

SigSys II Summary

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30.07.14

1. Fundamentals

\mathbb{Z} : integers $\{ \dots, -2, -1, 0, 1, 2, \dots \}$

\mathbb{N} : natural numbers $\{ 0, 1, 2, \dots \}$

\mathbb{C} : complex numbers $\{ z_1 + z_2 * i \mid z_1, z_2 \in \mathbb{R} \}$

$[a, b] = \{ x \in \mathbb{R} \mid a \leq x \leq b \}$ $(a, b) = \{ x \in \mathbb{R} \mid a < x < b \}$

$[a, b) = \{ x \in \mathbb{R} \mid a \leq x < b \}$ $(a, b] = \{ x \in \mathbb{R} \mid a < x \leq b \}$

$(a, \infty) = \{ x \in \mathbb{R} \mid a < x \}$ $(-\infty, b] = \{ x \in \mathbb{R} \mid x \leq b \}$

Time State	Discrete	Continuous	Hybrid
Discrete	Finite state machines, Turing machines	Queuing systems	
Continuous	z-transform $x_{k+1} = Ax_k + Bu_k$ $y_k = Cx_k + Du_k$	Laplace transform $\dot{x}(t) = Ax(t) + Bu(t)$ $y(t) = Cx(t) + Du(t)$	Impulse differential inclusions
Hybrid	Mixed Logic-Dynamical systems	Switching diffusions	Hybrid automata

System properties

Time invariant : A system in state space form is called time invariant if its dynamics do not depend explicitly on time

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

Autonomous : A system is called autonomous if it is time invariant and has no input variables

$$\dot{x}(t) = f(x(t)), \quad y(t) = h(x(t))$$

Linear : A system in state space form is called linear if the functions f and h are linear

$$f(a_1 x_1 + a_2 x_2) = a_1 f(x_1) + a_2 f(x_2)$$

Dimension & higher order differential equations

Dimension/Order n : number of states

Terms of higher order differential equations can be converted to state space form by defining lower order derivatives (all except the highest ones) as own states

Make time-invariant: add time as new state, $\dot{t} = 1$

Mechanical systems: position and velocity (Newton)

Electrical systems: capacitor voltages, inductor currents

$$u_L = L * \frac{di_L}{dt}, \quad i_C = C * \frac{du_C}{dt}$$

Newton's law

$$m \ddot{x} = \sum F_i + \gamma \dot{x}, \quad \gamma : \text{damping constant}$$

$$m * r * \ddot{\theta} = \sum F_{\text{tangential}}, \quad \text{rotation}$$

Laplace transformation & convulsion

Implicitly assume all signals = 0 for $t < 0$

$$\mathcal{L}\{u(t)\} = U(s) = \int_{-\infty}^{\infty} u(t) e^{-st} dt = \int_0^{\infty} u(t) e^{-st} dt$$

$$(u * h)(t) = \int_{-\infty}^{\infty} u(\tau) h(t - \tau) d\tau = \int_0^t u(\tau) h(t - \tau) d\tau$$

Lipschitz functions

Lipschitz functions are continuous but not necessarily differentiable. All differentiable functions with bounded derivatives are Lipschitz, therefore also all linear functions.

$$\exists \lambda > 0, \forall x, \hat{x}: \|f(x) - f(\hat{x})\| \leq \lambda \|x - \hat{x}\|$$

Unique solution: If f is Lipschitz, then is

$$\dot{x}(t) = f(x(t)) \text{ has a unique solution } \forall T \geq 0, x_0 \in \mathbb{R}^n$$

Continuity: If f is Lipschitz, then the solutions starting at

$x_0, \hat{x}_0 \in \mathbb{R}^n$ are such that for all $t \geq 0$

$$\|x(t) - \hat{x}(t)\| \leq e^{\lambda t} \|x_0 - \hat{x}_0\|$$

2. Linear algebra

Transpose $(A B)^T = B^T A^T$

2-norm $\|x\|^2 = x^T x = \sum x_i^2$
 $\|x + y\| \leq \|x\| + \|y\|, \|a * x\| = |a| * \|x\| \geq 0$

Inner product $\langle x, y \rangle = x^T y = \sum x_i y_i$
orthogonal: $\langle x, y \rangle = 0$; *orthonormal*: $\|x\| = 1$

Subspace $S \subseteq \mathbb{R}^n$, if $\forall x, y \in S : a x + b y \in S$

Range space $\text{range}(A) = \{ y \in \mathbb{R}^n \mid \exists x \in \mathbb{R}^m, y = A x \}$
 $= \text{span} \{ a_1, a_2, \dots, a_m \}$

Rank *dimension of range(A)*
number of linearly independent columns

Null space $\text{null}(A) = \{ x \in \mathbb{R}^m \mid A x = 0 \}$

Orthogonal $A A^T = A^T A = I \rightarrow A^T = A^{-1}$
 $\|A x\| = \|x\|$ (norm unchanged)

Symmetric $A = A^T$ (real E-Val., orthogonal E-Vect.)

Inverse of a matrix

$$A^{-1} A = A A^{-1} = I$$

If an inverse of A exists, it is unique and A is invertible.

If no inverse of A exists, A is called singular.

A is invertible: **det(A) ≠ 0**

$Ax = y$ has a unique solution

$\text{range}(A) = \mathbb{R}^n, \text{rang}(A) = n$ (full)

$\text{null}(A) = \{0\}$

all Eigenvalues are non – zero

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Eigenvalue & eigenvectors

$$A v = \lambda v, \quad v : \text{eigenvector}, \lambda : \text{eigenvalue}$$

$$\det(\lambda I - A) = \lambda^n + a_1 \lambda^{n-1} + \dots + a_n = 0$$

Spectrum $\text{span} \{ v_1, v_2, \dots, v_n \}$

Cayley-Hamilton: Every matrix A is a solution of its characteristic polynomial

$$A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I = 0$$

All powers A^m , $m = 0, 1, \dots$ can be written as linear combinations of A, A^2, \dots, A^{n-1} .

Diagonalizable Matrix: A is diagonalizable, if

$$i) W = \{ v_1 v_2 \dots v_n \} \text{ is invertible: } A = W \Lambda W^{-1}$$

$$ii) \text{ all eigenvalues are distinct } (\lambda_i \neq \lambda_j \forall i \neq j)$$

$$iii) \text{ all eigenvectors are linearly independent}$$

Positive definite

$$x^T A x > 0 \leftrightarrow \text{only real positive eigenvalues}$$

Positive semi-definite: $x^T A x \geq 0 \leftrightarrow \text{non-neg. eigenvalues}$

$$A \geq 0 : A \text{ symmetric \& positive semi-definite}$$

Coordinate transformation

Coordinate change: $\hat{x}(t) = T x(t), T \in \mathbb{R}^{n \times n}, \det(T) \neq 0$

$$\rightarrow \dot{\hat{x}}(t) = T A T^{-1} \hat{x}(t) + T B u(t)$$

$$\rightarrow y(t) = C T^{-1} \hat{x}(t) + D u(t)$$

Properties stay the same:

$$EV(A) = EV(\hat{A}) = EV(T A T^{-1}), \quad e^{\bar{\lambda}_i T} = \lambda_i$$

$$A \text{ observable} \leftrightarrow \hat{A} \text{ observable}$$

$$A \text{ controllable} \leftrightarrow \hat{A} \text{ controllable}$$

3. Continuous LTI systems: time domain

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

$$y(t) = C x(t) + D u(t) = h(x(t), u(t))$$

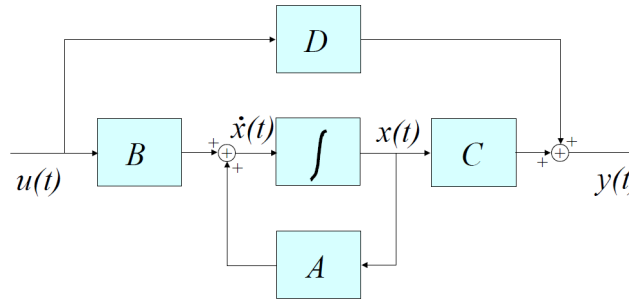
Energy stored in the system Q positive definite & symmetric

$$E(t) = \frac{1}{2} x(t)^T Q x(t)$$

Power (instantaneous energy change) for $u(t) = 0$

$$P(t) = \frac{1}{2} x(t)^T (A^T Q + Q A) x(t)$$

Block diagram representation



State transition matrix

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

Properties:

$$i) \quad \Phi(0) = I$$

$$ii) \quad \frac{d}{dt} \Phi(t) = A \Phi(t)$$

$$iii) \quad \Phi(-t) = [\Phi(t)]^{-1}$$

$$iv) \quad \Phi(t_1 + t_2) = \Phi(t_1) \Phi(t_2)$$

Computation through Eigenvalue decomposition

$$\Phi(t) = e^{At} = W e^{\Lambda t} W^{-1}$$

System solutions

State solution

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

Zero Input Transition (ZIT) : $u(t) = 0 \forall t \rightarrow x(t) = ZIT$

Zero State Transition (ZST) : $x_0 = 0 \rightarrow x(t) = ZST$

Output solution

$$y(t) = C \Phi(t) x_0 + \int_0^t C \Phi(t - \tau) B u(\tau) d\tau + D u(t)$$

Zero Input Response (ZIR) : $u(t) = 0 \forall t \rightarrow x(t) = ZIR$

Zero State Response (ZSR) : $x_0 = 0 \rightarrow x(t) = ZSR$

Stability

Stable: A system is called stable if $\forall \varepsilon > 0 \exists \delta > 0$

$$\|x_0\| < \delta \rightarrow \|x(t)\| \leq \varepsilon \quad \forall t \geq 0$$

Else, it is called **unstable** and diverges towards infinity.

Asymptotically stable: If it is stable and furthermore

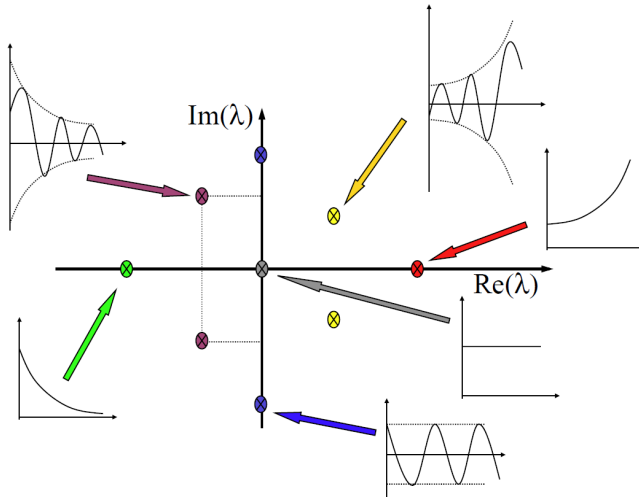
$$\|x(t)\| \rightarrow 0 \quad \text{for } t \rightarrow \infty$$

Diagonalisable matrix A: distinct E-Values

- *stable* $\Leftrightarrow \text{Re}[\lambda_i] \leq 0 \forall i$
- *asymptotically stable* $\Leftrightarrow \text{Re}[\lambda_i] < 0 \forall i$
- *unstable* $\Leftrightarrow \exists i : \text{Re}[\lambda_i] > 0$

Non-diagonalisable matrix A:

- *asymptotically stable* $\Leftrightarrow \text{Re}[\lambda_i] < 0 \forall i$
- *unstable* $\Leftrightarrow \exists i : \text{Re}[\lambda_i] > 0$



Impulse transition H(t)

$n = m = 1 : u(t) = \delta(t), x_0 = 0$

$$h(t) = \int_0^t \Phi(t-\tau) B \delta(\tau) d\tau = e^{at} b \rightarrow x(t) = (h * u)(t)$$

General n, m :

$$H(t) = \Phi(t) B \rightarrow x(t) = (H * u)(t)$$

Output impulse response K(t)

Input: $u(t) = \delta(t), x_0 = 0$

$$K(t) = C \Phi(t) B + D \delta(t) \rightarrow y(t) = (K * u)(t)$$

Stability with inputs

Assume that $\text{Re}[\lambda_i] < 0 \forall i$. Then there exists $\alpha \geq 0$ such that $ZST, x(t)$, satisfies

$$\|u(t)\| \leq M \quad \forall t \geq 0 \Rightarrow \|x(t)\| \leq \alpha M \quad \forall t \geq 0$$

$$u(t) \xrightarrow{t \rightarrow \infty} 0 \Rightarrow x(t) \xrightarrow{t \rightarrow \infty} 0$$

Small input leads to small states. Input goes to zero, so does the state.

4. Energy, Controllability, Observability

Energy stored in the system Q positive definite & symmetric

$$E(t) = \frac{1}{2} x(t)^T Q x(t)$$

Power (instantaneous energy change) for $u(t) = 0$

$$P(t) = \frac{1}{2} x(t)^T (A^T Q + Q A) x(t)$$

If R is positive definite, then energy always decreases.

Lyapunov equation

$$A^T Q + Q A = -R$$

Lyapunov function: energy-like function

$$V(x) = \frac{1}{2} x^T Q x$$

Minimum energy inputs

Assume that the system is controllable. Given $x_1 \in \mathbb{R}^n$, $t > 0$, the input that drives the system from $x(0) = 0$ to $x(t) = x_1$ and has the **minimum energy** is given by

$$u_m(\tau) = B^T e^{A^T(t-\tau)} W_C(t)^{-1} x_1 \quad \forall \tau \in [0, t]$$

The faster we go, the more energy & the bigger input is needed.

Controllability

The system is controllable over $[0, t]$, if

\forall initial conditions $x(0) = x_0$, terminal condition x_1

$$\exists u(\cdot) : [0, t] \text{ such that } x(t) = x_1$$

i.e. : **The system can be lead from any state to any state.**

For any x_0, x_1 , we can find $u(t)$ so that

$$x_1 = e^{At} x_0 + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau$$

Controllability gramian

$$W_C(t) = \int_0^t e^{A\tau} B B^T e^{A^T \tau} d\tau$$

system controllable over $[0, t] \Leftrightarrow W_C$ invertible

Controllability matrix P

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

system controllable over $[0, t] \Leftrightarrow \text{rank}(P)$ is n / full

Reachable states: $\{x_1 \mid \exists u(t) : 0 \rightarrow x_1\} = \text{Range}(P)$

Observability

The system is observable over $[0, t]$, if given input $u(\cdot)$ and output $y(\cdot)$, we can uniquely determine states $x(\cdot)$.

System is observable $\Leftrightarrow x = 0$ only unobservable state

$$x \in \mathbb{R}^n \text{ unobservable} \Leftrightarrow C e^{A\tau} x = 0 \quad \forall \tau \in [0, t]$$

Observability matrix

$$Q = \begin{bmatrix} C \\ C A \\ \vdots \\ C A^{n-1} \end{bmatrix} \in \mathbb{R}^{np \times n}$$

system observable over $[0, t] \Leftrightarrow \text{rank}(Q)$ is n / full

Observability gramian

$$W_O(t) = \int_0^t e^{A^T \tau} C^T C e^{A \tau} d\tau$$

system observable over all $[0, t'] \Leftrightarrow W_O$ invertible

Unobservable states: $\text{Null}(Q) = \{x \in \mathbb{R}^n \mid Qx = 0\}$

Observers

Filter that progressively constructs an estimate of the state

1. Start with an initial guess $\tilde{x}(t) \in \mathbb{R}^n$

2. Update estimate according to

$$\frac{d\tilde{x}(t)}{dt} = A\tilde{x}(t) + Bu(t) + L[y(t) - C\tilde{x}(t) - Du(t)]$$

Error dynamics:

$$e(t) = x(t) - \tilde{x}(t) \rightarrow \dot{e}(t) = (A - LC)e(t)$$

If the system is observable, L can be chosen such that eigenvalues of $(A - LC)$ have negative real parts.

→ system asymptotically stable

$$\rightarrow e(t) \xrightarrow{t \rightarrow \infty} 0, \quad \tilde{x}(t) \xrightarrow{t \rightarrow \infty} x(t)$$

Kalman decomposition

Optimal trade-off for L if noise corruption & linear system

There exists change of coordinates $T \in \mathbb{R}^{n \times n}$ invertible such that:

$$\hat{x}(t) = T\tilde{x}(t) = \begin{bmatrix} \hat{x}_1(t) \\ \hat{x}_2(t) \\ \hat{x}_3(t) \\ \hat{x}_4(t) \end{bmatrix} \begin{matrix} \leftarrow \text{controllable \& observable} \\ \leftarrow \text{controllable \& unobservable} \\ \leftarrow \text{uncontrollable \& observable} \\ \leftarrow \text{uncontrollable \& unobservable} \end{matrix}$$

$$\hat{A} = TAT^{-1} = \begin{bmatrix} \hat{A}_{11} & 0 & \hat{A}_{13} & 0 \\ \hat{A}_{21} & \hat{A}_{22} & \hat{A}_{23} & \hat{A}_{24} \\ 0 & 0 & \hat{A}_{33} & 0 \\ 0 & 0 & \hat{A}_{43} & \hat{A}_{44} \end{bmatrix}, \hat{B} = TB = \begin{bmatrix} \hat{B}_1 \\ \hat{B}_2 \\ 0 \\ 0 \end{bmatrix}$$

$$\hat{C} = CT^{-1} = \begin{bmatrix} \hat{C}_1 & 0 & \hat{C}_3 & 0 \end{bmatrix}$$

Detectable: all eigenvalues of \hat{A}_{22} & \hat{A}_{44} in the Kalman decomposition have negative real part.

→ can design observer with observation error decaying to zero

Stabilizable: all eigenvalues of \hat{A}_{33} & \hat{A}_{44} in the Kalman decomposition have negative real part.

→ can design controller that ensures system asymptotically stable

5. Continuous LTI systems: frequency domain

Laplace transform

$$F(s) = \mathcal{L}\{f(t)\} = \int_0^{\infty} f(t) e^{-st} dt, \quad f(t) = 0 \quad \forall t \leq 0$$

Properties

Linearity: $\mathcal{L}\{\alpha f(t) + \beta g(t)\} = \alpha F(s) + \beta G(s)$

S shift: $\mathcal{L}\{e^{-at} f(t)\} = F(s + a)$

Time derivative: $\mathcal{L}\left\{\frac{d}{dt} f(t)\right\} = sF(s) - f(0)$

$$\mathcal{L}\left\{\frac{d^2 f(t)}{dt^2}\right\} = s^2 F(s) - s f(0) - f'(0)$$

Convolution: $\mathcal{L}\{(f * g)(t)\} = F(s) G(s)$

Useful functions

Dirac: $\mathcal{L}\{\delta(t)\} = 1$

Step: $\mathcal{L}\{1\} = 1/s$

Exponential: $\mathcal{L}\{e^{-at}\} = \frac{1}{s+a}$
 $\mathcal{L}\{t e^{-at}\} = \frac{1}{(s+a)^2}$

Sinus: $\mathcal{L}\{\sin(\omega t)\} = \frac{\omega}{s^2 + \omega^2}$

Cosine: $\mathcal{L}\{\cos(\omega t)\} = \frac{s}{s^2 + \omega^2}$

Ramp: $\mathcal{L}\{u(t) = \begin{cases} 0, & t \leq 0 \\ t, & t > 0 \end{cases}\} = 1/s^2$

Inverse Laplace transform: use partial fraction

Initial / final value theorems

Whenever limits exist:

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

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

Steady state: $t \rightarrow \infty$

LTI systems in frequency domain

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

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

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

Transfer function

$$G(s) = C(sI - A)^{-1} B + D = \frac{(s - z_1) \dots (s - z_k)}{(s - p_1) \dots (s - p_n)}$$

For $x_0 = 0$: $Y(s) = G(s) U(s)$

Proper: numerator degree \leq denominator degree

Strictly proper: $k < n$

SISO (Single Input Single Output) transfer functions arising from state space descriptions of LTI systems are always *proper*. They are *strictly proper* if and only if $D = 0$.

If no pole-zero cancellations were performed, the **denominator is the characteristic polynomial of A**
↔ the poles are the eigenvalues of A

The transfer function is the Laplace transform of the output impulse response ($u(t) = \delta(t)$).

Stability

Provided there are no pole-zero cancellations!

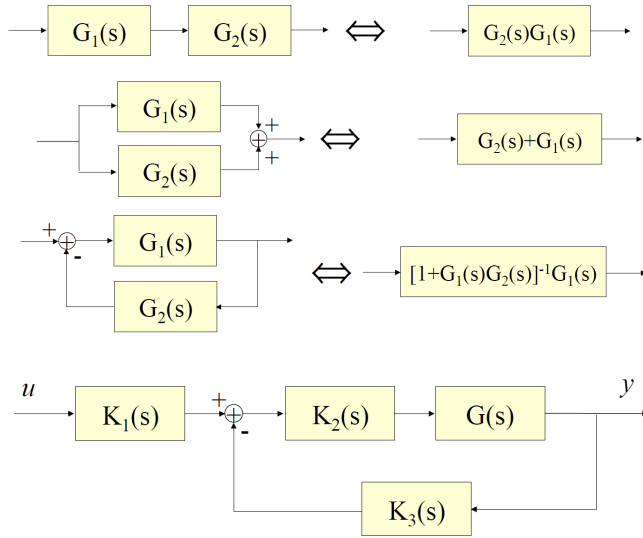
Distinct poles:

- Asymptotically stable $\Leftrightarrow \text{Re}[p_i] < 0 \quad \forall i$
- stable $\Leftrightarrow \text{Re}[p_i] \leq 0 \quad \forall i$
- unstable $\Leftrightarrow \exists i : \text{Re}[p_i] > 0$

Repeated poles:

- Asymptotically stable $\Leftrightarrow \text{Re}[p_i] < 0 \quad \forall i$
- unstable $\Leftrightarrow \exists i : \text{Re}[p_i] > 0$

Block diagrams



$$Y(s) = [1 + G(s)K_2(s)K_3(s)]^{-1} G(s)K_2(s)K_1(s)U(s)$$

Frequency response

Response to a sinusoidal input is also sinusoidal:

Frequency: ω (same)

Amplitude: $K = |G(j\omega)| = \sqrt{\text{Re}[G(j\omega)]^2 + \text{Im}[G(j\omega)]^2}$

Phase: $\Phi = \angle G(j\omega) = \tan^{-1} \left(\frac{\text{Im}[G(j\omega)]}{\text{Re}[G(j\omega)]} \right)$

Niquest plot: $G(j\omega)$ in polar coordinates, parameterized by ω

Resonance

Appears in second order systems (two poles)

$$G(s) = \frac{K \omega_n^2}{s^2 + 2\zeta \omega_n s + \omega_n^2}, \quad \omega_n > 0$$

Frequency response

$$|G(j\omega)| = \frac{K \omega_n^2}{\sqrt{(\omega_n^2 - \omega^2)^2 + (2\zeta \omega_n \omega)^2}}$$

$$\angle G(j\omega) = -\tan^{-1} \left(\frac{2\zeta \omega_n \omega}{\omega_n^2 - \omega^2} \right)$$

Properties

- $\zeta \geq 0$: required for stability
- $\zeta \geq 1$: poles real (over-damped system)
- $\zeta = 1$: poles real and equal (critical damp.)
- $0 < \zeta < 1$: poles complex (under-damp.)
- $\zeta = 0$: poles imaginary (undamped system)
- $\zeta \geq \frac{1}{\sqrt{2}}$: magnitude Bode plot decreasing in ω
- $0 \leq \zeta \leq \frac{1}{\sqrt{2}}$: magnitude Bode plot has Max. at

$$\omega = \omega_n \sqrt{1 - 2\zeta^2}, \quad |G(j\omega)| = \frac{K}{2\zeta \sqrt{1 - \zeta^2}}$$

Transfer function realization

Finding the state base description from the transfer funct.

$$G(s) = \frac{(s - z_1)(s - z_2) \cdots (s - z_k)}{(s - p_1)(s - p_2) \cdots (s - p_n)} \xrightarrow{?} \begin{cases} \frac{dx}{dt}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

SISO, strictly proper system

$$G(s) = \frac{b_1 s^{n-1} + b_2 s^{n-2} + \cdots + b_n}{s^n + a_1 s^{n-1} + a_2 s^{n-2} + \cdots + a_n}$$

$$\begin{array}{c|c} \begin{array}{l} \dot{x}(t) = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ -a_n & -a_{n-1} & -a_{n-2} & \cdots & -a_1 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} b_n & b_{n-1} & b_{n-2} & \cdots & b_1 \end{bmatrix} x(t) \end{array} & \begin{array}{l} \text{Controllable canonical form} \end{array} \end{array} \quad \left| \quad \begin{array}{c|c} \begin{array}{l} \dot{x}(t) = \begin{bmatrix} 0 & 0 & \cdots & -a_n \\ 1 & 0 & \cdots & -a_{n-1} \\ 0 & 1 & \cdots & -a_{n-2} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & -a_1 \end{bmatrix} x(t) + \begin{bmatrix} b_n \\ b_{n-1} \\ b_{n-2} \\ \vdots \\ b_1 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} 0 & 0 & \cdots & 1 \end{bmatrix} x(t) \end{array} & \begin{array}{l} \text{Observable canonical form} \end{array} \end{array}$$

This realisation is obviously *not unique*!

Pole-zero cancellation can render a system that was originally uncontrollable/unobservable into one that may look controllable/observable !

6. Discrete time LTI systems

$$x_{k+1} = A x_k + B u_k, \quad y_k = C x_k + D u_k$$

$$A = e^{\bar{A}t}, \quad B = \int_0^T e^{\bar{A}(T-\tau)} \bar{B} d\tau, \quad C = \bar{C}, \quad D = \bar{D}$$

Solution of discrete time linear systems

$$x_k = A^k \hat{x}_0 + \sum_{i=0}^{k-1} A^{k-i-1} B u_i$$

Diagonalizable matrix A

$$A = W \Lambda W^{-1} \rightarrow A^k = W \Lambda^k W^{-1}$$

- asymptotically stable $\Leftrightarrow |\lambda_i| < 1 \quad \forall i$
- stable $\Leftrightarrow |\lambda_i| \leq 1 \quad \forall i$
- unstable $\Leftrightarrow \exists i : |\lambda_i| > 1$

Non-diagonalizable matrix A

- asymptotically stable $\Leftrightarrow |\lambda_i| < 1 \quad \forall i$
- unstable $\Leftrightarrow \exists i : |\lambda_i| > 1$

Deadbeat response

The goal is to find the input that drives the system to the steady state in the smallest amount of time.

Nilpotent matrix: $A^k = 0 \quad \forall k \geq N \rightarrow x_k = 0$

Coordinate change

Transformation: $\hat{x}_k = T x_k, T$ invertible

$$\begin{aligned} \hat{x}_{k+1} &= T A T^{-1} \hat{x}_k + T B u_k \\ y_k &= C T^{-1} \hat{x}_k + D u_k \end{aligned}$$

Energy and Power

Energy ($Q = Q^T > 0$): $V(x) = \frac{1}{2} x^T Q x$

Power (change of energy): $V(x_{k+1}) = \frac{1}{2} x_{k+1}^T Q x_{k+1}$

Stability and energy

$$R = -(A^T Q A - Q)$$

If $R = R^T > 0$, then the energy decreases all the time.
Therefore, it is natural to assume that the system is stable.

If $u_k = 0$ (autonomous system)

$$V(x_{k+1}) - V(x_k) = -\frac{1}{2} x_k^T R x_k$$

Controllability

$$P = [B \ AB \ A^2 B \ \dots \ A^{n-1} B]$$

system is controllable $\Leftrightarrow \text{rank}(P)$ is n / full

Observability

$$Q = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}$$

system is observable $\Leftrightarrow \text{rank}(Q) = n$ / full

Z-Transform

$$F(z) = Z\{f_k\} = \sum_{k=0}^{\infty} f_k z^{-k}, \quad f_k = 0 \quad \forall k < 0$$

Properties

Linearity: $Z\{\alpha f_k + \beta g_k\} = \alpha F(z) + \beta G(z)$

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

Convolution: $Z\{(f * g)_k\} = Z\{\sum_{i=0}^k f_i g_{k-i}\} = F(z) G(z)$

Useful functions

Impulse: $Z\{\delta_k\} = 1 \quad (\delta_0 = 1, \delta_k = 0 \quad \forall k \neq 0)$

Step: $Z\{1_k\} = \frac{z}{z-1}$

Geometric progression: $Z\{a^k\} = \frac{z}{z-a} \quad (|a| < 1)$

Transfer function

Assume $x_0 = 0$

$$Y(z) = [C(zI - A)^{-1}B + D]U(z)$$

Transfer function is z-transform of “impulse response”

$$G(z) = C(zI - A)^{-1}B + D$$

system asymptotically stable $\Leftrightarrow |\text{Poles of } G(z)| < 1$

If there are pole – zero cancellations
 \Leftrightarrow system is uncontrollable/unobservable

Numerical approximations

Forward Euler method

$$x_{k+1} = (I + A\delta)x_k + \delta B u_k, \quad \delta = \frac{T}{N}$$

Backward Euler method

$$\begin{aligned} \dot{x} = Ax &\rightarrow x_{k+1} \approx x_k + \delta A x_{k+1} \\ \Rightarrow x_{k+1} &\approx (I - \delta A)^{-1} x_k \end{aligned}$$

Step width δ

Assume A diagonalizable, E-Values real & negative

asymptotically stable $\Leftrightarrow |1 + \lambda_i \delta| < 1 \quad \forall i$

$$\delta < \frac{2}{\max_{i=1,\dots,n} |\lambda_i|}$$

7. Nonlinear systems

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

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

We concentrate on autonomous, time-invariant systems:

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

f Lipschitz \Rightarrow existence and uniqueness of solution

Invariant: A set of states $S \subseteq \mathbb{R}^n$ is called invariant, if

$$\forall x_0 \in S, \forall t \geq 0: \quad x(t) \in S$$

Equilibria

A state \hat{x} is called **equilibrium**, if $f(\hat{x}) = 0 = \dot{x}(\hat{x})$

For linear systems, the linear subspace of equilibria coincides with the null space of A.

It is often convenient to “shift” an equilibrium to the origin before analysing the system behaviour:

$$w(t) = x(t) - \hat{x} \in \mathbb{R}^n$$

In the new coordinates, the system then becomes:

$$\dot{w}(t) = \dot{x}(t) = f(x(t)) = f(w(t) + \hat{x}) = \hat{f}(w(t))$$

Limit cycles

A solution $x(t)$ is called a **periodic orbit**, if

$$\exists T > 0, \forall t \geq 0: \quad x(t+T) = x(t)$$

An equilibrium defines a trivial periodic cycle.

Chaotic attractor

Given any two points in an invariant set, we can find a trajectory that starts arbitrarily close to one and ends arbitrarily close to the other.

Stability

If we start close, we stay close

An equilibrium \hat{x} is called **stable**, if $\forall \varepsilon > 0 \exists \delta > 0$:

$$\|x_0 - \hat{x}\| < \delta \Rightarrow \|x(t) - \hat{x}\| < \varepsilon \quad \forall t \geq 0$$

An equilibrium \hat{x} is called **locally asymptotically stable** if it is stable and there exists $M > 0$ such that:

$$\|x_0 - \hat{x}\| < M \Rightarrow \lim_{t \rightarrow \infty} x(t) = \hat{x}$$

It is called **globally asymptotically stable** if this holds for any $M > 0$.

Domain of attraction of \hat{x} : $\{x_0 \mid \lim_{t \rightarrow \infty} x(t) = \hat{x}\}$

Linearization

Nonlinear system are approximated by a linear system

$$\dot{x}(t) = f(x(t)), \quad f(\hat{x}) = 0$$

Take Taylor expansion around \hat{x} :

$$\begin{aligned} f(x) &= f(\hat{x}) + A(x - \hat{x}) + \text{higher order terms in } (x - \hat{x}) \\ &= A(x - \hat{x}) + \text{higher order terms in } (x - \hat{x}) \end{aligned}$$

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}, f(x) = \begin{bmatrix} f_1(x_1, \dots, x_n) \\ \vdots \\ f_n(x_1, \dots, x_n) \end{bmatrix}, A = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(\hat{x}) & \dots & \frac{\partial f_1}{\partial x_n}(\hat{x}) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}(\hat{x}) & \dots & \frac{\partial f_n}{\partial x_n}(\hat{x}) \end{bmatrix} \in \mathbb{R}^{n \times n}$$

Notice that after linearization, there is just one equilibrium

Distance of x to equilibrium

$$\delta x(t) = x(t) - \hat{x}$$

When x is close to the equilibrium, δx is small and

$$\frac{d \delta x(t)}{dt} \approx A \delta x(t)$$

Stability of \hat{x}

Locally asymptotically stable if the eigenvalues of the linearization have negative real part

Unstable if the linearization has at least one eigenvalue with positive real part

No conclusion if e-values are imaginary or zero

Lyapunov functions

Applying stability characteristics on nonlinear systems

Assume there exists an open set $S \subseteq \mathbb{R}^n$ with $\hat{x} \in S$:

- $V(\hat{x}) = 0$
- $V(x) > 0 \quad \forall x \in S \text{ with } x \neq \hat{x}$
- $\frac{d}{dt} V(x(t)) \leq 0 \quad \forall x \in S$

called "Lyapunov second / direct method"

Lie derivative

$$\frac{d}{dt} V(x(t)) = \nabla V(x(t)) * f(x(t))$$

Asymptotic stability

Assume there exists an open set $S \subseteq \mathbb{R}^n$ with $\hat{x} \in S$:

- $V(\hat{x}) = 0$
- $V(x) > 0 \quad \forall x \in S \text{ with } x \neq \hat{x}$
- $\frac{d}{dt} V(x(t)) < 0 \quad \forall x \in S \text{ with } x \neq \hat{x}$

Then the equilibrium \hat{x} is locally asymptotically stable.

If $S = \mathbb{R}^n$, then it is globally asymptotically stable.

La Salle's Theorem

Assume there exists a compact invariant set $S \subseteq \mathbb{R}^n$ and a differentiable function $V(\cdot) : \mathbb{R}^n \rightarrow \mathbb{R}$ such that:

$$\nabla V(x) f(x) \leq 0 \quad \forall x \in S$$

Let M be the largest invariant set contained in the set

$$\bar{S} = \{x \in S \mid \nabla V(x) f(x) = 0\} \subseteq \mathbb{R}^n$$

Then all trajectories starting in S tend to M as $t \rightarrow \infty$.

If \hat{x} only invariant set in

$$\{x \in S \mid \nabla V(x) * f(x) = 0\}$$

Then all trajectories starting in S tend to \hat{x} .

8. Various

Choose K of $u(t) = K * u(t)$

Try choosing such that K upper triangle \rightarrow easy λ

8. Tabellen

$i = \sqrt{1} = e^{i\frac{\pi}{2}}$
$\tan' x = 1 + \tan^2 x$
$\sin^2 x + \cos^2 x = 1$
$\cosh^2 x - \sinh^2 x = 1$
$\cos(z) = \cos(x) \cosh(y) - i \sin(x) \sinh(y)$
$\sin(z) = \sin(x) \cosh(y) + i \cos(x) \sinh(y)$

Grad	Rad	$\sin \varphi$	$\cos \varphi$	$\tan \varphi$
0°	0	0	1	0
30°	$\frac{1}{6}\pi$	$\frac{1}{2}$	$\frac{\sqrt{3}}{2}$	$\frac{\sqrt{3}}{3}$
45°	$\frac{1}{4}\pi$	$\frac{\sqrt{2}}{2}$	$\frac{\sqrt{2}}{2}$	1
60°	$\frac{1}{3}\pi$	$\frac{\sqrt{3}}{2}$	$\frac{1}{2}$	$\sqrt{3}$
90°	$\frac{1}{2}\pi$	1	0	
120°	$\frac{2}{3}\pi$	$\frac{\sqrt{3}}{2}$	$-\frac{1}{2}$	$-\sqrt{3}$
135°	$\frac{3}{4}\pi$	$\frac{\sqrt{2}}{2}$	$-\frac{\sqrt{2}}{2}$	-1
150°	$\frac{5}{6}\pi$	$\frac{1}{2}$	$-\frac{\sqrt{3}}{2}$	$-\frac{\sqrt{3}}{3}$
180°	π	0	-1	0

Additionstheoreme

$$\sin(\alpha \pm \beta) = \sin \alpha \cos \beta \pm \cos \alpha \sin \beta$$

$$\cos(\alpha \pm \beta) = \cos \alpha \cos \beta \mp \sin \alpha \sin \beta$$

$$\tan(\alpha \pm \beta) = \frac{\tan \alpha \pm \tan \beta}{1 \mp \tan \alpha \tan \beta}$$

Doppelter und halber Winkel

$$\sin 2\varphi = 2 \sin \varphi \cos \varphi \quad \sin^2 \frac{\varphi}{2} = \frac{1}{2}(1 - \cos \varphi)$$

$$\cos 2\varphi = \cos^2 \varphi - \sin^2 \varphi \quad \cos^2 \frac{\varphi}{2} = \frac{1}{2}(1 + \cos \varphi)$$

$$\tan 2\varphi = \frac{2 \tan \varphi}{1 - \tan^2 \varphi} \quad \tan^2 \frac{\varphi}{2} = \frac{1 - \cos \varphi}{1 + \cos \varphi}$$

Umformung einer Summe in ein Produkt

$$\sin \alpha + \sin \beta = 2 \sin \frac{\alpha + \beta}{2} \cos \frac{\alpha - \beta}{2}$$

$$\sin \alpha - \sin \beta = 2 \cos \frac{\alpha + \beta}{2} \sin \frac{\alpha - \beta}{2}$$

$$\cos \alpha + \cos \beta = 2 \cos \frac{\alpha + \beta}{2} \cos \frac{\alpha - \beta}{2}$$

$$\cos \alpha - \cos \beta = -2 \sin \frac{\alpha + \beta}{2} \sin \frac{\alpha - \beta}{2}$$

Umformung eines Produkts in eine Summe

$$2 \sin \alpha \sin \beta = \cos(\alpha - \beta) - \cos(\alpha + \beta)$$

$$2 \cos \alpha \cos \beta = \cos(\alpha - \beta) + \cos(\alpha + \beta)$$

$$2 \sin \alpha \cos \beta = \sin(\alpha - \beta) + \sin(\alpha + \beta)$$

Reihenentwicklungen

$$e^x = 1 + x + \dots = \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

$$\log(1+x) = x - \frac{x^2}{2} + \dots = \sum_{k=1}^{\infty} (-1)^{k-1} \frac{x^k}{k}$$

$$(1+x)^n = 1 + \binom{n}{1}x + \dots = \sum_{k=0}^{\infty} \binom{n}{k} x^k$$

$$\sin x = x - \frac{x^3}{3!} + \dots = \sum_{k=0}^{\infty} (-1)^k \frac{x^{2k+1}}{(2k+1)!}$$

$$\cos x = 1 - \frac{x^2}{2!} + \dots = \sum_{k=0}^{\infty} (-1)^k \frac{x^{2k}}{(2k)!}$$

$$\arctan x = x - \frac{x^3}{3} + \dots = \sum_{k=0}^{\infty} (-1)^k \frac{x^{2k+1}}{2k+1}$$

$$\sinh x = x + \frac{x^3}{3!} + \dots = \sum_{k=0}^{\infty} \frac{x^{2k+1}}{(2k+1)!}$$

$$\cosh x = 1 + \frac{x^2}{2!} + \dots = \sum_{k=0}^{\infty} \frac{x^{2k}}{(2k)!}$$

$$\operatorname{artanh} x = x + \frac{x^3}{3} + \dots = \sum_{k=0}^{\infty} \frac{x^{2k+1}}{2k+1}$$

Summe der ersten n-Zahlen

$$\sum_{k=1}^n k = 1 + 2 + \dots + n = \frac{n(n+1)}{2}$$

Geometrische Reihe

$$\sum_{k=0}^n x^k = 1 + x + \dots + x^n = \frac{1 - x^{n+1}}{1 - x}$$

Fourier-Korrespondenzen

$f(t)$	$\hat{f}(\omega)$
e^{-at^2}	$\sqrt{\frac{\pi}{a}} e^{-\frac{\omega^2}{4a}}$
$e^{-a t }$	$\frac{2a}{a^2 + \omega^2}$

Eigenschaften der Fourier-Transformation

Eigenschaft	$f(t)$	$\hat{f}(\omega)$
Linearität	$\lambda f(t) + \mu g(t)$	$\lambda \hat{f}(\omega) + \mu \hat{g}(\omega)$
Ähnlichkeit	$f(at) \quad a > 0$	$\frac{1}{ a } \hat{f}\left(\frac{\omega}{a}\right)$
Verschiebung	$f(t - a)$	$e^{-ai\omega} \hat{f}(\omega)$
	$e^{ait} f(t)$	$\hat{f}(\omega - a)$
Ableitung	$f^{(n)}(t)$	$(i\omega)^n \hat{f}(\omega)$
	$t^n f(t)$	$i^n \hat{f}^{(n)}(\omega)$
Faltung	$f(t) * g(t)$	$\hat{f}(\omega) \cdot \hat{g}(\omega)$

Partialbruchzerlegung (PBZ)

Reelle Nullstellen n-ter Ordnung:

$$\frac{A_1}{(x - a_k)} + \frac{A_2}{(x - a_k)^2} + \dots + \frac{A_n}{(x - a_k)^n}$$

Paar komplexer Nullstellen n-ter Ordnung:

$$\frac{B_1 x + C_1}{(x - a_k)(x - \overline{a_k})} + \dots + \frac{B_n x + C_n}{[(x - a_k)(x - \overline{a_k})]^n} +$$

$$(x - a_k)(x - \overline{a_k}) = (x - \operatorname{Re})^2 + \operatorname{Im}^2$$

Laplace- Korrespondenz

$f(t)$	$F(s)$	$f(t)$	$F(s)$
$\sigma(t)$	1	$H(t - a)$	$\frac{1}{s} e^{-as}$
1	$\frac{1}{s}$	e^{at}	$\frac{1}{s - a}$
t	$\frac{1}{s^2}$	te^{at}	$\frac{1}{(s - a)^2}$
t^n	$\frac{n!}{s^{n+1}}$	$t^n e^{at}$	$\frac{n!}{(s - a)^{n+1}}$
$\sin(at)$	$\frac{a}{s^2 + a^2}$	$\sinh(at)$	$\frac{a}{s^2 - a^2}$
$\cos(at)$	$\frac{s}{s^2 + a^2}$	$\cosh(at)$	$\frac{s}{s^2 - a^2}$

Eigenschaften der Laplace-Transformation

Eigenschaft	$f(t)$	$F(s)$
Linearität	$\lambda f(t) + \mu g(t)$	$\lambda F(s) + \mu G(s)$
Ähnlichkeit	$f(at) \quad a > 0$	$\frac{1}{a} F\left(\frac{s}{a}\right)$
Verschiebung im Zeitbereich	$f(t - t_0)$	$e^{-st_0} F(s)$
Verschiebung im Bildbereich	$e^{-at} f(t)$	$F(s + a)$
Ableitung im Zeitbereich	$f'(t)$	$sF(s) - f(0)$
	$f''(t)$	$s^2 F(s) - sf(0) - f'(0)$
	$f^{(n)}(t)$	$s^n F(s) - \sum_{k=0}^{n-1} f^{(k)}(0) s^{n-k-1}$
Ableitung im Bildbereich	$-tf(t)$	$F'(s)$
	$t^2 f(t)$	$F''(s)$
	$(-t)^n f(t)$	$F^{(n)}(s)$
Integration im Zeitbereich	$\int_0^t f(u) du$	$\frac{1}{s} F(s)$
Integration im Bildbereich	$\frac{1}{t} f(t)$	$\int_s^\infty F(u) du$
Faltung	$f(t) * g(t)$	$F(s) \cdot G(s)$
Periodische Funktion	$f(t) = f(t + T)$	$\frac{1}{1 - e^{-sT}} \int_0^T f(t) e^{-st} dt$

Ableitungen

Potenz- und Exponentialfunktionen			Trigonometrische Funktionen		Hyperbolische Funktionen	
$f(x)$	$f'(x)$	Bedingung	$f(x)$	$f'(x)$	$f(x)$	$f'(x)$
x^n	nx^{n-1}	$n \in \mathbb{Z}_{\geq 0}$	$\sin x$	$\cos x$	$\sinh x$	$\cosh x$
x^n	nx^{n-1}	$n \in \mathbb{Z}_{<0}, x \neq 0$	$\cos x$	$-\sin x$	$\cosh x$	$\sinh x$
x^a	ax^{a-1}	$a \in \mathbb{R}, x > 0$	$\tan x$	$\frac{1}{\cos^2 x}$	$\tanh x$	$\frac{1}{\cosh^2 x}$
$\log x$	$\frac{1}{x}$	$x > 0$	$\arcsin x$	$\frac{1}{\sqrt{1-x^2}}$	$\operatorname{arsinh} x$	$\frac{1}{\sqrt{x^2+1}}$
e^x	e^x		$\arccos x$	$-\frac{1}{\sqrt{1-x^2}}$	$\operatorname{arcosh} x$	$\frac{1}{\sqrt{x^2-1}}$
a^x	$a^x \cdot \log a$	$a > 0$	$\arctan x$	$\frac{1}{1+x^2}$	$\operatorname{artanh} x$	$\frac{1}{1-x^2}$

Stammfunktionen

$f(x)$	$F(x)$	Bedingung	$f(x)$	$F(x)$	$f(x)$	$F(x)$
x^n	$\frac{1}{n+1}x^{n+1}$	$n \in \mathbb{Z}_{\geq 0}$	$\frac{1}{x}$	$\log x $	$\sin(\omega t) \sin(\omega t)$	$\frac{t}{2} - \frac{\sin(2\omega t)}{4\omega}$
x^n	$\frac{1}{n+1}x^{n+1}$	$n \in \mathbb{Z}_{\leq -2}, x \neq 0$	$\tan x$	$-\log \cos x $	$\sin(\omega t) \cos(\omega t)$	$-\frac{\cos(2\omega t)}{4\omega}$
x^a	$\frac{1}{a+1}x^{a+1}$	$a \in \mathbb{R}, a \neq -1, x > 0$	$\tanh x$	$\log(\cosh x)$	$\sin(\omega t) \sin(n\omega t)$	$\frac{n \cos(\omega t) \sin(n\omega t) - \sin(\omega t) \cos(n\omega t)}{\omega(n^2-1)}$
$\log x$	$x \log x - x$	$x > 0$	$\sin^2 x$	$\frac{1}{2}(x - \sin x \cos x)$	$\sin(\omega t) \cos(n\omega t)$	$\frac{n \sin(\omega t) \sin(n\omega t) + \cos(\omega t) \cos(n\omega t)}{\omega(n^2-1)}$
e^{ax}	$\frac{1}{a}e^{ax}$	$a \neq 0$	$\cos^2 x$	$\frac{1}{2}(x + \sin x \cos x)$	$\cos(\omega t) \sin(n\omega t)$	$\frac{\sin(\omega t) \sin(n\omega t) + n \cos(\omega t) \cos(n\omega t)}{\omega(1-n^2)}$
a^x	$\frac{a^x}{\log a}$	$a > 0, a \neq 1$	$\tan^2 x$	$\tan x - x$	$\cos(\omega t) \cos(n\omega t)$	$\frac{\sin(\omega t) \cos(n\omega t) + n \cos(\omega t) \sin(n\omega t)}{\omega(1-n^2)}$

Standard-Substitutionen

Integral	Substitution	Ableitung	Bemerkung
$\int f(x, x^2 + 1) dx$	$x = \tan t$	$dx = \tan^2 t + 1 dt$	$t \in \bigcup_{k \in \mathbb{Z}} (k\pi - \frac{\pi}{2}, k\pi + \frac{\pi}{2})$
$\int f(x, \sqrt{ax+b}) dx$	$x = \frac{t^2-b}{a}$	$dx = \frac{2}{a}t dt$	$t \geq 0$
$\int f(x, \sqrt{ax^2+bx+c}) dx$	$x + \frac{b}{2a} = t$	$dx = dt$	$t \in \mathbb{R}$, quadratische Ergänzung
$\int f(x, \sqrt{a^2-x^2}) dx$	$x = a \sin t$	$dx = a \cos t dt$	$-\frac{\pi}{2} < t < \frac{\pi}{2}$, $1 - \sin^2 x = \cos^2 x$
$\int f(x, \sqrt{a^2+x^2}) dx$	$x = a \sinh t$	$dx = a \cosh t dt$	$t \in \mathbb{R}$, $1 + \sinh^2 x = \cosh^2 x$
$\int f(x, \sqrt{x^2-a^2}) dx$	$x = a \cosh t$	$dx = a \sinh t dt$	$t \geq 0$, $\cosh^2 x - 1 = \sinh^2 x$
$\int f(e^x, \sinh x, \cosh x) dx$	$e^x = t$	$dx = \frac{1}{t} dt$	$t > 0$, $\sinh x = \frac{t^2-1}{2t}$, $\cosh x = \frac{t^2+1}{2t}$
$\int f(\sin x, \cos x) dx$	$\tan \frac{x}{2} = t$	$dx = \frac{2}{1+t^2} dt$	$-\frac{\pi}{2} < t < \frac{\pi}{2}$, $\sin x = \frac{2t}{1+t^2}$, $\cos x = \frac{1-t^2}{1+t^2}$