

Feedback Control Theory: Architectures and Tools for Real-Time Decision Making



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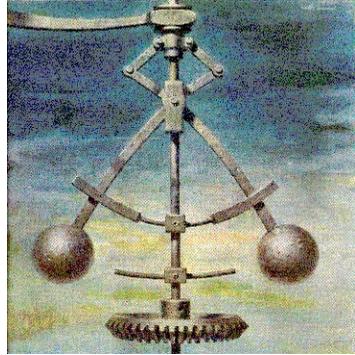
**Real-Time Decision Making Bootcamp
Simons Institute for the Theory of Computing**

24 January 2018

Goals for this lecture

- Give an *overview* of key ideas from control theory that might be relevant for applications in real-time decision making
- Encourage you to come find me if you want to learn more or work on a joint project applying ideas from control theory
- RMM schedule: Tue-Thu most weeks from now to end of Mar

What is “Control”?



Traditional view

- Use of feedback to provide stability, performance, robustness

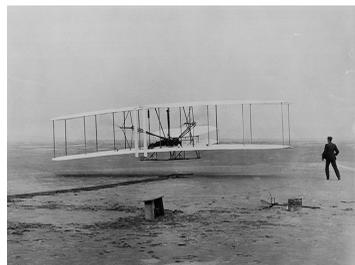
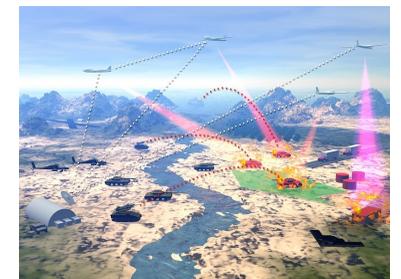
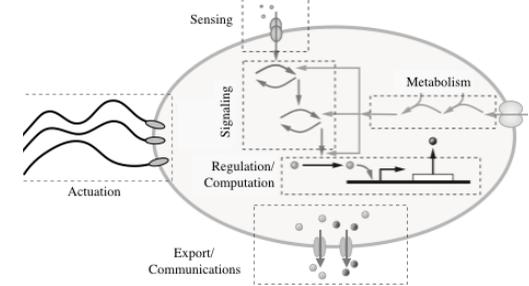
Emerging view

- Collection of tools and techniques for analyzing, designing, implementing complex systems in presence of uncertainty
- Combination of dynamics, interconnection (feedback/feedforward), communications, computing and software

Control = dynamics ⊗ uncertainty ⊗ feedforward ⊗ feedback

Key principles for control systems

- Principle #1: Feedback is a tool for **managing uncertainty** (system and environment) [no uncertainty ⇒ don't bother]
- Principle #2: Feedforward & feedback are tools for **design of dynamics** via integration of sensing, actuation & computation
- Corollary: Feedback enables subsystem **modularity** and **interoperability** ⇒ ability to **manage complexity at scale**



Important Trends in Control in the Last 15 Years

(Online) Optimization-based control

- Increased use of online optimization (MPC/RHC)
- Use knowledge of (current) constraints & environment to allow performance and adaptability

Layering, architectures, networked control systems

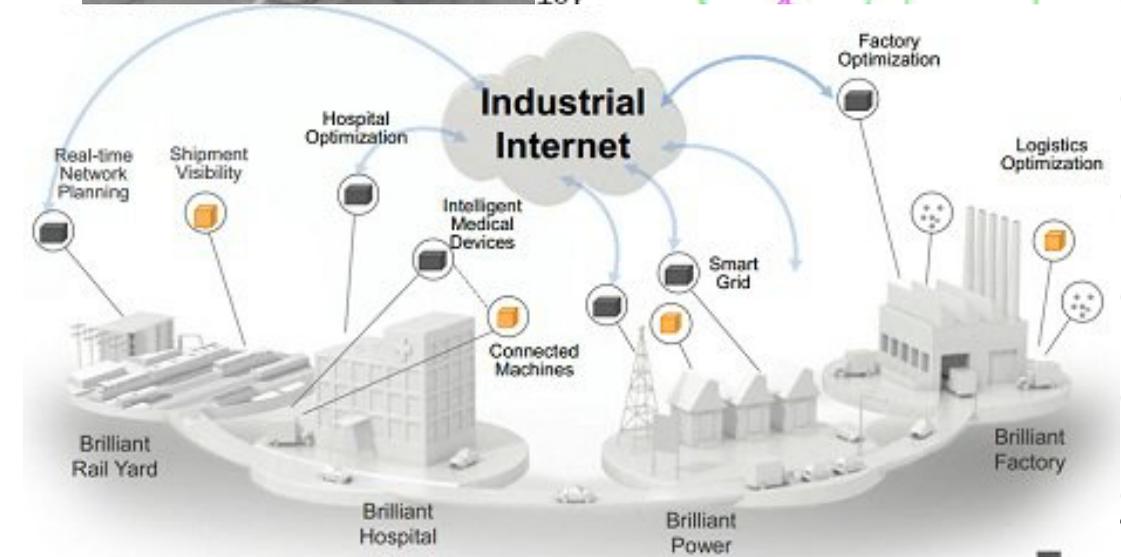
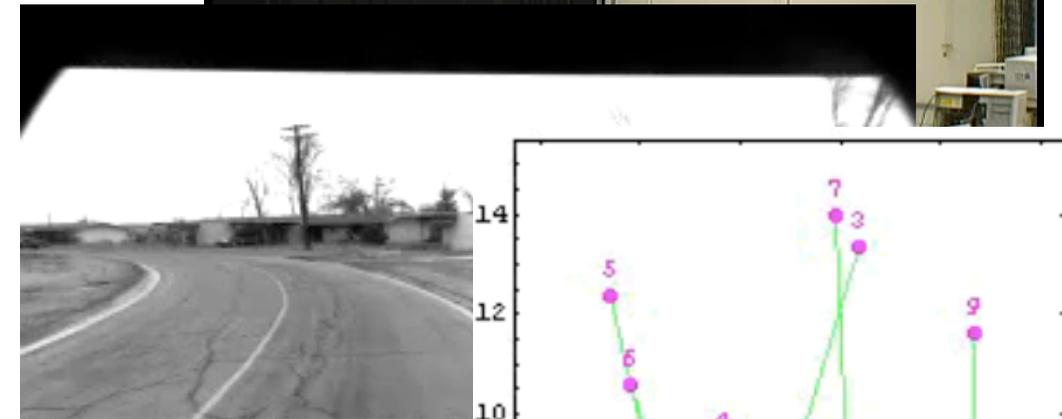
- Command & control at multiple levels of abstraction
- Modularity in product families via layers

Formal methods for analysis, design and synthesis

- Build on work in hybrid and discrete event systems
- Formal methods from computer science, adapted for “cyberphysical” (computing + control) systems

Components → Systems → Enterprise

- Increased scale: supply chains, smart grid, IoT
- Use of modeling, analysis and synthesis techniques at all levels. Integration of “software” with “controls”



Outline

Control Systems: Architectures and Examples

- “Standard model” (for control systems)
- Examples and relationship to real-time decision making

Control System Design Patterns

Design of Feedback Systems

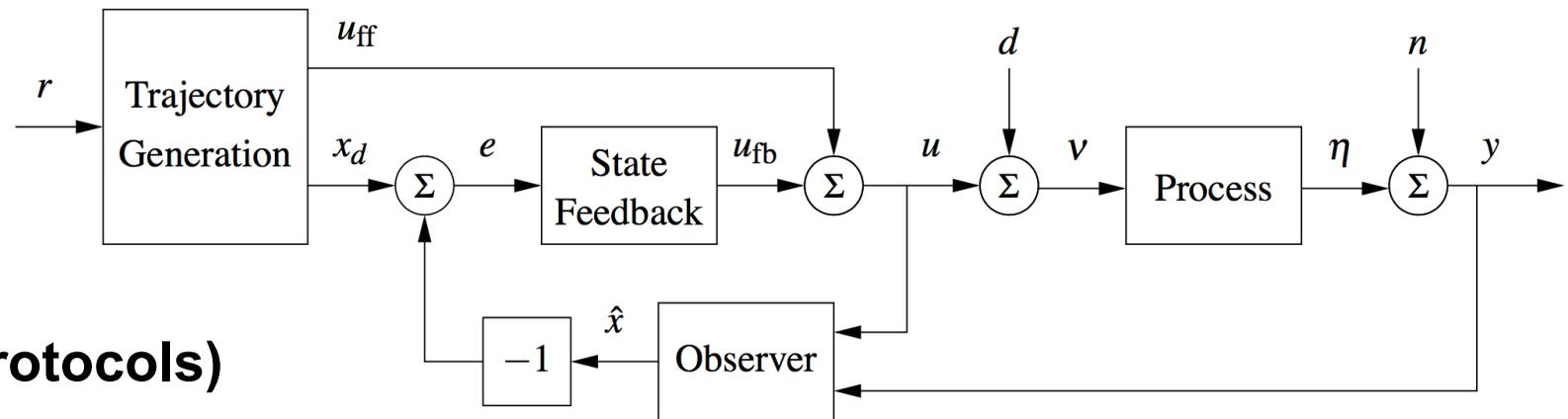
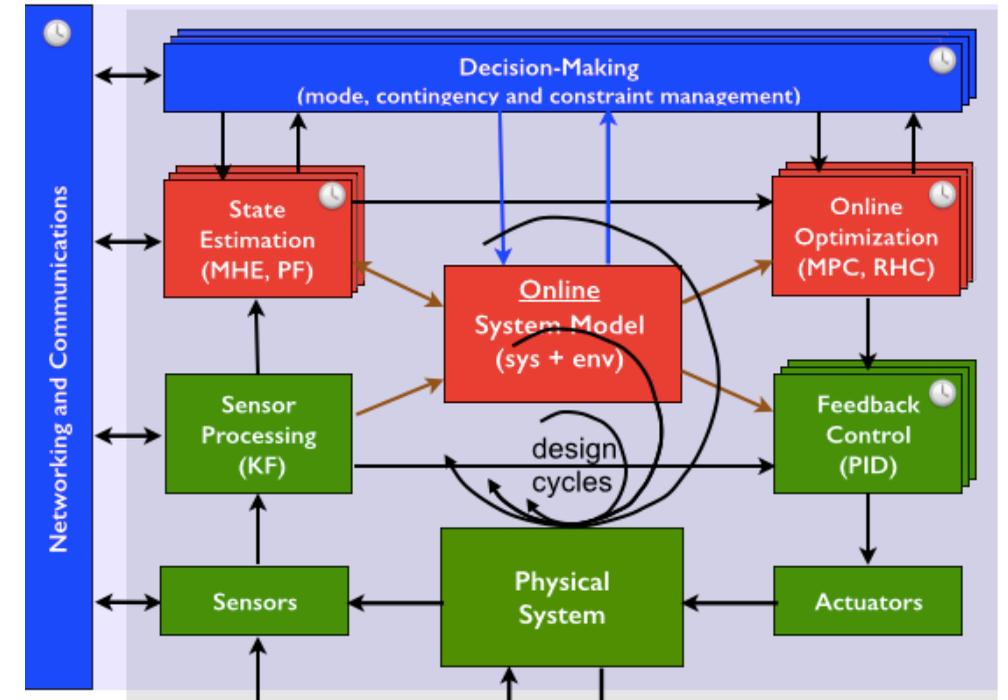
- Specifications for control systems
- Integral feedback (and PID)
- State feedback

Design of Feedforward Systems

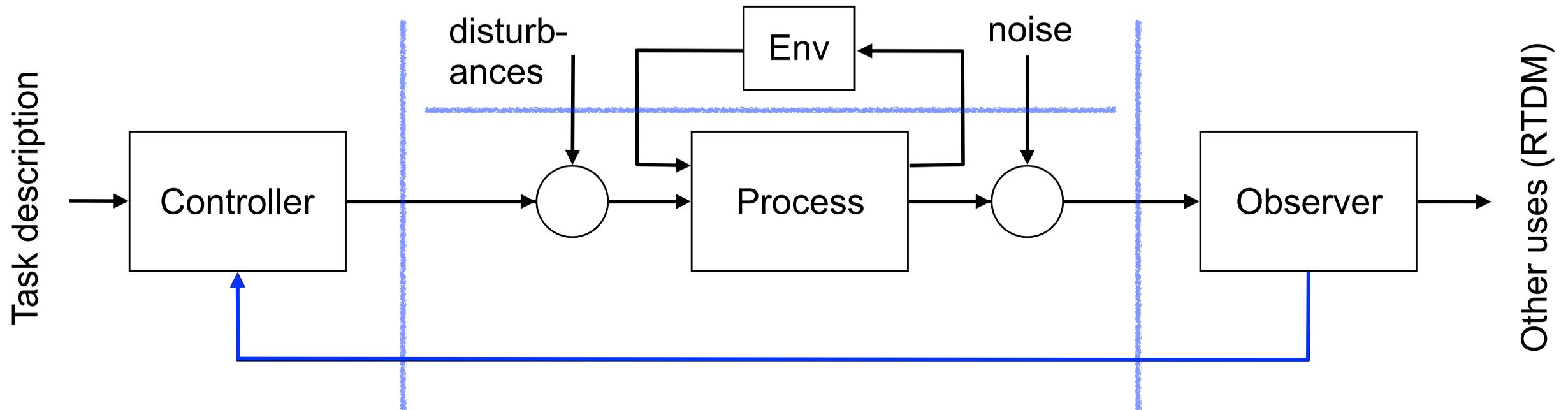
- Real-time optimization
- Receding horizon control

Layered architectures

Discrete state systems (reactive protocols)



Control System “Standard Model”



Key elements

- Process: input/output system w/ dynamics (memory)
- Environment: description of the uncertainty present in the system (bounded set of inputs/behaviors)
- Observer: real-time processing of process data
- Controller: achieve desired task via data, actions

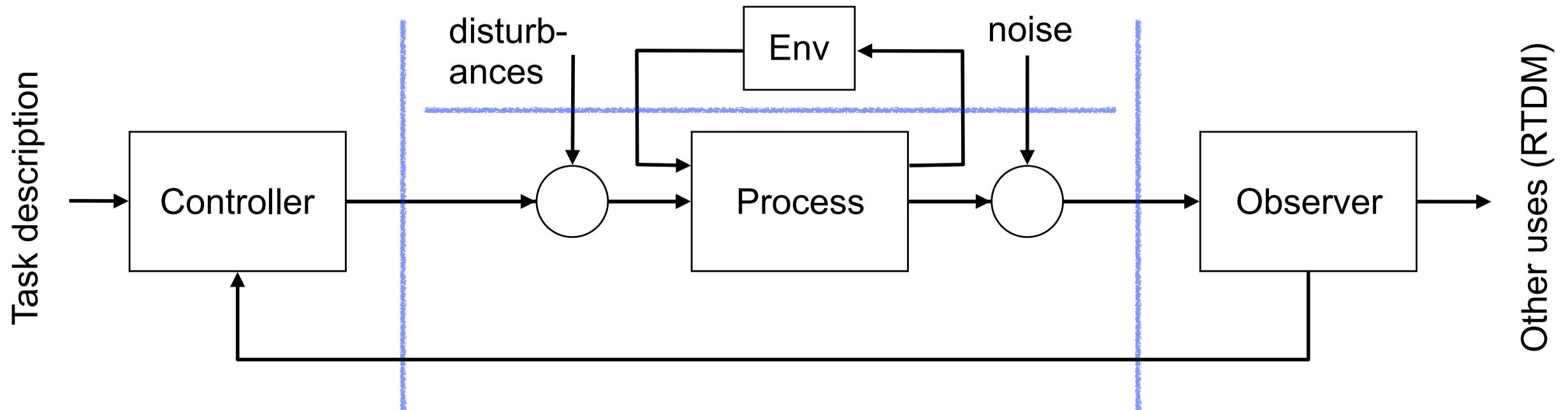
Disadvantages of feedback

- Increased complexity
- Potential for instability
- Amplification of noise

Advantages of feedback

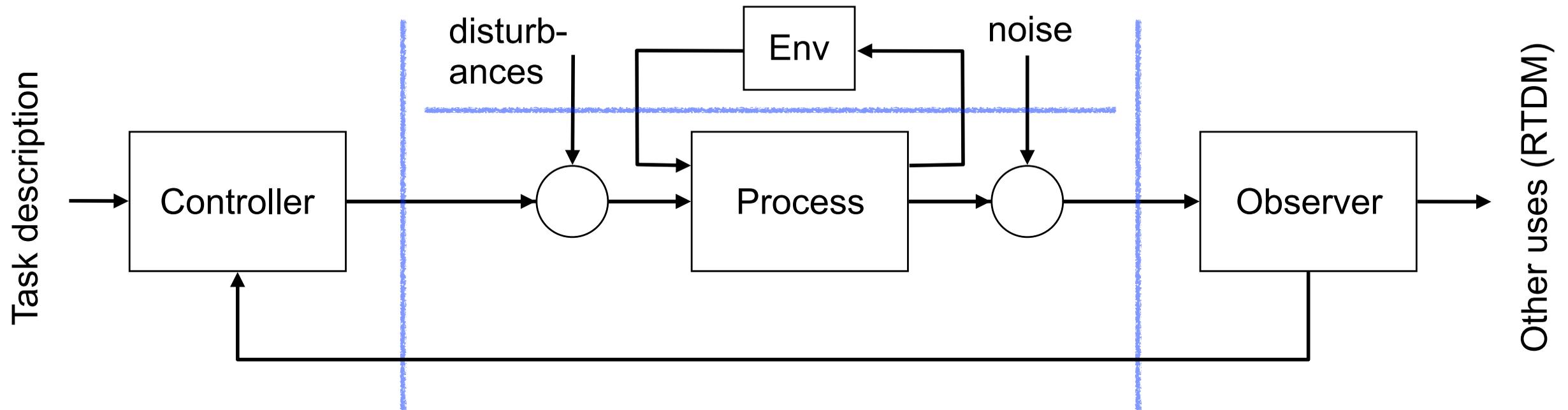
- Robustness to uncertainty
- Modularity and interoperability

Control System Examples



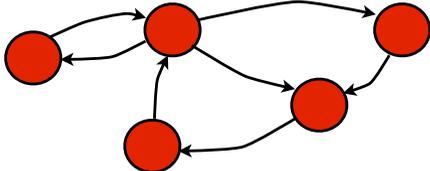
Application	Process description (input/output)	Specifications	Comments
Scientific detection	Actuator: N/A Sensor: instruments	Detect events	Observer only
Transportat'n networks	Actuator: schedules, incentives Sensor: demand, supply	Optimize utility function	Open-loop at fast time scale; closed-loop at slower rates
Autonomous vehicles	Actuator: gas, steer Sensor: cameras, radar, LIDAR, traffic	A to B w/out hitting anyone	Multiple decision-making loops; very complex environment

Different Types of Control Systems



System type	Modeling approaches	Specifications	Comments
Continuous states	Ordinary and partial differential equations; difference equations	Integrated cost over time/space	Well-studied; excellent tools avail (especially LTI systems)
Discrete state systems	Finite state automata, timed automata, Petri nets	Temporal logic formulae	Good tools for verification; design/synthesis is harder
Probabilistic systems	Stochastic ODEs, Kolmogorov equations, Markov chains	Expected values and moments	Well-studied; excellent tools avail (especially LTI, MDPs)

Control System Specifications

Level	Model	Specification
Regulation	$y = P_{yu}(s) u + P_{yd}(s) d$ $\ W(s)d(s)\ \leq 1$	$\ W_1 S + W_2 T\ _\infty < \gamma$
Optimization (planning)	$\dot{x} = f_\alpha(x, u)$ $g_\alpha(x, u, z) \leq 0$	$\min J = \int_0^T L_\alpha(x, u) dt$ $+ V(x(T))$
Decision-Making		$(\phi_{\text{init}} \wedge \square \phi_{\text{env}}) \implies$ $(\square \phi_{\text{safe}} \wedge \square \diamond_{\leq T} \phi_{\text{live}})$

Transient: initial response to input

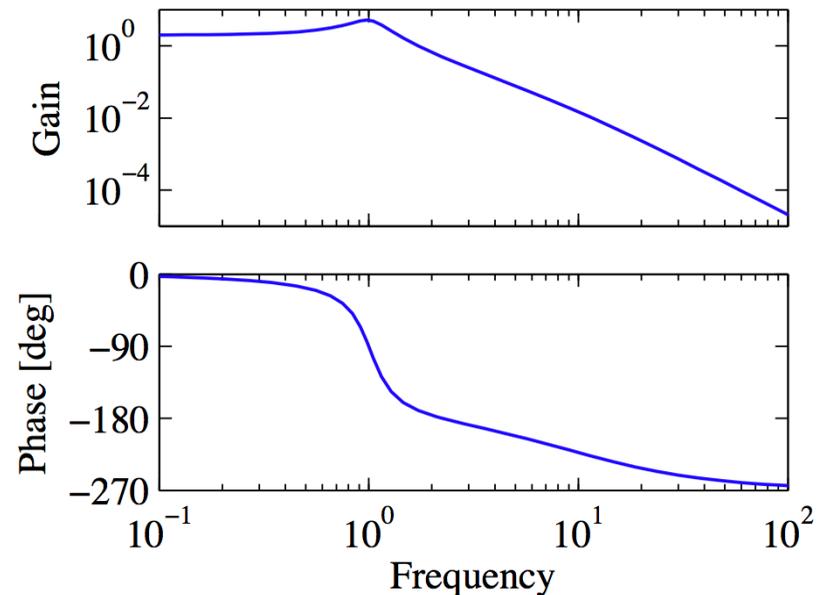
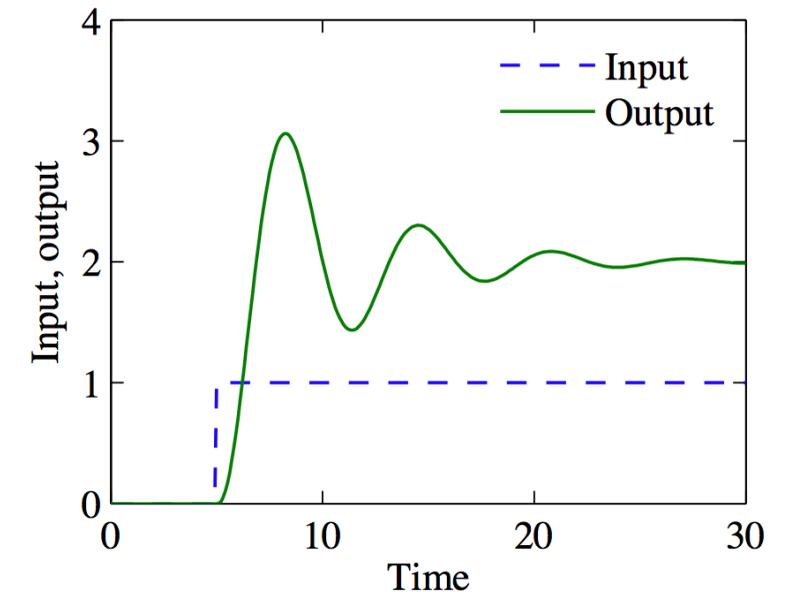
- Step response: rise time, overshoot, settling time, etc

Steady state: response after the transients have died out

- Frequency response: magnitude and phase for sinusoids

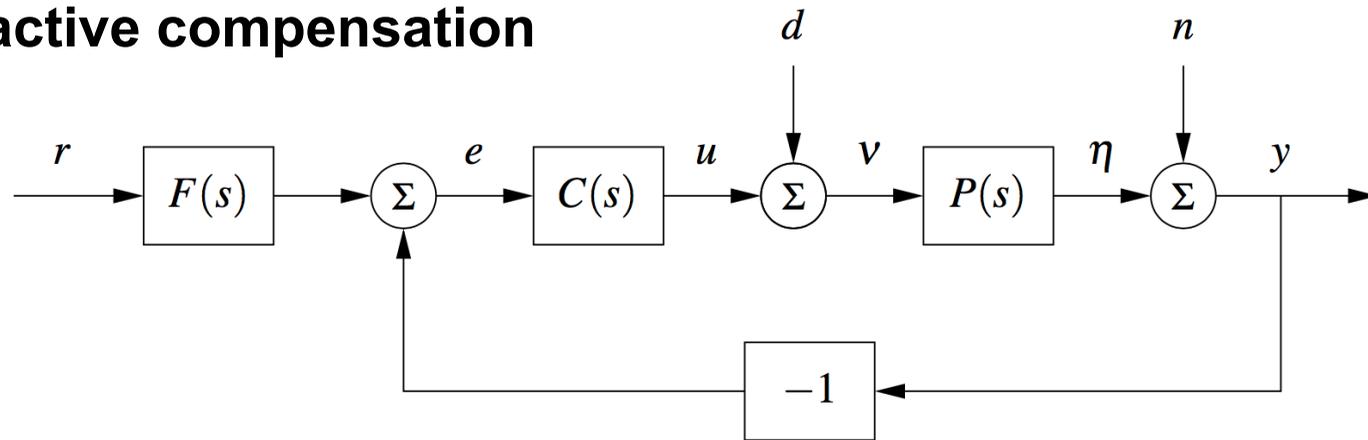
Safety: constraints that the system should never violate

Liveness: conditions that system should satisfy repeatedly



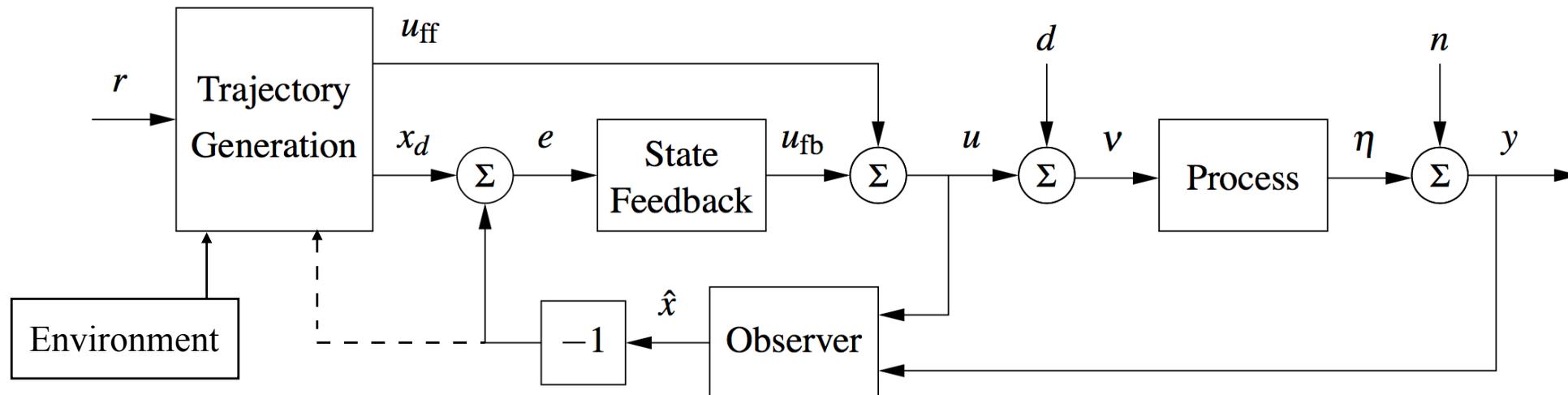
Design Patterns for Control Systems

Reactive compensation



- Reference input shaping
- Feedback on output error
- Compensator dynamics shape closed loop response
- *Uncertainty* in process dynamics + external disturbances and noise
- Goals: stability, performance (tracking), robustness

Predictive compensation

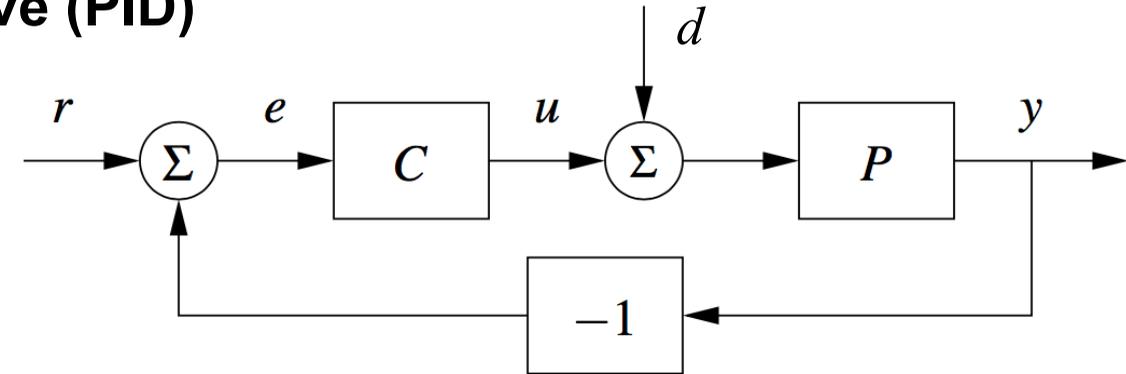


- Explicit computation of trajectories given a model of the process and environment

Feedback Design Tools: PID Control

Three term controller: proportional, integral, derivative (PID)

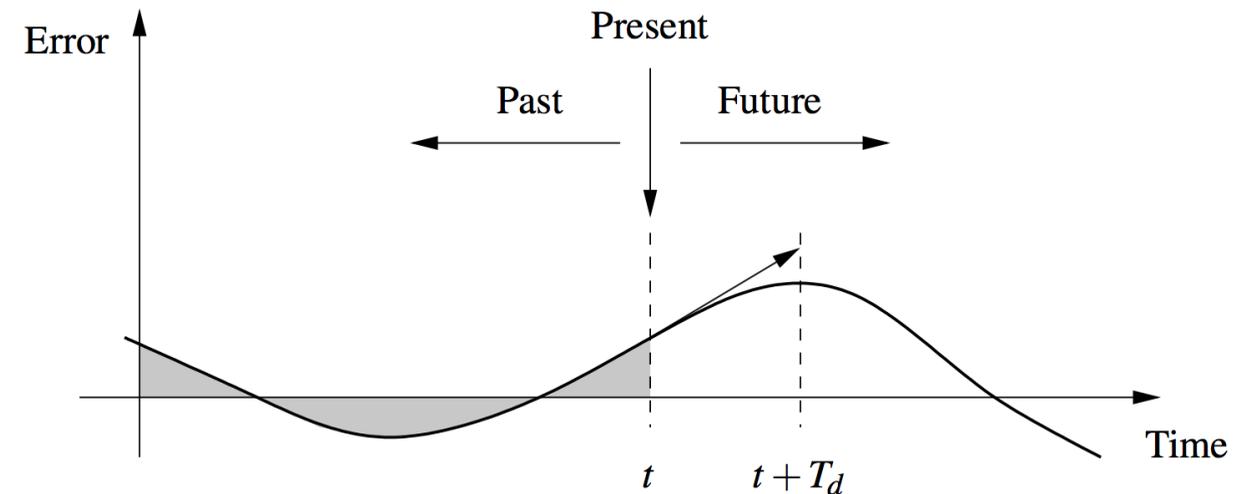
- **Present:** feedback proportional to current error
- **Past:** feedback proportional to integral of past error
 - Insures that error eventual goes to 0
 - Automatically adjusts setpoint of input
- **Future:** derivative of the error
 - Anticipate where we are going



$$u(t) = ke(t) + k_i \int_0^t e(\tau) d\tau + k_d \frac{de(t)}{dt}$$

PID design

- Choose gains k , k_i , k_d to obtain desired behavior
- **Stability:** solutions converge to equilibrium point
- **Performance:**
 - output of system, y , should track reference
 - disturbances d should be attenuated
- **Robustness:** stability and performance properties should hold in face of disturbances & process uncertainty



Feedback Design Tools: State Space Control

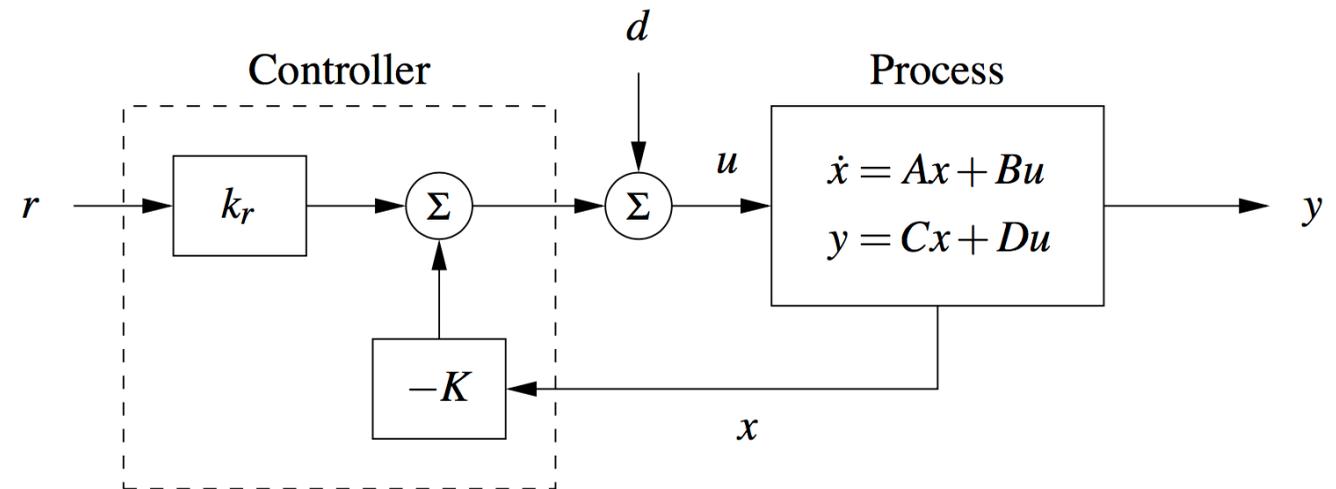
$$\begin{aligned} \dot{x} &= Ax + Bu & x &\in \mathbb{R}^n, x(0) \text{ given} \\ y &= Cx + Du & u &\in \mathbb{R}, y \in \mathbb{R} \end{aligned}$$

$$x(T) = e^{AT} x_0 + \int_{\tau=0}^T e^{A(T-\tau)} Bu(\tau) d\tau$$

Goal: find a linear control law $u = -Kx + k_r r$ such that the closed loop system

$$\dot{x} = Ax + Bu = (A - BK)x + Bk_r r$$

is stable at equilibrium point x_e with $y_e = r$.



Remarks

- If $r = 0$, control law simplifies to $u = -Kx$ and system becomes $\dot{x} = (A - BK)x$
- Stability based on eigenvalues \Rightarrow use K to make eigenvalues of $(A - BK)$ stable
- Can also link eigenvalues to *performance* (eg, initial condition response)
- Q: Can we place the eigenvalues anyplace that we want? A: Yes, if *reachable*

MATLAB/Python: $K = \text{place}(A, B, \text{eigs})$, $K = \text{lqr}(A, B, Q, R)$, ...

Note: this is *design of dynamics*

Feedforward Design Tools: Real-Time Trajectory Generation

Goal: find a feasible trajectory that satisfies dynamics/constraints

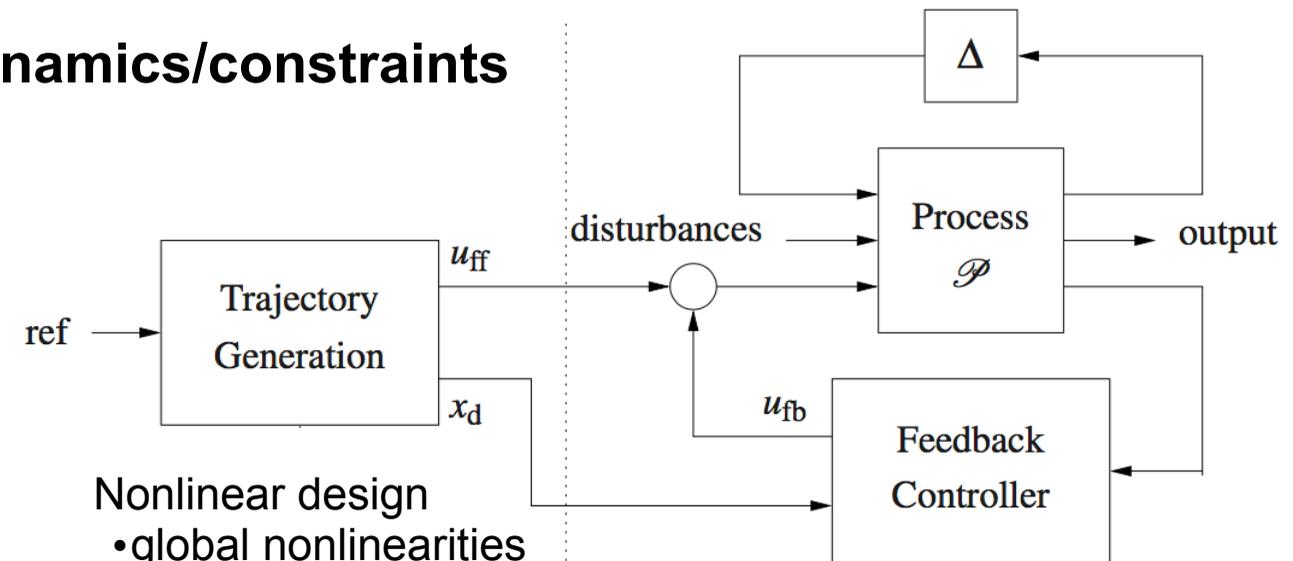
$$\min J = \int_{t_0}^T q(x, u) dt + V(x(T), u(T))$$
$$\dot{x} = f(x, u) \quad lb \leq g(x, u) \leq ub$$

Solve as a constrained optimization problem

- Various tricks to get very fast calculations
- Need to update solutions at the rate at which the reference (task description) is modified

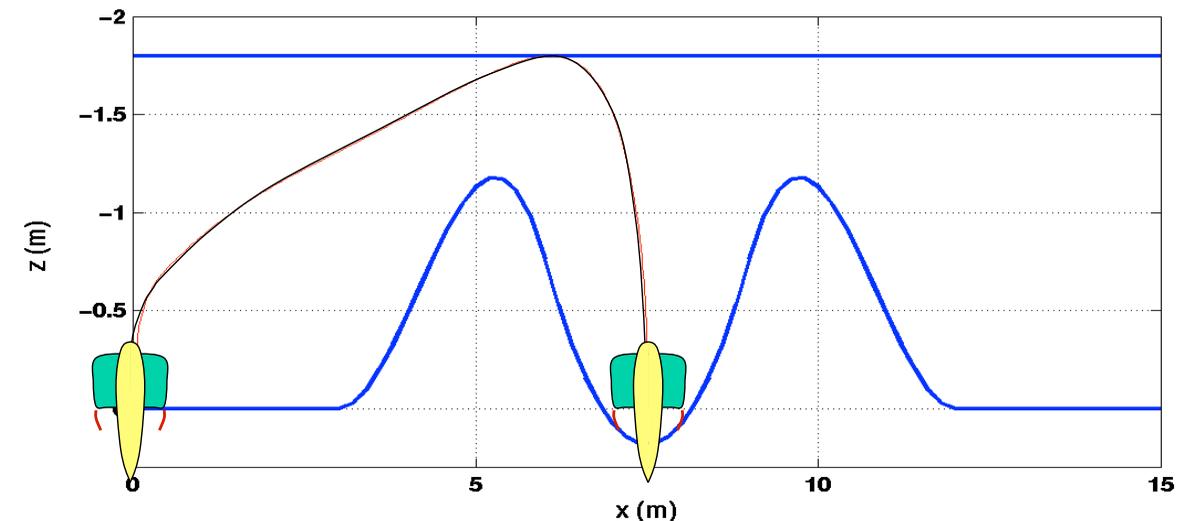
Use feedback (“inner loop”) to track trajectory

- Trajectory generation provides feasible trajectory plus nominal input
- Feedback used to correct for disturbances and model uncertainties
- Example of “two degree of freedom” design



Nonlinear design

- global nonlinearities
- input saturation
- state space constraints



Feedforward Design Tools: Receding Horizon Control

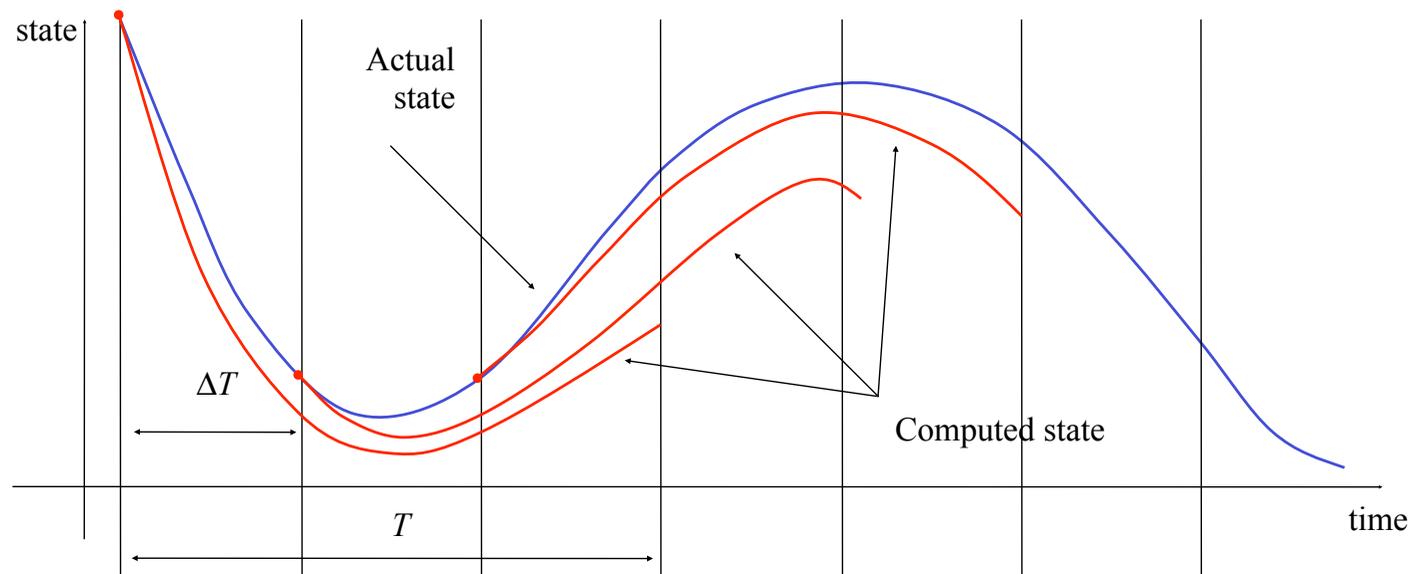
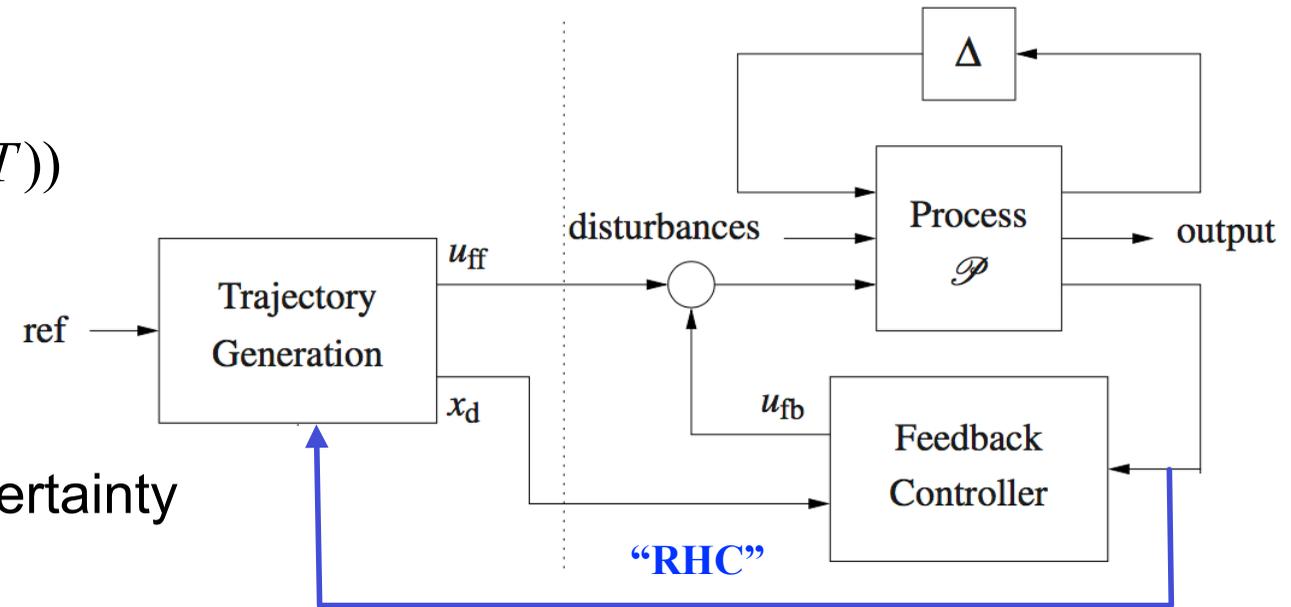
Basic idea: recompute solutions

$$u_{[t,t+\Delta T]} = \arg \min \int_t^{t+T} L(x(\tau), u(\tau)) d\tau + V(x(t+T))$$

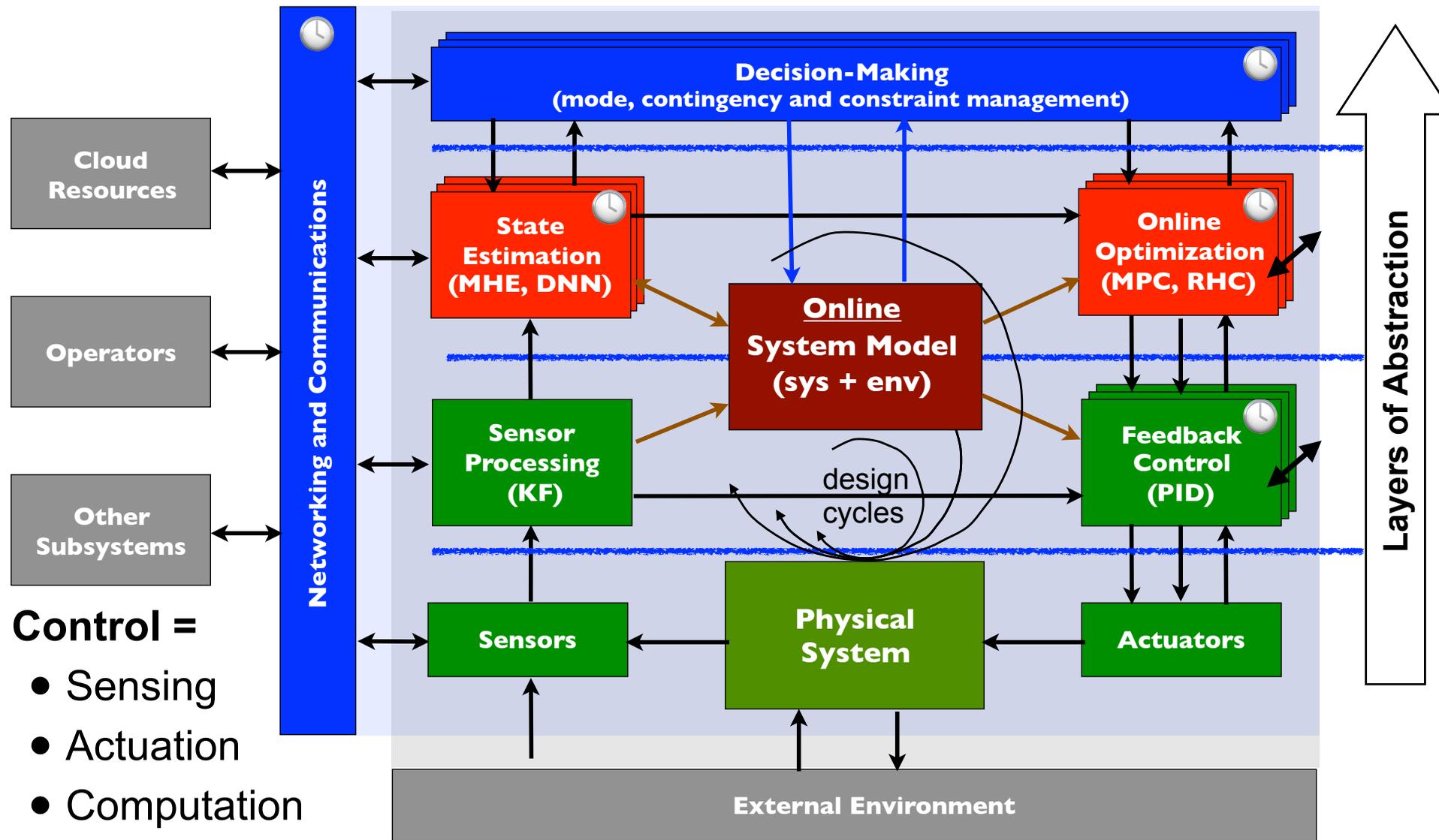
$$x_0 = x(t) \quad x_f = x_d(t+T)$$

$$\dot{x} = f(x, u) \quad g(x, u) \leq 0$$

- Provides second feedback loop to manage uncertainty
- Need to be careful about terminal constraints



Design of Modern (Networked) Control Systems



- Control =**
- Sensing
 - Actuation
 - Computation

Control = dynamics, uncertainty, feedforward, feedback

Examples

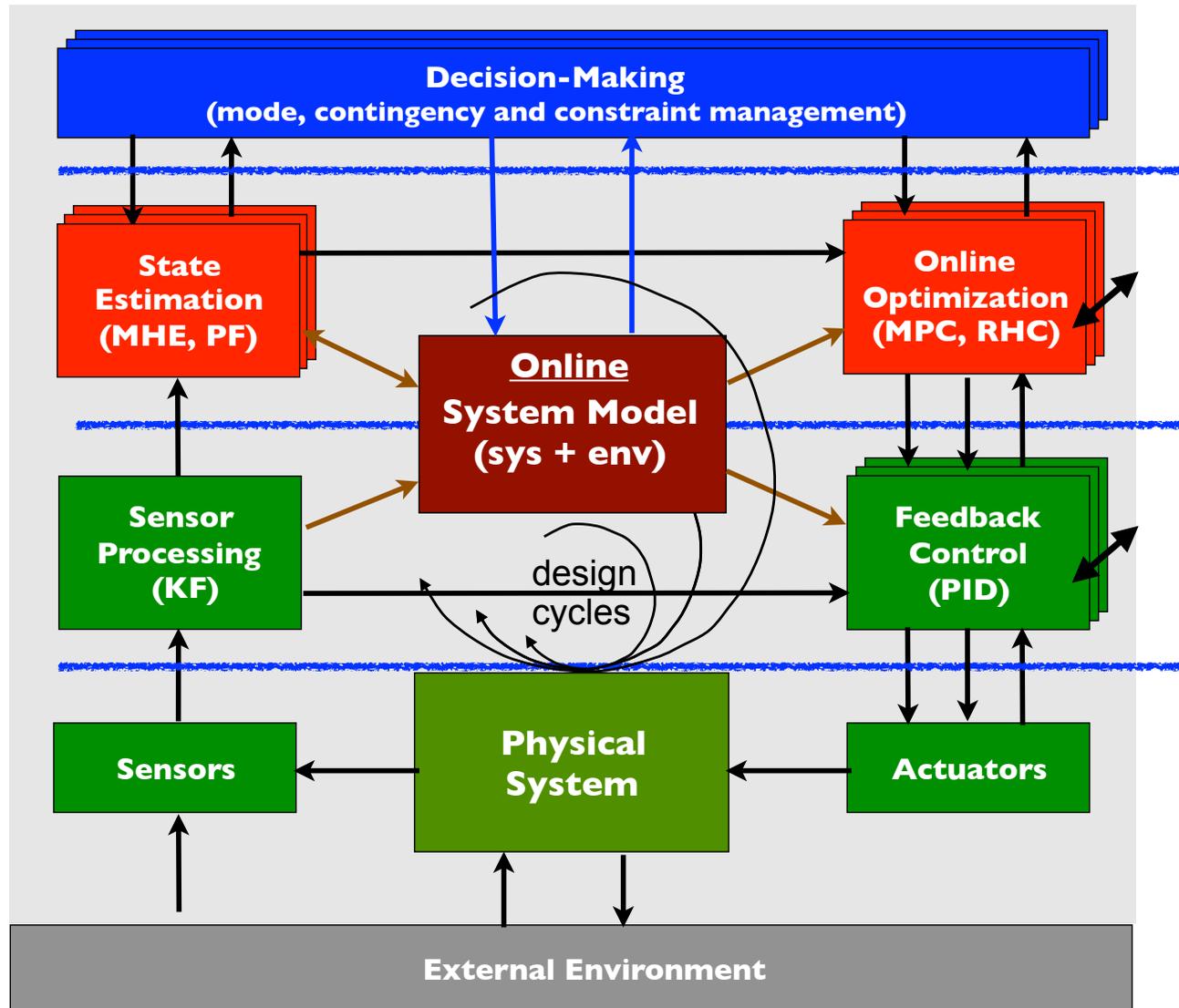
- Aerospace systems
- Self-driving cars
- Factory automation/ process control
- Smart buildings, grid, transportation

Challenges

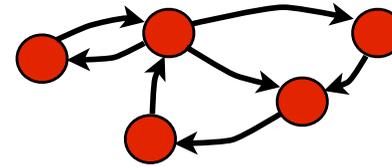
- How do we define the layers/interfaces (vertical contracts)
- How do we scale to *many* devices (horizontal contracts)
- Stability, robustness, security, privacy

Layered Approaches to Design

Multi-layer Networked Control System



Model



$$\dot{x} = f_\alpha(x, u)$$

$$g_\alpha(x, u, z) \leq 0$$

$$y = P_{yu}(s)u + P_{yd}(s)d$$

$$\|W(s)d(s)\| \leq 1$$

$$\dot{x}^i = f_\alpha(x^i, u^i, d^i)$$

$$x \in \mathcal{X}, u \in \mathcal{U}, d \in \mathcal{D}$$

Specs

$$(\phi_{\text{init}} \wedge \square \phi_{\text{env}}) \implies$$

$$(\square \phi_{\text{safe}} \wedge \square \diamond_{\leq T} \phi_{\text{live}})$$

$$\min J = \int_0^T L_\alpha(x, u) dt$$

$$+ V(x(T))$$

$$\|W_1 S + W_2 T\|_\infty < \gamma$$

Operating Envelope
Energy Efficiency
Actuator Authority

Specifying Discrete Behavior Using Temporal Logic

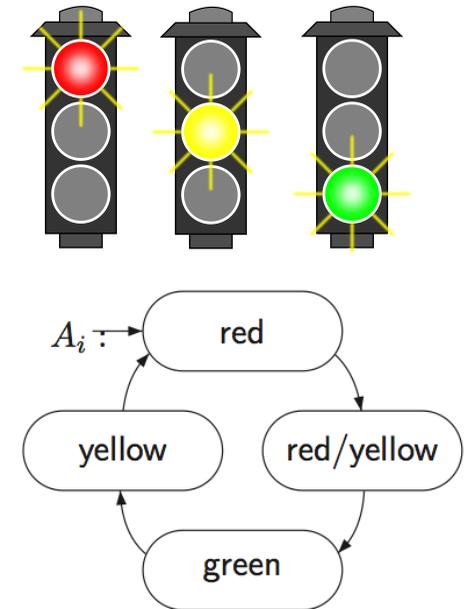
Linear temporal logic (LTL)

- ◇ “eventually” - satisfied at some point in the future
- “always” - satisfied now and forever into the future
- “next” - true at next step

Signal temporal logic (STL)

- Allow predicates that compare values
- Allow temporal bounds

- $p \rightarrow \diamond q$ p implies eventually q (response)
- $p \rightarrow q \text{ U } r$ p implies q until r (precedence)
- $\square \diamond p$ always eventually p (progress)
- $\diamond \square p$ eventually always p (stability)
- $\diamond p \rightarrow \diamond q$ eventually p implies eventually q (correlation)
- $V < V_{\max}$ $V(t)$ less than threshold (V_{\max})
- $\square_{[t_1, t_2]} p$ p true for all time in $[t_1, t_2]$
- $p \rightarrow \diamond_{[0, t]} q$ if p occurs, q will occur w/in time t



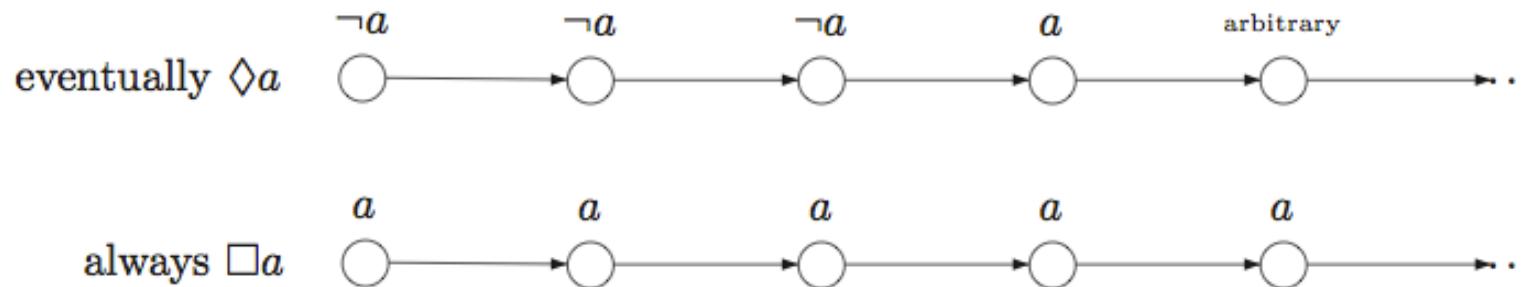
Baier and Katoen, *Principles of Model Checking*, 2007

$\square \diamond \text{green}$

$\square (\text{green} \rightarrow \neg \bigcirc \text{red})$

$\square (\text{red} \rightarrow (\diamond \text{green}$

$\wedge (\neg \text{green} \text{ U } \text{yellow}))$)



Baier and Katoen, *Principles of Model Checking*, 2007

Synthesis of Reactive Controllers

Reactive Protocol Synthesis

- Find control action that insures that specification is always satisfied
- For LTL, complexity is doubly exponential (!) in the size of system specification

GR(1) synthesis for reactive protocols

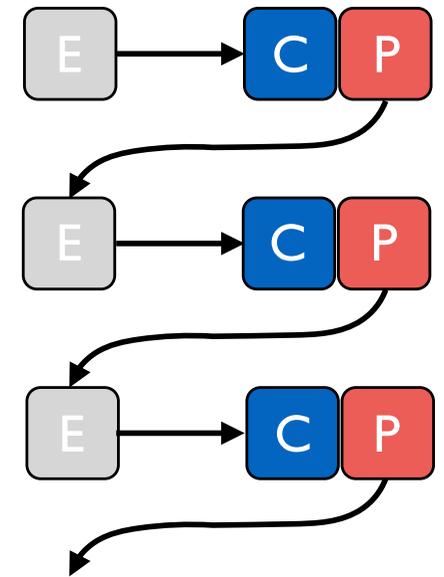
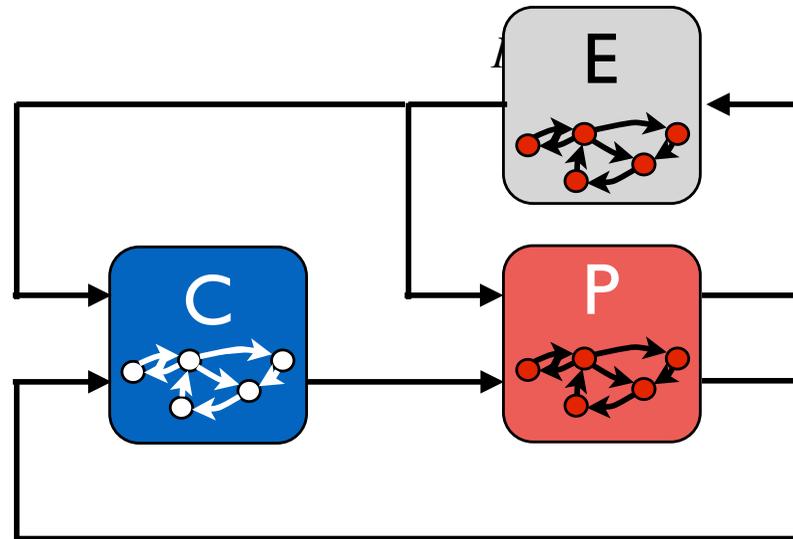
- Piterman, Pnueli and Sa'ar, 2006
- Assume environment fixes action before controller (breaks symmetry)
- For certain class of specifications, get complexity cubic in # of states (!)

$$(\phi_{\text{init}}^e \wedge \square \phi_{\text{safe}}^e \wedge \square \diamond \phi_{\text{prog}}^e) \rightarrow (\phi_{\text{init}}^s \wedge \square \phi_{\text{safe}}^s \wedge \square \diamond \phi_{\text{prog}}^s)$$

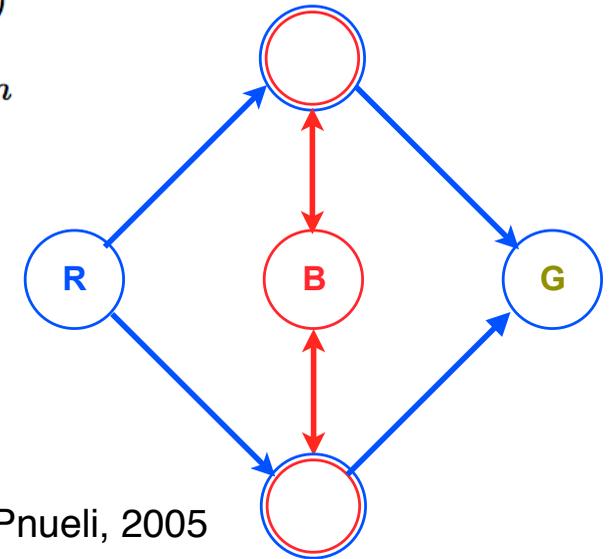
Environment assumption

System guarantee

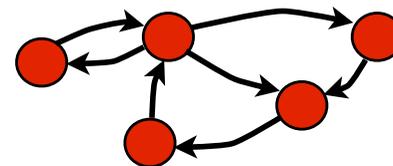
- GR(1) = general reactivity formula
- Assume/guarantee style specification



$$\begin{aligned} \mathcal{L}_m(S/G) &= \mathcal{L}(S/G) \cap \mathcal{L}_m(G) \\ &\subseteq \overline{L_{am}} \cap \mathcal{L}_m(G) = L_{am} \end{aligned}$$



A. Pnueli, 2005



Summary: Feedback Control Theory

Two main principles of (feedback) control theory

- Feedback is a tool to **provide robustness to uncertainty**
 - Uncertainty = noise, disturbances, unmodeled dynamics
 - Useful for modularity: consistent behavior of subsystems
- Feedback is a tool to **design the dynamics of a system**
 - Convert unstable systems to stable systems
 - Tune the performance of a system to meet specifications
- Combined, these principles **enable modularity and hierarchy**

Control theory: past, present and future

- Tools originally developed to design low-level control systems
- Increasing application to networked (hybrid) control systems
- New challenges: systematic design of layered architectures and control protocols, security and privacy, data-driven (AI/ML)

More information

- *Feedback Systems* (free download): <https://fbsbook.org>

