Introduction to the Science of Control

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Objectives of Talk

- Learn some control terminology
- Develop some intuition about control concepts
 - Details occasionally (and intentionally) omitted
- Understand the multiple objectives of control



Model-Based Control Design Process

- 1. Make system model
- 2. Verify model predicts behavior of system
- 3. Design controller
- 4. Test using models in closed-loop simulation
- 5. Implement and test implementation
- 6. Deploy in operation
- Using only 5-6 is feasible and often successful why do steps 1-4?
 - Requires empirical tuning, cost = \$50,000 \$100,000 per day on present devices
 - Performance:
 - Large systems (many inputs / outputs) difficult to tune properly for best control
 - Nonlinear systems require retuning over many equilibrium states.
 - Even if Steps 5-6 is chosen approach, studying models is useful to understand how control affects system
 - Next Generation devices (e.g. ITER) will not allow empirical tuning



Introduction to System Representation - Block Diagrams

A Block Diagram consists of two parts:

- Signals (arrows in diagram)
- Operations (blocks in diagram)
- Example (poloidal field system producing plasma shape)



System Representation – Ordinary Differential Equations

u(t)

- State Space Models
 - General (x is "state"):

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

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

- Note ordinary differential equation (ODE) is 1st order
- Linear, time-invariant (LTI) system:

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

System

y(t)

- Example (plasma + conductors):

$$M_* \frac{d(\delta I)}{dt} + R\delta I = U\delta v \qquad \Rightarrow \qquad d(\delta I)/dt = A\delta I + Bv$$

$$y = CI \qquad \qquad \Rightarrow \qquad y = CI$$

$$(A = -M_*^{-1}R, \quad B = M_*^{-1}U)$$



- I(t) = toroidal conductor currents (perturbations δl from equilibrium ⇔ states x);
 M_{*}=mutual inductance matrix (modified by plasma response), R=resistance matrix
- -y(t) = coil currents, flux and field in vacuum region; C=green functions
- v(t) = input voltage from power supplies (δv from equilibrium); U = ones for coils, zeros for vessel conductors



System Representation – Laplace Transform

• Definition: For a given function f(t) with f(0)=0, Laplace transform of f is:

$$F(s) = \mathcal{L}\left\{f(t)\right\} = \int_0^\infty e^{-st} f(t) dt, \qquad s = \sigma + j\omega$$

Nice properties:

$$\mathcal{L}\left\{\frac{df}{dt}\right\} = sF(s), \quad \mathcal{L}\left\{\frac{d^2f}{dt^2}\right\} = s^2F(s), \quad \dots etc \dots$$

$$\mathcal{L}\left\{\int_0^t f(\tau)d\tau\right\} = \frac{1}{s}F(s)$$

• For an example of how it's used, apply to :

 $\dot{x}(t) = Ax(t) + Bu(t) \qquad \Rightarrow \qquad sX(s) = AX(s) + BU(s)$ $\Rightarrow \qquad Y(s) = CX(s) + DU(s)$

 $sX(s) = AX(s) + BU(s) \implies (s\mathbf{I} - A)X(s) = BU(s) \implies X(s) = (s\mathbf{I} - A)^{-1}BU(s)$

$$Y(s) = CX(s) + DU(s)$$

$$= C(s\mathbf{I} - A)^{-1}BU(s) + DU(s) \implies Y(s) = \left(C(s\mathbf{I} - A)^{-1}B + D\right)U(s)$$



σ

∧ jω

complex ("s") plane

("frequency domain")

System Representation - Transfer Functions

- Transfer Function = ratio of Laplace Transforms of (scalar) output and input signals: $\frac{Y(s)}{U(s)}$
- Example (simple mechanical system; x is displacement):

 $m\ddot{x}(t) + d\dot{x}(t) + kx(t) = u(t) \implies (ms^2 + ds + k)X(s) = U(s) \implies \frac{X(s)}{U(s)} = \frac{1}{(ms^2 + ds + k)}$

• Example (lowpass RC filter):



- General LTI case, from previous page: $Y(s) = (C(s\mathbf{I} A)^{-1}B + D)U(s)$
- If Y, U are scalars: $\frac{Y(s)}{U(s)} = (C(s\mathbf{I} A)^{-1}B + D)$ (Single-Input-Single Output (SISO) system)
- If Multi-Input-Multi-Output (MIMO) system, each element in matrix $C(s\mathbf{I} A)^{-1}B + D$ is a scalar transfer function, so still called "transfer function"



System Representation - Equivalent Representations

Block Diagram

force mechanical displacement x



 $y = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ x \end{bmatrix}$

 $\begin{bmatrix} m & 0 \\ 0 & 1 \end{bmatrix} \frac{d}{dt} \begin{bmatrix} v \\ x \end{bmatrix} + \begin{bmatrix} d & k \\ -1 & 0 \end{bmatrix} \begin{bmatrix} v \\ x \end{bmatrix} = \begin{bmatrix} u \\ 0 \end{bmatrix}$

Transfer Function

$$\frac{Y(s)}{U(s)} = \frac{1}{(ms^2 + ds + k)}$$



 $m\ddot{x} + d\dot{x} + kx = u$

$$\dot{V}_{out}(t) = -\frac{1}{RC}V_{out}(t) + \frac{1}{RC}V_{in}(t) \qquad \qquad \frac{V_{out}(s)}{V_{in}(s)} = \frac{1}{RCs+1}$$
$$y(t) = Vout(t)$$



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

$$Y(s) = \left(C(s\mathbf{I} - A)^{-1}B + D\right)U(s)$$



System Representation – Feedforward/Feedback



Analysis of Dynamics (Time Dependent Behavior)

- What is undriven "natural" behavior of system? $\dot{x}(t) = Ax(t) + B(t)$
- Defined by the eigenvalues λ :

 $\lambda x = Ax$

• An arbitrary vector v can be expressed as sum of eigenvectors: $v = \sum_{k=1}^{n} \alpha_k x_k$

• Then:
$$Av = \sum_{k=1}^{n} \alpha_k A x_k = \sum_{k=1}^{n} \alpha_k \lambda_k x_k \implies \dot{x} = \sum_{k=1}^{n} \alpha_k \dot{x}_k = \sum_{k=1}^{n} \alpha_k \lambda_k x_k$$

• That is, we can analyze as n scalar ODE's:

$$\dot{x}_k = \lambda_k x_k \implies x_k(t) = e^{\lambda_k t} x_k(0)$$

• To determine stability of the system:

$$\sigma_k = real(\lambda_k) < 0 \implies x_k(t) \to 0, \ t \to \infty \quad \text{(stable)}$$

$$\sigma_k = real(\lambda_k) > 0 \implies x_k(t) \to \infty, \ t \to \infty \quad \text{(unstable)}$$

- If ANY eigenvalue has $Re(\lambda)>0 =>$ system is UNSTABLE.
- Otherwise, system is STABLE.





complex-plane

Analysis of Dynamics (Laplace Domain)



LHP/RHP = Left/Right Half Plane



Analysis of Dynamics (Time vs. Laplace Domains)

- Eigenvalue is a complex number λ satisfying:
 - $-(\lambda I A)x = 0$ for some $x \neq 0$
 - $\Leftrightarrow (\lambda I A)^{-1}$ does not exist
 - \Leftrightarrow determinant $|\lambda I A| = 0$

polynomial in

• Note similarity to portion of Transfer Function:

$$Y(s) = \left(C(s\mathbf{I} - A)^{-1}B + D\right)U(s)$$

• In fact,

$$(s\mathbf{I} - A)^{-1} = \frac{1}{|s\mathbf{I} - A|} Adj(s\mathbf{I} - A)$$
 where: $\begin{vmatrix} X \end{vmatrix} = \text{determinant of } X \\ Adj(X) = \text{adjugate of } X \text{ (matrix of cofactors)} \end{vmatrix}$

• A common situation is D=0, so that the transfer function is:

s
$$\frac{1}{(CAdj(s\mathbf{I} - A)B)}$$
 matrix of polynomials in s

That is, the POLES of the transfer function = roots of determinant of (sI-A)
 = EIGENVALUES of A



Understanding System Response – Correspondence Between Eigenvalue (Pole) Location and Time Response





Understanding System Response – Frequency Response



Understanding System Response – Bode Plots of Frequency Response

• Examples:



NOTE: Bode gain plot is ratio of <u>powers</u> (20log₁₀(amplitude ratio)).



Objectives of Control – Tracking and Regulation

• Control plasma major radius:

- Assume plasma current (I_p) is positive
- Radial hoop force F_R pushes plasma outward
- Vertical field (B_z) produced by outer coils holds it in desired location (**regulation**) ...
- ... or moves plasma in/out to match a timedependent request (tracking)



tokamak positive current sign convention (viewed from above)





Objectives of Control – Tracking and Regulation

Control plasma elongation:

 Increasing elongation (κ) has been shown to improve performance, so we want to control:

$$\kappa = \frac{b}{a}$$

- Control accomplished by "pulling" on top and bottom of plasma
- However, elongating plasma introduces destabilizing field curvature (explained in a moment)





Objectives of Control – Tracking and Regulation

• Derivation of Closed-Loop Transfer Function:





Objectives of Control - Stabilization

• Open-loop instability:



 Plasma vertical instability (caused by destabilizing curvature):



Anti-symmetric coils provide radial field to apply force that opposes plasma vertical motion





Objectives of Control – Avoid Closed Loop Instability



Objectives of Control – Closed Loop Stability





Objectives of Control – Disturbance & Noise Rejection



Disturbance rejection means ratio of norms of errors to input is small:



• Noise rejection means ratio of norms of errors to input noise is small:



 $\frac{\|e(s)\|}{\|n(s)\|} \ll 1$ (attenuate effect of noise)

- These are ensured by making norms of transfer functions small, e.g.: $\frac{\|e(s)\|}{\|d_{V}(s)\|} \le \left\| -(I + CTPK)^{-1}CT \right\| << 1$
- For example, large gains in controller K can make this small.



Performance Requirements – Time Domain

- Typical Specifications on Step Response:
 - Rise Time < X seconds
 - Percent Overshoot < Y %</p>
 - Settling Time < Z seconds (within ε %)





Performance Requirements - Stability





System Representation – Sampled Data Systems

• Modern plasma control mixes discrete- and continuous-time systems:



• Approach (1) to Control Design:

- Treat entire system as continuous time. Develop continuous controller K(s), then convert to discrete controller K(z).
- Issues: Close to original physics models, but sampling rate must be fast enough to justify treating discrete controller as continuous.

• Approach (2) to Control Design:

- Treat entire system as discrete and develop discrete controller directly. (Methods exist to convert mixed continous/discrete to all discrete system.)
- Issues: Direct production of discrete controller with given sample rate, but difficult to retain physical intuition.



System Representation – Discrete Time Systems

- Time now represented by integers k=1,2,... (i.e., time = sample number)
- State-Space models are difference equations: x(k+1) = Ax(k) + Bu(k)

y(k) = Cx(k) + Du(k)

Now we have Z-transform instead of Laplace transform

$$F(z) = Z\{f(k)\} = \sum_{k=-\infty}^{\infty} f(k)z^{-k}$$

u(**k**)

• Nice properties:

 $\begin{array}{c|c} u(k)=f(k) \\ \hline U(z)=F(z) \end{array} \begin{array}{c|c} 1\text{-sample} & y(k)=f(k-1) \\ \hline delay & Y(z)=z^{-1}F(z) \end{array}$

• Transfer functions now defined on "z"-plane:

$$zX(z) = AX(z) + BU(z)$$

$$Y(z) = CX(z) + DU(z)$$

$$\Rightarrow Y(z) = C((zI - A)^{-1}BU(z)) + DU(z)$$





v(k)

System

Controllers – Example Digital Implementations

Simple gain multiplier:

- Command signal u(k) = K * e(k) (error e(k) = r(k) y(k))
- K can be scalar (SISO) or matrix (MIMO)
- Digital filter (SISO): only previous samples $u(k) = a_{1}u(k-1) + \dots + a_{n}u(k-n) \\ + b_{0}e(k) + b_{1}e(k-1) + \dots + b_{m}e(k-m) \Rightarrow \frac{U(z)}{E(z)} = \frac{b_{0} + b_{1}z^{-1} + \dots + b_{m}z^{-m}}{1 - a_{1}z^{-1} - \dots - a_{n}z^{-n}}$ present and previous samples

• State Space:

- Either SISO or MIMO: $u(k) = C_c x_c (k-1) + D_c e(k)$ $x_c(k) = A_c x_c(k-1) + B_c e(k)$
- Output computed from present error and previous state
- Controller state is updated at each time step



Next – some examples of types of controllers

• Why different controller types?

- Simple versus difficult to use
- SISO versus MIMO system
- Highly coupled versus mostly diagonal system
- How problem is posed (what you "care about")
- Noise characteristics of system
- Disturbance sources/effects and characteristics
- Level of knowledge of system dynamics (model uncertainty)
- Guaranteed stability including uncertainty versus nominal stability (not accounting for uncertainty)
- Guaranteed performance including uncertainty versus nominal performance (not accounting for uncertainty)



Controller Types – PID controllers

- PID = Proportional, Derivative, Integral feedback
 - Ideal: $u(t) = K_p e(t) + K_D \dot{e}(t) + K_I \int e(t) dt$
 - e(t) = error signal, u(t) = command to control actuator
- Simple and often all that is needed (DO NOT confuse "often" with "always")
- Purpose of each term:
 - K_P : Tracking ($K_PG/(1+K_PG) \sim 1$ over control bandwidth)
 - K_I: Regulation (gain is infinite at $j\omega=0 \Rightarrow$ steady-state error = 0)
 - K_D: Damping, phase lead
- Issues:
 - K_P : can destabilize if too large
 - K_I: integrator windup
 - K_D:amplifies noise at high frequencies

(implemented as simple gain multiplier)

(implemented as digital filter)

(implemented as digital filter)

- Advantage:
 - Simple, tunable
- Disadvantage
 - Difficult to determine gains in highly coupled systems



Controller Types – LQG controllers

- LQG= Linear, Quadratic, Gaussian ("optimal control")
 - Assume the linear system has Gaussian noise v(t), w(t): $\dot{x}(t) = Ax(t) + Bu(t) + v(t)$

y(t) = Cx(t) + w(t)

- Minimize objective functional J ... where O > O R>O (superscript cost) $J = \int_0^\infty x(t)^T Q x(t) + u(t)^T R u(t) dt$
- ... where Q>0, R>0 (quadratic cost)
- Typically, states x are variations around a stable equilibrium x_0
- Sometimes J has terms for output y or error e = reference output
- Main idea: keep signals small "on average" (variation due to noise)
- Optimal controller is given by: $\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + K(y(t) C\hat{x}(t))$

 $u(t) = -L\hat{x}(t)$

- First equation is the Kalman Filter, which provides an optimal estimate for x
- If state measured directly, insert x in place of x-hat and use 2nd equation only
- Advantage:
 - Straightforward to generate controller optimal against "noise", once J is defined
- Disadvantage
 - Matrices Q and R typically determined through trial and error



Controller Types – H-infinity ("robust") controllers

- \mathbf{H}^{∞} = method for synthesizing robust controllers ("Hardy space, infinity norm")
- Robust = guaranteed stability/performance with unknown (but bounded) uncertainty in plant model
 - Infinity ("worst case") norm : $\|\Delta\|_{\infty} < bound$



• Main idea

- Remove Δ from picture ...
- ... and make transfer function from Δ_{out} to Δ_{in} as small as possible

• Advantage:

- Guarantees on stability and performance in the deployed feedback system

• Disadvantage:

- More difficult to understand and to use; some tools produce conservative designs





Summary

• Control Terminology and Concepts:

 Linear/Nonlinear systems, Linear-Time-Invariant system, Discrete time system, System gain/phase, s-plane, z-plane, poles, zeros, pure delay, phase lag, phase lead, SISO, MIMO, feedforward, feedback, open-loop instability, control-driven instability, LHP, RHP, frequency response, roll-off, gain margin, phase margin, stability margin, disturbance, overshoot, rise time, settling time

Control Tools and Methods:

 Block Diagrams, Transfer Functions, State Space Models, Laplace Transform, Z-Transform, Fourier Transform, Bode plot, derivation of closed-loop transfer function, Root Locus, PID controllers, LQG controllers, H-infinity controllers

• Multiple Objectives of Control:

- Stability,
- Tracking and Regulation
- Disturbance Rejection
- Noise Rejection
- Robustness



Further Reading

Free downloadable books:

- Wikibook of automatic control systems, http://en.wikibooks.org/wiki/ Control_Systems (not how you would want to learn control, but useful as a reference)
- Kwaakernak and Sivan, Linear Optimal Control Systems, http:// www.ieeecss.org/PAB/classics/
- Wikibook of signals and systems, http://en.wikibooks.org/wiki/ Signals_and_Systems
- Matlab documentation at http://www.mathworks.com/access/helpdesk/ help/helpdesk.html
 - Control System Toolbox, Robust Control Toolbox

Good entry-level control books:

- Franklin, Powell, Emami-Naeini, Feedback Control of Dynamic Systems
- Friedland, Control System Design: An Introduction to State-Space Methods

