



CDS 101/110a: Lecture 10-1

Robust Performance



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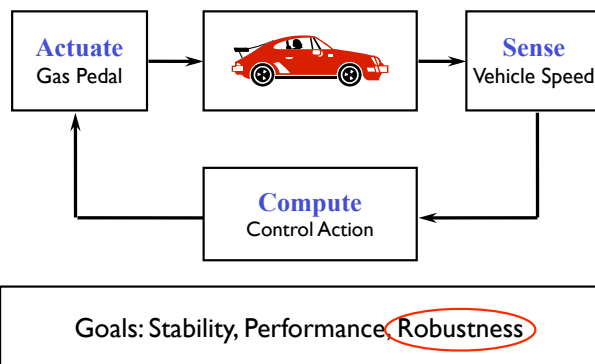
Goals:

- Describe how to represent uncertainty in process dynamics
- Describe how to analyze a system in the presence of uncertainty
- Review the main principles and tools for the course

Reading:

- Åström and Murray, Feedback Systems, Ch 12
- Advanced: Lewis, Chapter 9
- CDS 210: DFT, Chapters 4-6

Control = Sensing + Computation + Actuation



Stability: bounded inputs produce bounded outputs

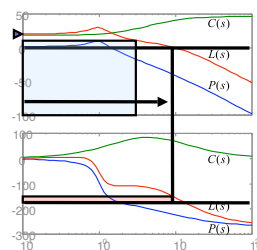
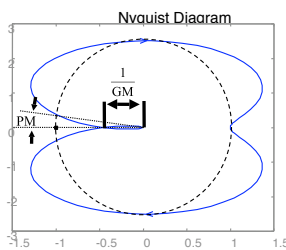
- Need to check all input/output pairs (Gang of Four/Six)
- Necessary and sufficient condition: check for nonzero solutions around feedback loop

Performance: desirable system response

- Step response: rise time, settling time, overshoot, etc
- Frequency response: tracking signals over given range

Robustness: stability/performance in presence of uncertainties

- Need to check stability for set of disturbances & system models



Modeling Uncertainty

Noise and disturbances

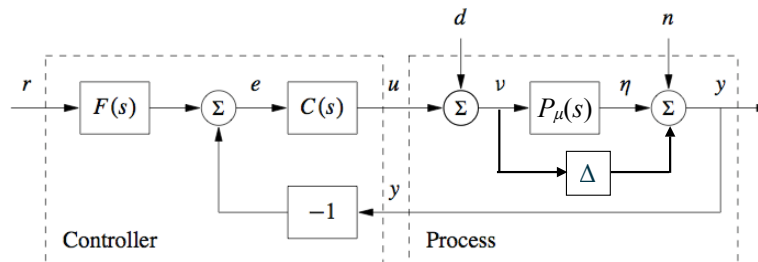
- Model the amount of noise by its signal strength in different frequency bands
- Can model signal strength by peak amplitude, average energy, and other norms
- Typical example: Dryden gust models (filtered white noise)

Parametric uncertainty

- Unknown parameters or parameters that vary from plant to plant
- Typically specified as tolerances on the basic parameters that describe system

Unmodeled dynamics

- High frequency dynamics (modes, etc) can be excited by control loops
- Use bounded operators to account for effects of unmodeled modes:



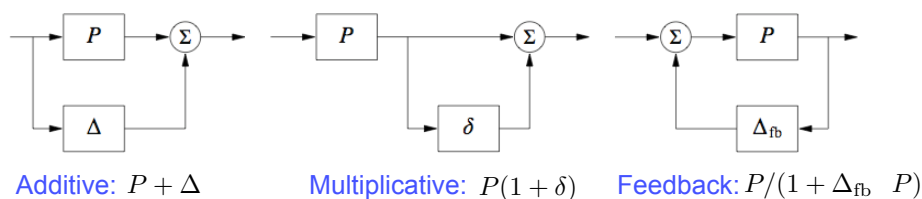
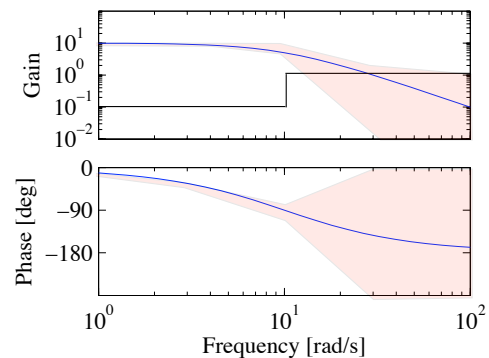
Unmodeled Dynamics

“Model” unknown dynamics through bounded transfer function

- Simplest case: additive uncertainty
- $$\hat{P}(s) = P(s) + \Delta(s), \quad |\Delta(s)| < W_2(s)$$
- $\Delta(s)$ is unknown, but bounded in magnitude
 - Magnitude bound can depend on frequency; typically have a good model at low frequency

Different types of uncertainty can be used depending on where uncertainty enters

- Multiplicative: good for unknown gain or actuator dynamics
- Feedback: good for “leakage” effects (eg, in electrical circuits)



Stability in the Presence of Uncertainty

Characterize stability in terms of stability margin s_m

- Stability margin = distance on Nyquist plot to -1 point
- Stability margin = $1/M_s$ (M_s = maximum sensitivity)

$$M_s = \max |S(i\omega)| = \max \left| \frac{1}{1+L} \right|,$$

$$s_m = \min |(-1) - L| = \min |1 + L| = 1/M_s$$

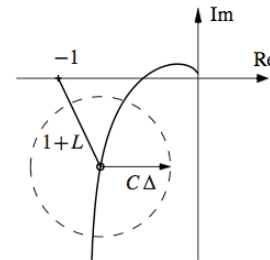
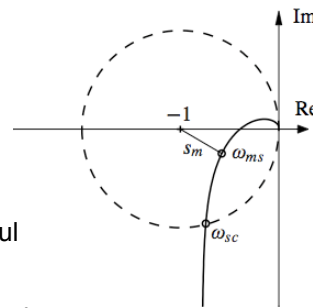
- For robustness analysis, stability margin is more useful than classical gain and phase margins

Robust stability: verify no new net encirclements occur

- Assume that nominal system is stable
- New loop transfer function: $\tilde{L} = (P + \Delta)C = L + C\Delta$
- No net encirclements as long as $|C\Delta| < |1 + L|$
- Can rewrite as bound on allowable perturbation

$$|\Delta| < \left| \frac{1+PC}{C} \right| = \left| \frac{P}{T} \right| \quad \text{or} \quad |\delta| = \left| \frac{\Delta}{P} \right| < \frac{1}{|T|}$$

- If condition is satisfied, then sm will never cross to zero
=> no new net encirclements

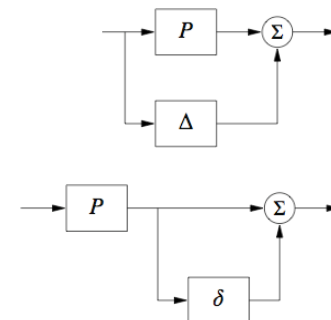
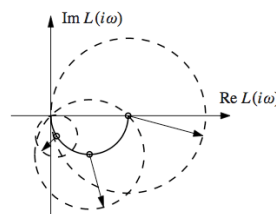
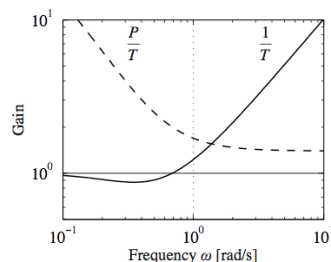


Example: Cruise Control

Question: how accurate does our model have to be

- Start with very simple model (25 m/s, 4th gear)
- Ignores details of engine dynamics, sensor delays, etc
- Model unknown dynamics as additive or multiplicative uncertainty (can convert bounds from one to the other)
- System will remain stable as long as

$$|\Delta| < \left| \frac{1+PC}{C} \right| = \left| \frac{P}{T} \right| \quad |\delta| = \left| \frac{\Delta}{P} \right| < \frac{1}{|T|}$$



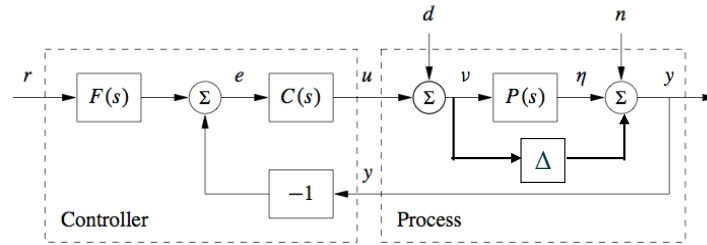
Remarks

- Conditions show why we can use simple models for designs
- Models must be accurate near critical point (eg, crossover)

Robust Performance Using Sensitivity Functions

Performance conditions

- What happens to performance in the presence of uncertainty?
- Start by looking at the *sensitivity* of transfer functions to process model



$$G_{yd} = \frac{P}{1 + PC} = PS$$

$$\frac{dG_{yd}}{dP} = \frac{1}{(1 + PC)^2} = \frac{SP}{P(1 + PC)} = S \frac{G_{yd}}{P}$$

$$G_{yd} + \Delta_{yd} \approx \frac{P + \Delta}{1 + (P + \Delta)C}$$

$$\approx \frac{P}{1 + PC} + \frac{\Delta}{1 + PC} = G_{yd} + S\Delta$$

Other measures

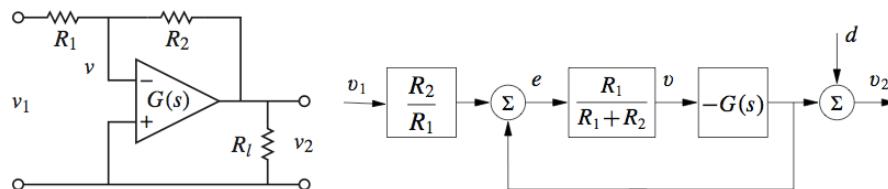
$$\frac{dG_{yr}}{G_{yr}} = S \frac{dP}{P}$$

$$\frac{dG_{yd}}{G_{yd}} = S \frac{dP}{P}$$

$$\frac{dG_{un}}{G_{un}} = T \frac{dP}{P}$$

- Sensitivity functions actually come from this analysis

Example: Operational Amplifier



Note: disturbance enters in different location

Basic amplifier with unmodeled (high frequency) dynamics

- Start with low frequency model:

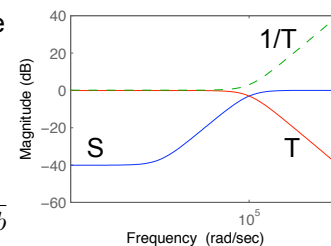
$$G_{v_2 v_1} = -\frac{R_2}{R_1} \frac{G(s)}{G(s) + R_2/R_1 + 1}. \quad G(s) = \frac{b}{s + a}, \quad \alpha = \frac{R_2}{R_1}$$

- Typical parameters $b \gg a \gg 1$
- Robust stability: see how much uncertainty we can handle

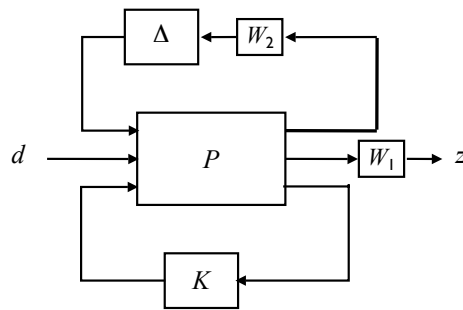
$$|\delta| = \left| \frac{\Delta}{P} \right| < \frac{1}{|T|} \quad T = \frac{\alpha b}{s + a + \alpha b}$$

- Robust performance: effect on tracking, disturbance rej

$$\frac{dG_{yr}}{G_{yr}} = S \frac{dP}{P} \quad \frac{dG_{yd}}{G_{yd}} = T \frac{dP}{P} \quad S = \frac{s + a}{s + a + \alpha b}$$



Alternative Formulation for Robust Performance



d disturbance signal
 z output signal
 Δ uncertainty block
 W_2 performance weight
 W_2 uncertainty weight

Goal: guaranteed performance in presence of uncertainty

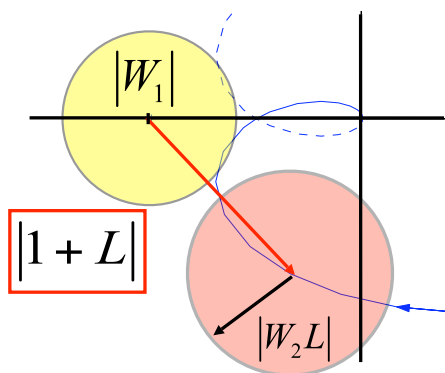
$$\|z\|_2 \leq \gamma \|d\|_2 \quad \text{for all } \|\Delta\|_\infty \leq 1$$

$$\|u\|_2 = \sqrt{\int_0^\infty |u(t)|^2 dt}$$

$$\|H\|_\infty = \max_\omega |H(i\omega)|$$

- Compare energy in disturbances to energy in outputs
- Use frequency weights to change performance/uncertainty descriptions
- “Can I get X level of performance even with Y level of uncertainty?”

Design for Robust Performance (ala DFT)



Basic idea: interpret conditions on Nyquist

- Performance: keep sensitivity function small
 $|W_1 S| < 1 \implies |W_1| < |1 + L|$
- W_1 serves as performance weighting function
- Stability: avoid additional encirclements
 $|W_2 \delta| < \frac{1}{|T|} \implies |W_2 L| < |1 + L|, |\delta| < 1$
- W_2 serves as uncertainty weighting function
- Individual conditions provide robust stability and (nominal) performance

Design loop shape to satisfy robust stability and performance conditions

- Nominal performance:
 $|W_1| < |1 + L|$ for all ω
- Robust stability:
 $|W_2 L| < |1 + L|$ for all ω
- Missing: robust performance...

Theorem: robust performance if circles don't intersect on Nyquist plot:

$$|W_1| + |W_2 L| < |1 + L| \quad \text{for all } \omega$$

- Holds for multiplicative uncertainty + weighted sensitivity (cf DFT)

Tools for Analyzing and Synthesizing Controllers

Robust Multi-Variable Control Theory

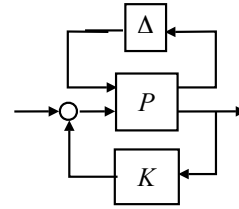
- Generalizes gain/phase margin to MIMO systems
- Uses operator theory to handle uncertainty, performance
- Uses state space theory to performance computations (LMIs)

Analysis Tools

- H^∞ gains for multi-input, multi-output systems
- μ analysis software
 - Allow structured uncertainty descriptions (fairly general)
 - Computes upper and lower bounds on performance
 - Wide usage in aerospace industry
- SOSTOOLS: Nonlinear extensions

Synthesis Tools

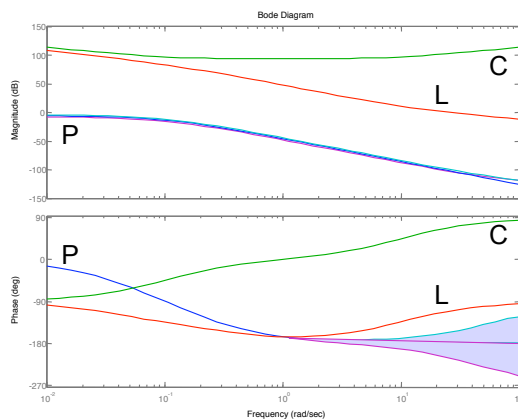
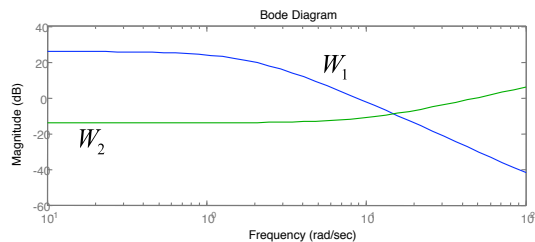
- LQR/LQG + H^∞ "loop shaping"; modern tools for control engineers
- μ synthesis software; tends to generate high order controllers
- Model reduction software for reducing order of plants, controllers



Example: Robust Cruise Control

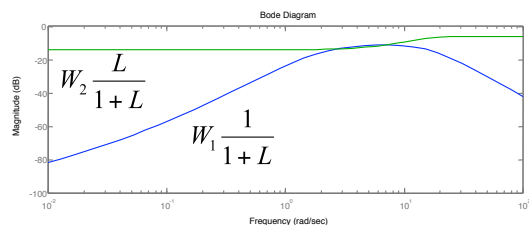


$$P(s) = \frac{1/m}{s + b/m} \cdot \frac{r}{s + a} (1 + W_2 \Delta)$$



Theorem

- Performance: $\left| W_1 \frac{1}{1+L} \right| < 1 \quad \forall \omega$
- Robust Stability: $\left| W_2 \frac{L}{1+L} \right| < 1 \quad \forall \omega$



Course Summary: Two Main Principles of Control

Design of Dynamics through Feedback

- Feedback allows the dynamics of a system to be modified
- Key idea: interconnection gives closed loop that modifies natural behavior
- Tools: eigenvalue assignment, loop shaping

Robustness to Uncertainty through Feedback

- Feedback allows high performance in the presence of uncertainty
- Key idea: accurate sensing to compare actual to desired, correction through computation and actuation
- Tools: stability margins, sensitivity functions

