

Lecture 13

Endogenous variables (n)

$$x_t = \begin{bmatrix} z_t \\ p_t \end{bmatrix} \rightarrow \begin{matrix} \text{state (predetermined)} \\ \text{at } t \end{matrix}$$

$n \times 1$

exogenous variables (r)

$$u_t$$

$r \times 1$

Under perfect foresight

$$x_{t+1} = A \cdot x_t + B \cdot u_t$$

Solving forwards:

SOLUTION

$$x_{t+k} = \underbrace{A^k}_{\text{solution of homogeneous}} x_t + \sum_{j=0}^{k-1} \underbrace{A^{k-1-j}}_{\text{particular solution}} B u_{t+j}$$

At time $t \rightarrow z_t$ is known
 $\rightarrow p_t$ is not known

$$x_{t+1} = A \cdot x_t + B \cdot u_t$$

$n \times 1$ $n \times n$ $n \times 1$ $n \times r$ $r \times 1$

n eigenvalues of A

- l unstable $|z_i| > 1$ $i=1, \dots, l$
- $n-l$ stable $|z_i| < 1$ $i=l+1, \dots, n$

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Assume $A = T^{-1} \cdot \Lambda \cdot T$, $\Lambda = \text{diag}(z_1, z_2, \dots, z_n)$

Partition: $\Lambda = \begin{bmatrix} z_1 & & \emptyset \\ & z_2 & \\ \emptyset & & z_n \end{bmatrix} = \begin{bmatrix} \Lambda_1 & \emptyset \\ \emptyset & \Lambda_2 \end{bmatrix}$

$\Lambda_1 \rightarrow$ stable eigenvalues $(n-l) \times (n-l)$
 $\Lambda_2 \rightarrow$ unstable $l \times l$

matrix of eigenvector $T = \begin{bmatrix} T_1 \\ T_2 \end{bmatrix}$

$(n-l) \times n$ $l \times n$ $n \times n$

$$T^{-1} = \begin{bmatrix} S_1 & S_2 \end{bmatrix}$$

$n \times (n-l)$ $n \times l$ $n \times n$

Solution: $x_{t+k} = A^k x_t + \sum_j A^{k-1-j} B u_{t+j}$

$$= T^{-1} \cdot \begin{bmatrix} \Lambda_1^k & \emptyset \\ \emptyset & \Lambda_2^k \end{bmatrix} \cdot \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} x_t + T^{-1} \cdot \left\{ \sum_j \begin{bmatrix} \Lambda_1^{k-1-j} & \emptyset \\ \emptyset & \Lambda_2^{k-1-j} \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \end{bmatrix} \cdot B \cdot u_{t+j} \right\}$$

$$\underbrace{A^k}_{\text{particular solution}} = T^{-1} \Lambda^k T = [S_1 \ S_2] \cdot \begin{bmatrix} \Lambda_1^k & \emptyset \\ \emptyset & \Lambda_2^k \end{bmatrix} \cdot \begin{bmatrix} T_1 \\ T_2 \end{bmatrix}$$

$$= T^{-1} \begin{bmatrix} z_1^k & & \emptyset \\ & z_2^k & \\ \emptyset & & z_n^k \end{bmatrix} T = \left. \begin{matrix} = S_1 \Lambda_1^k T_1 + S_2 \Lambda_2^k T_2 \end{matrix} \right\} =$$

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$n \times 1$

exogenous variables (r)

$$u_t$$

$r \times 1$

Under perfect foresight

$$x_{t+1} = A \cdot x_t + B \cdot u_t$$

Solving forwards:

SOLUTION

$$x_{t+k} = A^k \cdot x_t + \sum_{j=0}^{k-1} A^{k-1-j} \cdot B \cdot u_{t+j}$$

solution of homogeneous

particular solution

At time $t \rightarrow z_t$ is known

$\rightarrow p_t$ is not known

$$x_{t+1} = A \cdot x_t + B \cdot u_t$$

$n \times 1$ $n \times n$ $n \times 1$ $n \times r$ $r \times 1$

n eigenvalues of A

$\begin{cases} l \text{ unstable } | \lambda_i | > 1 \\ i=1, \dots, l \\ n-l \text{ stable } | \lambda_i | < 1 \\ i=l+1, \dots, n \end{cases}$

Therefore

$$x_{t+k} = [S_1 \Lambda_1^k T_1 + S_2 \Lambda_2^k T_2] \cdot x_t + \sum_j [S_1 \Lambda_1^{k-1-j} T_1 + S_2 \Lambda_2^{k-1-j} T_2] \cdot B \cdot u_{t+j}$$

$$\Rightarrow x_{t+k} = S_1 \Lambda_1^k \left\{ T_1 \cdot x_t + \sum_j \Lambda_1^{-1-j} T_1 \cdot B \cdot u_{t+j} \right\} + S_2 \Lambda_2^k \left\{ T_2 \cdot x_t + \sum_j \Lambda_2^{-1-j} T_2 \cdot B \cdot u_{t+j} \right\}$$

$\lim_{k \rightarrow \infty}$

$\lim_{k \rightarrow \infty} \Lambda_2^k \rightarrow 0$ (I) $\rightarrow 0$

$\lim_{k \rightarrow \infty} \Lambda_1^k \rightarrow \infty$, (II) $\rightarrow \infty$

Unless $\{ T_2 \cdot x_t + \sum_j \Lambda_2^{-1-j} T_2 \cdot B \cdot u_{t+j} \} = 0$

In this case (II) = 0

Partition: $\Lambda = \begin{bmatrix} \Lambda_1 & \emptyset \\ \emptyset & \Lambda_2 \end{bmatrix}$

matrix of eigenvector $T = \begin{bmatrix} T_1 \\ T_2 \end{bmatrix}$

$n \times n$ $n \times n$

$T^{-1} = \begin{bmatrix} S_1 & S_2 \end{bmatrix}$

$n \times (n-l)$ $n \times l$ $n \times n$

$$A^k = T^{-1} \Lambda^k T = T^{-1} \begin{bmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_n^k \end{bmatrix} T = T^{-1} \begin{bmatrix} \Lambda_1^k & \emptyset \\ \emptyset & \Lambda_2^k \end{bmatrix} T$$

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endogenous variables (n)

$$x_t = \begin{bmatrix} z_t \\ p_t \end{bmatrix} \rightarrow \begin{array}{l} \text{State (predetermined)} \\ \text{at } t \end{array}$$

$$n \times 1 \quad \rightarrow m \text{ control (jump / free to choose at time } t)$$

exogenous variables (r)

$$u_t$$

$$r \times 1$$

$$x_{t+1} = A \cdot x_t + B \cdot u_t$$

$$n \times 1 \quad n \times n \quad n \times r \quad n \times 1 \quad r \times 1$$

λ_i
n eigenvalues
of A

l unstable $|\lambda_i| > 1$
 $i = 1, \dots, l$
 $n-l$ stable $|\lambda_i| < 1$
 $i = l+1, \dots, n$

$$T_2 = \begin{bmatrix} T_{21} & T_{22} \end{bmatrix}$$

$$T = \begin{bmatrix} T_1 & \\ & T_2 \end{bmatrix} = \begin{bmatrix} T_{11} & T_{12} \\ T_{21} & T_{22} \end{bmatrix}$$

$$T = \begin{bmatrix} T_{11} & T_{12} \\ T_{21} & T_{22} \end{bmatrix} = \begin{bmatrix} T_1 & \\ & T_2 \end{bmatrix}$$

$$\begin{matrix} (n-l) \times l & (n-l) \times l \\ l \times (n-m) & l \times m \\ (n-l) \times n \\ l \times n \end{matrix}$$

(3)

So for stability, I need

$$T_2 \cdot x_t + \sum \Lambda_2^{-1-j} T_2 B u_{t+j} = 0 \Leftrightarrow$$

$$\Leftrightarrow T_2 \cdot x_t = - \sum_{j=0}^{\infty} \Lambda_2^{-1-j} T_2 B u_{t+j} \Leftrightarrow$$

$$\Leftrightarrow T_2 \cdot x_t = - \Lambda_2^{-1} \cdot \sum_{j=0}^{\infty} \Lambda_2^{-j} T_2 B u_{t+j} \Leftrightarrow$$

$$\Leftrightarrow T_{21} z_t + T_{22} p_t = - \Lambda_2^{-1} \sum_{j=0}^{\infty} \Lambda_2^{-j} T_2 B u_{t+j} \Leftrightarrow T_{22} \text{ to be a square matrix}$$

$$\Leftrightarrow T_{22} p_t = - T_{21} z_t - \Lambda_2^{-1} \sum_{j=0}^{\infty} \Lambda_2^{-j} T_2 B u_{t+j}$$

I need to be able to define T_{22}^{-1}

$$\begin{matrix} T_{22} \cdot p_t \\ l \times m & m \times 1 \\ l = m \text{ for} \end{matrix}$$

For T_{22} to be a square matrix

$l = m$
Saddle path stability condition

unstable roots = # control (free/jump) variables

$$p_t = - (T_{22}^{-1} T_{21}) z_t - T_{22}^{-1} \Lambda_2^{-1} \sum_{j=0}^{\infty} \Lambda_2^{-j} T_2 B u_{t+j}$$

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(4)

$$y_t = N_1 \cdot y_{t-1} + N_2 \cdot y_{t-2} + N_3 \cdot y_{t-3} + N_4 \cdot y_{t-4} + \dots + N_k \cdot y_{t-k}$$

k -th order linear difference equation

equivalent to a system of k first order d.e.

Define $(k-1)$ auxiliary variables

$$x_{1t} = y_{t-1}$$

$$x_{2t} = y_{t-2} = x_{1,t-1}$$

$$x_{3t} = y_{t-3} = x_{2,t-1}$$

and so on

⋮

$$x_{k-1,t}$$

$$x_t = \begin{bmatrix} y_t \\ x_{1t} \\ x_{2t} \\ \vdots \\ x_{k-1,t} \end{bmatrix}$$

$$x_t = N \cdot x_{t-1}$$

$$= \begin{bmatrix} N_1 & N_2 & \dots & N_k & N_n \\ 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & 0 \end{bmatrix} \cdot \begin{bmatrix} y_{t-1} \\ x_{1,t-1} \\ x_{2,t-1} \\ \vdots \\ x_{k-1,t-1} \end{bmatrix}$$

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Phase Diagrams

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

eigenvalues $\lambda_1 = 1$
 $\lambda_2 = -1$
 eigenvectors $\lambda_1 \rightarrow [1, 0]$
 $\lambda_2 \rightarrow [1, -1]$

$$\dot{x}_i \equiv \frac{dx_i}{dt}$$

$$\dot{x} = A \cdot x + b$$

$$\dot{x}_1 = 0 \Leftrightarrow x_1 + 2x_2 + 1 = 0 \Leftrightarrow x_1 = -\frac{1}{2} - \frac{1}{2}x_2$$

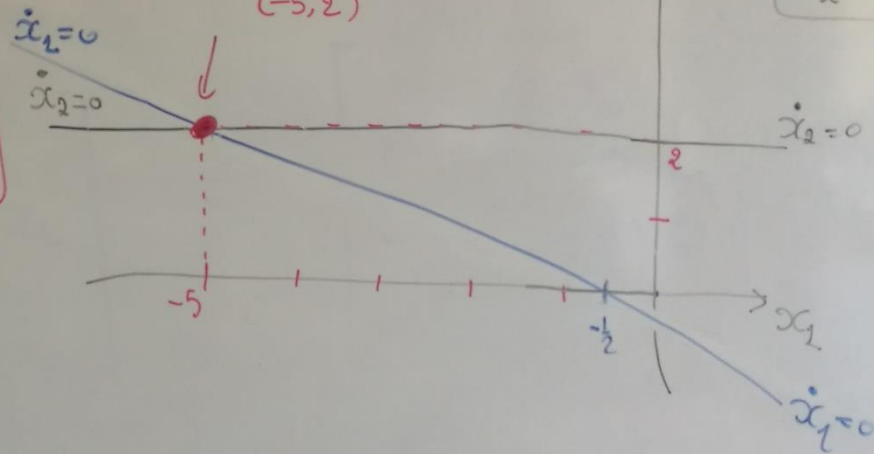
$$\dot{x}_2 = 0 \Leftrightarrow -x_2 + 2 = 0 \Leftrightarrow x_2 = 2$$

Define the **equilibrium** point $\dot{x}_1 = \dot{x}_2 = 0$

$$2 = -\frac{1}{2} - \frac{1}{2}x_1$$

$$\Leftrightarrow \boxed{x_1^* = -5}$$

$$\boxed{x_2^* = 2}$$



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Phase Diagrams

If I find myself in region
 α , c
 NEVER move
 towards $(-5, 2)$

b or d in general
 move away from
 equilibrium
 unless
 I am in some special
 position

$$\dot{x}_1 = 0 \Leftrightarrow x_1 + 2x_2 + 1 = 0 \Leftrightarrow x_2 = -\frac{1}{2} - \frac{1}{2}x_1 \quad \textcircled{b}$$

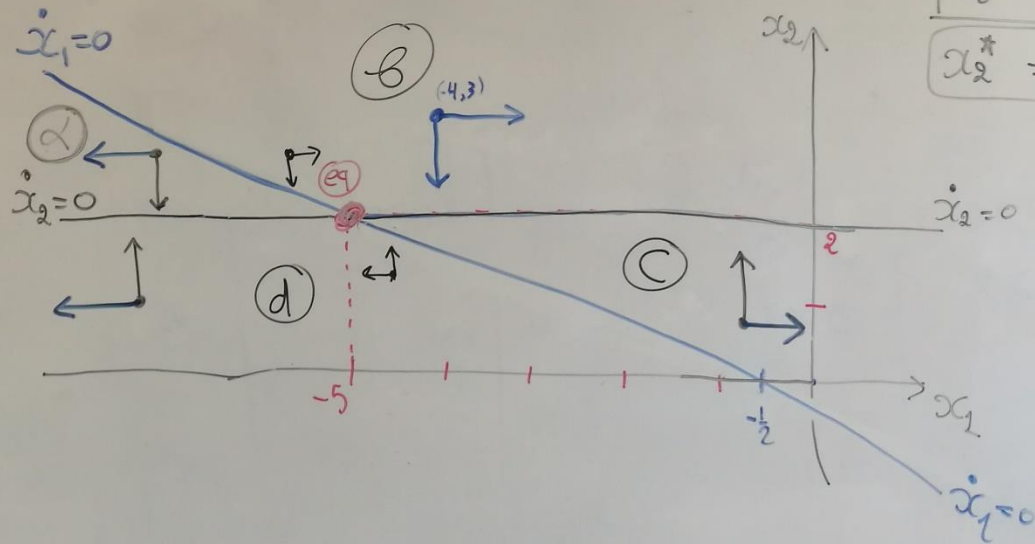
$$\dot{x}_2 = 0 \Leftrightarrow -x_2 + 2 = 0 \Leftrightarrow x_2 = 2$$

Define the equilibrium point $\dot{x}_1 = \dot{x}_2 = 0$

$$2 = -\frac{1}{2} - \frac{1}{2}x_1 \Leftrightarrow$$

$$\boxed{x_1^* = -5}$$

$$\boxed{x_2^* = 2}$$



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Set $K_s=0$
 ↓
 unstable manifold

$$\dot{x} = A \cdot x + b$$

$$\lambda_s < 0, v^s$$

$$\lambda_u > 0, v^u$$

eigenvalues eigenvectors

x^e particular solution

Solution

$$x(t) = K_s \cdot v^s \cdot e^{\lambda_s t} + K_u \cdot v^u \cdot e^{\lambda_u t} + x^e$$

unstable unless $K_u=0$ } $x(t) = K_s \cdot v^s \cdot e^{\lambda_s t} + x^e$

$$\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} - \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = K_s \cdot \begin{pmatrix} v_1^s \\ v_2^s \end{pmatrix} \cdot e^{\lambda_s t}$$

$$\begin{aligned} x_1(t) - \bar{x}_1 &= K_s \cdot v_1^s \cdot e^{\lambda_s t} \\ x_2(t) - \bar{x}_2 &= K_s \cdot v_2^s \cdot e^{\lambda_s t} \end{aligned}$$

Stable Manifold

(7)

If I start from a point $(\tilde{x}_1, \tilde{x}_2)$ on this line then my path will converge to x^e

$$\frac{x_1(t) - \bar{x}_1}{x_2(t) - \bar{x}_2} = \frac{v_1^s}{v_2^s}$$

all other unstable paths = integral curves

$$\lambda^s = -1, v^s = [1, -1]$$

$$\lambda^u = 1, v^u = [0, 1]$$

In the example:

$$\frac{x_1 + 5}{x_2 - 2} = -1$$

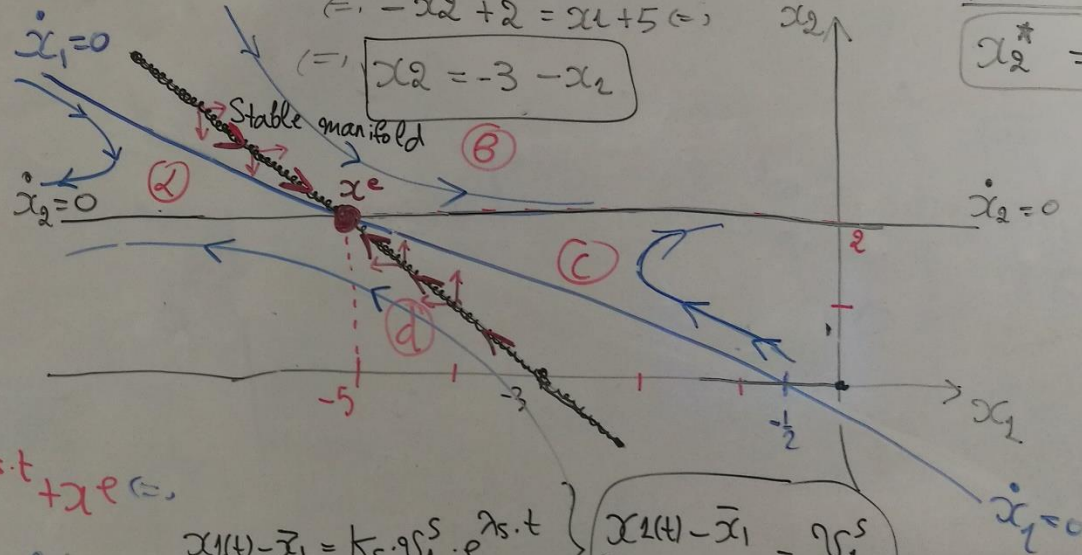
$$x_1^* = -5$$

$$x_2^* = 2$$

$$\Leftrightarrow -(x_2 - 2) = x_1 + 5$$

$$\Leftrightarrow -x_2 + 2 = x_1 + 5$$

$$\Leftrightarrow x_2 = -3 - x_1$$



$$\frac{x_1(t) - \bar{x}_1}{x_2(t) - \bar{x}_2} = \frac{v_1^s}{v_2^s}$$