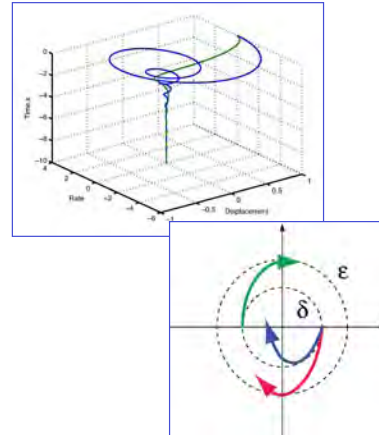


Stability of Dynamic Systems

Robert Stengel

Optimal Control and Estimation, MAE 546
Princeton University, 2015

- Bounds on the system norm
- Lyapunov criteria for stability
- Eigenvalues
- Transfer functions
- Continuous- and discrete-time systems

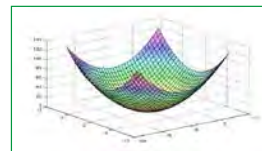


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<http://www.princeton.edu/~stengel/MAE546.html>
<http://www.princeton.edu/~stengel/OptConEst.html>

1

Vector Norms for Real Variables

- “**Norm**” = Measure of length or magnitude of a vector, **x**
- **Euclidean or Quadratic Norm**



$$L^2 \text{ norm} = \|\mathbf{x}\|_2 = (\mathbf{x}^T \mathbf{x})^{1/2} = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2}$$

- **Weighted Euclidean Norm**

$$\|\mathbf{y}\|_2 = (\mathbf{y}^T \mathbf{y})^{1/2} = (y_1^2 + y_2^2 + \dots + y_m^2)^{1/2}$$

$$= (\mathbf{x}^T \mathbf{D}^T \mathbf{D} \mathbf{x})^{1/2} = \|\mathbf{D} \mathbf{x}\|_2$$

$$\mathbf{x}^T \mathbf{D}^T \mathbf{D} \mathbf{x} \triangleq \mathbf{x}^T \mathbf{Q} \mathbf{x}$$

$$\mathbf{Q} \triangleq \mathbf{D}^T \mathbf{D} = \text{Defining matrix}$$

2

Uniform Stability

- Autonomous dynamic system

- Time-invariant
- No forcing input

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t)]$$

- Uniform stability about $\mathbf{x} = 0$

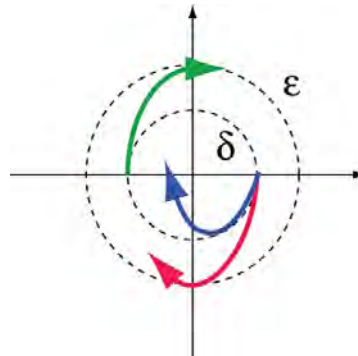
$$\|\mathbf{x}(t_0)\| \leq \delta, \quad \delta > 0$$

Let $\delta = \delta(\varepsilon)$

If, for every $\varepsilon \geq 0$,

$$\|\mathbf{x}(t)\| \leq \varepsilon, \quad \varepsilon \geq \delta > 0, \quad t \geq t_0$$

Then the system is **uniformly stable**



- If system response is bounded, then the system possesses uniform stability

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Local and Global Asymptotic Stability

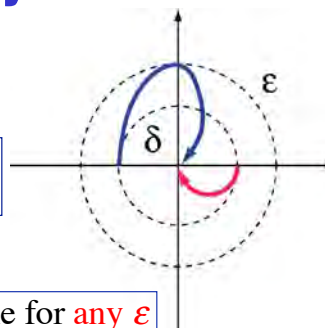
- Local asymptotic stability

- Uniform stability plus

$$\|\mathbf{x}(t)\| \xrightarrow{t \rightarrow \infty} 0$$

- Global asymptotic stability

System is asymptotically stable for **any** ε



- If a linear system has uniform asymptotic stability, it also is globally stable

$$\dot{\mathbf{x}}(t) = \mathbf{F} \mathbf{x}(t)$$

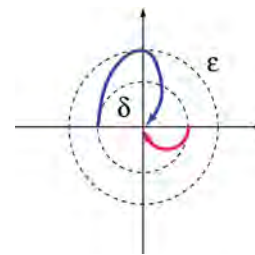
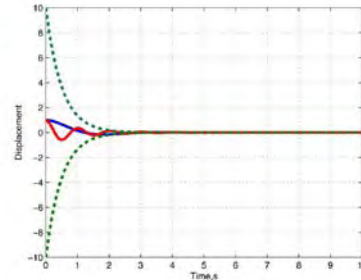
4

Exponential Asymptotic Stability

- Uniform stability about $\mathbf{x} = 0$ plus

$$\|\mathbf{x}(t)\| \leq k e^{-\alpha t} \|\mathbf{x}(0)\|; \quad k, \alpha \geq 0$$

- If norm of $\mathbf{x}(t)$ is contained within an exponentially decaying envelope with convergence, system is *exponentially asymptotically stable (EAS)*
- Linear system that is stable is EAS



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Exponential Asymptotic Stability

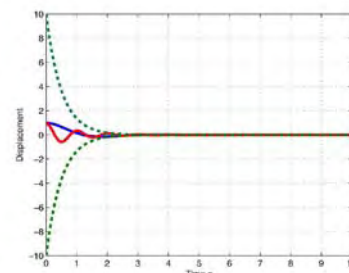
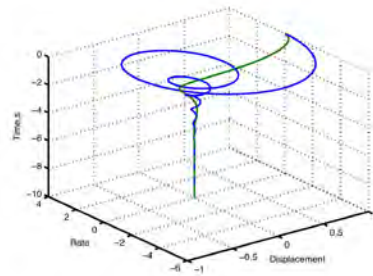
$$k \int_0^{\infty} e^{-\alpha t} dt = -\left(\frac{k}{\alpha}\right) e^{-\alpha t} \Big|_0^{\infty} = \frac{k}{\alpha}$$

Therefore, time integrals of the norm of an EAS system are bounded

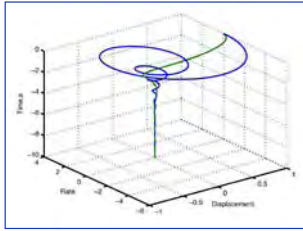
$$\int_0^{\infty} \|\mathbf{x}(t)\| dt = \int_0^{\infty} [\mathbf{x}^T(t) \mathbf{x}(t)]^{1/2} dt \leq \left(\frac{k}{\alpha}\right) \|\mathbf{x}(0)\|$$

and

$$\int_0^{\infty} \|\mathbf{x}(t)\|^2 dt \text{ is bounded}$$



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Exponential Asymptotic Stability

Weighted Euclidean norm and its square are bounded if system is EAS

$$\int_0^{\infty} \|\mathbf{D}\mathbf{x}(t)\| dt = \int_0^{\infty} [\mathbf{x}^T(t) \mathbf{D}^T \mathbf{D} \mathbf{x}(t)]^{1/2} dt \quad \text{is bounded}$$

with $\infty > \mathbf{Q} = \mathbf{D}^T \mathbf{D} > 0$

$$\int_0^{\infty} [\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t)] dt \quad \text{is bounded}$$

Conversely, if the weighted Euclidean norm is bounded, the system is EAS

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Initial-Condition Response of an EAS Linear System

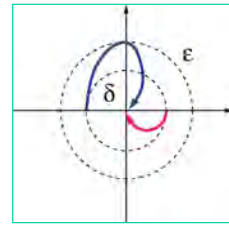
$$\mathbf{x}(t) = \Phi(t, 0) \mathbf{x}(0) = e^{\mathbf{F}(t)} \mathbf{x}(0)$$

$$\|\mathbf{x}(t)\|^2 = \mathbf{x}^T(0) \Phi^T(t, 0) \Phi(t, 0) \mathbf{x}(0) \quad \text{is bounded}$$

- **To be shown**
 - Continuous-time LTI system is stable if all of its eigenvalues have negative real parts
 - Discrete-time LTI system is stable if all of its eigenvalues lie within the unit circle

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Lyapunov's First Theorem



- A nonlinear system is asymptotically stable at the origin if its linear approximation is stable at the origin, i.e.,
 - for all trajectories that start “close enough”
 - within a stable manifold

$$\dot{\mathbf{x}}(t) = \mathbf{f}[\mathbf{x}(t)] \text{ is stable at } \mathbf{x}_o = 0 \text{ if}$$

$$\Delta \dot{\mathbf{x}}(t) = \left. \frac{\partial \mathbf{f}[\mathbf{x}(t)]}{\partial \mathbf{x}} \right|_{\mathbf{x}_o=0} \Delta \mathbf{x}(t) \text{ is stable}$$

“At the origin” is a fuzzy concept

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Lyapunov's Second Theorem*

Define a scalar Lyapunov function, a positive definite function of the state in the region of interest

$$V^*[\mathbf{x}^*(t)] \geq 0$$

Examples

$$V = E = \frac{mV^2}{2} + mgh; \quad \frac{E}{mg} = \frac{E}{\text{weight}} = \frac{V^2}{2g} + h$$

$$V = \frac{1}{2} \mathbf{x}^T \mathbf{x}; \quad V = \frac{1}{2} \mathbf{x}^T \mathbf{P} \mathbf{x}$$



* Who was Lyapunov? see http://en.wikipedia.org/wiki/Aleksandr_Lyapunov

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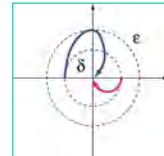
$$V^*[\mathbf{x}^*(t)] \geq 0$$

Lyapunov's Second Theorem

Evaluate the time derivative of the Lyapunov function

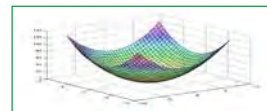
$$\begin{aligned} \frac{dV}{dt} &= \frac{\partial V}{\partial t} + \frac{\partial V}{\partial \mathbf{x}} \dot{\mathbf{x}} \\ &= \frac{\partial V}{\partial \mathbf{x}} \dot{\mathbf{x}} \text{ for autonomous systems} \end{aligned}$$

- If $\frac{dV}{dt} < 0$ in the neighborhood of the origin, the origin is asymptotically stable



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Quadratic Lyapunov Function



Lyapunov function

Linear, Time-Invariant System

$$V[\mathbf{x}(t)] = \mathbf{x}^T(t) \mathbf{P} \mathbf{x}(t)$$

$$\dot{\mathbf{x}}(t) = \mathbf{F} \mathbf{x}(t)$$

Rate of change for quadratic Lyapunov function

$$\begin{aligned} \frac{dV}{dt} &= \frac{\partial V}{\partial \mathbf{x}} \dot{\mathbf{x}} = \mathbf{x}^T(t) \mathbf{P} \dot{\mathbf{x}}(t) + \dot{\mathbf{x}}^T(t) \mathbf{P} \mathbf{x}(t) \\ &= \mathbf{x}^T(t) (\mathbf{P} \mathbf{F} + \mathbf{F}^T \mathbf{P}) \mathbf{x}(t) \triangleq -\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) \end{aligned}$$

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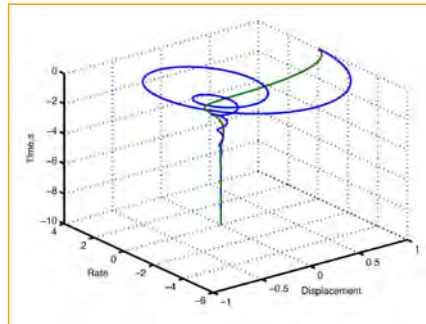
Lyapunov Equation

The LTI system is stable if the Lyapunov equation is satisfied with positive-definite **P** and **Q**

$$\mathbf{P}\mathbf{F} + \mathbf{F}^T\mathbf{P} = -\mathbf{Q}$$

with

$$\mathbf{P} > 0, \quad \mathbf{Q} > 0$$



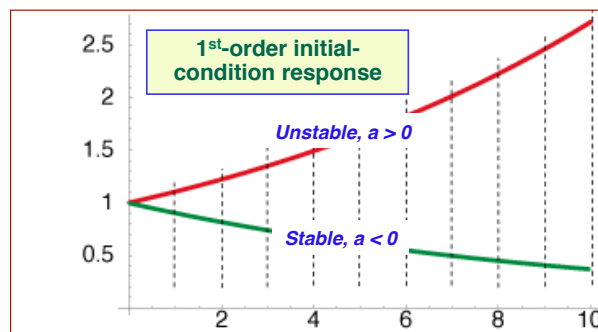
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Lyapunov Stability: 1st-Order Example

$$\Delta\dot{x}(t) = a\Delta x(t), \Delta x(0) \text{ given}$$

$$\mathbf{F} = a, \mathbf{P} = p, \mathbf{Q} = q$$

$$\begin{aligned} \Delta x(t) &= \int_0^t \Delta\dot{x}(t) dt = \int_0^t a\Delta x(t) dt \\ &= e^{at} \Delta x(0) \end{aligned}$$



$$\mathbf{P}\mathbf{F} + \mathbf{F}^T\mathbf{P} = -\mathbf{Q}$$

with $p > 0, a < 0$

$2pa < 0$ and $q > 0$

\therefore system is stable

$$\mathbf{P}\mathbf{F} + \mathbf{F}^T\mathbf{P} = -\mathbf{Q}$$

with $p > 0, a > 0$

$2pa < 0$ and $q < 0$

\therefore system is unstable

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Lyapunov Stability and the HJB Equation

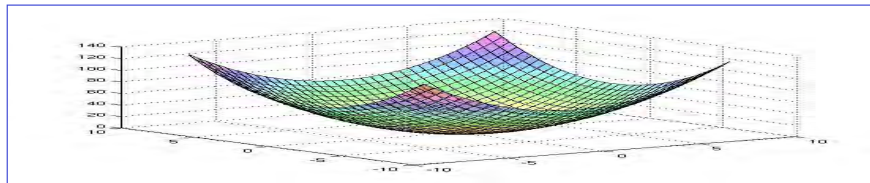
$$V[\mathbf{x}(t)] = \mathbf{x}^T(t) \mathbf{P} \mathbf{x}(t)$$

Lyapunov stability

$$\frac{dV}{dt} < 0$$

Dynamic programming optimality

$$\frac{\partial V^*}{\partial t} = -\min_{\mathbf{u}(t)} H$$



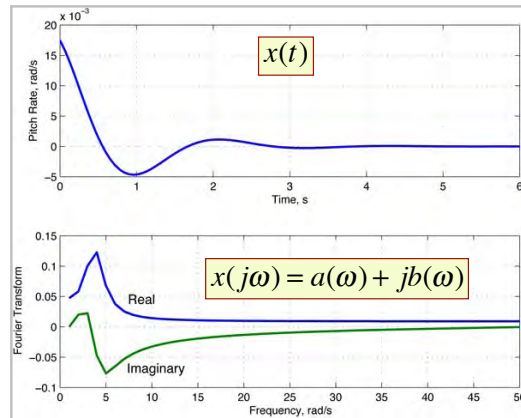
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Laplace Transforms and Linear System Stability

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Fourier Transform of a Scalar Variable

$$F[x(t)] = x(j\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t} dt, \quad \omega = \text{frequency, rad / s}$$



$x(t)$: real variable
 $x(j\omega)$: complex variable
 $= a(\omega) + jb(\omega)$
 $= A(\omega)e^{j\varphi(\omega)}$

A : amplitude
 φ : phase angle

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Laplace Transforms of Scalar Variables

Laplace transform of a scalar variable is a complex number
 s is the Laplace operator, a complex variable

$$L[x(t)] = x(s) = \int_0^{\infty} x(t)e^{-st} dt, \quad s = \sigma + j\omega, \quad (j = i = \sqrt{-1})$$

Laplace transformation is a linear operation

Multiplication by a constant

$$L[ax(t)] = ax(s)$$

Sum of Laplace transforms

$$L[x_1(t) + x_2(t)] = x_1(s) + x_2(s)$$

$x(t)$: real variable
 $x(s)$: complex variable
 $= a(\omega) + jb(\omega)$
 $= A(\omega)e^{j\varphi(\omega)}$

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Laplace Transforms of Vectors and Matrices

Laplace transform of a
vector variable

$$L[\mathbf{x}(t)] = \mathbf{x}(s) = \begin{bmatrix} x_1(s) \\ x_2(s) \\ \dots \end{bmatrix}$$

Laplace transform of a
matrix variable

$$L[\mathbf{A}(t)] = \mathbf{A}(s) = \begin{bmatrix} a_{11}(s) & a_{12}(s) & \dots \\ a_{21}(s) & a_{22}(s) & \dots \\ \dots & \dots & \dots \end{bmatrix}$$

Laplace transform of a time-derivative

$$L[\dot{\mathbf{x}}(t)] = s\mathbf{x}(s) - \mathbf{x}(0)$$

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Transformation of the System Equations

Time-Domain System Equations

$$\dot{\mathbf{x}}(t) = \mathbf{F} \mathbf{x}(t) + \mathbf{G} \mathbf{u}(t)$$

Dynamic Equation

$$\mathbf{y}(t) = \mathbf{H}_x \mathbf{x}(t) + \mathbf{H}_u \mathbf{u}(t)$$

Output Equation

Laplace Transforms of System Equations

$$s\mathbf{x}(s) - \mathbf{x}(0) = \mathbf{F} \mathbf{x}(s) + \mathbf{G} \mathbf{u}(s)$$

Dynamic Equation

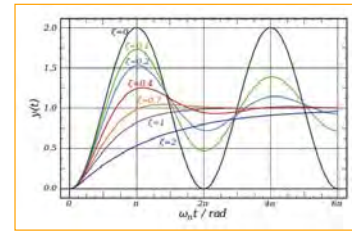
$$\mathbf{y}(s) = \mathbf{H}_x \mathbf{x}(s) + \mathbf{H}_u \mathbf{u}(s)$$

Output Equation

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Second-Order Oscillator

Differential Equations for 2nd-Order System



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

$$\begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u(t) = \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$$

Dynamic Equation

Output Equation

Laplace Transforms of 2nd-Order System

$$\begin{bmatrix} sx_1(s) - x_1(0) \\ sx_2(s) - x_2(0) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & -2\zeta\omega_n \end{bmatrix} \begin{bmatrix} x_1(s) \\ x_2(s) \end{bmatrix} + \begin{bmatrix} 0 \\ \omega_n^2 \end{bmatrix} u(s)$$

$$\begin{bmatrix} y_1(s) \\ y_2(s) \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1(s) \\ x_2(s) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u(s) = \begin{bmatrix} x_1(s) \\ x_2(s) \end{bmatrix}$$

Dynamic Equation

Output Equation

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Laplace Transform of the State Vector Response to Initial Condition and Control

Rearrange Laplace Transform of Dynamic Equation

$$s\mathbf{x}(s) - \mathbf{F}\mathbf{x}(s) = \mathbf{x}(0) + \mathbf{G}\mathbf{u}(s)$$

$$[s\mathbf{I} - \mathbf{F}]\mathbf{x}(s) = \mathbf{x}(0) + \mathbf{G}\mathbf{u}(s)$$

$$\mathbf{x}(s) = [s\mathbf{I} - \mathbf{F}]^{-1} [\mathbf{x}(0) + \mathbf{G}\mathbf{u}(s)]$$

The matrix inverse is

$$[s\mathbf{I} - \mathbf{F}]^{-1} = \frac{\text{Adj}(s\mathbf{I} - \mathbf{F})}{|s\mathbf{I} - \mathbf{F}|} \quad (n \times n)$$

$\text{Adj}(s\mathbf{I} - \mathbf{F})$: Adjoint matrix $(n \times n)$ Transpose of matrix of cofactors

$|s\mathbf{I} - \mathbf{F}| = \det(s\mathbf{I} - \mathbf{F})$: Determinant (1×1)

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Characteristic Polynomial of a Dynamic System

Matrix Inverse

$$[s\mathbf{I} - \mathbf{F}]^{-1} = \frac{\text{Adj}(s\mathbf{I} - \mathbf{F})}{|s\mathbf{I} - \mathbf{F}|} \quad (n \times n)$$

Characteristic matrix of the system

$$(s\mathbf{I} - \mathbf{F}) = \begin{pmatrix} (s - f_{11}) & -f_{12} & \dots & -f_{1n} \\ -f_{21} & (s - f_{22}) & \dots & -f_{2n} \\ \dots & \dots & \dots & \dots \\ -f_{n1} & -f_{n2} & \dots & (s - f_{nn}) \end{pmatrix} \quad (n \times n)$$

Characteristic polynomial of the system

$$\begin{aligned} |s\mathbf{I} - \mathbf{F}| &= \det(s\mathbf{I} - \mathbf{F}) \\ &\equiv \Delta(s) = s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0 \end{aligned}$$

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Eigenvalues

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Eigenvalues of the System

Characteristic equation of the system

$$\Delta(s) = s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0 = 0$$

$$= (s - \lambda_1)(s - \lambda_2)(\dots)(s - \lambda_n) = 0$$

Eigenvalues, λ_i , are solutions (roots) of the polynomial, $\Delta(s) = 0$

$$\lambda_i = \sigma_i + j\omega_i$$

$$\lambda_i^* = \sigma_i - j\omega_i$$



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Factors of a 2nd-Degree Characteristic Equation

$$|s\mathbf{I} - \mathbf{F}| = \begin{vmatrix} (s - f_{11}) & -f_{12} \\ -f_{21} & (s - f_{22}) \end{vmatrix} \triangleq \Delta(s)$$

$$= s^2 - (f_{12} + f_{21})s + (f_{11}f_{22} + f_{12}f_{21})$$

$$= (s - \lambda_1)(s - \lambda_2) = 0 \text{ [real or complex roots]}$$

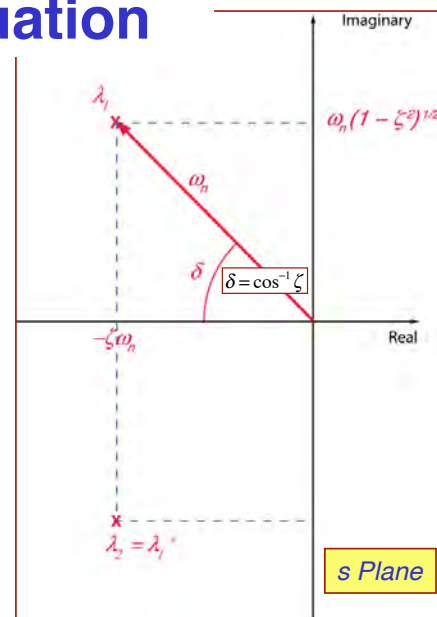
$$= s^2 + 2\zeta\omega_n s + \omega_n^2 \text{ with complex-conjugate roots}$$

$$\lambda_1 = \sigma_1, \quad \lambda_2 = \sigma_2$$

$$\lambda_1 = \sigma_1 + j\omega_1$$

$$\lambda_2 = \sigma_1 - j\omega_1$$

ω_n : natural frequency, rad/s
 ζ : damping ratio, dimensionless



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z Transforms and Discrete-Time Systems

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Application of Dirac Delta Function to Sampling Process

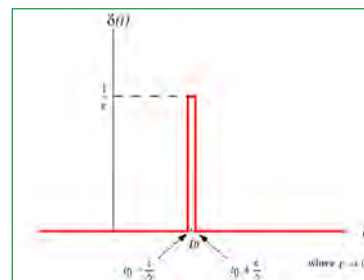
- Periodic sequence of numbers

$$\Delta x_k = \Delta x(t_k) = \Delta x(k\Delta t)$$

- Dirac delta function

$$\delta(t_0 - k\Delta t) = \begin{cases} \infty, & (t_0 - k\Delta t) = 0 \\ 0, & (t_0 - k\Delta t) \neq 0 \end{cases}$$

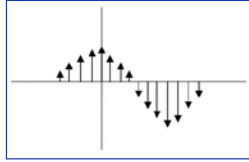
$$\int_{(t_0 - k\Delta t) - \varepsilon}^{(t_0 - k\Delta t) + \varepsilon} \delta(t_0 - k\Delta t) dt = 1$$



- Periodic sequence of scaled delta functions

$$\Delta x(k\Delta t) \delta(t_0 - k\Delta t)$$

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Laplace Transform of a Periodic Scalar Sequence

- **Periodic sequence of numbers**
- **Periodic sequence of scaled delta functions**
- **Laplace transform of the delta function sequence**

$$\Delta x_k = \Delta x(t_k) = \Delta x(k\Delta t)$$

$$\Delta x(k\Delta t)\delta(t - k\Delta t)$$

$$\begin{aligned} L[\Delta x(k\Delta t)\delta(t - k\Delta t)] &= \Delta x(z) = \int_0^{\infty} \Delta x(k\Delta t)\delta(t - k\Delta t)e^{-s\Delta t} dt \\ &= \sum_{k=0}^{\infty} \Delta x(k\Delta t)e^{-sk\Delta t} \triangleq \sum_{k=0}^{\infty} \Delta x(k\Delta t)z^{-k} \end{aligned}$$

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z Transform of the Periodic Sequence

z transform is the Laplace transform of the delta function sequence

$$L[\Delta x(k\Delta t)\delta(t - k\Delta t)] = \sum_{k=0}^{\infty} \Delta x(k\Delta t)e^{-sk\Delta t} \triangleq \sum_{k=0}^{\infty} \Delta x(k\Delta t)z^{-k}$$

z Transform (time-shift) Operator

$$\begin{aligned} z &\triangleq e^{s\Delta t} \quad [\text{advance by one sampling interval}] \\ z^{-1} &\triangleq e^{-s\Delta t} \quad [\text{delay by one sampling interval}] \end{aligned}$$

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z Transform of a Discrete-Time Dynamic System

System equation in sampled time domain

$$\Delta \mathbf{x}_{k+1} = \Phi \Delta \mathbf{x}_k + \Gamma \Delta \mathbf{u}_k + \Lambda \Delta \mathbf{w}_k$$

Laplace transform of sampled-data system equation
(“**z Transform**”)

$$z\Delta \mathbf{x}(z) - \Delta \mathbf{x}(0) = \Phi \Delta \mathbf{x}(z) + \Gamma \Delta \mathbf{u}(z) + \Lambda \Delta \mathbf{w}(z)$$

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z Transform of a Discrete-Time Dynamic System

Rearrange

$$z\Delta \mathbf{x}(z) - \Phi \Delta \mathbf{x}(z) = \Delta \mathbf{x}(0) + \Gamma \Delta \mathbf{u}(z) + \Lambda \Delta \mathbf{w}(z)$$

Collect terms

$$(z\mathbf{I} - \Phi) \Delta \mathbf{x}(z) = \Delta \mathbf{x}(0) + \Gamma \Delta \mathbf{u}(z) + \Lambda \Delta \mathbf{w}(z)$$

Pre-multiply by inverse

$$\Delta \mathbf{x}(z) = (z\mathbf{I} - \Phi)^{-1} [\Delta \mathbf{x}(0) + \Gamma \Delta \mathbf{u}(z) + \Lambda \Delta \mathbf{w}(z)]$$

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Characteristic Matrix and Determinant of Discrete-Time System

$$\Delta \mathbf{x}(z) = (z\mathbf{I} - \Phi)^{-1} [\Delta \mathbf{x}(0) + \Gamma \Delta \mathbf{u}(z) + \Lambda \Delta \mathbf{w}(z)]$$

Inverse matrix

$$(z\mathbf{I} - \Phi)^{-1} = \frac{\text{Adj}(z\mathbf{I} - \Phi)}{|z\mathbf{I} - \Phi|} \quad (n \times n)$$

Characteristic polynomial of the discrete-time model

$$\begin{aligned} |z\mathbf{I} - \Phi| &= \det(z\mathbf{I} - \Phi) \equiv \Delta(z) \\ &= z^n + a_{n-1}z^{n-1} + \dots + a_1z + a_0 \end{aligned}$$

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Eigenvalues (or Roots) of the Discrete-Time System

Characteristic equation of the system

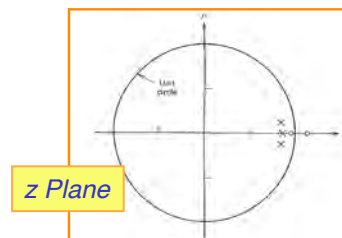
$$\begin{aligned} \Delta(z) &= z^n + a_{n-1}z^{n-1} + \dots + a_1z + a_0 \\ &= (z - \lambda_1)(z - \lambda_2)(\dots)(z - \lambda_n) = 0 \end{aligned}$$

Eigenvalues, λ_i , of the state transition matrix, Φ , are solutions (roots) of the polynomial, $\Delta(z) = 0$

Eigenvalues are complex numbers that can be plotted in the z plane

$$\lambda_i = \sigma_i + j\omega_i$$

$$\lambda_i^* = \sigma_i - j\omega_i$$



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Laplace Transforms of Continuous- and Discrete-Time State-Space Models

Initial condition and disturbance effect neglected

$$\begin{aligned}\Delta \mathbf{x}(s) &= (s\mathbf{I} - \mathbf{F})^{-1} \mathbf{G} \Delta \mathbf{u}(s) \\ \Delta \mathbf{y}(s) &= \mathbf{H} (s\mathbf{I} - \mathbf{F})^{-1} \mathbf{G} \Delta \mathbf{u}(s)\end{aligned}$$

Equivalent discrete-time model

$$\begin{aligned}\Delta \mathbf{x}(z) &= (z\mathbf{I} - \Phi)^{-1} \Gamma \Delta \mathbf{u}(z) \\ \Delta \mathbf{y}(z) &= \mathbf{H} (z\mathbf{I} - \Phi)^{-1} \Gamma \Delta \mathbf{u}(z)\end{aligned}$$

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Scalar Transfer Functions of Continuous- and Discrete-Time Systems

$$\begin{aligned}\dim(\mathbf{H}) &= 1 \times n \\ \dim(\mathbf{G}) &= n \times 1\end{aligned}$$

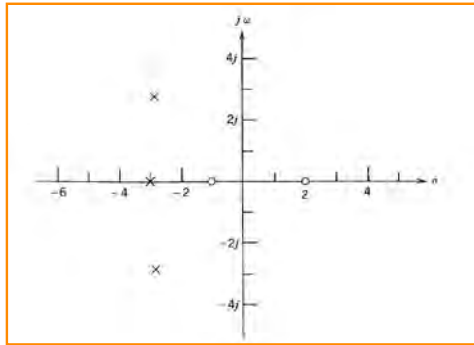
$$\frac{\Delta y(s)}{\Delta u(s)} = \mathbf{H} (s\mathbf{I} - \mathbf{F})^{-1} \mathbf{G} = \frac{\mathbf{H} \text{Adj}(s\mathbf{I} - \mathbf{F}) \mathbf{G}}{|s\mathbf{I} - \mathbf{F}|} = Y(s)$$

$$\frac{\Delta y(z)}{\Delta u(z)} = \mathbf{H} (z\mathbf{I} - \Phi)^{-1} \Gamma = \frac{\mathbf{H} \text{Adj}(z\mathbf{I} - \Phi) \Gamma}{|z\mathbf{I} - \Phi|} = Y(z)$$

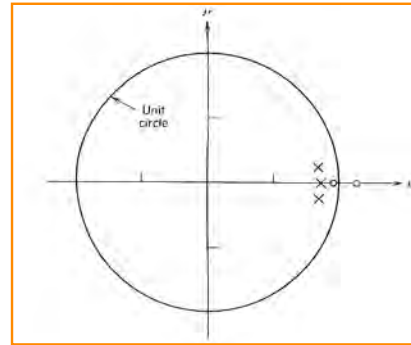
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Comparison of s-Plane and z-Plane Plots of Poles and Zeros

- **s-Plane Plot of Poles and Zeros**
 - Poles in **left-half-plane** are **stable**
 - Zeros in **left-half-plane** are **minimum phase**
- **z-Plane Plot of Poles and Zeros**
 - Poles within **unit circle** are **stable**
 - Zeros within **unit circle** are **minimum phase**



Note correspondence of configurations



Increasing sampling rate moves poles and zeros toward the (1,0) point

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Next Time:
Time-Invariant Linear-Quadratic Regulators

SUPPLEMENTARY MATERIAL

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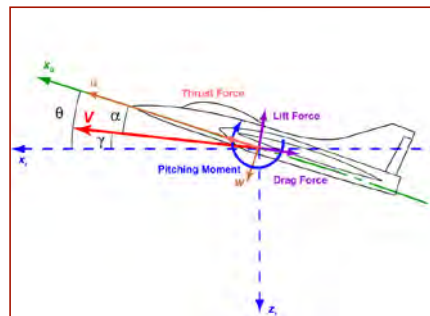
Small Perturbations from Steady, Level Flight

$$\Delta \dot{\mathbf{x}}(t) = \mathbf{F} \Delta \mathbf{x}(t) + \mathbf{G} \Delta \mathbf{u}(t) + \mathbf{L} \Delta \mathbf{w}(t)$$

$$\Delta \mathbf{x}(t) = \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \\ \Delta x_4 \end{bmatrix} = \begin{bmatrix} \Delta V \\ \Delta \gamma \\ \Delta q \\ \Delta \alpha \end{bmatrix} \quad \begin{array}{l} \text{velocity, m/s} \\ \text{flight path angle, rad} \\ \text{pitch rate, rad/s} \\ \text{angle of attack, rad} \end{array}$$

$$\Delta \mathbf{u}(t) = \begin{bmatrix} \Delta u_1 \\ \Delta u_2 \end{bmatrix} = \begin{bmatrix} \Delta \delta E \\ \Delta \delta T \end{bmatrix} \quad \begin{array}{l} \text{elevator angle, rad} \\ \text{throttle setting, \%} \end{array}$$

$$\Delta \mathbf{w}(t) = \begin{bmatrix} \Delta w_1 \\ \Delta w_2 \end{bmatrix} = \begin{bmatrix} \Delta V_w \\ \Delta \alpha_w \end{bmatrix} \quad \begin{array}{l} \sim \text{horizontal wind, m/s} \\ \sim \text{vertical wind}/V_{\text{nom}}, \text{ rad} \end{array}$$



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Eigenvalues of Aircraft Longitudinal Modes of Motion



$$\begin{aligned}
 |s\mathbf{I} - \mathbf{F}| &= \det(s\mathbf{I} - \mathbf{F}) \equiv \Delta(s) = (s - \lambda_1)(s - \lambda_2)(s - \lambda_3)(s - \lambda_4) \\
 &= (s - \lambda_P)(s - \lambda_P^*)(s - \lambda_{SP})(s - \lambda_{SP}^*) \\
 &= (s^2 + 2\zeta_P\omega_{n_P}s + \omega_{n_P}^2)(s^2 + 2\zeta_{SP}\omega_{n_{SP}}s + \omega_{n_{SP}}^2) = 0
 \end{aligned}$$

Eigenvalues determine the damping and natural frequencies of the linear system's modes of motion

$$\begin{aligned}
 (\zeta_P, \omega_{n_P}) &: \text{phugoid (long-period) mode} \\
 (\zeta_{SP}, \omega_{n_{SP}}) &: \text{short-period mode}
 \end{aligned}$$

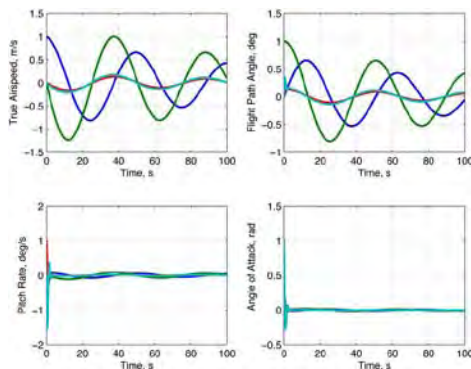
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Initial-Condition Response of Business Jet at Two Time Scales

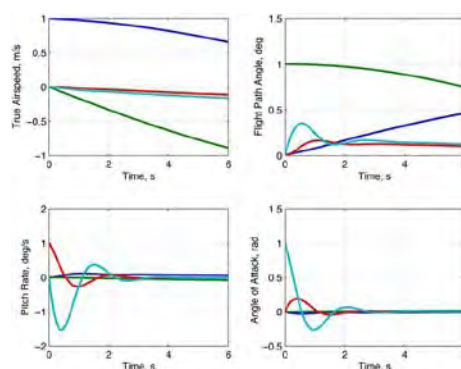
$$\Delta \mathbf{x}(t) = \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \\ \Delta x_4 \end{bmatrix} = \begin{bmatrix} \Delta V \\ \Delta \gamma \\ \Delta q \\ \Delta \alpha \end{bmatrix} \quad \begin{array}{l} \text{velocity, m/s} \\ \text{flight path angle, rad} \\ \text{pitch rate, rad/s} \\ \text{angle of attack, rad} \end{array}$$

Same 4th-order responses viewed over different periods of time

- 0 - 100 sec
- Reveals Long-Period Mode



- 0 - 6 sec
- Reveals Short-Period Mode



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