

Linear Algebra & ODEs

General Matrix Properties: Let $A \in \mathbb{R}^{n \times m}$

- $\text{range}(A) = \text{span}\{a_1, \dots, a_m\}$ is a subspace of \mathbb{R}^n
- Perform Gaussian elimination: For all pivots take the corresponding original
- $\text{rank}(A) = \dim(\text{range}(A))$ Column in A
- $\text{null}(A) = \{x \in \mathbb{R}^n : Ax = 0\}$ is a subspace of \mathbb{R}^n
- Set of vectors Orthogonal to the rows of A
- Perform Gaussian elimination for homogeneous linear system:
- Introduce a parameter (s, t, \dots) for every linearly dependent set of columns
- Span of solution with parameters in \mathbb{R} is the nullspace.

A invertible $\Leftrightarrow \det(A) \neq 0 \Leftrightarrow Ax = y$ has a unique sol.
 $\Leftrightarrow \text{null}(A) = \{0\} \Leftrightarrow \text{range}(A) = \mathbb{R}^n \Leftrightarrow$ All e-values are nonzero

Matrix Inverse \Leftarrow Always unique if it exists

2x2 Matrix $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{\det(A)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

3x3 Matrix $\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}^{-1} = \frac{1}{\det(A)} \begin{bmatrix} ei-fh & ch-bf & cf-bd \\ fg-di & ai-gf & af-gd \\ dh-eg & bh-ah & ad-bd \end{bmatrix}$

$\det(A) = ad-bc$

Block Diagonal Matrix

$A = \begin{bmatrix} A_1 & 0 \\ 0 & A_k \end{bmatrix}^{-1} = \begin{bmatrix} A_1^{-1} & 0 \\ 0 & A_k^{-1} \end{bmatrix}$

$\det(A) = a \cdot \det \begin{bmatrix} b & c \\ d & e \end{bmatrix} + b \cdot \det \begin{bmatrix} d & e \\ g & h \end{bmatrix} + c \cdot \det \begin{bmatrix} g & h \\ d & e \end{bmatrix}$

Determinant with Laplace expansion

E-Values & Vectors: $Ax = \lambda x, \lambda \in \mathbb{C}$

$\rightarrow \det(A - \lambda I) = \text{chp}(\lambda) = 0 \rightarrow$ solve for λ ;

\rightarrow insert found λ_i in $A v_i = \lambda v_i \rightarrow$ solve for v_i ;

Hurwitz-Criterion: (only for 2x2 matrix x)

$\text{chp}(\lambda) = \alpha \lambda^2 + \beta \lambda + \gamma = 0$

If α, β, γ have same sign $\Leftrightarrow \text{Re}\{\lambda_i\} < 0 \forall i$

If α, β, γ not same sign $\Leftrightarrow \exists i : \text{Re}\{\lambda_i\} > 0$

Diagonalizability: (AM=Algebraic Multiplicity, GM=Geometric Multiplicity)

If $AM_i = GM_i \forall \lambda_i \Rightarrow A$ is diagonalizable

Thm: $AM \geq GM \geq 1$ always \Rightarrow if no λ_i repeated \Rightarrow diagonalizable

Cayley-Hamilton Thm:

Every Matrix $A \in \mathbb{R}^{n \times n}$ satisfies its characteristic polynomial.

$A^n + a_{n-1}A^{n-1} + \dots + a_1A + a_0I = 0 \Leftrightarrow A^n = -(a_{n-1}A^{n-1} + \dots + a_1A + a_0I)$

\Rightarrow If all powers of A are 0 until A^{n-1} , all higher terms $A^n, A^{n+1}, \dots \rightarrow 0$

Symmetric Matrices have real e-values & orthogonal e-vectors

\Rightarrow are always diagonalizable through Orthogonal transformation.

$x^T A x > 0 \forall x \neq 0 \Leftrightarrow A$ positive definite $\Leftrightarrow \lambda_i > 0 \forall i$

$x^T A x \geq 0 \forall x \neq 0 \Leftrightarrow A$ positive semidefinite $\Leftrightarrow \lambda_i \geq 0 \forall i$

Canonical Jordan Forms:

$\begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_k \end{bmatrix}$ Jordan matrices are diagonalizable iff all Jordan blocks (blocks with same diagonal entries) are 1x1 matrices.

$GM_i = \# \text{Blocks for } \lambda_i, AM_i = \# \lambda_i$

Lipschitz, Existence & Uniqueness of solutions:

Def: $\|f(x, u) - f(\tilde{x}, \tilde{u})\| \leq \|Ax - A\tilde{x}\| \leq \alpha \|x - \tilde{x}\| \Leftrightarrow f$ Lipschitz

$\frac{df}{dx}$ exists & is bounded $\Rightarrow f$ Lipschitz $\bullet f$ linear $\Rightarrow f$ Lipschitz

Thm: If either:

- $f(x(t))$ (autonomous) is Lipschitz in $x(t)$
- $f(x(t), u(t), t)$ is Lipschitz in $x(t)$, continuous in $u(t)$ & t and $u(t)$ continuous for almost all t

\Rightarrow State solution exists & is unique

Complex Analysis:

$\sin(z) = \frac{1}{2j}(e^{jz} - e^{-jz}), \cos(z) = \frac{1}{2}(e^{jz} + e^{-jz})$

Partial fraction decompos:

double poles: $\frac{A_1}{x-p} + \frac{A_2}{(x-p)^2}$

complex poles: $\frac{Ax+B}{x^2+px+q}$

Complex Analysis:

	$\frac{1}{z}$	$\frac{1}{z^2}$	$\frac{1}{z^3}$	$\frac{1}{z^4}$	$\frac{1}{z^5}$	$\frac{1}{z^6}$	$\frac{1}{z^7}$	$\frac{1}{z^8}$	$\frac{1}{z^9}$	$\frac{1}{z^{10}}$
\sin	0	$\frac{j}{2}$	$\frac{j}{2}$	$\frac{j}{6}$	$\frac{j}{24}$	$\frac{j}{120}$	$\frac{j}{720}$	$\frac{j}{5040}$	$\frac{j}{362880}$	$\frac{j}{3628800}$
\cos	1	$-\frac{1}{2}$	$-\frac{1}{2}$	$-\frac{1}{24}$	$-\frac{1}{24}$	$-\frac{1}{720}$	$-\frac{1}{720}$	$-\frac{1}{5040}$	$-\frac{1}{362880}$	$-\frac{1}{3628800}$
\tan	0	$\frac{j}{3}$	1	$\frac{j}{5}$	$\frac{2}{15}$	$\frac{j}{7}$	$\frac{17}{315}$	$\frac{j}{31}$	$\frac{62}{2835}$	$\frac{j}{2835}$

Modeling

General System:

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

System Classification

Time-Invariant no explicit dependence on time

Linear Additivity & Homogeneity (A, B, C, D, E can depend on time)

Autonomous time & input invariant: $\dot{x}(t) = f(x(t))$

To make a system time invariant, introduce t as a new state with $\dot{t} = 1$. The new system has dimension $n+1$

LTI Systems (State Space Representation)

System Dynamics

$\dot{x}(t) = Ax(t) + Bu(t)$

Output measurement

$y(t) = Cx(t) + Du(t)$

states \leftrightarrow energy in the system

$A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, C \in \mathbb{R}^{p \times n}, D \in \mathbb{R}^{p \times m}$

$\begin{bmatrix} \dot{x} \\ y \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} x \\ u \end{bmatrix} + \begin{bmatrix} 0 \\ D \end{bmatrix} u$

Coordinate Transforms invertible

$\hat{x}(t) = T^{-1}x(t), \hat{u}(t) = T^{-1}u(t), \hat{y}(t) = Ty(t)$

$\hat{A} = TAT^{-1}, \hat{B} = TB, \hat{C} = CT^{-1}, \hat{D} = D$

System properties remain unchanged

Modeling equations: Physical background

Electrical circuit

Inductor: Equation: $v_L(t) = L \frac{di_L(t)}{dt}$ State: $x(t) = i_L(t)$ Energy: $E_L(t) = \frac{1}{2} L i_L^2(t)$

Capacitor: Equation: $i_C(t) = C \frac{dv_C(t)}{dt}$ State: $x(t) = v_C(t)$ Energy: $E_C(t) = \frac{1}{2} C v_C^2(t)$

\rightarrow Kirchhoff eq \rightarrow State Space representation

Mechanical system

Spring: Free body: Equation: $F_s(t) = kx(t)$ State: $x(t) = p(t)$ Energy: $E_s(t) = \frac{1}{2} kx^2(t)$

Damper: Free body: Equation: $F_d(t) = d\dot{x}(t)$ State: $x(t) = \dot{p}(t)$ Energy: $E_d(t) = \frac{1}{2} m\dot{p}^2(t)$

Continuous LTI in Time Domain

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

$y(t) = Cx(t) + Du(t) = h(x(t), u(t), t)$

$x(0) = x_0$ (all LTI Systems have a unique solution)

State & Output Solution

$x(t) = \Phi(t)x_0 + \int_0^t \Phi(t-\tau)Bu(\tau)d\tau$

$y(t) = C\Phi(t)x_0 + \int_0^t C\Phi(t-\tau)Bu(\tau)d\tau + Du(t)$

$\Phi(t) = e^{At} = \sum_{k=0}^{\infty} \frac{(At)^k}{k!} = I + At + \frac{A^2 t^2}{2!} + \dots + \frac{A^k t^k}{k!} + \dots$

State Transition Matrix

$\Phi(t) = e^{At}$

If A diagonalizable: $A = W\Lambda W^{-1} \Rightarrow e^{At} = W e^{\Lambda t} W^{-1}, \Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$

If $A = N + D$ and N, D commute: $e^{At} = e^{Dt} e^{Nt}$

If A nilpotent, calculate $e^{At} = \sum_{k=0}^{n-1} \frac{(At)^k}{k!}, A^n = 0$, all lower powers $\neq 0$

Properties of State Transition Matrix

$\Phi(0) = I, \Phi(-t) = \Phi(t)^{-1} \Rightarrow \Phi(t)\Phi(-t) = I$

$\frac{d\Phi(t)}{dt} = A\Phi(t), \Phi(t_1+t_2) = \Phi(t_1)\Phi(t_2)$

Impulse Transition & Impulse Response

Let $u(t) = \delta(t)$ and $x_0 = 0$

Impulse Transition: $x(t) \big|_{u(t)=\delta(t)} = H(t) = \Phi(t)B$ ZST: $x(t) = (H * u)(t)$

Impulse Response: $y(t) \big|_{u(t)=\delta(t)} = K(t) = C\Phi(t)B + D\delta(t)$ ZSR: $y(t) = (K * u)(t)$

Stability

Definitions: A system is called

\rightarrow stable if $\forall \epsilon > 0 \exists \delta > 0 : \|x_0\| < \delta \Rightarrow \|x(t)\| < \epsilon$

\rightarrow asymptotically stable if the system is stable and $\lim_{t \rightarrow \infty} \|x(t)\| = 0$

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Controllability

Def: $\forall x(0) = x_0 \forall x, \exists u(\cdot) : [0, t] \rightarrow \mathbb{R}^m$ s.t. $x(t) = x_1$

$\Leftrightarrow \forall x(0) = x_0 \exists u(\cdot) : [0, t] \rightarrow \mathbb{R}^m$ s.t. $x(t) = 0$

$\Leftrightarrow \forall x_1 \in \mathbb{R}^n \exists u(\cdot) : [0, t] \rightarrow \mathbb{R}^m$ s.t. $x(t) = x_1$

\Leftrightarrow controllable over $[0, T] \Leftrightarrow$ controllable over $[0, T] \forall T > 0$

Controllability Gramian symmetric, positive semi-definite

$\mathbb{R}^{n \times n} \ni W_c(t) = \int_0^t e^{A(t-\tau)} B B^T e^{A^T(t-\tau)} d\tau = W_c^T(t) \geq 0$

controllable over $[0, t] \Leftrightarrow W_c(t)$ invertible $\Leftrightarrow \det(W_c(t)) \neq 0$

Controllability Matrix

$P = [B \ AB \ A^2B \ \dots \ A^{n-1}B] \in \mathbb{R}^{n \times (n \cdot m)}$

Controllable over $[0, t] \Leftrightarrow \text{rank}(P) = n \Leftrightarrow \det(P) \neq 0$

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LTI-Systems in Frequency Domain

Laplace Transform: $F(s) = \mathcal{L}\{f(t)\}(s) = \int_0^\infty f(t)e^{-st} dt$

$a f(t) + b g(t) \rightarrow a F(s) + b G(s)$ $t f(t) \rightarrow -\frac{d}{ds} F(s)$ $f(at) \rightarrow \frac{1}{a} F(\frac{s}{a})$ $f(t-a) \rightarrow e^{-as} F(s)$ $f'(t) \rightarrow s F(s) - f(0)$ $f''(t) \rightarrow s^2 F(s) - s f(0) - f'(0)$ $f(t) \rightarrow \frac{1}{s} F(\frac{s}{a})$ $f(t-a) \rightarrow e^{-as} F(s)$ $f''(t) \rightarrow s^2 F(s) - s f(0) - f'(0)$ $f(t) \rightarrow \frac{1}{s} F(\frac{s}{a})$ $f(t-a) \rightarrow e^{-as} F(s)$

Initial value thm: $f(0) = \lim_{t \rightarrow 0} f(t) = \lim_{s \rightarrow \infty} s \cdot F(s)$
Final value thm: $\lim_{t \rightarrow \infty} f(t) = \lim_{s \rightarrow 0} s \cdot F(s)$

Important Transformations:

$1 \rightarrow \frac{1}{s}$	$e^{at} \rightarrow \frac{1}{s-a}$
$t^n \rightarrow \frac{n!}{s^{n+1}}$	$e^{at} \sin(bt) \rightarrow \frac{b}{(s-a)^2 + b^2}$
$\sin(at) \rightarrow \frac{a}{s^2 + a^2}$	$e^{at} \cos(bt) \rightarrow \frac{s-a}{(s-a)^2 + b^2}$
$\cos(at) \rightarrow \frac{s}{s^2 + a^2}$	$t^n e^{at} \rightarrow \frac{n!}{(s-a)^{n+1}}$

LTI-Systems:

$\mathcal{L}\{e^{At}\}(s) = \mathcal{L}\{\Phi(t)\}(s) = (sI - A)^{-1} \in \mathbb{C}^{n \times n}$

$X(s) = (sI - A)^{-1} x_0 + (sI - A)^{-1} B U(s)$
 $Y(s) = C X(s) + D U(s) = C(sI - A)^{-1} x_0 + G(s) U(s)$

For $x_0 = 0$: $Y(s) = [C(sI - A)^{-1} B + D] U(s) = G(s) U(s)$

G(s) transfer function

$G(s) = \frac{(s-z_1)(s-z_2)\dots(s-z_k)}{(s-p_1)(s-p_2)\dots(s-p_n)}$ $k \leq n \Rightarrow G(s)$ proper $\Rightarrow \lim_{\omega \rightarrow \infty} |G(j\omega)| < \infty$
 $k < n \Rightarrow G(s)$ strictly proper $\Rightarrow \lim_{\omega \rightarrow \infty} |G(j\omega)| = 0$

Thm:
• SISO $G(s)$ from LTI-Systems are always proper
• SISO $G(s)$ from LTI-Systems are strictly proper iff $D=0$

(Single) Frequency Response: $G(s) = |G(s)| \cdot e^{i\angle G(s)}$
 $u(t) = \sin(\omega t)$, $y(t) = K \sin(\omega t + \varphi) = K \cdot e^{i\varphi}$

Stability: (Requirement: no pole-zero cancellations!)

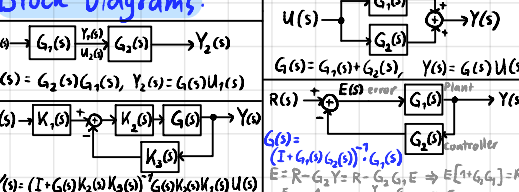
distinct poles: $\begin{cases} \text{asym. stable} & \text{iff } \operatorname{Re}\{p_i\} < 0 \forall i \\ \text{stable} & \text{iff } \operatorname{Re}\{p_i\} \leq 0 \forall i \\ \text{unstable} & \text{iff } \exists i: \operatorname{Re}\{p_i\} > 0 \end{cases}$

repeated poles: $\begin{cases} \text{asym. stable} & \text{iff } \operatorname{Re}\{p_i\} < 0 \forall i \\ \text{unstable} & \text{iff } \exists i: \operatorname{Re}\{p_i\} > 0 \end{cases}$
else: further investigation necessary

Marginally stable $\Rightarrow \operatorname{Re}\{p_i\} \leq 0$, $\operatorname{Re}\{p_i\} = 0 \Rightarrow \operatorname{Im}\{p_i\} \neq 0$

\Rightarrow If we have **pole-zero cancellations**, only BIBO-stability is determinable. Pole-zero cancellation corresponds to a loss of controllability and/or observability
 \Rightarrow If $G(s)$ has less poles than the dimension of A , we know that pole-zero cancellations happened
 \Rightarrow Find number of pole-zero cancellations:
Calculate which modes λ are uncontrollable and/or unobservable using PHB-test. If observability and/or controllability PHB-Matrix for a λ do not have full rank $\Rightarrow \lambda$ gets pole-zero cancelled. Count those 2.

Block Diagrams:



Closed Loop Systems:

$Y(s) = \frac{K G(s)}{1 + K G(s)} R(s)$
 $E(s) = \frac{1}{1 + K G(s)} R(s)$

Deduction: See Block diagrams with $G_1(s) = K G(s)$, $G_2(s) = I$
We want $e(t) \rightarrow 0$ (asymptotic stability)
 \Rightarrow Closed loop stability determined by roots of $1 + K G(s)$
Fact: Open loop poles = closed loop zeros

Principle of the Argument:

Assume clockwise D-curve that does not pass through poles/zeros of $G(s)$

$N = \#$ clockwise encirclements of $(0,0)$ by L-curve
 $Z = \#$ of (pos.) zeros in D
 $P = \#$ of (pos.) poles in D

$N = Z - P$

Nyquist Stability Criterion for $K(s) \in \mathbb{R}$

Consider Nyquist plot of $1 + K G(s) / K G(s) / G(s)$

$\rightarrow N = \#$ times L encircles $(0,0)$ / $(-1,0)$ / $(-\frac{1}{K}, 0)$ clockwise
 $\rightarrow Z = \#$ closed loop unstable (RHP) poles (roots of $1 + K G(s) = 0$)
 $\rightarrow P = \#$ open loop unstable (RHP) poles (poles of $G(s)$)

Principle of the Argument $\Rightarrow N = Z - P$
For closed loop stability we aim for $Z = 0$
Closed loop system is **asym. stable** iff $N = -P$

Corollary: If open loop system is stable, the closed loop system is stable iff Nyquist plot makes no encirclements of $(-1,0)$ / $(-\frac{1}{K}, 0)$

Bode Plots: $G_{dB} = 20 \cdot \log_{10} |G|$

	Magnitude	Phase
$1/s^N$ Integrator	-20 · N dB/dec everywhere	-90° · N everywhere
s^N Differentiator	+20 · N dB/dec everywhere	+90° · N everywhere
$1/(s + \sigma)^N$ LHP pole	-20 · N dB/dec	-90° · N $0.9\omega - 10\omega$
$1/(s - \sigma)^N$ RHP pole	-20 · N dB/dec	+90° · N $0.9\omega - 10\omega$
$(s + \sigma)^N$ LHP zero	+20 · N dB/dec	+90° · N $0.9\omega - 10\omega$
$(s - \sigma)^N$ RHP zero	+20 · N dB/dec	-90° · N $0.9\omega - 10\omega$

Resonance: $G(s) = \frac{K \omega_n^2}{s^2 + 2\zeta \omega_n s + \omega_n^2}$

- For stability need $\zeta > 0$
- For $\zeta \geq 1$ overdamped (real poles)
- For $\zeta = 1$ critically damped (real & equal)
- For $0 < \zeta < 1$ underdamped (complex conj)
- For $\zeta = 0$ undamped (imaginary p)
- For $\zeta \geq \frac{1}{2}$ magnitude plot decreasing w/ ω
- For $0 < \zeta < \frac{1}{2}$ magnitude plot has minimum (Resonance) at $\omega_r = \omega_n \sqrt{1 - 2\zeta^2}$, $|G(j\omega_r)| = \frac{K}{2\zeta \omega_n}$

Ex. spring-mass-damper: $m \ddot{x} + d \dot{x} + k x = F \sin(\omega t)$
Natural freq $\omega_n = \sqrt{k/m}$, $\zeta = \frac{d}{2m\omega_n}$

Bode Stability Criterion:

Assume $G(s)$ asym. stable and magnitude & phase bode plot are monotonically decreasing. Then the closed loop system (CLS) is asymptotically stable iff

$|K G(j\omega)| < 1$ at the frequency where $\angle G(j\omega) = -180^\circ$

Small Gain Thm: Under above conditions, if $|K G(j\omega)| < 1 \forall \omega \Rightarrow$ CLS is asym. stable.

For GM, PM: Assume OLS G is asym. stable.

Gain Margin: = maximum gain K_m s.t. CLS still stable

$GM = \frac{1}{|G(j\omega_c)|}$ where $\angle G(j\omega_c) = -180^\circ$
 ω_c Phase crossover frequency

Phase Margin: = maximum phase ϕ_m s.t. CLS still stable

$PM = \angle G(j\omega_c) + 180^\circ$ where $|G(j\omega_c)| = 1 = 0dB$

If ω_u does not exist $\Rightarrow GM \rightarrow \infty$, If ω_c does not exist $\Rightarrow PM \rightarrow \infty$

Discrete LTI-Systems in Time Domain

State solution:

$$\begin{cases} x_{k+1} = A x_k + B u_k \\ y_k = C x_k + D u_k \\ x(0) = x_0 \end{cases}$$

$x_k = A^k x_0 + \sum_{i=0}^{k-1} A^{k-i-1} B u_i$

• A diagonalizable: $A = W \Lambda W^{-1} \Rightarrow A^k = W \Lambda^k W^{-1}$
• at equilibrium: $x_{k+1} = x_k$

Stability:

A diagonalizable, The System is...
 \rightarrow stable iff $|\lambda_i| \leq 1 \forall i$
 \rightarrow asym. stable iff $|\lambda_i| < 1 \forall i$
 \rightarrow unstable iff $\exists i: |\lambda_i| > 1$

A non-diagonalizable, The System is...
 \rightarrow asym. stable iff $|\lambda_i| < 1 \forall i$
 \rightarrow unstable iff $\exists i: |\lambda_i| > 1$
 \rightarrow else: No statement. H depends on e-value with $|\lambda_i| = 1$

Lyapunov Stability: (in the discrete case)

$|\lambda_i| < 1 \forall i$ iff $\forall R = R^T > 0$ the eq. $A^T Q A - Q = -R$ has a unique solution with $Q = Q^T$

Deadbeat Response: e-values of $A: \lambda_i = 0 \forall i$

(not possible in continuous) $\Rightarrow A^N = 0$ $N \leq n$ (nilpotent Matrix)
 $\Rightarrow \exists T: x_k = A^k x_0 = 0 \forall k \geq N$

Controllability & Observability (Power is the same as in cont.)

check the same way as in continuous case. Difference: m, n ∈ N
Controllability for $[0, m]$ \Leftrightarrow controllability for $[0, m]$ as we require at least $k \geq n$ (system dim.) steps to steer system from initial to final state

Sampling: $x((k+1)T) = e^{A T} x(kT) + \int_0^T e^{A(T-\tau)} B u(kT) d\tau$

sampling autonomous cont. time systems never leads to nilpotent discrete time sys

Forward Euler approx: $x_{k+1} = (I + \Delta t A) x_k + \Delta t B u_k$

If autonomous: $x((k+1)T) = e^{A \Delta t} x(kT) \approx (I + A \Delta t) x_k$ $\Delta t = T/N = \text{step}$

Minimum Energy input: $U = P(P^T P)^{-1} (S^T - A^T x_0)$ $P = [B \ A B \ \dots \ A^{n-1} B]$

Z-Transform $F(z) = \mathcal{Z}\{f_k\} = \sum_{k=0}^{\infty} f_k z^{-k}$

Linearity: $\mathcal{Z}\{a f_k + b g_k\} = a F(z) + b G(z)$

Time shift: $\mathcal{Z}\{f_{k-k_0}\} = z^{-k_0} F(z)$

Convolution: $\mathcal{Z}\{f * g\}_k = F(z) \cdot G(z)$

Final value thm: $\lim_{k \rightarrow \infty} x_k = \lim_{z \rightarrow 1} (1 - z^{-1}) X(z)$

Initial value thm: $\lim_{k \rightarrow 0} x_k = \lim_{z \rightarrow \infty} z X(z)$

Transfer function: $G(z) = \frac{Y(z)}{U(z)} = C(zI - A)^{-1} B + D$

no pole-zero canc. \Rightarrow denominator if $G(s) = \frac{Y(s)}{U(s)} \Rightarrow$ (asym. stability) $\Leftrightarrow |R| < 1 \forall i$

\Rightarrow system is both observable AND controllable

Geometric Series: $\sum_{k=0}^{\infty} z^k = \frac{1}{1-z}$, $\sum_{k=0}^{\infty} k z^k = \frac{z}{(1-z)^2}$

Nonlinear Systems

We only look at autonomous, time-invariant systems

$\dot{x}(t) = f(x(t))$ $x(t) \in \mathbb{R}^n$, f Lipschitz \Leftrightarrow sol. exists & is unique

Invariant Sets

A set of states $S \subseteq \mathbb{R}^n$ is called invariant if $\forall x \in S \forall t \geq 0: x(t) \in S$

Equilibrium Let \hat{x} be an equilibrium

Conditions for \hat{x} : continuous $\dot{\hat{x}} = 0$ | discrete $\hat{x}_{k+1} = \hat{x}_k = \hat{x}$

If you start at \hat{x} , you will stay there forever

all equilibrium points of a system form an invariant set

Linear sys: A regular \Rightarrow 1 eq. pt. A singular \Rightarrow ∞ eq. pts.

nonlinear sys: $f(x) = 0$ with $\dot{x}(t) = f(x(t))$ has ∞ solutions: 0, 1, finitely many, count., ∞

Shifting Equilibria to the origin: $\omega(t) := x(t) - \hat{x}$

Periodic Orbits: sol. $x(t)$ s.t. $\exists T > 0, \forall t \geq 0: x(t+T) = x(t)$

Van der Pol Oscillator: $\ddot{\theta}(t) - \epsilon(1 - \theta^2(t))\dot{\theta}(t) + \theta(t) = 0$

Stability of Equilibria: Let \hat{x} be an equilibrium

Def: \hat{x} is stable $\Leftrightarrow \forall \epsilon > 0 \exists \delta > 0: \|x_0 - \hat{x}\| < \delta \Rightarrow \|x(t) - \hat{x}\| < \epsilon \forall t \geq 0$

Otherwise the equilibrium is unstable

Local asymptotic stability (LAS): The equilibrium \hat{x} is called LAS if (i) \hat{x} is stable and

and (ii) $\exists M > 0: \|x_0 - \hat{x}\| < M \Rightarrow \lim_{t \rightarrow \infty} \|x(t) - \hat{x}\| = 0$

Global asymptotic stability (GAS): Same as LAS but must hold $\forall M > 0$

System has more than one equilibrium $\Rightarrow \nexists$ a GAS equilibrium

Lyapunov first/indirect Method (by Linearization)

$\dot{x}(t) = f(x(t)) = \begin{bmatrix} f_1(x_1, \dots, x_n) \\ f_n(x_1, \dots, x_n) \end{bmatrix} = f(\hat{x}) + \frac{d}{dx} f(\hat{x}) \cdot (x(t) - \hat{x}) + O(\|x(t) - \hat{x}\|^2)$

$A := J[f(x)]|_{\hat{x}} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(\hat{x}) & \dots & \frac{\partial f_1}{\partial x_n}(\hat{x}) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}(\hat{x}) & \dots & \frac{\partial f_n}{\partial x_n}(\hat{x}) \end{bmatrix} \Rightarrow \dot{x}(t) \approx A \delta x(t)$ linear

If $\operatorname{Re}\{\lambda_i\} < 0 \forall i \Rightarrow$ linearized system asym. stable \Rightarrow nonlinear system is locally asym. stable around \hat{x}

If $\exists i: \operatorname{Re}\{\lambda_i\} > 0 \Rightarrow$ unstable

Inconclusive if $\exists i: \operatorname{Re}\{\lambda_i\} = 0$ & others have negative real part (no information about domain of attraction)

Lyapunov second/direct Method

Assume \exists open set $S \subseteq \mathbb{R}^n$ with $\hat{x} \in S$, $V(\cdot): \mathbb{R}^n \rightarrow \mathbb{R}$ differentiable

(i) $V(\hat{x}) = 0$ (ii) $V(x) > 0 \forall x \in S \setminus \{\hat{x}\}$ (iii) $\frac{d}{dt} V(x(t)) \leq 0 \forall x \in S$

\Rightarrow the equilibrium \hat{x} is stable

If additionally (iv) $\frac{d}{dt} V(x(t)) < 0 \forall x \in S \setminus \{\hat{x}\} \Rightarrow \hat{x}$ is locally asym. stable

If additionally (v) $\|x\| \rightarrow \infty \Rightarrow \|V(x)\| \rightarrow \infty \Rightarrow \hat{x}$ is globally asym. stable

$\frac{d}{dt} V(x(t)) = (\nabla V(x))^T \cdot f(x)$

La Salle's Theorem

Assume \exists compact (bounded & closed) invariant set $S \subseteq \mathbb{R}^n$

and $\exists V(\cdot): S \rightarrow \mathbb{R}$ differentiable such that

1) $\frac{d}{dt} V(x(t)) = (\nabla V(x))^T \cdot f(x) \leq 0 \forall x \in S$

2) Let $\bar{S} = \{x \in S: \frac{d}{dt} V(x(t)) = 0\} \subseteq S$

3) Let M be the largest invariant set in \bar{S}

Then: All trajectories $x(t) \in S \xrightarrow{t \rightarrow \infty} M \Rightarrow$ LAS

If $S \rightarrow \mathbb{R}^n$ and all conditions above still hold \Rightarrow GAS

We either get: $S_k = \{x(t) \in \mathbb{R}^n: V(x) \leq K\}$

$\frac{d}{dt} V(x(t)) \leq 0 \forall x \in S_k \Rightarrow S_k$ is invariant

\Rightarrow find $\bar{S} \subseteq S_k \Rightarrow$ find $M \subseteq \bar{S}$ (mostly $\{0, 0\}$)

\Rightarrow show that M is the maximal invariant set in \bar{S}

i.e. show that all $x \in \bar{S} \setminus M$ would leave the invariant set

La Salle $\forall x(t) \in S_k \xrightarrow{t \rightarrow \infty} M$, If $K > 0$ can be chosen arbitrarily

when $\|x(t)\| \rightarrow \infty \Rightarrow V(x) \rightarrow \infty \Rightarrow$ GAS

Or we get: $S = \{x: x_1 \in [a, b], x_2 \in [c, d]\}$

Prove S invariant by calculating derivatives on boundary

\Rightarrow Trajectories starting in S never leave $S \Rightarrow S$ is invariant

Find \bar{S} , Find M . If $\|x(t)\| \rightarrow \infty \Rightarrow V(x(t)) \rightarrow \infty \Rightarrow$ GAS

Nyquist from Bode: Set $\omega = 0$ and read Magn. & Phase from Bode

Set $\omega \rightarrow \infty$ and read Magn. (0) & Phase from Bode

3. Find intersections with real imaginary axis

4. Analyse phase as $\omega \rightarrow 0$ if $a < 0$ & $b < 0$ $\Rightarrow \arg(-\frac{a}{b}) = \arg(-1) = \pi$ if $a > 0$ & $b < 0$ $\Rightarrow \arg(\frac{a}{b}) = 0$ if $a < 0$ & $b > 0$ $\Rightarrow \arg(-\frac{a}{b}) = \pi$ if $a > 0$ & $b > 0$ $\Rightarrow \arg(\frac{a}{b}) = 0$

5. $\arg(a + ib) = \arctan(\frac{b}{a})$

6. read Magn. & Phase from Bode

7. plot in Nyquist: $\arg(a + ib) \rightarrow a + ib$

8. $\arg(a + ib) = \arctan(\frac{b}{a})$

9. $\arg(a + ib) = \arctan(\frac{b}{a})$