

# Difference Equations and Dynamic Optimization

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# Fixed Point

- A **Fixed Point** of a function  $f(x)$  is a point  $x_0$  such that  $f(x_0) = x_0$ .
- $f(x) = x + 1$  doesn't have any fixed point,  $f(x) = .4x + 1$  has a unique fixed point, and  $f(x) = x^3$  has multiple fixed points.
- A fixed point is stable (unstable) if after a perturbation, the variable moves back to (away from) the fixed point.
- $f(x) = 0.4x + 1$  has a stable FP, but  $f(x) = 2x + 1$  has an unstable one.

# Log-Linearization

- Using the first order Taylor expansion around the steady state (a point at which  $x_{t+1} = x_t \ \forall t$ ), one may study a nonlinear system of equations with a linear approximation.
- Recall  $f(x) = f(x^*) + f'(x^*)(x - x^*) + \dots$
- $\log(x) \approx \log(x^*) + \frac{1}{x^*}(x - x^*)$
- $\log\left(\frac{x}{x^*}\right) \approx \frac{x - x^*}{x^*}$

# Eigenvalues and Eigenvector

- Log-Linearization allows us to focus on linear transformations.
- In particular, we are interested in the principal axes associated with a linear transformation; the vectors that are only rescaled by the transformation  $T$ :

$$T(v) = \lambda v$$

Where the vector  $v$  and the scalar  $\lambda$  are called eigenvector and eigenvalue, respectively.

# Eigenvalues and Eigenvector

- Let  $A$  be a square matrix. An eigenvalue of  $A$  is a number  $\lambda$  which when subtracted from each of the diagonal entries of  $A$ , converts  $A$  into a singular (noninvertible) matrix.
- In particular,  $\lambda$  is an eigenvalue of  $A$  if  $A - \lambda I$  singular

## Theorem

*The following are equivalent:*

- 1  $A$  is singular
- 2  $|A| = 0$
- 3  $\lambda = 0$  is an eigenvalue of  $A$

# Eigenvalues and Eigenvector

- Note that for  $A$   $n \times n$  matrix,

$$\det(A - \lambda I) = 0$$

is a  $n$ -th order function

- We call such a polynomial (the left hand side of the equation) the characteristic polynomial of  $A$

## Theorem

*$\lambda$  is an eigenvalue of  $A$  if and only if  $\lambda$  is a root of the characteristic polynomial of  $A$ .*

- For example, if  $A$  is  $2 \times 2$ , then the characteristic polynomial is given by

$$\lambda^2 - \operatorname{tr}(A)\lambda + \det(A)$$

# Eigenvalues and Eigenvector

- For  $\lambda$  an eigenvalue of  $A$ , a nonzero vector  $v$  such that

$$(A - \lambda I)v = 0$$

is called an eigenvector of  $A$  corresponding to the eigenvalue  $\lambda$ .

# Eigenvalues and Eigenvector

- $\lambda = 0$  defines the Kernel of the transformation,  $Av = 0$ .
- Repeated eigenvalues can define different eigenvectors.
- A real-valued matrix may have (conjugate) complex eigenvalues. Multiplying a vector by a complex eigenvalue can be interpreted as “rescaling plus a rotation”.
- All of the eigenvalues of a symmetric matrix are real-valued.
- For the Diagonalization (below) we assume the eigenvalues are real-valued.



# Eigenvalues and Eigenvector

- Let  $S = \{v_1, v_2 \dots v_n\}$  be a subset of  $n$  vectors. These vectors are said to be linearly dependent if there exist a finite number of distinct vectors  $v_1, \dots v_k \in S$  and scalars  $a_1 \dots a_k$  such that

$$\sum_{i=1}^k a_i v_i = 0$$

## Theorem

*Let  $A$  be a  $n \times n$  matrix and let  $\lambda_1, \lambda_k$  be  $k$  distinct eigenvalues of  $A$ . If  $v_1, \dots v_k$  are the corresponding eigenvectors, then they are all linearly independent.*

# Difference Equations

- We will focus on linear difference equations of the form

$$x_{t+1} = ax_t + b$$

- A solution to such a difference equation is of the form

$$x_t = f(t)x_0 + g(t)$$

# Difference Equations

- Now consider the system

$$x_{t+1} = a_1x_t + a_2y_t + b_1$$

$$y_{t+1} = a_3x_t + a_4y_t + b_2$$

- Let

$$A = \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix}$$

Then,

$$z_{t+1} = \begin{bmatrix} x_{t+1} \\ y_{t+1} \end{bmatrix} = A \begin{bmatrix} x_t \\ y_t \end{bmatrix} + B = Az_t + B$$

where  $B = [b_1, b_2]'$ .

# Difference Equations: Diagonalization

- Let  $A$  be a  $n \times n$  matrix with eigenvalues  $\lambda_i$ , and eigenvectors  $v_i$ .
- Let  $Q = [v_1, \dots, v_k]$  and  $\Lambda$  be a diagonal matrix of the eigenvalues:

$$AQ = Q\Lambda$$

- If  $Q$  is invertible, then

$$A = Q\Lambda Q^{-1}$$

$$\Lambda = Q^{-1}AQ$$

# Difference Equations

- We can express an  $n$ –dimensional system

$$z_{t+1} = Az_t + B$$

as

$$z_{t+1} = Q\Lambda Q^{-1}z_t + B$$

or

$$Q^{-1}z_{t+1} = \Lambda Q^{-1}z_t + Q^{-1}B$$

- Now define  $w_t = Q^{-1}z_t$  and solve the decoupled system in the variable  $w_t$