic system

#### Local dynamics

• The first-order Taylor series expansion of the decision rule gives

$$k' = g(k) \approx g(k^*) + g'(k^*)(k - k^*)$$
  
 $k' - k^* = g'(k^*)(k - k^*)$ 

interpret  $g'(k^*)$  as a measure of the rate of convergence

- stationary if  $g'(k^*) < 1$
- If g'(k\*) close to zero, convergence is fast
   ( gap decreases significantly each period)

Linearization for a general dynamic system

- The task is now to find  $g'(k^*)$
- Linearize the Euler equation
  - Lead to difference equation in  $k_t$
  - One of the solutions to this difference equation will be the one we are looking for



#### Local convergence cnditions

#### Theorem

Let  $x_t \in R^n$ . Given  $x_{t+1} = h(x_t)$  with a stationary point  $x^* : x^* = h(x^*)$  if

 h is continuously differentiable with Jacobian H(x\*) around x\*
 I - H(x\*) is non-singular

then there is a set of initial conditions  $x_0$ , of dimension equal to the number of eigenvalues of  $H(x^*)$  that are less than 1 in absolute value, for which  $x_t \rightarrow x^*$ (SUFE) Advanced Macro Spring 2011 7 / 31

#### Jacobian matrix

• First derivatives: it changes the value of  $x_t - x^*$ , but does not change the direction

$$\mathcal{H} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \dots & \frac{\partial y_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_m}{\partial x_1} & \dots & \frac{\partial y_m}{\partial x_n} \end{bmatrix}$$

•  $x_t - x^*$  is the eigenvector of H,  $H(x_t - x^*) = \lambda (x_t - x^*)$ , eigenvalues:  $|H - \lambda I| = 0$ 

#### Converging Dynamics - n=1

#### Examples

There is only one eigenvalue:  $k' - k^* = g'(k^*)(k - k^*): \lambda = g'(k^*)$ 1.  $|\lambda| \ge 1$ , no initial condition leads to  $x_t$ converging to  $x^*$ In this case, only for  $x_0 = x^*$  will the system stay in  $x^*$  if  $\lambda = 1$ 

2. 
$$|\lambda| < 1, x_t 
ightarrow x^*$$
 for any value of  $x_0$ 

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#### Converging Dynamics - n=2

#### Examples

There are two eigenvalues:  $\lambda_1$ ,  $\lambda_2$ 

$$k' - k^{*} = g_{1}(k^{*}, h^{*}) (k - k^{*}) + g_{2}(k^{*}, h^{*}) (h - h^{*}) h' - h^{*} = f_{1}(k^{*}, h^{*}) (k - k^{*}) + f_{2}(k^{*}, h^{*}) (h - h^{*})$$

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#### Converging Dynamics - n=2

$$\binom{k'-k^*}{h'-h^*} = \begin{bmatrix} g_1(k^*, h^*) & g_2(k^*, h^*) \\ f_1(k^*, h^*) & f_2(k^*, h^*) \end{bmatrix} \binom{k-k^*}{h-h^*}$$
1.  $|\lambda_1|$ ,  $|\lambda_2| \ge 1$ , no initial condition  $x_0 = \binom{k_0}{h_0}$ 
leads to  $x_t = \binom{k_t}{h_t}$  converging to  $x^* = \binom{k^*}{h^*}$ 
2.  $|\lambda_1| < 1$ ,  $|\lambda_2| \ge 1$ , dimention of  $x'_0s$  leading to convergence is 1
3.  $|\lambda_1| < 1$ ,  $|\lambda_2| < 1$ , dimention of  $x'_0s$  leading to convergence is 2,  $x_t \to x^*$  for any value of  $x_0$ 

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#### Interpret the eigenvalues

- Let the number of eigenvalues less than 1 in absolute value be denoted by *m*.
  - dimension of the set of initial  $x_0$  leading to  $x^*$
- Interpret m as the degrees of freedom: converge with any initial value x<sub>0</sub>
- Let the number of economic restrictions on initial conditions be denoted by  $\hat{m}$ 
  - the restrictions coming from physical conditions in our economic model

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- An interpretation: we have m̂ equations and m unknowns
- $m = \hat{m}$ : there is a unique convergent solution to the difference equation system
- **2**  $m < \hat{m}$ : No convergent solution obtains
- m > m̂: There is "indeterminacy", i.e. many solutions

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# Solving for the speed of convergence

 Derive the Euler equation:
 F(k<sub>t</sub>, k<sub>t+1</sub>, k<sub>t+2</sub>) = 0
 u' [f(k<sub>t</sub>) - k<sub>t+1</sub>] βu' [f(k<sub>t+1</sub>) - k<sub>t+2</sub>] f'(k<sub>t+1</sub>) = 0
 k\* is a steady state ⇔ F(k\*, k\*, k\*) = 0

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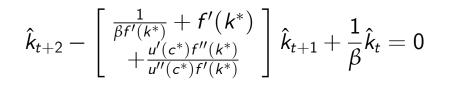
#### Linearize the Euler equation

Define  $\hat{k}_t = k_t - k^*$ , and using first-order Taylor approximation derive  $a_0$ ,  $a_1$ , and  $a_2$  such that

$$a_2 \hat{k}_{t+2} + a_1 \hat{k}_{t+1} + a_0 \hat{k}_t = 0$$

$$\beta u''(c^*) f'(k^*) \hat{k}_{t+2} - \begin{bmatrix} u''(c^*) + \beta u''(c^*) [f'(k^*)]^2 \\ + \beta u'(c^*) f''(k^*) \end{bmatrix} \hat{k}_{t+1} + u''(c^*) f'(k^*) \hat{k}_t = 0$$

#### Difference equation



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Write the Euler equation as a first-order system: define  $x_t = {\hat{k}_{t+1} \choose \hat{k}_t}$  and then

$$x_{t+1} = H x_t$$

$$egin{pmatrix} \hat{k}_{t+2} \ \hat{k}_{t+1} \end{pmatrix} = Higg( egin{matrix} \hat{k}_{t+1} \ \hat{k}_{t} \end{pmatrix}$$



#### A first-order system

$$egin{aligned} \mathcal{H} = \left(egin{array}{c} 1 + rac{1}{eta} + rac{u'(c^*)f''(k^*)}{u''(c^*)f'(k^*)} & -rac{1}{eta} \ 1 & 0 \end{array}
ight) \ \mathcal{B}f'(k^*) = 1 \end{aligned}$$

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#### Look for eigenvalues

- From the characteristic polynomial given by  $|H \lambda I| = 0$
- Decompose *H* as follows:

$$H = V\Lambda V^{-1} \Longrightarrow \Lambda = \left(\begin{array}{cc} \lambda_1 & 0\\ 0 & \lambda_2 \end{array}\right)$$

λ<sub>1</sub> and λ<sub>2</sub> are (distinct) eigenvalues of
 H; V is a matrix of eigenvectors of H

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#### A change of variables

• Given 
$$x_{t+1} = H x_t$$
, multiply by  $V^{-1}$ 

$$V^{-1}x_{t+1} = V^{-1}Hx_t = V^{-1}V\Lambda V^{-1}x_t$$
  
=  $\Lambda V^{-1}x_t$ 

• let 
$$z_t \equiv V^{-1}x_t$$
 and  $z_{t+1} = V^{-1}x_{t+1}$ , where  $z_t = {z_{1t} \choose z_{2t}}$ 

$$m{z}_{t+1} = \Lambda m{z}_t = \Lambda^t m{z}_0$$

• So  $z_{1t} = z_{10}\lambda_1^t$  and  $z_{2t} = z_{20}\lambda_2^t$ 

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• We can go back to  $x_t$  by premultiplying  $z_t$  by V :

$$x_t = V z_t = V \begin{pmatrix} z_{1t} \\ z_{2t} \end{pmatrix}$$

$$\begin{aligned} x_t &= z_{10}\lambda_1^t \begin{pmatrix} V_{11} \\ V_{21} \end{pmatrix} + z_{20}\lambda_2^t \begin{pmatrix} V_{12} \\ V_{22} \end{pmatrix} \\ &= \begin{pmatrix} \hat{k}_{t+1} \\ \hat{k}_t \end{pmatrix} \end{aligned}$$

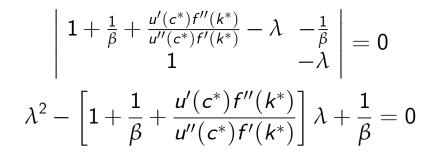
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#### Eigenvalues



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#### Eigenvalues

Let

$$egin{aligned} F(\lambda) &= \lambda^2 - \ & \left[1 + rac{1}{eta} + rac{u'(c^*)f''(k^*)}{u''(c^*)f'(k^*)}
ight]\lambda \ & + rac{1}{eta} \end{aligned}$$

 $F(\lambda)$  is a continuous function of  $\lambda$ 

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#### Eigenvalues

# $\left. \begin{array}{l} F(0) = \frac{1}{\beta} > 0 \\ F(1) = -\frac{u'(c^*)f''(k^*)}{u''(c^*)f'(k^*)} < 0 \\ F(\infty) = \infty > 0 \end{array} \right\}$

 $\Longrightarrow \lambda_1 < 1$ ,  $\lambda_2 > 1$ 



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#### A convergent solution

$$\begin{pmatrix} \hat{k}_{t+1} \\ \hat{k}_t \end{pmatrix} = z_{10}\lambda_1^t \begin{pmatrix} V_{11} \\ V_{21} \end{pmatrix} + z_{20}\lambda_2^t \begin{pmatrix} V_{12} \\ V_{22} \end{pmatrix}$$
$$\hat{k}_t = c_1\lambda_1^t + c_2\lambda_2^t$$

A convergent solution to the system requires  $c_2 = 0$ . The remaining constant,  $c_1$ , will be determined from

$$k_0-k^*\equiv \hat{k}_0=c_1\lambda_1^0=c_1$$

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$$k_t - k^* = \lambda_1^t (k_0 - k^*)$$
  
 $k_{t+1} - k^* = \lambda_1 (k_t - k^*)$ 

The eigenvalue  $\lambda_1$  has a particular meaning: it measures the (inverse of the) rate of convergence to the steady state

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#### Empirical Evidence of Convergence

- Whether poor conutries tend to grow faster than rich countries?
  - Baumol (1986) examines convergence from 1870 to 1979 among 16 industralized countries

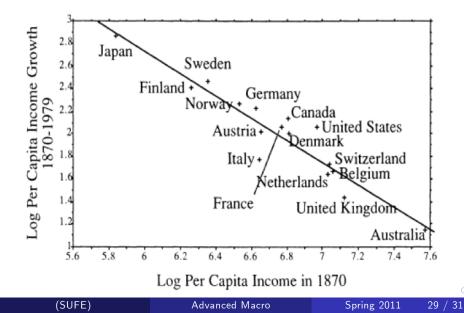
$$\ln\frac{y_{i,t}}{y_{i,0}} = a + b \ln y_{i,0} + e_i$$

a = 8.457 and b = -0.995 $R^2 = 0.87$ 

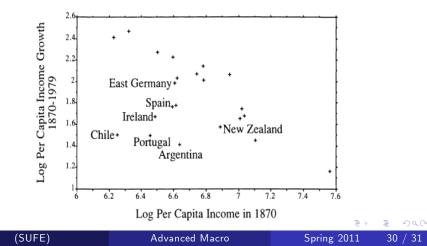
## • b = -1 corresponds to perfect convergence:

- higher initial income on average lowers subsequent growth one-for-one
- and so output per capita in 1979 is uncorrelated with its value in 1870

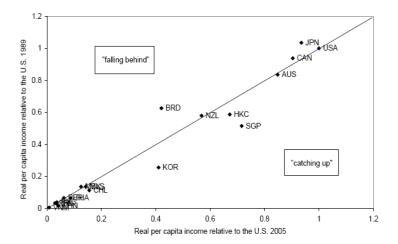




Delong (1988) criticizes: sample selection (only industrialized countries in the sample, while they were poor 100 years ago)



#### Catching up



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