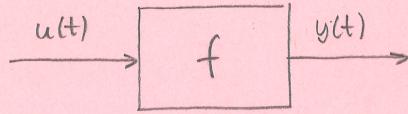


INTRO/REFRESHER DYNAMICAL SYSTEMS



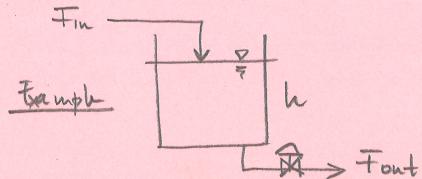
INPUT SIGNAL $u(t) \in \mathbb{R}^{N_u}$
 OUTPUT SIGNAL $y(t) \in \mathbb{R}^{N_y}$
 DYNAMICS + MEASUREMENT f

The representation can be regarded as MAPPING IN TIME

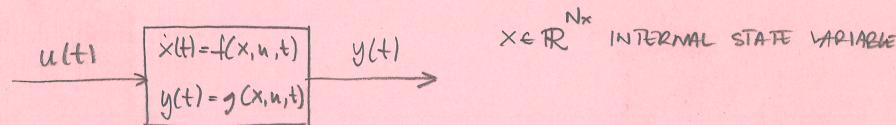
$$u: t \rightarrow u(t)$$

$$y: t \rightarrow y(t)$$

$$u \mapsto y \text{ as } f \{ u \}$$



If the system dynamics are given as ordinary differential equations and the measurement process is given as an algebraic equation,



$$\dot{x}(t) = f(x(t), u(t), t)$$

DYNAMICS OR STATE EQUATION
 (determine the time evolution of the state v.)

$$y(t) = g(x(t), u(t), t)$$

MEASUREMENT OR OUTPUT EQUATION
 (map the state to the output)



SINGLE STATE VARIABLE

$\dot{x} = f(x) = dx/dt$, we approximate the nonlinearity using a Taylor series

NEGLIGE H.O.T.

$$\rightarrow \text{FIXED POINT } x_s \quad \rightarrow f(x) = f(x_s) + \frac{\partial f}{\partial x} \Big|_{x_s} (x - x_s) + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} \Big|_{x_s} (x - x_s)^2 + \dots$$

$$\rightarrow f(x) \approx f(x_s) + \frac{\partial f}{\partial x} \Big|_{x_s} (x - x_s) \text{ with } f(x_s) = \frac{dx}{dt} \Big|_{x_s} = 0, \text{ AS WE ARE AT STEADY STATE}$$

$$\rightarrow f(x) \approx \frac{\partial f}{\partial x} \Big|_{x_s} (x - x_s) \approx \frac{dx}{dt} \text{ AND ALSO } \frac{dx}{dt} = \frac{d(x - x_s)}{dt}$$

$$\text{LET } x' = x - x_s \Rightarrow \frac{dx'}{dt} = \frac{\partial f}{\partial x} \Big|_{x_s} x' \quad (\text{PERTURBATION VARIABLES})$$

SINGLE STATE VARIABLE + SINGLE INPUT VARIABLE

$$\rightarrow \dot{x}(t) = f(x, u), \text{ FIXED POINT } (x_s, u_s)$$

$$\rightarrow f(x, u) = f(x_s, u_s) + \frac{\partial f}{\partial x} \Big|_{x_s, u_s} (x - x_s) + \frac{\partial f}{\partial u} \Big|_{x_s, u_s} (u - u_s) + \text{H.O.T.}$$

$$\rightarrow f(x, u) \approx \frac{\partial f}{\partial x} \Big|_{x_s, u_s} (x - x_s) + \frac{\partial f}{\partial u} \Big|_{x_s, u_s} (u - u_s) \approx \frac{dx}{dt} = \frac{d(x - x_s)}{dt} = 0$$

$$\text{LET } x' = x - x_s, u' = u - u_s \Rightarrow \frac{dx'}{dt} = \frac{\partial f}{\partial x} \Big|_{x_s, u_s} x' + \frac{\partial f}{\partial u} \Big|_{x_s, u_s} u'$$

(2) MULTIPLE STATE VARIABLES + MULTIPLE INPUT VARIABLES (1)

$$\begin{aligned} x(t) &= f(x, u) \text{ with } x = (x_1, x_2), & \dot{x}_1(t) &= f_1(x_1, x_2, u), & \text{FIXED POINT} \\ & & & x_2(t) &= f_2(x_1, x_2, u) & (x_s, u_s) \end{aligned}$$

$$\rightarrow f_1(x_1, x_2, u) \approx f(x_1^{ss}, x_2^{ss}, u^{ss}) + \frac{\partial f_1}{\partial x_1} \Big|_{x_s, u_s} (x_1 - x_1^{ss}) + \frac{\partial f_1}{\partial x_2} \Big|_{x_s, u_s} (x_2 - x_2^{ss}) + \frac{\partial f_1}{\partial u} \Big|_{x_s, u_s} (u - u^{ss})$$

$$f_2(x_1, x_2, u) \approx f_2(x_1^{ss}, x_2^{ss}, u^{ss}) + \frac{\partial f_2}{\partial x_1} \Big|_{x_s, u_s} (x_1 - x_1^{ss}) + \frac{\partial f_2}{\partial x_2} \Big|_{x_s, u_s} (x_2 - x_2^{ss}) + \frac{\partial f_2}{\partial u} \Big|_{x_s, u_s} (u - u^{ss})$$

$$\text{and } \frac{dx_1}{dt} = \frac{d(x_1 - x_1^{ss})}{dt} \approx f_1$$

LINEAR-TIME INVARIANT SYSTEMS

A dynamical system f is said to be linear if the following holds

- SUPERPOSITION $f(u_1 + u_2) = f(u_1) + f(u_2)$

- AMPLIFICATION $f(cu) = c \cdot f(u)$

A dynamical system is said to be time-invariant if the dynamics f are not dependent on time: $f(x, u, t')$

The general linear and time-invariant system:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases} \quad \begin{cases} A \in \mathbb{R}^{Nx \times Nx} \\ B \in \mathbb{R}^{Nx \times Nu} \\ C \in \mathbb{R}^{Ny \times Nx} \\ D \in \mathbb{R}^{Ny \times Nu} \end{cases}$$

No initial cond.

LINEARISATION: The idea is to approximate the behaviour of a generally nonlinear system around a reference or steady-state point

→ LINEARIZATION OF THE ODE

$$\begin{cases} \dot{x}(t) = f(x, u) \\ y(t) = g(x, u) \end{cases} \xrightarrow{\text{Linearization}} \begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

Perturbation vars

→ APPROXIMATION VALID AROUND A FIXED POINT

- STEADY-STATE
- EQUILIBRIUM

FOR THE LINEAR TIME-INVARIANT CASE →
→ det. $\phi(t, t_0) = e^{\int_{t_0}^t A(t') dt'}$

SOLUTION OF THE STATE-SPACE ODE

Dynamical Systems 3

Consider the LTI STATE EQUATION $\dot{x}(t) = Ax(t) + Bu(t)$ with initial conditions $x(t_0) = x_0$

HOMOGENEOUS SOLUTION (NO CONTROLS)

$$x(t) = \underbrace{e^{A(t-t_0)} x_0}_{\text{solves } \dot{x}(t) = Ax(t), x(0) = x_0}$$

We used the MATRIX EXPONENTIAL FUNCTION

$$\underbrace{e^{A(t-t_0)}}_{\text{matrix exponential}} = \sum_{v=0}^{\infty} \frac{A^v (t-t_0)^v}{v!}$$

The derivative

$$\begin{aligned} \frac{d}{dt} e^{A(t-t_0)} &= \frac{d}{dt} \sum_{v=0}^{\infty} \frac{A^v (t-t_0)^v}{v!} = \sum_{v=0}^{\infty} \frac{A^v v (t-t_0)^{v-1}}{v!} \\ &= A \sum_{v=0}^{\infty} \frac{A^{v+1} (t-t_0)^{v-1}}{(v+1)!} = A e^{A(t-t_0)} \end{aligned}$$

Thus, by computing the derivative of the solution, we have

$$\dot{x}(t) = A \underbrace{e^{A(t-t_0)} x_0}_{x(t)} = Ax(t) \quad \boxed{\phi(t,t_0) = e^{A(t-t_0)}}$$

GENERAL SOLUTION (WITH CONTROLS)

$$x(t) = \underbrace{\phi(t,t_0)x_0}_{\text{HOMOGENEOUS SOLUTION}} + \underbrace{\int_{t_0}^t \phi(t,\tau) Bu(\tau) d\tau}_{\text{CONVOLUTION INTEGRAL OF THE INPUT}} \quad \boxed{\phi(t,\tau) = e^{A(t-\tau)}}$$

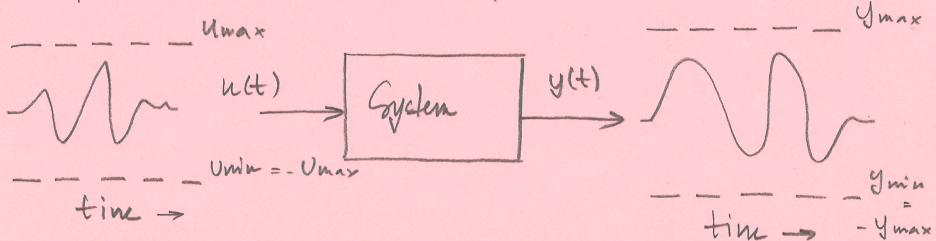
IS THIS REALLY THE SOLUTION?

$$\begin{aligned} \dot{x}(t) &= A\phi(t,t_0)x_0 + \phi(t,t)Bu(t) + \int_{t_0}^t \frac{d}{dt} \phi(t,\tau) Bu(\tau) d\tau \\ &= I \\ &= A(\phi(t,t_0)x_0 + \int_{t_0}^t Bu(\tau) d\tau) + Bu(t) \quad \text{with } \phi(t,t_0) = e^{A(t-t_0)} \\ &\quad \boxed{Ax(t)} \end{aligned}$$

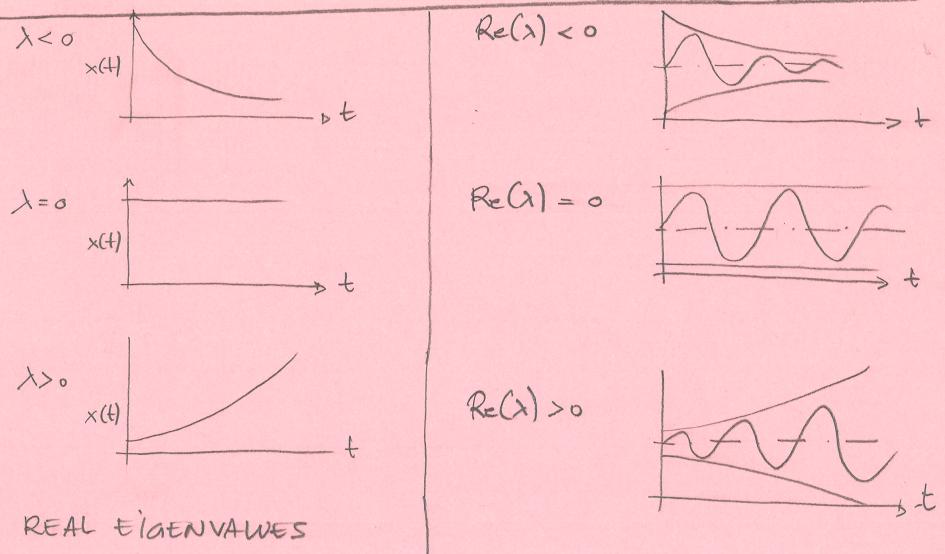
DYNAMICS AND STABILITY

BIBO (Bounded Input, Bounded Output) STABILITY

- A system is said to have BIBO stability, if every bounded input results in a bounded output

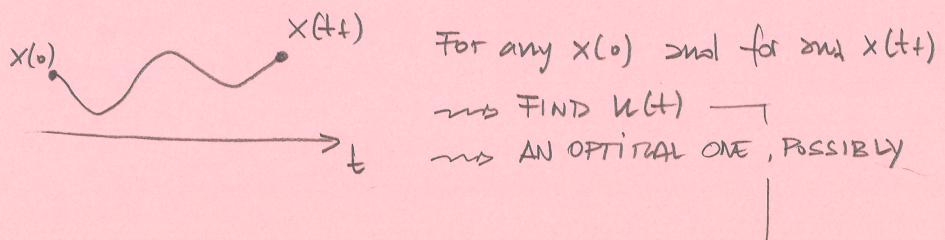


A LTI INVARIANT SYSTEM IS BIBO STABLE IF $\operatorname{Re}(\lambda_i) < 0$
FOR ALL $\lambda_i \in \sigma(A)$ (all the eigenvalues of A)



CONTROLLABILITY

CONTROLLABILITY (GENERAL CASE) - A system is controllable if it can be steered from any initial state $x(0)$ to any final state $x(t_f)$ in a finite time (t_f) by an appropriate choice of the input $u(t)$, $0 \leq t \leq t_f$



FOR LINEAR AND TIME-INVARIANT SYSTEMS THERE EXIST A NUMBER OF CRITERIA FOR CHECKING CONTROLLABILITY

→ CONTROLLABILITY MATRIX
 $C = [B \ AB \ AB^2 \ \dots \ AB^{N-1}]$

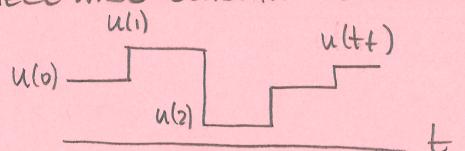
(A, B) CONTROLLABLE
 IF C IS FULL RANK
 $\det(C) \neq 0$

→ POPOV - BELEVITCH - HANTUS TEST

$$\text{rank}[\lambda I - A | B] = N_x, \forall \lambda \in \sigma(A)$$

↑ THE SPECTRUM
OF MATRIX A

CONTROLS, IN PRACTICE, CAN ONLY
BE DISCRETE, FOR EXAMPLE
PIECEWISE CONSTANT FUNCTIONS



OBSERVABILITY

OBSERVABILITY (GENERAL CASE) — A system is observable if it is possible to uniquely determine its initial state from a sequence of measurements over a finite time interval

FOR LINEAR AND TIME-INVARIANT SYSTEMS THERE EXIST A NUMBER OF CRITERIA FOR CHECKING OBSERVABILITY

→ OBSERVABILITY MATRIX

$$\mathbf{O} = [C^T \quad A^T C^T \quad (A^T)^2 C^T \quad \dots \quad (A^T)^{N_x-1} C^T]^T$$

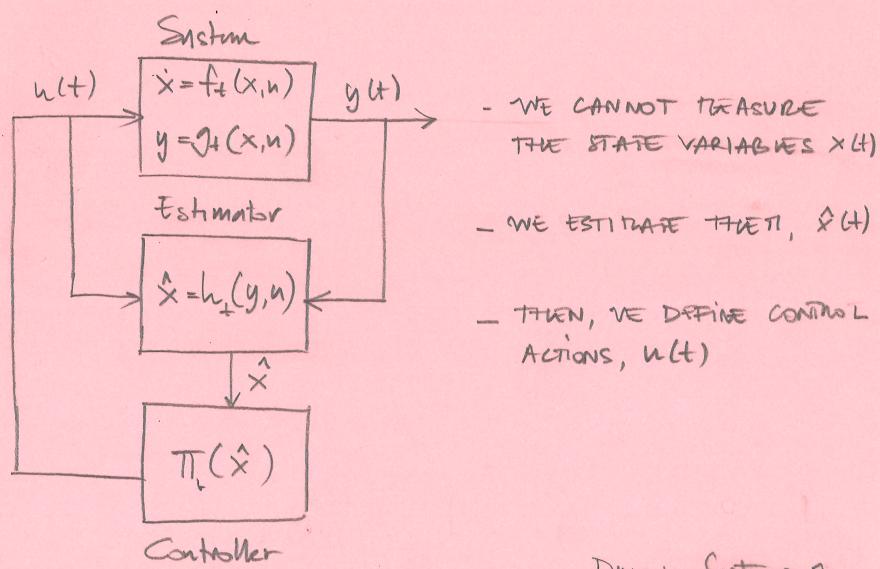
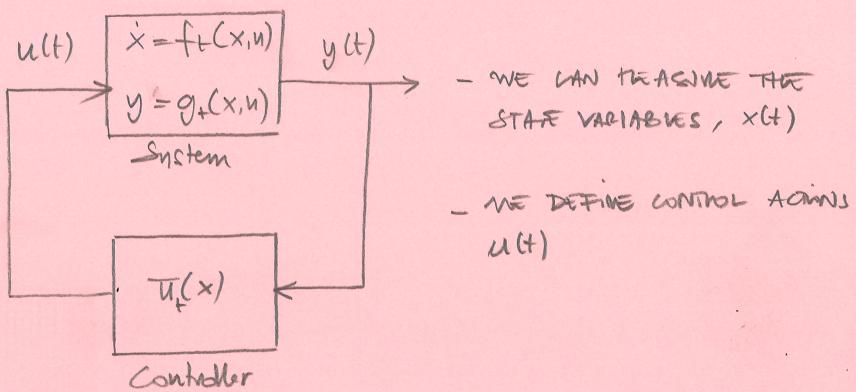
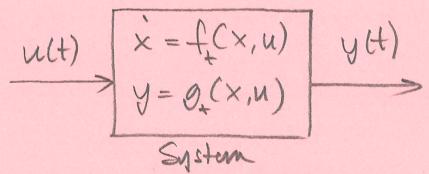
(A, C) OBSERVABLE IF \mathbf{O} IS FULL-RANK $\det(\mathbf{O}) \neq 0$

→ POPOV - BELEVITCH TEST

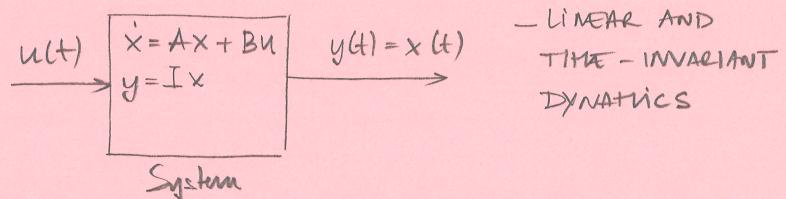
$$\text{rank } [\lambda I - A^T \mid C^T]^T = N_x, \quad \forall \lambda \in \sigma(A)$$

↓
THE SPECTRUM OF
MATRIX A

STATE FEEDBACK CONTROL



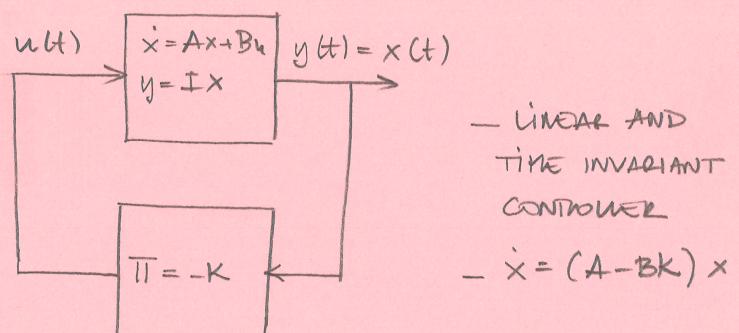
LINEAR QUADRATIC REGULATOR (LQR)



- LINEAR AND
TIME-INVARIANT
DYNAMICS

→ STABLE OR UNSTABLE
→ CONTINUABLE

→ OPTIMAL PLACEMENT OF THE EIGENVALUES OF THE
CONTINUOUS SYSTEM



OPTIMAL IN THE FOLLOWING SENSE

$$J = \sqrt{\int_0^{\infty} (x^T Q x + u^T R u) dt}$$

Q
 R } WEIGHTING
MATRICES

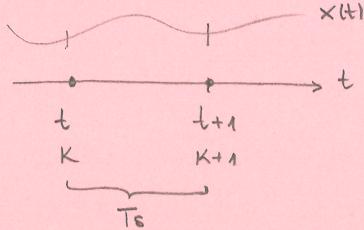
- x IS SHORT HAND FOR $x - x_{ss}$, x_{ss} STEADY STATE
- u IS SHORT HAND FOR $u - u_{ss}$, u_{ss} STEADY STATE

Dynamical Systems⁹

DISCRETE-TIME SYSTEMS, LINEAR CASE

- THIS TYPE OF MODELS ARE OFTEN CREATED BY DISCRETISING CONTINUOUS-TIME MODELS
- WE FOCUS ON LINEAR-TIME-INVARIANT MODELS

THE CONTINUOUS-TIME SYSTEM IS DISCRETISED AT EQUIDISTANT TIME POINTS, THE SAMPLING TIME IS T_s

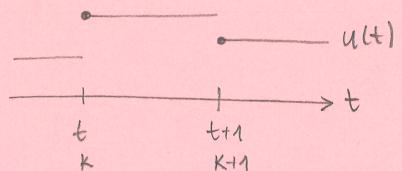


T_s IS THE TIME BETWEEN TWO CONSECUTIVE SAMPLING TIMES

EXAMPLE: THE STATE VECTOR $x(t)$ AT THE k -TH SAMPLING TIME IS THE VALUE OF THE VECTOR AT TIME $t = kT_s$

$$\text{and } x_k = x(kT_s)$$

THE INPUT VECTOR IS ASSUMED CONSTANT BETWEEN SAMPLING TIMES



EXAMPLE: THE INPUT VECTOR $u(t)$ IS ASSUMED PIECEWISE CONSTANT

(it could have been piecewise linear
(piecewise quadratic, ...))

CONSIDER THE GENERAL LTI SYSTEM \rightarrow STATE-SPACE CONTINUOUS-TIME

$$\begin{aligned}\dot{x}(t) &= A_c x(t) + B_c u(t) && (c' \text{ denotes continuous-time}) \\ y(t) &= C x(t) + D_c u(t)\end{aligned}$$

THE DISCRETE-TIME REPRESENTATION IS OBTAINED FROM THE CONTINUOUS-TIME ONE, BY MEANS OF SIMULATION (INTEGRATION)

FOR LTI SYSTEMS WE CAN DERIVE AN ANALYTICAL EXPRESSION FOR THE DISCRETE-TIME FORM OF THE MODEL UNDER THE PIECEWISE CONSTANT ASSUMPTION FOR THE INPUT

and we do not need numerical integration

WE OBTAIN, THE TIME INVARIANT REPRESENTATION

$$\begin{cases} \dot{x}_{k+1} = A x_k + B u_k \\ y_k = C x_k + D u_k \end{cases}, \text{ for all } k=0, \dots, K-1$$

A, B, C, D constant with k

CONSIDER TIME $t_0 = 0$ AND $t = T$, WE CAN WRITE

$$x_1 = x(T_s) = e^{A_c(T_s - 0)} x(0) + \int_0^{T_s} e^{A_c(T_s - t)} B_c u(t) dt$$

$$= e^{A c T_s} x(0) + \int_0^{T_s} e^{A c t} dt \underbrace{\left(B_c u(t) \right)}_{\text{IT IS PIECEWISE CONSTANT IN } [0, T_s] \text{ AND THUS CAN BE TAKEN OUTSIDE THE INTEGRAL}} = A x_0 + B u_0$$

CHANGE OF VARIABLE

$$t = T_s - \tau$$

IT IS PIECEWISE CONSTANT
IN $[0, T_s]$ AND THUS CAN
BE TAKEN OUTSIDE THE
INTEGRAL

THE DISCRETE-TIME MATRICES WRT THE CONTINUOUS-TIME ONES

$$\left\{ \begin{array}{l} A = e^{At_s} \\ B = \int_0^{T_s} e^{Act} dt \quad (B_c) \\ C = C_c \\ D = D_c \end{array} \right. \quad \left. \begin{array}{l} \text{UNCHANGED, THE TIPPING FROM STATE} \\ \text{VARS TO MEASUREMENTS IS STATIC} \end{array} \right.$$

THE HOMOGENEOUS RESPONSE (only A is needed)

$$x_1 = Ax_0$$

$$x_2 = Ax_1 = A(Ax_0) = A^2x_0$$

=

$$x_k = Ax_{k-1} = A(A^{k-1}x_0) = A^kx_0$$

FORCED RESPONSE

The forced response with arbitrary input can be computed by induction

$$\begin{aligned} x_{k+1} &= Ax_{k+1} + Bu_{k+1} \\ &= A(Ax_k + bu_k) + bu_{k+1} \\ &= A^2x_k + Abu_k + bu_{k+1} \end{aligned}$$

It generalizes to $x_k = A^kx_0 + \sum_{m=0}^{k-1} A^{k-m-1}Bu_m \quad (k \geq 0)$

FOR THE OUTPUT, WE HAVE

$$y_k = CA^kx_0 + \sum_{m=0}^{k-1} CA^{k-m-1}Bu_m + Du_k$$

Dynamical System 10b

STABILITY IN DISCRETE TIME

The eigenvalues of the state matrix A are defined as

$$\lambda_i v_i = A v_i \quad \text{for } v_i \neq 0$$

We can write $(\lambda_i I - A)v_i = 0$, which implies that

$$\det(\lambda_i I - A) = 0 \quad (\text{characteristic polynomial})$$

Let A be $n \times n$, then the characteristic polynomial is

$$\lambda^n + a_{n-1}\lambda^{n-1} + \dots + a_1\lambda + a_0 = 0$$

By factorization, we get

$$(\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_{n-1}) = 0 \quad \text{for } \lambda \in \mathbb{C}$$

We define $V = [v_1, v_2, \dots, v_n]$

Then, we define

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_{n-1} \end{bmatrix}$$

Then, we also define

$$\Lambda^k = (\Lambda)^k$$

$$\text{And, finally } A^k = (V \Lambda V^{-1})^k = V \Lambda^k V^{-1}$$