

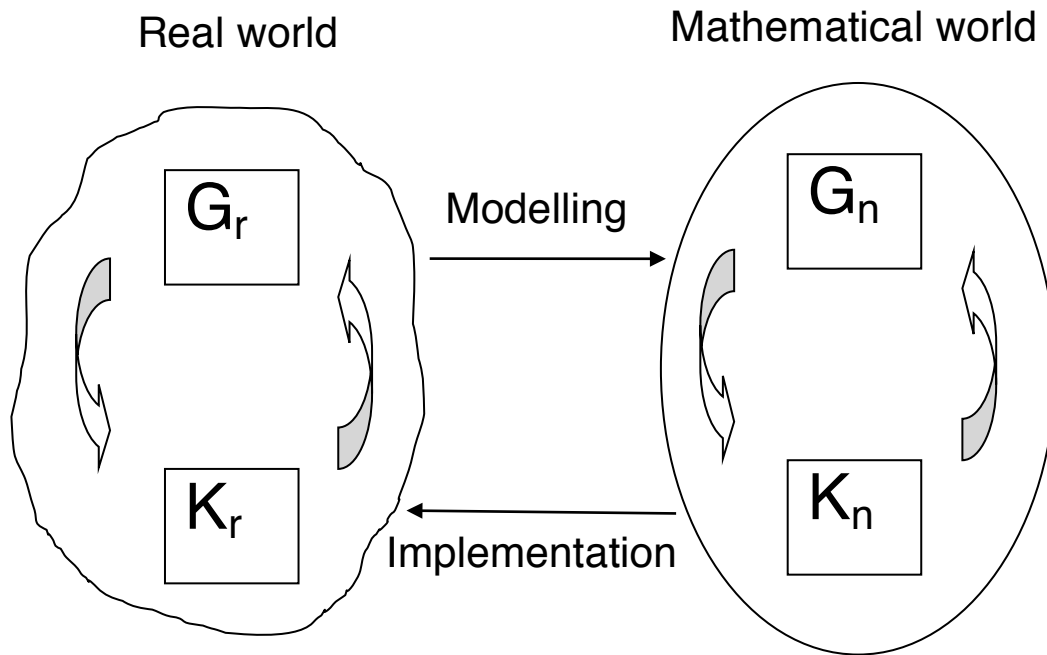
MULTIVARIABLE CONTROL SYSTEMS DESIGN^{*°}

by Ian K. Craig

* These viewgraphs are based on notes prepared by Prof. Michael Athans of MIT for the course "Multivariable Control Systems 1 & 2"

° These viewgraphs should be read in conjunction with the textbook:
S Skogestad, I Postlethwaite, Multivariable Feedback Control,
Second Edition, Wiley, Chichester, 2005.

General Control Problem (I)



General Control Problem (II)

Summary

- Analyze the plant for control purposes
- Obtain an adequate mathematical model for the plant
- Controller design and control system analysis
- Controller implementation
- Controller evaluation

Analyze the plant (G_r) for control purposes

- obtain process knowledge
- perform an initial cost benefit analysis
- determine initial control objectives
- determine measurements, manipulated and control variables
- determine the role of the operator, before and after the implementation of the controller

Obtain an adequate mathematical model (G_n) for the plant

- use first principles and empirical relationships and/or plant input-output data
- simplify the model to fit purpose
- *analyse the resulting model (input-output analysis)*

General Control Problem (III)

Controller design (K_n) and control system analysis

- select controller configuration and type
- decide performance specifications
- *design controller and analyse it to see if specifications are satisfied*

Controller implementation (K_r)

- simulate controller (hardware-in-loop; pilot plant)
- select hardware and software and implement

Controller evaluation

- test and validate for functional and economic specifications

References:

- S. Skogestad, I. Postlethwaite, *Multivariable Feedback Control*, Second Edition, Wiley, Chichester, 2005. Page 1.
- Craig I.K., and Henning, R.G.D., Evaluation of advanced industrial control projects: a framework for determining economic benefits, *Control Engineering Practice*, Vol. 8, No. 7, 2000, pp. 769-780.

Main thrust of the course

- Description of:
 - Analysis tools (block 1)
 - Synthesis tools (block 2)
- Analysis and synthesis as part of the engineering design process
- Analysis: the process of determining whether a given system has the desired characteristics

Tools

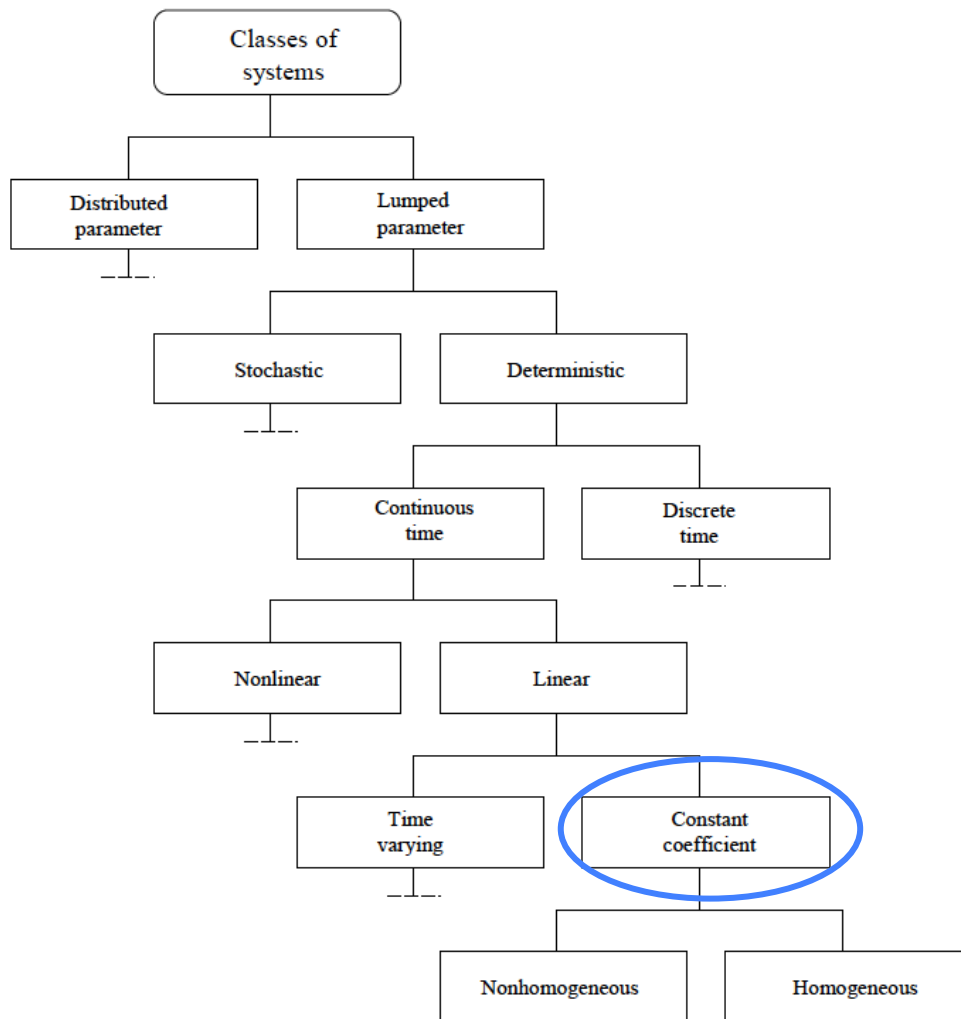
- linear systems theory
- linear algebra
- functional analysis

- Synthesis: the process of finding a particular system component to achieve desirable characteristics

Tools

- dynamic optimal control theory
- optimal estimation theory

Classes of system equations



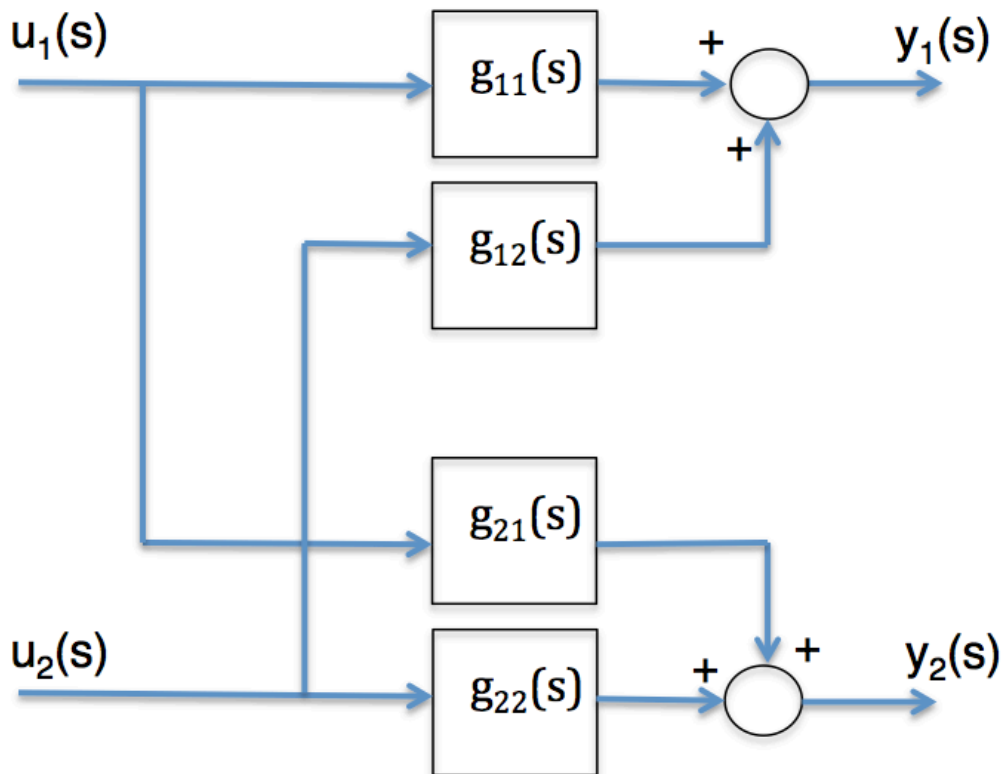
This course deals with:
Lumped parameter, deterministic,
continuous-time, linear, constant coefficient
systems

What is a dynamic system?

- Physical property
 - physical dynamic systems contain energy-storage elements
 - energy-storage elements are usually interconnected
- Physical energy changes as a function of time
 - Potential energy \leftrightarrow kinetic energy
 - Dissipation of energy
 - Change in energy from external inputs
- Dynamic evolution of energy is the key to understanding the behaviour of dynamic systems

What is a MIMO system?

- MIMO: Multi-Input Multi-Output
- Plant model example



- Mathematical description

$$\begin{pmatrix} y_1(s) \\ y_2(s) \end{pmatrix} = \begin{pmatrix} g_{11}(s) & g_{12}(s) \\ g_{21}(s) & g_{22}(s) \end{pmatrix} \begin{pmatrix} u_1(s) \\ u_2(s) \end{pmatrix}$$

$$y(s) = G(s)u(s)$$

$G(s)$: Transfer function matrix

Physical MIMO systems

- Aerospace
 - aircraft
 - missiles
 - satellites
 - space platforms

- Biological
 - immune system
 - glucose–insulin feedback system
 - disease transmission

- Chemical
 - reactors
 - distillation columns

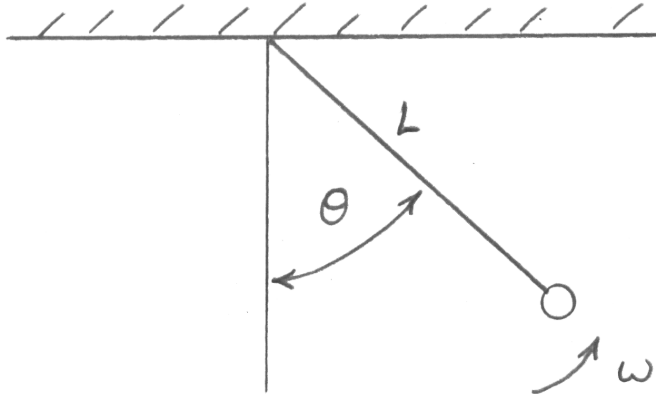
- Electrical
 - power systems
 - motors

- Mechanical
 - robots
 - automotive

- Metallurgical
 - grinding milling circuits
 - furnances
 - rolling machines

Simple dynamic systems

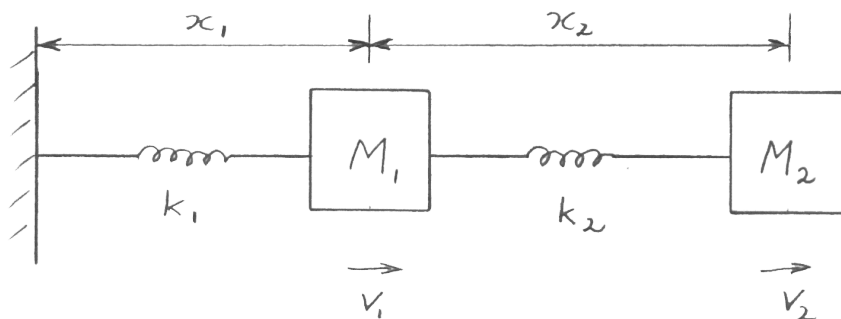
- Pendulum



$\theta \Rightarrow$ potential energy

$\omega \Rightarrow$ kinetic energy

- Mass-spring system



$x_1, x_2 \Rightarrow$ potential energy

$v_1, v_2 \Rightarrow$ kinetic energy

Physical state variables

- Associate one state with each energy storage element

- Mechanical system state variables

Positions \Rightarrow potential energy

Velocities \Rightarrow kinetic energy

- Electrical circuit state variables

Inductor currents \Rightarrow kinetic energy

Capacitor voltages \Rightarrow potential energy

- Thermodynamic systems

Pressures \Rightarrow "potential energy"

Temperatures \Rightarrow "kinetic energy"

Properties of dynamic systems

- Physical systems are interconnections of energy storage and dissipative elements
- State dynamics are described by coupled first order differential equations
- Unforced dynamic systems (no external inputs)
 - Assume valid model
 - If we know the numerical values of the state variables now, then we can calculate all future values of:
 - all state variables
 - other variables of interest
 - State variables are continuous functions of time

Forced dynamic systems

- External inputs (controls, disturbances)
 - can modify temporal evolution of energy (state) variables
 - add or subtract finite energy to/from system
 - bounded, piece-wise continuous
- If we know
 - state variables now
 - inputs from now on

Then we can calculate:

- future values of all state variables
- future values of all output variables

Forced state variable models (I)

- Finite dimensional time invariant models
- Notation
 - State vector: $x'(t) = [x_1(t), x_2(t), \dots, x_n(t)]'$
 - Input vector: $u'(t) = [u_1(t), u_2(t), \dots, u_m(t)]'$
 - Output vector: $y'(t) = [y_1(t), y_2(t), \dots, y_p(t)]'$
- Nonlinear model description
 - State dynamics
 - $dx_1(t)/dt = f_1(x(t), u(t))$
 - $dx_2(t)/dt = f_2(x(t), u(t))$
 - \vdots
 - \vdots
 - $dx_n(t)/dt = f_n(x(t), u(t))$
 - Collectively:** $dx(t)/dt = f(x(t), u(t))$
 - Output equations
 - $y_1(t) = g_1(x(t), u(t))$
 - $y_2(t) = g_2(x(t), u(t))$
 - \vdots
 - \vdots
 - $y_p(t) = g_p(x(t), u(t))$
 - Collectively:** $y(t) = g(x(t), u(t))$

Forced state variable models (II)

- Linear model description

- State dynamics: $i = 1, 2, \dots, n$

$$\frac{dx_i(t)}{dt} = \sum_{j=1}^n a_{ij} x_j(t) + \sum_{k=1}^m b_{ik} u_k(t)$$

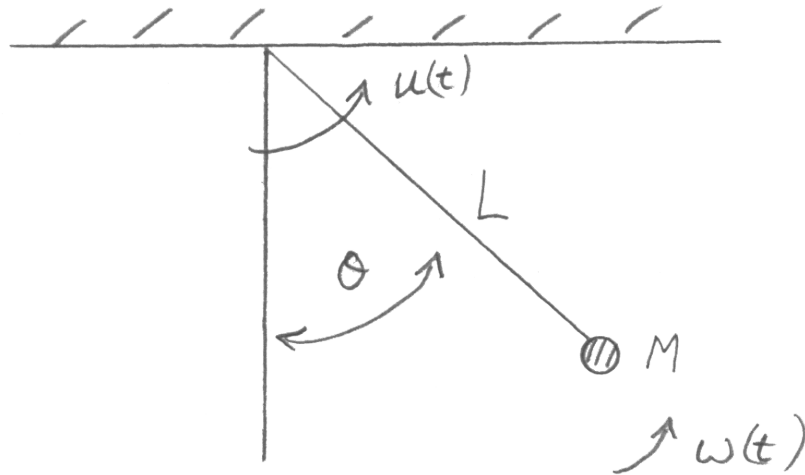
$$\Rightarrow dx(t)/dt = Ax(t) + Bu(t)$$

- Output equation: $q = 1, 2, \dots, p$

$$y_q(t) = \sum_{i=1}^n c_{qi} x_i(t) + \sum_{k=1}^m d_{qk} u_k(t)$$

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

Pendulum example



- State variables

$\theta(t) \equiv x_1(t)$ = angular position

$\omega(t) \equiv x_2(t)$ = angular velocity

- Control variable: torque $u(t)$
- State dynamics (without friction)

$$\frac{dx_1(t)}{dt} = x_2(t)$$

$$\frac{dx_2(t)}{dt} = (g/L)\sin(x_1(t)) + (1/ML)u(t)$$

Linearization of nonlinear dynamics

- Most dynamic systems are nonlinear
 - quantitative and qualitative properties not transparent
- Motivation for linearizing
 - lack of systematic design methodology for direct design of nonlinear feedback control systems
 - linearized dynamic models are useful
 - analysis: qualitative and quantitative insight
 - design: general and integrated CAD methodologies
- Linearized models have obvious limitations
- Pendulum example
Linear approximation for small $x_1(t)$

$$\sin \theta = \theta - \frac{\theta^3}{3!} + \frac{\theta^5}{5!} - K \approx \theta$$

Thus $dx_2(t)/dt = (g/L)x_1(t) + (1/ML)u(t)$

Linear Dynamic Systems: Issues to be discussed

- Open-loop stability : MIMO poles
- Transient response : MIMO zeros
 - nature provides poles
 - control engineers regulate the zeros
- Modes of dynamic systems
- Solutions: time- and frequency-domain

Structure

- Work with finite dimensional linear time invariant (FDLTI) models
- LTI model describes some nonlinear system near steady-state equilibrium

- Notation

State: $x(t) \in \mathbb{R}^n$

Control: $u(t) \in \mathbb{R}^m$

Output: $y(t) \in \mathbb{R}^p$

- Dynamics

$$\frac{dx(t)}{dt} = Ax(t) + Bu(t) \quad ;x(0) = \xi$$

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

Transfer function matrix

- Time-domain state-space model

$$\begin{aligned} dx(t)/dt &= Ax(t) + Bu(t) && ;x(0) = 0 \\ y(t) &= Cx(t) + Du(t) \end{aligned}$$

- Vector Laplace transforms

$$\begin{aligned} x(s) &= \mathcal{L}\{x(t)\} \quad \text{with} \quad \mathcal{L}\{dx(t)/dt\} = sIx(s) - 0 \\ u(s) &= \mathcal{L}\{u(t)\} \\ y(s) &= \mathcal{L}\{y(t)\} \end{aligned}$$

- Laplace transform of state-space

$$\begin{aligned} sIx(s) &= Ax(s) + Bu(s) \\ y(s) &= Cx(s) + Du(s) \end{aligned}$$

or $x(s) = (sI - A)^{-1}Bu(s)$

- Input-output description

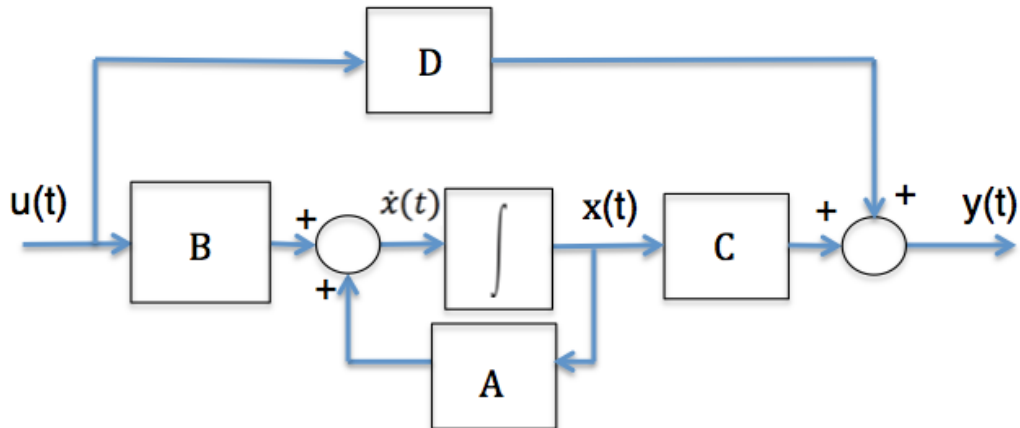
$$\begin{aligned} y(s) &= [C(sI - A)^{-1}B + D]u(s) \\ y(s) &= G(s)u(s) \end{aligned}$$

$G(s)$ is a $p \times m$ TFM

Finite dimensional LTI models

- State-space block diagram

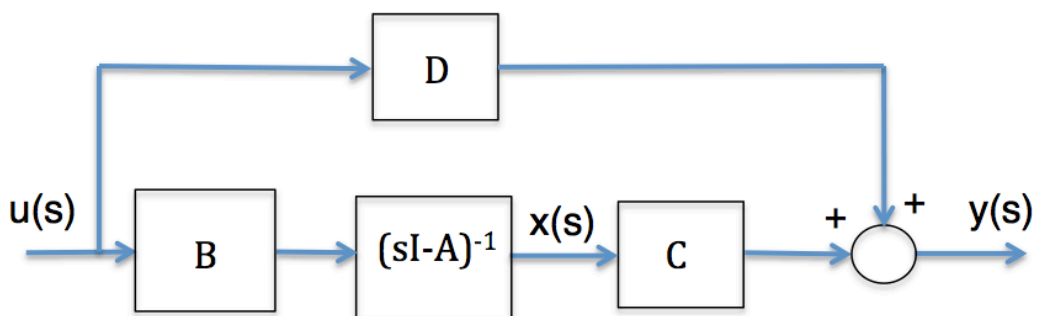
$$\begin{aligned} dx(t)/dt &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t) \end{aligned}$$



- Frequency-domain block diagram

$$y(s) = G(s)u(s)$$

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



Unforced LTI systems

- Set inputs to zero: ignore outputs

$$\frac{dx(t)}{dt} = Ax(t) \quad ;x(0) = \xi$$

- Time-domain solution

$$x(t) = e^{At} \xi$$

$$\text{with } e^{At} = \sum_{k=0}^{\infty} \frac{1}{k!} (At)^k$$

- Laplace transform of state-space model

$$sIx(s) - \xi = Ax(s)$$

$$\text{or } x(s) = (sI - A)^{-1} \xi$$

$$\text{thus } \mathcal{L}\{e^{At}\} = (sI - A)^{-1} \quad ;(n \times n)$$

$$\mathcal{L}^{-1}\{(sI - A)^{-1}\} = e^{At}$$

Modal analysis (I)

- Eigenstructure of A ; $i = 1, 2, \dots, n$

$$Av_i = \lambda_i v_i \quad ; \quad w_i' A = \lambda_i w_i'$$

- Dyadic formula

$$A = \sum_{i=1}^n \lambda_i v_i w_i'$$

- Fact:

$$A^k v_i = \lambda_i^k v_i \quad ; \quad w_i' A^k = \lambda_i^k w_i'$$

$$A^k = \sum_{i=1}^n \lambda_i^k v_i w_i'$$

- Fact:

$$e^{At} = \sum_{j=0}^{\infty} \frac{1}{j!} (At)^j = \sum_{j=0}^{\infty} \sum_{i=1}^n \frac{1}{j!} (\lambda_i t)^j v_i w_i'$$

thus
$$e^{At} = \sum_{i=1}^n e^{\lambda_i t} v_i w_i'$$

with
$$e^{\lambda_i t} = \sum_{j=0}^{\infty} \frac{1}{j!} (\lambda_i t)^j$$

Modal analysis (II)

- Unforced dynamics solution

$$x(t) = e^{At} \xi = \sum_{i=1}^n e^{\lambda_i t} v_i [w_i' \xi]$$

- Definition of i^{th} mode:

$$e^{\lambda_i t} v_i$$

- State consists of sum of modes associated with right eigenvector directions
- $[w_i' \xi]$: Degree that initial state ξ excites i^{th} mode

- Important relation

$$w_j' v_i = \delta_{ij} \quad ; (\text{Kronecker delta})$$

$$w_j' v_i = 1 \quad ; i = j$$

$$w_j' v_i = 0 \quad ; i \neq j$$

- Laplace transform

$$\mathcal{L}\{e^{\lambda_i t}\} = \frac{1}{s - \lambda_i}$$

$$x(s) = \sum_{i=1}^n \frac{1}{s - \lambda_i} v_i [w_i' \xi]$$

thus $(sI - A)^{-1} = \sum_{i=1}^n \frac{1}{s - \lambda_i} v_i w_i'$

Eigenvector interpretation

- Unforced dynamics

$$dx(t)/dt = Ax(t) \quad ;x(0) = \xi$$

- To excite single mode

pick initial state ξ colinear with eigenvector v_j ,
i.e.: $\xi = kv_j$

- Solution

$$\begin{aligned} x(t) &= \sum_{i=1}^n e^{\lambda_i t} v_i [w_i' \xi] \\ &= \sum_{i=1}^n k e^{\lambda_i t} v_i [w_i' v_j] \\ &= k e^{\lambda_j t} v_j \end{aligned}$$

- Solution contains single mode!

Stability and multivariable poles

- Open-loop dynamics

$$\frac{dx(t)}{dt} = Ax(t) \quad ; x(0) = \xi$$

$$x(t) = e^{At} \xi = \sum_{i=1}^n e^{\lambda_i t} v_i [w_i' \xi]$$

- If $\text{Re}\{\lambda_i\} > 0$ then i^{th} mode $e^{\lambda_i t} v_i$ is unstable

- System is stable iff

$$\text{Re}\{\lambda_i[A]\} < 0 \quad \text{for all } i = 1, 2, \dots, n$$

$$\text{thus } x(t) \rightarrow 0 \quad \text{for all } \xi \neq 0$$

- Definition: multivariable poles are the eigenvalues of A

- definition makes sense in terms of natural frequencies
- MIMO poles are roots of characteristic polynomial: $\det(\lambda I - A)$

Forced LTI systems

- Model

$$\begin{aligned} dx(t)/dt &= Ax(t) + Bu(t) & ;x(0) = \xi \\ y(t) &= Cx(t) + Du(t) \end{aligned}$$

- Standing assumption:

Components of control vector $u(t)$

- bounded
- piece-wise continuous functions of time

- Complete solution

$$x(t) = e^{At}\xi + \int_0^t e^{A(t-\tau)} Bu(\tau) d\tau$$

- Proof: use vector calculus facts

$$(d/dt)e^{At} = Ae^{At}$$

$$(d/dt) \int_0^t f(\tau) d\tau = f(t)$$

and differentiate $x(t) = e^{At} \left[\xi + \int_0^t e^{-A\tau} Bu(\tau) d\tau \right]$

to get $dx(t)/dt = Ax(t) + Bu(t)$

The complete solution

- Dynamics

$$\begin{aligned} dx(t)/dt &= Ax(t) + Bu(t) && ;x(0) = \xi \\ y(t) &= Cx(t) + Du(t) \end{aligned}$$

- Laplace transform

$$sIx(s) - \xi = Ax(s) + Bu(s)$$

$$\text{thus } x(s) = (sI - A)^{-1}\xi + (sI - A)^{-1}Bu(s)$$

- Compare with time-domain solution

$$\mathcal{L}\left\{\int_0^t e^{A(t-\tau)} Bu(\tau) d\tau\right\} = (sI - A)^{-1}Bu(s)$$

- Complete solution

Time-domain

$$y(t) = Ce^{At}\xi + \int_0^t Ce^{A(t-\tau)} Bu(\tau) d\tau + Du(t)$$

Frequency-domain

$$y(s) = C(sI - A)^{-1}\xi + [C(sI - A)^{-1}B + D]u(s)$$

Modal forms (I)

- State dynamics : $x(0) = 0$

$$\begin{aligned} dx(t)/dt &= Ax(t) + Bu(t). \\ &= Ax(t) + \sum_{k=1}^m b_k u_k(t) \end{aligned}$$

with $B = [b_1 \ b_2 \ \dots \ b_m]$

Solution $x(t) = \sum_{k=1}^m \int_0^t e^{A(t-\tau)} b_k u_k(\tau) d\tau$

Recall $e^{A(t-\tau)} = \sum_{i=1}^n e^{\lambda_i(t-\tau)} v_i w_i'$

- Time-domain solution

$$x(t) = \sum_{i=1}^n \sum_{k=1}^m v_i (w_i' b_k) \int_0^t e^{\lambda_i(t-\tau)} u_k(\tau) d\tau$$

- Frequency-domain solution

$$x(s) = \sum_{i=1}^n \sum_{k=1}^m v_i (w_i' b_k) \frac{1}{s - \lambda_i} u_k(s)$$

Modal forms (II)

- Solution with $x(0) = \xi \neq 0$

$$x(t) = \sum_{i=1}^n (w_i' \xi) v_i e^{\lambda_i t} + \sum_{i=1}^n \sum_{k=1}^m v_i (w_i' b_k) \int_0^t e^{\lambda_i(t-\tau)} u_k(\tau) d\tau$$

- Alternate form

$$x(t) = \sum_{i=1}^n v_i e^{\lambda_i t} \left[(w_i' \xi) + \sum_{k=1}^m (w_i' b_k) \int_0^t e^{-\lambda_i \tau} u_k(\tau) d\tau \right]$$

- Modal directions preserved
- Natural time evolution ($e^{\lambda_i t}$) is changed by controls
- Insight

$(w_i' \xi)$: degree that initial state ξ excites i^{th} mode

$(w_i' b_k)$: degree that k^{th} control u_k influences i^{th} mode

Output response

- LTI model

$$\frac{dx(t)}{dt} = Ax(t) + \sum_{k=1}^m b_k u_k(t) \quad ; x(0) = \xi.$$

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

- Individual outputs: $q = 1, 2, \dots, p$

$$y_q(t) = c'_q x(t) + d'_q u(t)$$

- Output response

$$\begin{aligned} y_q(t) = & \sum_{i=1}^n (c'_q v_i)(w'_i \xi) e^{\lambda_i t} \\ & + \sum_{i=1}^n \sum_{k=1}^m (c'_q v_i)(w'_i b_k) \int_0^t e^{\lambda_i(t-\tau)} u_k(\tau) d\tau \\ & + d'_q u(t) \end{aligned}$$

- Insight

$(c'_q v_i)$: degree to which i^{th} mode will be visible in q^{th} output

Frequency domain solutions

- Recall

$$(sI - A)^{-1} = \sum_{i=1}^n \frac{1}{s - \lambda_i} v_i w_i'$$

- State

$$x(s) = \sum_{i=1}^n \frac{1}{s - \lambda_i} v_i [w_i' \xi] + \sum_{i=1}^n \frac{1}{s - \lambda_i} v_i w_i' B u(s)$$

- Output

$$y(s) = \sum_{i=1}^n \frac{1}{s - \lambda_i} C v_i [w_i' \xi] + \sum_{i=1}^n \frac{1}{s - \lambda_i} C v_i w_i' B u(s) + D u(s)$$

Controllability

Background

- Formalized by R.E. Kalman in 1960
- Key concept in dynamic systems and control theory
- Formalizes intuitive notions about being able to control state variables and modes
- Will present "modern" and classical controllability tests
- Used with observability to understand MIMO input-output properties
 - MIMO pole-zero cancellations
 - minimum realizations

Controllability definition

- Deals only with state dynamics
- Applicable to both linear and nonlinear systems
- Definition:

The system $\dot{x}(t) = f(x(t), u(t))$; $x(0) = \xi$

is called controllable if for any initial state $\xi \in \mathbb{R}^n$ and any terminal state $\theta \in \mathbb{R}^n$, we can find a piece-wise continuous function $u(t)$, $0 \leq t \leq T$, with T finite, such that

$$x(T) = \theta$$

Otherwise the system is called uncontrollable

Remarks

- No easy test for general nonlinear systems
- Easy test exist for finite-dimensional linear-time-invariant (FDLTI) dynamic systems
- Two tests

"modern" - modal approach

"classical" - Via Caley-Hamilton theorem

- Warning: A dynamic model which is mathematically controllable, might be uncontrollable from a practical point of view
 - Test does not say how states behave, e.g. might not be possible to hold states at given value θ
 - The required inputs $u(t)$ may be very large
 - Some of the states may be of no practical importance
 - Definition does not provide a degree of controllability
- What does the controllability result tell us?
 - If our model includes states that we cannot affect
 - If we can save on computer time by deleting uncontrollable (stable!) states

Modal solutions

- State dynamics with $x(0) = \xi$

$$dx(t)/dt = Ax(t) + Bu(t)$$

$$dx(t)/dt = Ax(t) + \sum_{k=1}^m b_k u_k(t)$$

- Recall eigenstructure

$$Av_i = \lambda_i v_i \quad ; \quad w'_i A = \lambda_i w'_i; \quad w'_j v_i = \delta_{ij}$$

- State response

$$x(t) = \sum_{i=1}^n (w'_i \xi) v_i e^{\lambda_i t} + \sum_{i=1}^n \sum_{k=1}^m v_i (w'_i b_k) \int_0^t e^{\lambda_i(t-\tau)} u_k(\tau) d\tau$$

- Modal uncontrollability

If $w'_i b_k = 0$ for some k , then mode i is uncontrollable from control $u_k(t)$

Modal uncontrollability

- Mode i is uncontrollable (from all inputs) iff for all $k = 1, 2, \dots, m$

$$w'_i b_k = 0 \quad \text{or} \quad w'_i B = 0$$

- System is uncontrollable iff one or more of its modes are uncontrollable
- Reasoning using state response

Pick initial state $\xi = 0$

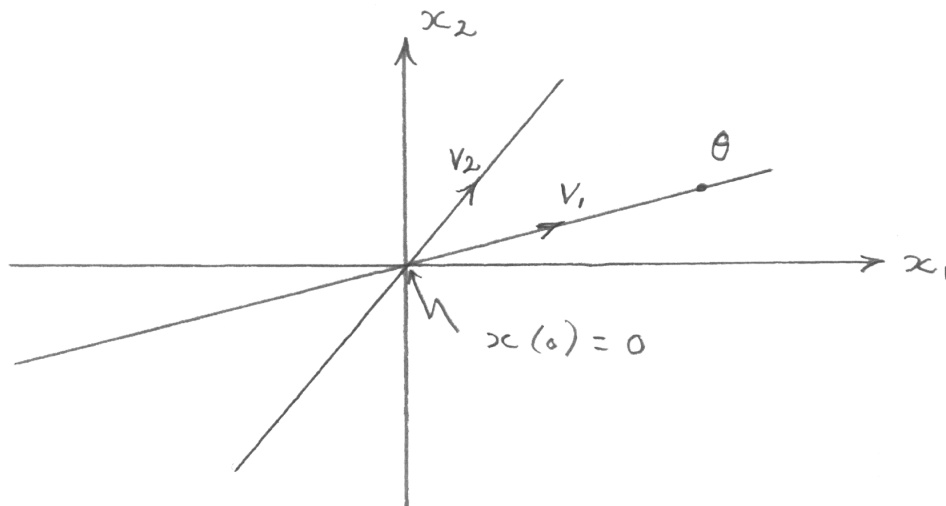
Pick terminal state θ colinear to uncontrollable mode i , i.e. $\theta = kv_i$

$$x(T) = \sum_{j=1}^n \sum_{k=1}^m v_j (w'_j b_k) \int_0^T e^{\lambda_j(t-\tau)} u_k(\tau) d\tau$$

$x(T) \neq kv_i$ - directions do not match

Visualization

- 2 modes and 2 controls: $n = 2$; $k = 2$
- Suppose mode $v_1 e^{\lambda_1 t}$ is uncontrollable



- Response restricted along v_2 direction. θ cannot be reached
- Time-domain solution

$$\begin{aligned}
 x(t) = & v_1 (w_1' b_1 = 0) \int_0^t e^{\lambda_1(t-\tau)} u_1(\tau) d\tau \\
 & + v_1 (w_1' b_2 = 0) \int_0^t e^{\lambda_1(t-\tau)} u_2(\tau) d\tau \\
 & + v_2 (w_2' b_1) \int_0^t e^{\lambda_2(t-\tau)} u_1(\tau) d\tau \\
 & + v_2 (w_2' b_2) \int_0^t e^{\lambda_2(t-\tau)} u_2(\tau) d\tau
 \end{aligned}$$

Modal controllability

- The i^{th} mode is controllable (from one or more inputs) iff

$$w'_i B \neq 0$$

w'_i : left eigenvector associated with i^{th} mode,

$$w'_i A = \lambda_i w'_i$$

- System is controllable iff all the modes of the system are controllable

Test: $w'_i B \neq 0$ for all $i = 1, 2, \dots, n$

- Notation

Refer to the controllability of a matrix pair:
[A, B]

A = n x n matrix

B = n x m matrix

Complex modes

- If $\lambda_i = \lambda_j^*$

then $v_i = v_j^*$ and $w_i = w_j^*$

- i^{th} mode is uncontrollable from k^{th} input

$$\Rightarrow w_i' b_k = 0$$

$$\text{let } w_i = \alpha_i + j\beta_i$$

$$\text{thus } (\alpha_i + j\beta_i)' b_k = 0$$

$$\Rightarrow \alpha_i' b_k = 0 \quad \text{and} \quad \beta_i' b_k = 0$$

$$\text{but } w_j = \alpha_i - j\beta_i = w_i^*$$

$$\Rightarrow (\alpha_i - j\beta_i)' b_k = 0$$

$$\Rightarrow w_j' b_k = 0$$

$\Rightarrow j^{\text{th}}$ mode is also uncontrollable from k^{th} input

- Complex mode is thus uncontrollable

Stabilizability

- Useful concept for design
- If mode $v_i e^{\lambda_i t}$ is uncontrollable but $\text{Re}\{\lambda_i\} < 0$, then mode i is stabilizable
- If all uncontrollable modes are stabilizable, then $[A, B]$ is called stabilizable
- Notes
 - If $[A, B]$ is controllable, then it is stabilizable
 - If every unstable mode is controllable, then $[A, B]$ is stabilizable

Classical controllability test

- Form the $n \times (m \times n)$ controllability matrix M_C

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

- If out of the $m \times n$ columns M_C there are n that are linearly independent, i.e.

$$\text{Rank}(M_C) = n$$

Then $[A, B]$ is controllable

- If $\text{Rank}(M_C) < n$

Then $[A, B]$ is uncontrollable

- it may be stabilizable

- No modal information
- No stabilizability information