

Control

- Applying input to cause system variables to conform to desired values called

Feedback (close-loop) Control **Open-loop control** Compute control input without continuous variable measurement **Controlled System** Simple Controller Need to know EVERYTHING ACCURATELY to work right + Cruise-control car: friction(t), ramp_angle(t) manipulated control Actuator E-commerce server: Workload (request arrival rate? resource consumption?); system (service time? failures?) input variable function Open-loop control fails when error We don't know everythingWe make errors in estimation/modeling controlled Monitor ⊕← variable Things change reference

Feedback (close-loop) Control

- Measure variables and use it to compute control input
 - More complicated (so we need control theory)
 - Continuously measure & correct
 - + Cruise-control car: measure speed & change engine force
 - Ecommerce server: measure response time & admission control
 Embedded network: measure collision & change backoff window
- Feedback control theory makes it possible to control well even if
 - We don't know everything
 - We make errors in estimation/modeling
 Things change

Why feedback control? Open, unpredictable environments

- Deeply embedded networks: interaction with physical environments
 - Number of working nodes
 - Number of interesting events
 - Number of hops
 - Connectivity Available bandwidth
 - Congested area
- Internet: E-business, on-line stock broker
- Unpredictable off-the-shelf hardware

Why feedback control? We want QoS guarantees

Deeply embedded networks

- Update intruder position every 30 sec Report fire <= 1 min</p>
- E-business server
 - Purchase completion time <= 5 sec
- Throughput >= 1000 transaction/sec The problem: provide QoS guarantees in open, unpredictable
- environments

Advantage of feedback control theory

- Adaptive resource management heuristics
 - Laborious design/tuning/testing iterations Not enough confidence in face of untested workload

Queuing theory

- Doesn't handle feedbacks
- Not good at characterizing transient behavior in overload Feedback control theory

- Systematic theoretical approach for analysis and design
- Predict system response and stability to input

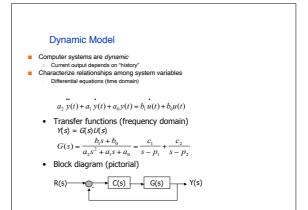
Outline

- Introduction
- What is feedback control? Why do today's computer systems need feedback control?
- Control design methodology
- System modeling Performance specs/metrics
- Controller design
- Summary

Control design methodology Controller Modeling Design analytical system IDs Control algorithm Root-Locus PI Control Satisfy Requirement Performance Specifications Analysis



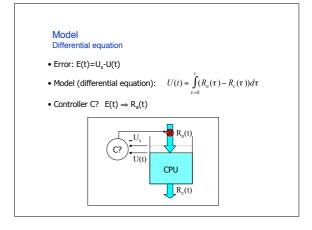
- Linear vs. non-linear (differential eqns)
- Deterministic vs. Stochastic
- **Time-invariant** vs. Time-varying
- Are coefficients functions of time?
- Continuous-time vs. Discrete-time
- System ID vs. First Principle

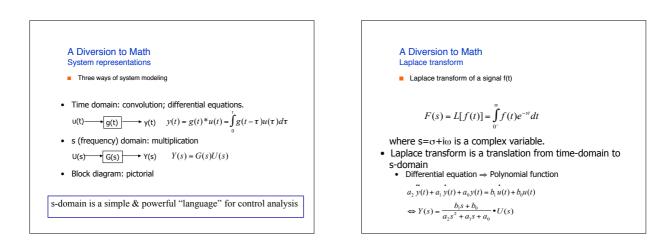


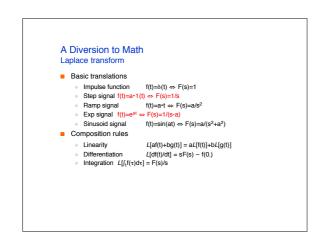
Example Utilization control in a video server

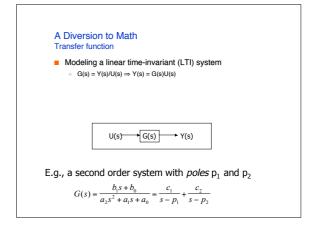
- Periodic task T_i corresponding to each video stream i

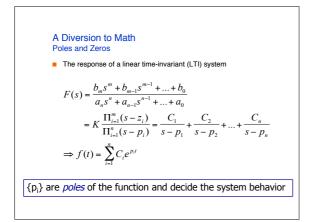
 - c[i]: processing time, p[i]: period
 Stream i's requested CPU utilization: u[i]=c[i]/p[i]
 Total CPU utilization: U(t)=2_(k)u[k], {k} is the set of active streams
- Completion rate: $R_{c}(t)=(\Sigma_{(tc)}u[m])/\Delta t$, where {m} is the set of terminated video streams during [t, t+ Δt]
- Unknown Admission rate: $R_a(t) = (\Sigma_{\{ka\}}u[j])/\Delta t$, where $\{j\}$ is the set of admitted streams during [t, t]t+∆t]
- Problem: design an admission controller to guarantee U(t)=U_a regardless of B_a(t).

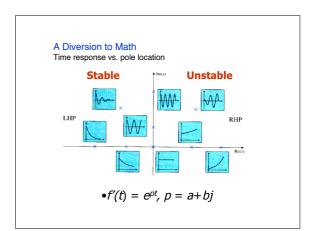


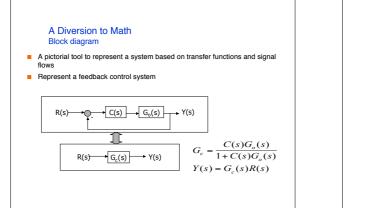


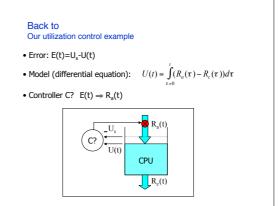


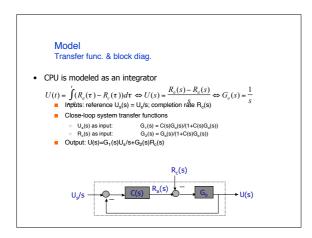


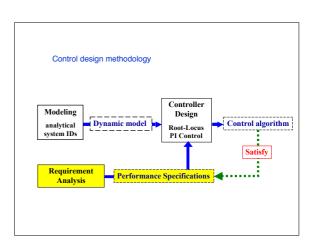


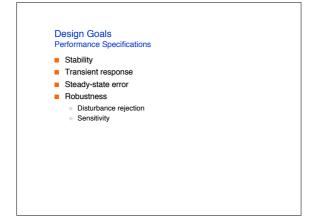


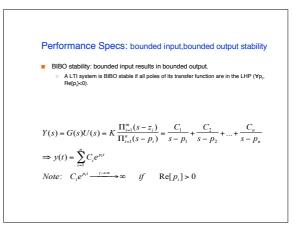


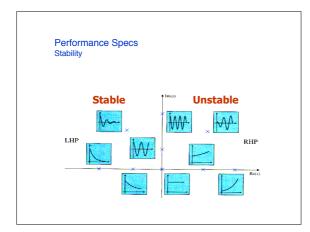


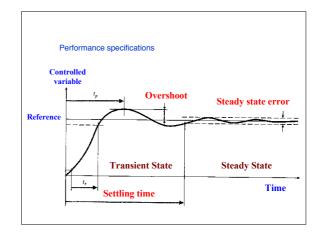


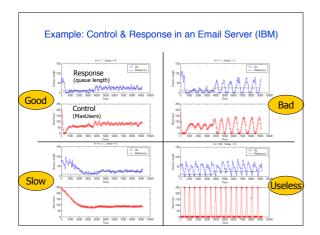


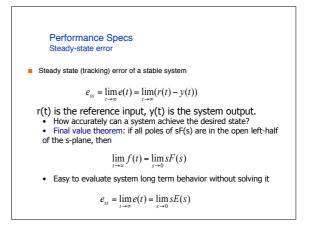


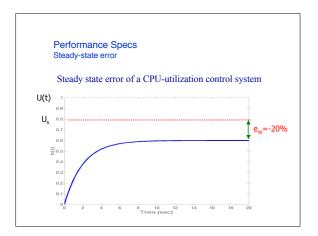


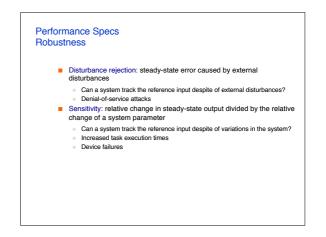


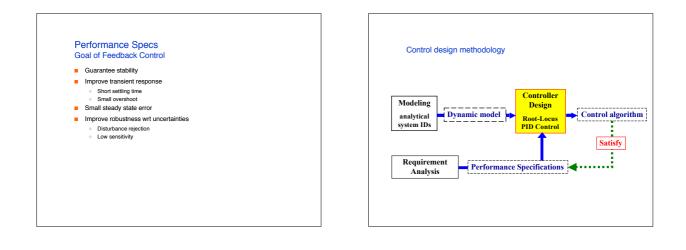


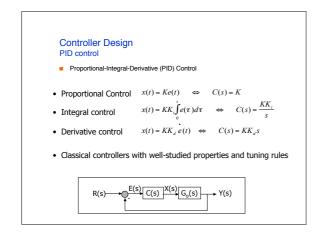


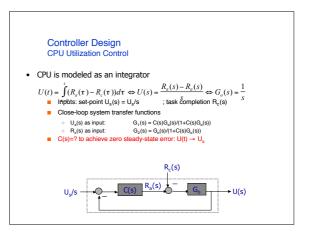


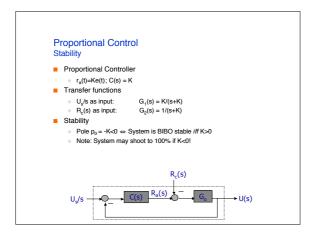


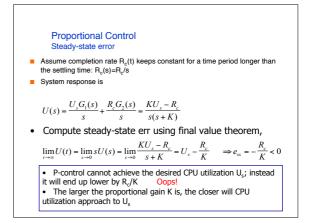


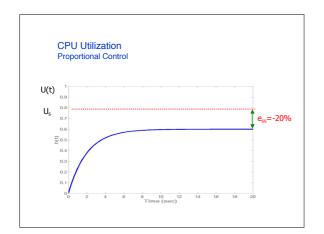


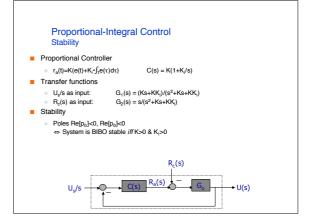


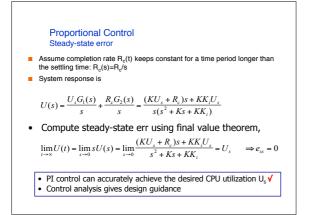


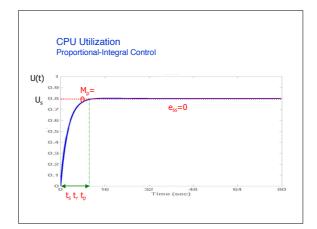












Controller Design Summary & pointers

- PID control: simple, works well in many systems

 - P control: may have non-zero steady-state error
 I control: improves steady-state tracking
 D control: may improve stability & transient response
 Linear continuous time control

- Root-locus design Frequency-response design State-space design G. F. Franklin et. al., *Feedback control of dynamic systems*

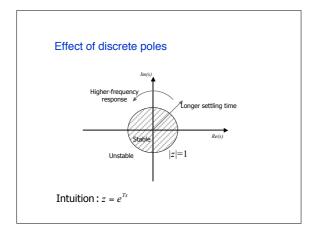
Discrete Control More useful for computer systems Time is discrete; sampled system denoted k instead of t Main tool is z-transform f(k) → F(z), where z is complex Analogous to Laplace transform for s-domain $\mathbf{Z}[f(k)] = F(z) = \sum_{k=0}^{\infty} f(k) z^{-k}$

Discrete Modeling

- Difference equation
 - + $V(m) = a_1 V(m-1) + a_2 V(m-2) + b_1 U(m-1) + b_2 U(m-2)$ + z domain: $V(z) = a_1 z^{-1} V(z) + a_2 z^2 V(z) + b_1 z^{-1} U(z) + b_2 z^2 U(z)$ + Transfer function $G(z) = (b_1 z + b_2)/(z^2 - a_1 z - a_2)$
- V(m): output in mth sampling window
- U(m): input in mth sampling window
- Order n: #sampling-periods in history affects current performance SP = 30 sec, and n = 2 \rightarrow Current system performance depends on previous 60 sec

Root Locus analysis of Discrete Systems

- Stability boundary: |z|=1 (Unit circle)
- Settling time = distance from Origin
- Speed = location relative to Im axis
 - Right half = slower Left half = faster



Feedback control works in CS

- U.Mass: network flow controllers (TCP/IP RED)
- IBM: Lotus Notes admission control
 UIUC: Distributed visual tracking
- UVA

 - A Web Caching QoS Apache Web Server QoS differentiation Active queue management in networks Processor thermal control Online data migration in network storage (with HP) Real-line embedded networking Control middleware Feedback control real-time scheduling

Advanced Control Topics

- Robust Control
- Can the system tolerate noise?
 Adaptive Control
 Controller changes over time (adapts)
 MIMO Control
- Multiple inputs and/or outputs
 Stochastic Control
- Controller minimizes variance
 Optimal Control
- Optimal Control
 Controller minimizes a cost function of error and control energy
 Nonlinear systems

 - Neuro-fuzzy control
 Challenging to derive analytic results

Issues for Computer Science

- Most systems are non-linear
 - But linear approximations may do + eg, fluid approximations
- First-principles modeling is difficult Use empirical techniques
- Mapping control objectives to feedback control loops
- ControlWare paper
- Deeply embedded networking
 - Massively decentralized control problem
 - Modelling
 Node failures