

①

# DIFFERENTIAL EQUATIONS.

What's the difference?

$l_0, r_0$     $l_1$     $r_2$     $l_3$     $r_3$    ...

left and right feet.

time DISCRETE.

$h(t)$  — continuous fun.  
of time.

$$l_{2m+1} = \lambda(r_{2m})$$

$$r_{2m+2} = p(l_{2m+1})$$

↓  
coupled equations.

DIFFERENCE equation:

①

$$x_{t+1} - x_t = \Delta x_t = d_t(x_t).$$

Recurrence relation

$$x_{t+1} = r_t(x_t).$$

$x_0, x_1, \dots, x_T.$

as functions of  $x_0$ .  $t=0, \dots, T-1$

LINEAR CASE.

$$x_{t+1} - x_t = d_t \quad (t=0, \dots, T-1).$$

T equations in  $T+1$  unknowns.

$(x_0, x_1, \dots, x_T).$

Rank T  
(check).  
(upper triangular)

One degree of freedom.

BUT. if an initial condition is given ( $x_0$ )

or a terminal condition ( $x_T$ ).

then

$$x_t = \cancel{x_0} + \sum_{s=0}^{t-1} d_s.$$

or

$$x_t = x_T - (\text{exercise}).$$

2

Wealth equation (in discrete time).

$$w_{t+1} = (1 + r_t)(w_t + y_t - e_t).$$

rate of interest. income expenditure

$$w_{t+1} = p_t (w_t - x_t)$$

gross return

net expenditure.

$$= p_t (w_t + s_t)$$

net saving.

Interest compounded:

$$R_t = \prod_{k=0}^{t-1} (1 + r_k) = \prod_{k=0}^{t-1} p_k.$$

gross return factor

$(1+r)^t$  when  $r_t = r$  (all t).

Discounted wealth:  $R_{t+1} = (1+r_t)R_t.$

$$w_t^d = w_t / R_t.$$

$$w_{t+1}^d = \frac{w_{t+1}}{R_{t+1}} = \frac{(1+r_t)(w_t - x_t)}{R_{t+1}} = \frac{1}{R_t} (w_t - x_t)$$

Present discounted

~~$w_t$~~

$$s_t := x_t / R_t.$$

$$R_0 = 1.$$

value of future expenditure.

$$w_{t+1} = w_t - s_t$$

$$w_t = w_0 - \sum_{k=0}^{t-1} s_k$$

product of zero terms

$$w_t = R_t w_0 - \sum_{k=0}^{t-1} \prod_{j=k}^{t-1} R_j s_k$$

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# VECTOR Linear difference equation in $n$ variables

$$x_{t+1} = A_t x_t + d_t.$$

$\uparrow$   $n$  vector                       $n \times n$  matrix.                       $\uparrow$   $n$  vector.

Soln:  $x_t = P_{t,0} x_0 + \sum_{k=0}^{t-1} P_{t,k} d_{k-1}$

When  $t=0$ ,  $P_{0,0} = I$ ,  ~~$P_{t,k}$~~

~~When  $t=1$ ,  $P_{1,0} = I$ ,  $P_{1,1} = A_1$~~

$x_1 = A_0 x_0 + d_0$      $P_{1,0} = A_0$ ,  $P_{1,1} = I$ .

$x_2 = A_1 x_1 + d_1 = A_1 A_0 x_0 + A_1 d_0 + d_1$

$P_{2,0} = A_1 A_0$ ,  $P_{2,1} = A_1$ ,  ~~$P_{2,2} = I$~~

$P_{2,2} = I$ .

$P_{t,0} = \prod_{s=0}^{t-1} A_s = A_{t-1} A_{t-2} \dots A_0$

$P_{t,k} = A_{t-1} A_{t-2} \dots A_k$

$P_{t,t} = I$ .

4

Constant coefficients

$$x_{t+1} = Ax_t + d_t.$$

$$x_{t+1} = A^t x_0 + \sum_{k=1}^t A^{t-k} d_k$$

$$A^0 = I.$$

Autonomous

Constant right hand side

$$x_{t+1} = Ax_t + d$$

$$(x_{t+1} = x_t)$$

Steady state  $x^*$ ,  $x^* = Ax^* + d.$

$$x^* = (I - A)^{-1} d \quad (y = (I - A)^{-1} d)$$

Condition for stability

INSERT 4

$$x_t \rightarrow x^* \text{ as } t \rightarrow \infty.$$

~~Generally~~ Requires  $A^t d \rightarrow 0$  as  $t \rightarrow \infty.$

~~$A^t x_0 \rightarrow 0$  as  $t \rightarrow \infty.$~~   
 $A^t \rightarrow 0$  as sufficient condition for all  $d.$

Characteristic roots

$$|A - \lambda I| = 0.$$

Each eigenvalue has modulus  $< 1$  (even if complex).

$$|\lambda + i\mu| = \sqrt{\lambda^2 + \mu^2} \text{ - distance from origin in Argand diagram}$$

(4 bis)

$$x_{t+1} = Ax_t + d$$

$$\begin{aligned} \text{soln: } x_t &= A^t x_0 + \sum_{k=1}^t A^{t-k} d \\ &= A^t x_0 + S_t d \end{aligned}$$

$$\begin{aligned} \text{where } S_t &= I + A + \dots + A^{t-1} \\ &= \sum_{k=1}^t A^{t-k} \end{aligned}$$

geometric series  
in the

$$1 + a + a^2 + \dots + a^{t-1} = \frac{1 - a^t}{1 - a}$$

square matrix  $A$

(multiply ~~eq~~ each side by  $(1 - A)$ ).

$$(I - A)S_t = \cancel{I} + A - A + A^2 - A^2 \dots - A^t$$

$$= I - A^t$$

assuming

$$S_t = (I - A)^{-1} (I - A^t) \quad (I - A)^{-1} \text{ exists.}$$

$$x_t = A^t x_0 + (I - A)^{-1} (I - A^t) d.$$

PROVIDED  $(I - A)^{-1}$  exists.

$$x_t \rightarrow (I - A)^{-1} d = x^* \text{ as } t \rightarrow \infty.$$

provided  
 $A^t \rightarrow 0.$

# EXAMPLE.

4-ter.

$$A^t d \rightarrow 0$$

$$\Rightarrow A^t \rightarrow 0$$

$$A = \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & 2 \end{pmatrix}, \quad d = (1, 0)$$

$$A^t = \begin{pmatrix} 2^{-t} & 0 \\ 0 & 2^t \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \text{ as } t \rightarrow \infty$$

$$A^t d = \begin{pmatrix} 2^{-t} & 0 \\ 0 & 2^t \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2^{-t} \\ 0 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix} \text{ as } t \rightarrow \infty.$$

BUT  $A^t d = 0$  for all  $d \Rightarrow A^t \rightarrow 0$ .

$A^t d \rightarrow 0$  for  $d = e_j$  (j-th unit vector)  $\Rightarrow$  just suff. of  $A^t \rightarrow 0$

SUPPOSE

A has n distinct eigenvalues (real or complex)

A may not be symmetric.

Then the eigenvectors form a (complex) matrix P such that

$$P^* A P = \text{diag}(\lambda_1, \dots, \lambda_n)$$

↑  
adjoint matrix
eigenvalues  
(real or complex)

$P_{jk}^*$  is the complex conjugate of  $P_{ji}$ .

$$p_{jk} + i q_{jk} = p_{kj} - i q_{kj}$$

Take ~~conjugate complex~~ of transpose.  
complex conjugate

$$P^* P = I \quad (\text{Spectral Theorem})$$



$$x_{t+1} = Ax_t + d$$

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~~Define  $y_t = Px_t$ , ?~~

Define  $x_t = Py_t$ .

~~$P^* x_{t+1} =$~~

$$y_t = P^* x_t$$

$$y_{t+1} = P^* x_{t+1} = P^* A P y_t + P^* d$$

$$\Rightarrow y_{t+1} = \text{diag}(\lambda_1, \dots, \lambda_n) y_t + P^* d$$

UNCOUPLED equations

$$y_{t+1}^j = \lambda_j y_t^j + (P^* d)_j$$

$n$  separated equations

STABILITY iff  $\lambda_j^t \rightarrow 0$   
as  $t \rightarrow \infty$  (all  $j$ )

$$\Leftrightarrow |\lambda_j| < 1.$$

A diagonalizable matrix is stable  
( $A^t \rightarrow 0$ ) iff its eigenvalues  
are all  $< 1$  in absolute value  
(modulus)

Also for non-diagonalizable matrices.

Second-order equation:  
(linear, constant coefficients, homogeneous), (7)

$$x_{t+1} + a x_t + b x_{t-1} = 0$$

Define  $y_t = x_{t-1}$

Then 
$$\left. \begin{aligned} x_{t+1} &= -a x_t - b y_t \\ y_{t+1} &= x_t \end{aligned} \right\}$$

Coupled pair of 1st order equations

$$\begin{pmatrix} x_{t+1} \\ y_{t+1} \end{pmatrix} = \begin{pmatrix} -a & -b \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_t \\ y_t \end{pmatrix}$$

A.

Instead of treating coupled pair, consider 2nd equation directly.

Guess <sup>that</sup> ~~when~~  $x_t = \lambda^t x_0$  could be a solution, for suitable  $\lambda$ ,  $x_0$ .

$$\lambda^{t+1} x_0 + a \lambda^t x_0 + b \lambda^{t-1} x_0 = 0.$$

Cancel  $\lambda^{t-1} x_0$  ( $\neq 0$ , assume).

$$\lambda^2 + a \lambda + b = 0.$$

Roots  $\lambda_1, \lambda_2$ . (Case of distinct roots,  $\lambda_1 \neq \lambda_2$ ).

Possible  
~~Bar~~ Solutions :

(8)

$$x_t = \lambda_1^t, \quad x_t = \lambda_2^t. \quad \text{if } r_0 = 1.$$

Two degrees of freedom.

General sol<sup>n</sup>:  $x_t = \alpha \lambda_1^t + \beta \lambda_2^t.$

COINCIDENT ROOTS.

If  $\lambda_1 = \lambda_2 = \lambda$ , sol<sup>n</sup>s  $x_t = \lambda^t, t\lambda^t.$

General sol<sup>n</sup>:  $x_t = (\alpha + \beta t) \lambda^t$

Two degrees of freedom.

to HOMOGENEOUS equation

$$x_{t+1} + ax_t + bx_{t-1} = 0$$

INHOM. eq., general RHS,  $d_t.$

Find a particular solution  $x_t^P.$

General sol<sup>n</sup>  $x_t = x_t^P + \text{sol}^n \text{ to homog. equation.}$

two degrees of freedom.

LOCAL STABILITY of <sup>AUTONOMOUS</sup> non-linear system of difference equations:

$$x_{t+1} = F(x_t). \quad F: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

stationary state, if it exists, ~~must~~ must be a fixed point of F.

$$x^* = F(x^*)$$

$$x_{t+1} - x^* = F(x_t) - F(x^*) \approx F'(x^*) (x_t - x^*) \quad (\text{if } x_t \rightarrow x^*)$$

$(n \times n)$  - matrix of derivatives  $\frac{\partial F_i}{\partial x_j}$ .  
**JACOBIAN matrix.**

**LOCAL** Stability requires  $F'(x^*)$  to be a stable matrix, (all eigenvalues have modulus or absolute value  $< 1$ ).

$$\underline{x}_{t+1} = A \underline{x}_t$$

$$\underline{x}_{t+1} - A \underline{x}_t = \underline{0}$$

homogeneous equation in  $n$  variables

$\uparrow$   
constant  $n \times n$  matrix

General solution: (?)

$$\underline{x}_t = \sum_{j=1}^n c_j \lambda_j^t$$

sum of  $n$  linearly independent power solutions.

$$c_j \lambda_j^t$$

When are these  $n$  solutions linearly independent?

Answer: iff  $\sum_{j=1}^n c_j \lambda_j^t \equiv 0 \Rightarrow$  each  $c_j = 0$

$\iff$  all  $\lambda_j$  different ( $n$  distinct roots).

When is  $\sum_{j=1}^n c_j \lambda_j^t$  a solution??

A2

(of  $x_{t+1} = A x_t$ ). vector eqn.

Answer: iff

~~$$\sum_{j=1}^n c_j \lambda_j^{t+1} - A \sum_{j=1}^n c_j \lambda_j^t \equiv 0$$~~

~~$$\sum_{j=1}^n c_j (\lambda_j^{t+1} - A \lambda_j^t)$$~~

When is  $c_j \lambda_j^t$  a solution?

Answer: iff  $c_j \lambda_j^{t+1} - A c_j \lambda_j^t \equiv 0$

~~or (assuming  $c_j \neq 0$ )~~

~~$$c_j \lambda_j^t (\lambda_j - A) = 0$$~~

iff  $\lambda_j = 0$  or  $A c_j = \lambda_j c_j$

either  $c_j = 0$  or  $\lambda_j$  an eigenvalue and  $c_j$  and eigenvector.

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Providing the eigenvalues  $\lambda_j$  ( $j=1, \dots, n$ ) are distinct (possibly one  $\lambda_j = 0$ ), the general solution is  $x_t = \sum_{j=1}^n c_j \lambda_j^t$ , where  $c_j$ 's are eigenvectors.

What if eigenvalues are not distinct?

Example  $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$

$$|A - \lambda I| = \begin{vmatrix} a_{11} - \lambda & a_{12} \\ a_{21} & a_{22} - \lambda \end{vmatrix}$$

$$= (a_{11} - \lambda)(a_{22} - \lambda) - a_{12}a_{21}$$

$$= |A| - \lambda \text{tr} A + \lambda^2$$

$$= \left(\lambda - \frac{a_{11} + a_{22}}{2}\right)^2 = \lambda^2 - 2\lambda \frac{\text{tr} A}{2} + \frac{1}{4} (\text{tr} A)^2$$

$$\lambda_1 = \frac{1}{2} \text{tr} A = \frac{1}{2} (a_{11} + a_{22})$$

$$\lambda_1^2 = \det A.$$

~~$$x_{t+1} = a_{11}x_t + a_{12}y_t$$~~

~~$$y_{t+1} = a_{21}x_t + a_{22}y_t.$$~~

Make  $a_{12} = a_{21} = 0$  (diagonal).

$$a_{11} = a_{22} = a.$$

~~$$x_{t+1} = ax_t,$$~~

~~$$y_{t+1} = ay_t.$$~~

AK

~~$x_{t+1} - 2\lambda x_t +$~~

$x_{t+1} - a x_t + b x_{t-1} = 0.$

$x_t = c \lambda^t,$

$c \lambda^{t+1} + a c \lambda^t + b c \lambda^{t-1} = 0$

If  $c \neq 0, \lambda \neq 0,$   $\lambda^2 + a\lambda + b = 0$   
( $\div c \lambda^{t-1}$ )

Two roots,  $\lambda_1, \lambda_2.$   
All well if  $\lambda_1 \neq \lambda_2$

general sol<sup>n</sup>:  $x_t = c_1 \lambda_1^t + c_2 \lambda_2^t.$

BUT, what if  $\lambda_1 = \lambda_2 = \lambda^*$ ?

$x_{t+1} - 2\lambda^* x_t + \lambda^{*2} x_{t-1} = 0.$

$(\lambda^2 - 2\lambda^* \lambda + \lambda^{*2} = 0)$   
 $(\lambda - \lambda^*)^2 = 0.$

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Put  $y_t = \lambda^{*-t} x_t$  ( $x_t = \lambda^{*t} y_t$ )  
are sol<sup>n</sup>.

~~$x_{t+1} + a$~~   
 $(\lambda^*)^{-t-1} y_{t+1} - 2\lambda^* (\lambda^*)^{-t} y_t + (\lambda^{*2}) (\lambda^*)^{-t-1} y_{t-1} = 0$

If  $\lambda^* \neq 0,$   $y_{t+1} - 2y_t + y_{t-1} = 0.$



Put  $z_t = y_{t+1} - y_t$ .

(A5)

Then  $z_t - z_{t-1} = 0$ .

So

$z_t = c, y_t = \alpha + \beta t$

$x_t = y_t \lambda^{*t} = (\alpha + \beta t) \lambda^{*t}$

lin. in  $t$ .

Solution with a double eigenvalue.

With a triple eigenvalue,

$x_t = (\alpha + \beta t + \gamma t^2) \lambda^{*t}$

COMPLEX EIGENVALUES.

~~$x_t = (\alpha + i\beta) (\lambda + i\mu)^t + (\gamma + i\delta) (\lambda - i\mu)^t$~~

~~Complex conjugates eigenvalues~~

$\lambda_1, \lambda_2 = \lambda \pm i\mu$

$x_t = (\alpha + i\beta) (\lambda + i\mu)^t + (\gamma + i\delta) (\lambda - i\mu)^t$

$\lambda + i\mu = r e^{i\theta} = r(\cos\theta + i\sin\theta)$   
 $\lambda - i\mu = r e^{-i\theta} = r(\cos\theta - i\sin\theta)$

DE MOIVRE.

$$x_t = (\alpha + i\beta) r^t e^{it\theta} + (\gamma + i\delta) r^t e^{-it\theta}$$

$$= (\alpha + i\beta) r^t (\cos t\theta + i \sin t\theta) + (\gamma + i\delta) r^t (\cos t\theta - i \sin t\theta)$$

$$= (\alpha + \gamma) r^t \cos t\theta - (\beta - \delta) r^t \sin t\theta$$

$i r^t \beta \cos t\theta + (\beta + \delta) \cos t\theta + (\alpha - \gamma) \sin t\theta$   
 ~~$i r^t \beta \cos t\theta$~~  + complex terms (=0)

$$\beta + \delta = 0, \quad \alpha - \gamma = 0$$

$$= 2\alpha r^t \cos t\theta - 2\beta r^t \sin t\theta$$

$$\rightarrow 0 \iff |r| < 1$$

$$\rightarrow \infty \iff |r| > 1$$

$$\text{circles} \iff |r| = 1$$