

A systematic procedure for economic plantwide control

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Outline

- Our paradigm based on time scale separation
- Plantwide control procedure based on economics
- Example: Runner
- Selection of primary controlled variables ($CV_1 = H y$)
 - Optimal is gradient, $CV_1 = J_u$ with setpoint=0
 - General $CV_1 = Hy$. Nullspace and exact local method
- Throughput manipulator (TPM) location
- Examples
- Conclusion

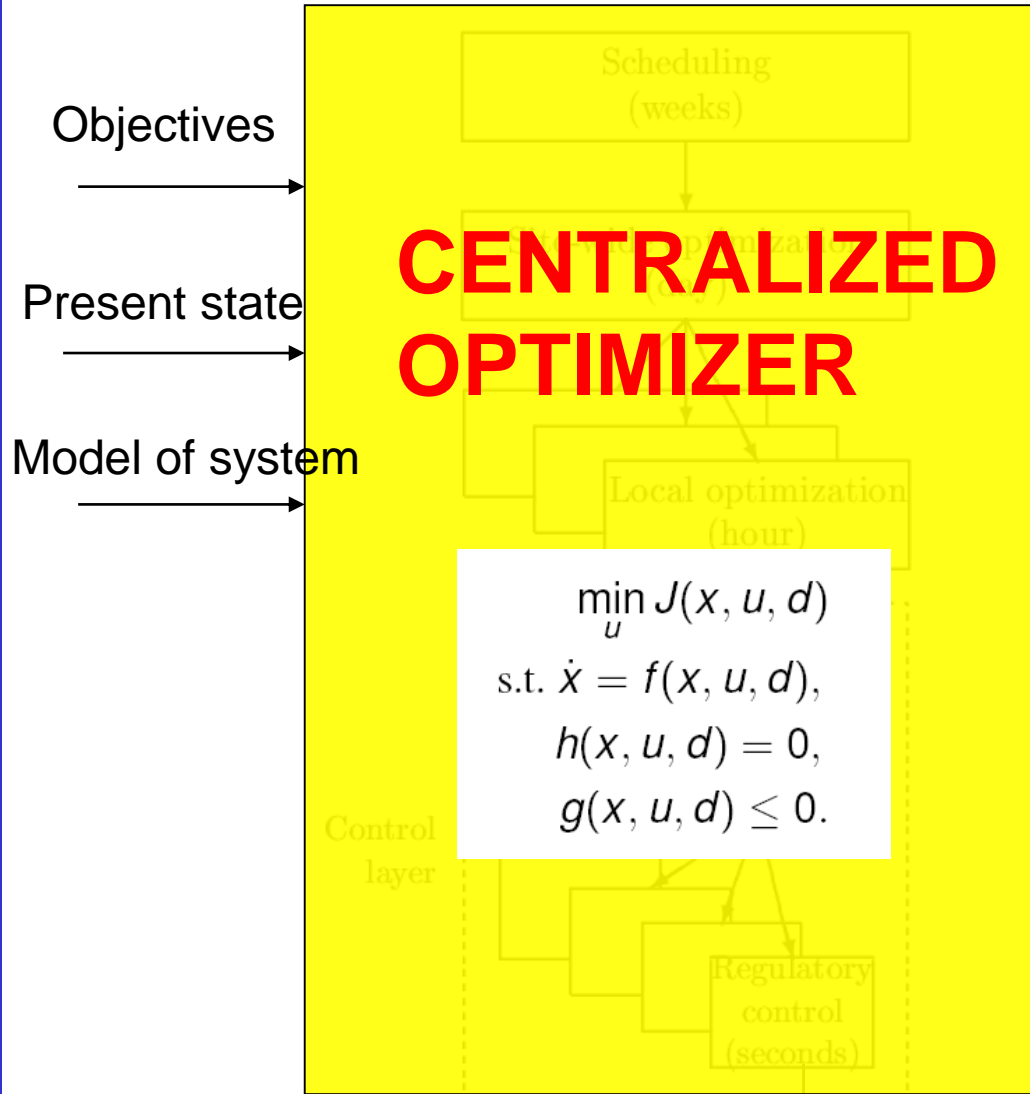
**This is
the truth
and
the only truth**



How we design a control system for a complete chemical plant?

- Where do we start?
- What should we control? and why?
- etc.
- etc.

In theory: Optimal control and operation



Approach:

- Model of overall system
- Estimate present state
- Optimize all degrees of freedom

Process control:

- Excellent candidate for centralized control

Problems:

- Model not available
- Objectives = ?
- Optimization complex
- Not robust (difficult to handle uncertainty)
- Slow response time

(Physical) Degrees of freedom

Practice: Engineering systems

- Most (all?) large-scale engineering systems are controlled using hierarchies of quite simple controllers
 - Large-scale chemical plant (refinery)
 - Commercial aircraft
- 100's of loops
- Simple components:
 - PI-control + selectors + cascade + nonlinear fixes + some feedforward

Same in biological systems

But: Not well understood

- Alan Foss (“Critique of chemical process control theory”, AIChE Journal, 1973):

*The central issue to be resolved ... is the determination of control system structure. **Which variables should be measured, which inputs should be manipulated and which links should be made between the two sets?***

There is more than a suspicion that the work of a genius is needed here, for without it the control configuration problem will likely remain in a primitive, hazily stated and wholly unmanageable form. The gap is present indeed, but contrary to the views of many, it is the theoretician who must close it.

Previous work on plantwide control:

- Page Buckley (1964) - *Chapter on “Overall process control”* (still industrial practice)
- Greg Shinskey (1967) – *process control systems*
- Alan Foss (1973) - *control system structure*
- Bill Luyben et al. (1975-) – *case studies ; “snowball effect”*
- George Stephanopoulos and Manfred Morari (1980) – *synthesis of control structures for chemical processes*
- Ruel Shinnar (1981-) - *“dominant variables”*
- Jim Downs (1991) - *Tennessee Eastman challenge problem*
- Larsson and Skogestad (2000): *Review of plantwide control*

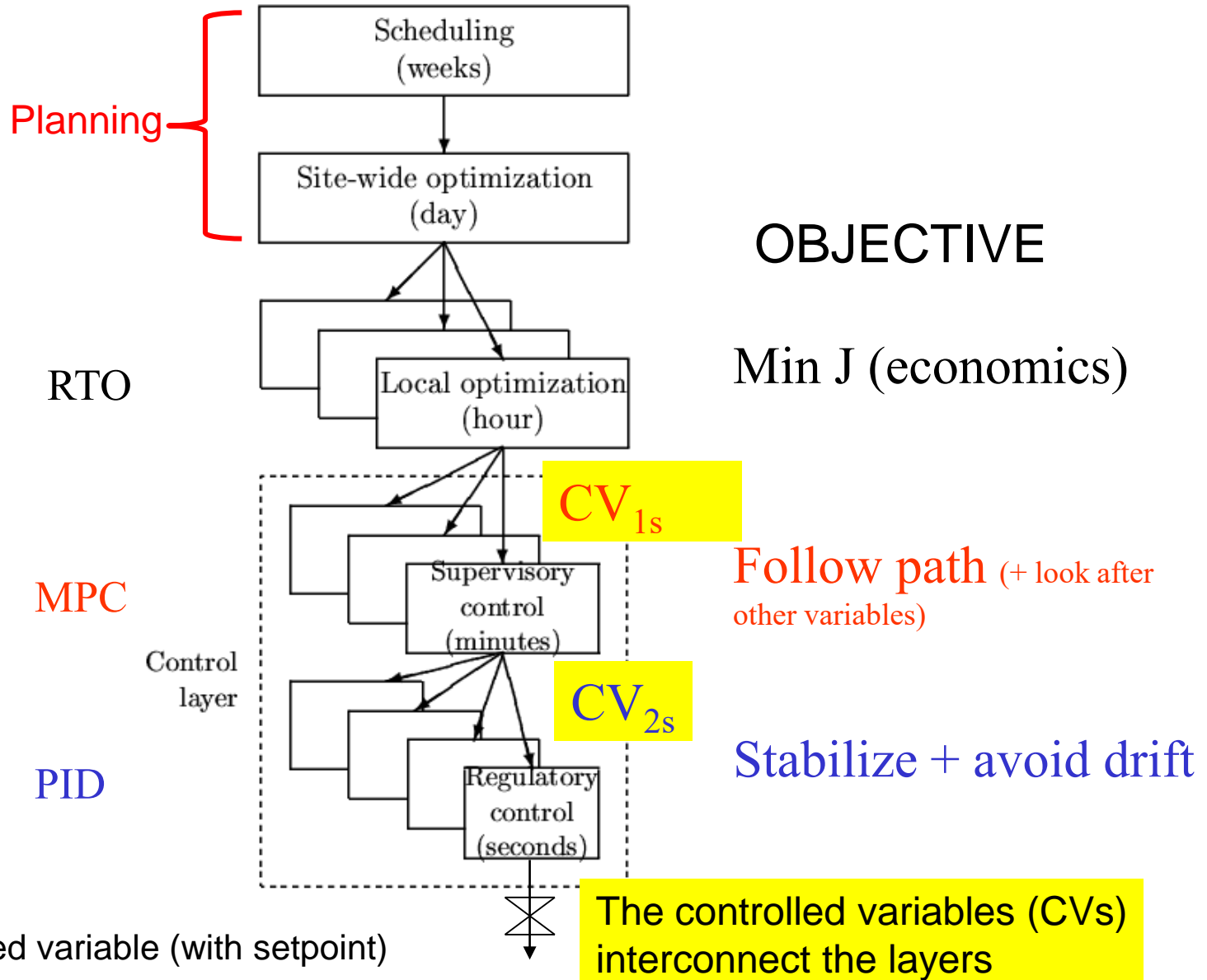
Main objectives control system

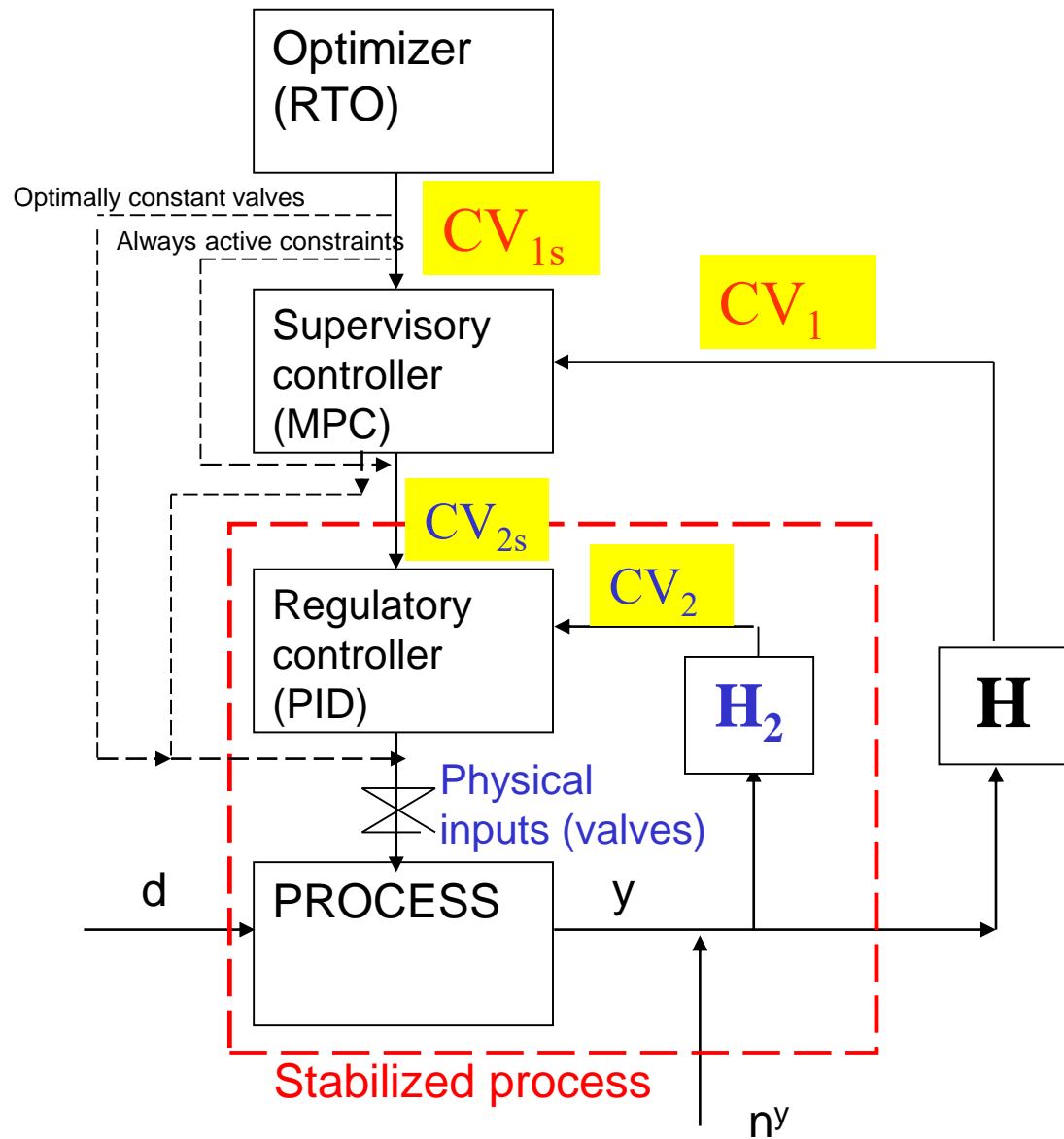
1. **Economics:** Implementation of acceptable (near-optimal) operation
2. **Regulation:** Stable operation

ARE THESE OBJECTIVES CONFLICTING?

- **Usually NOT**
 - Different time scales
 - Stabilization fast time scale
 - Stabilization doesn't "use up" any degrees of freedom
 - Reference value (setpoint) available for layer above
 - But it "uses up" part of the time window (frequency range)

Practical operation: Hierarchical structure





Degrees of freedom for optimization (usually steady-state DOFs), $MV_{opt} = CV_{1s}$

Degrees of freedom for supervisory control, $MV_1 = CV_{2s} + \text{unused valves}$

Physical degrees of freedom for stabilizing control, $MV_2 = \text{valves (dynamic process inputs)}$

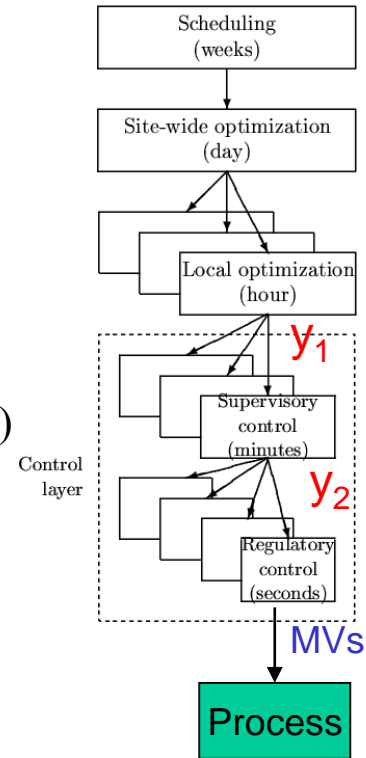
Control structure design procedure

I Top Down (mainly steady-state economics, y_1)

- **Step 1:** Define operational objectives (optimal operation)
 - Cost function J (to be minimized)
 - Operational constraints
- **Step 2:** Identify degrees of freedom (MVs) and optimize for expected disturbances
 - Identify **Active constraints**
- **Step 3:** Select primary “economic” controlled variables $c=y_1$ (CV1s)
 - **Self-optimizing variables (find H)**
- **Step 4:** Where locate the **throughput manipulator (TPM)?**

II Bottom Up (dynamics, y_2)

- **Step 5:** Regulatory / stabilizing control (PID layer)
 - What more to control (y_2 ; local CV2s)? **Find H_2**
 - Pairing of inputs and outputs
- **Step 6:** Supervisory control (MPC layer)
- **Step 7:** Real-time optimization (Do we need it?)



Step 1. Define optimal operation (economics)

- What are we going to use our degrees of freedom u (MVs) for?
- Define scalar cost function $J(u,x,d)$
 - u : degrees of freedom (usually steady-state)
 - d : disturbances
 - x : states (internal variables)

Typical cost function:

$$J = \text{cost feed} + \text{cost energy} - \text{value products}$$

- Optimize operation with respect to u for given d (usually steady-state):

$$\min_u J(u,x,d)$$

subject to:

$$\text{Model equations:} \quad f(u,x,d) = 0$$

$$\text{Operational constraints:} \quad g(u,x,d) < 0$$

Step S2. Optimize

(a) Identify degrees of freedom

(b) Optimize for expected disturbances

- Need good model, usually steady-state
- Optimization is time consuming! But it is offline
- Main goal: Identify ACTIVE CONSTRAINTS
- A good engineer can often guess the active constraints

Step S3: Implementation of optimal operation

- Have found the optimal way of operation.
How should it be implemented?
- **What to control ?** (CV_1).
 1. Active constraints
 2. Self-optimizing variables (for unconstrained degrees of freedom)

Optimal operation of runner

- Cost to be minimized, $J=T$
- One degree of freedom ($u=\text{power}$)
- What should we control?



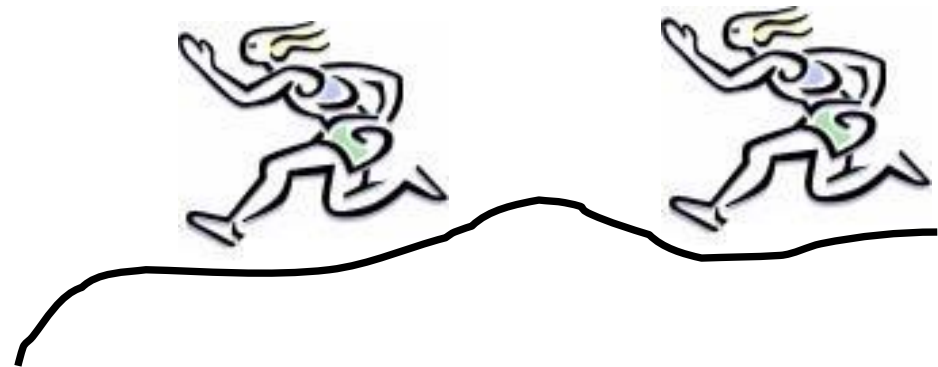
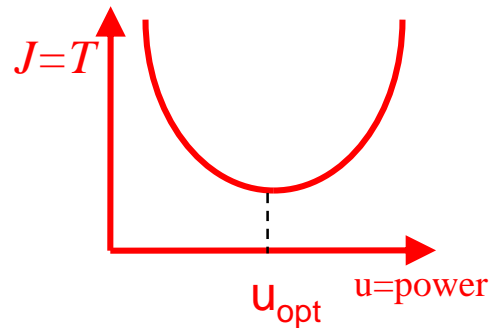
1. Optimal operation of Sprinter

- 100m. $J=T$
- **Active constraint control:**
 - Maximum speed (“no thinking required”)
 - $CV = \text{power (at max)}$



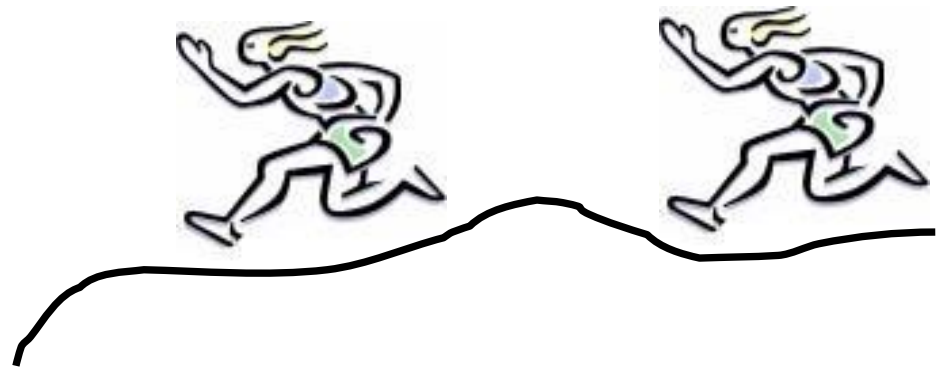
2. Optimal operation of Marathon runner

- 40 km. $J=T$
- What should we control? $CV=?$
- **Unconstrained optimum**

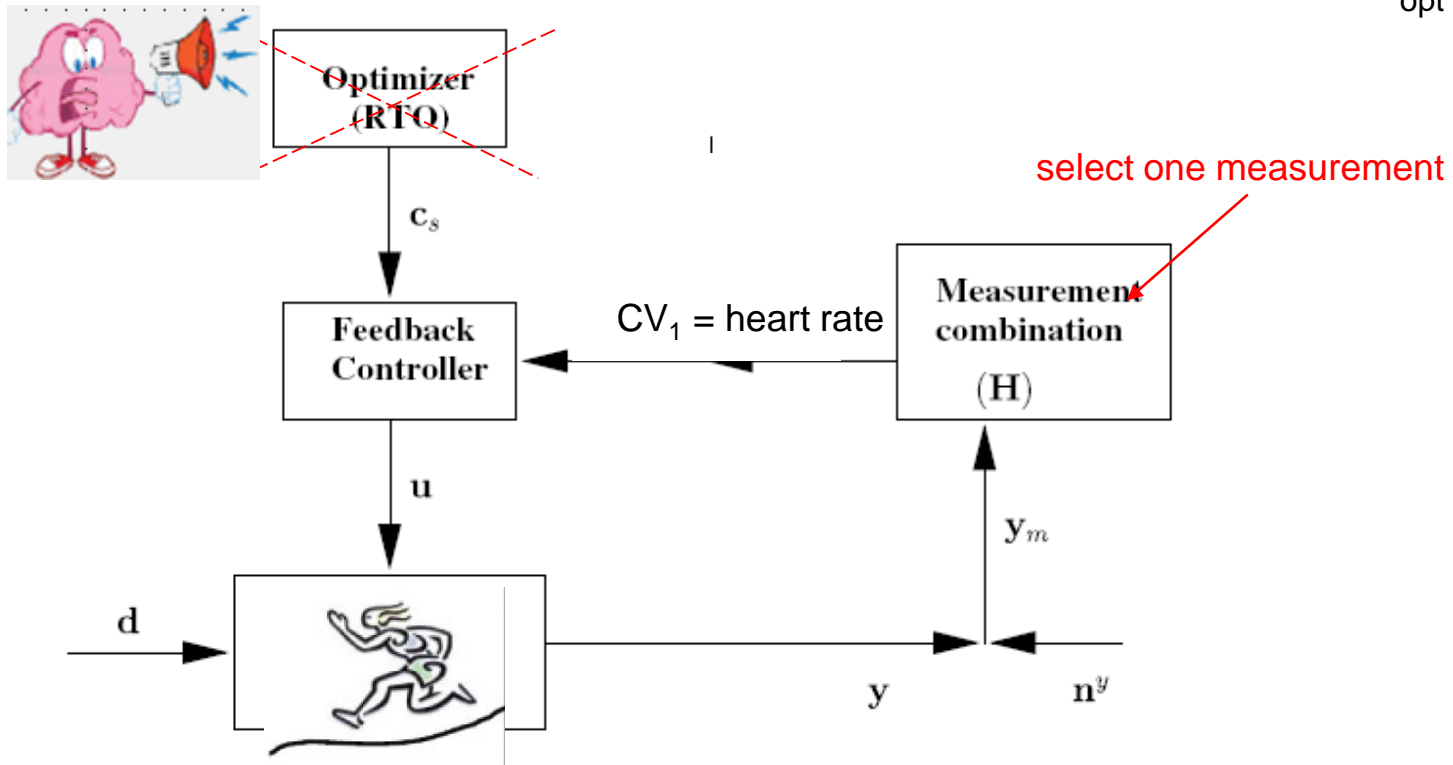
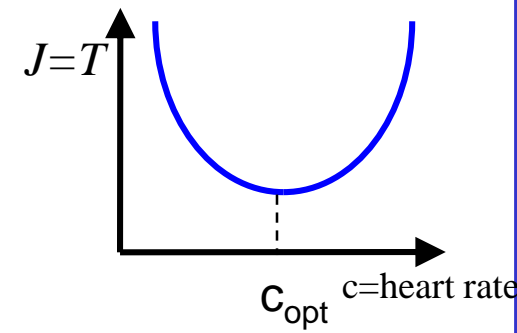


Self-optimizing control: Marathon (40 km)

- Any **self-optimizing variable** (to control at constant setpoint)?
 - c_1 = distance to leader of race
 - c_2 = speed
 - c_3 = heart rate
 - c_4 = level of lactate in muscles



Conclusion Marathon runner



- CV = heart rate is good “self-optimizing” variable
- Simple and robust implementation
- Disturbances are indirectly handled by keeping a constant heart rate
- May have infrequent adjustment of setpoint (c_s)

Summary Step 3.

What should we control (CV_1)?

Selection of primary controlled variables $c = CV_1$

- 1. Control active constraints!**
- 2. Unconstrained variables: Control self-optimizing variables!**

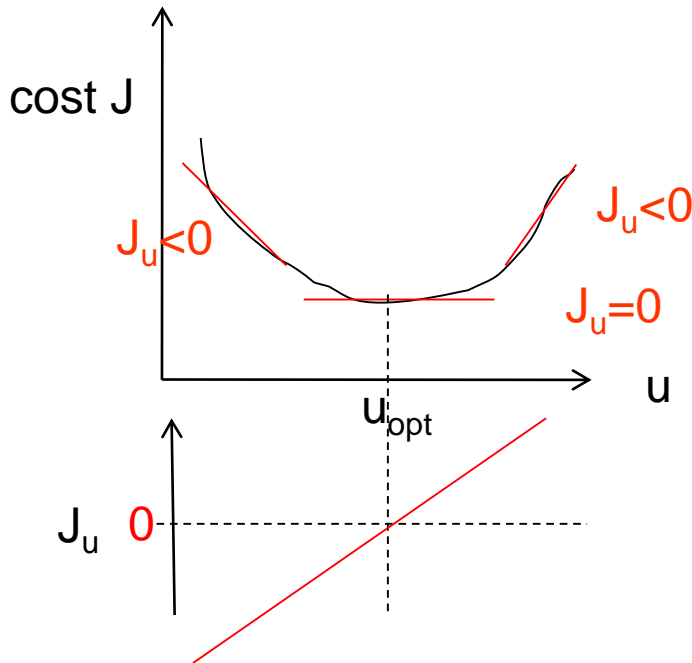
- Old idea (Morari *et al.*, 1980):

“We want to find a function c of the process variables which when held constant, leads automatically to the optimal adjustments of the manipulated variables, and with it, the optimal operating conditions.”

The ideal “self-optimizing” variable is the gradient, J_u

$$\mathbf{c} = \partial J / \partial \mathbf{u} = J_u$$

- Keep gradient at zero for all disturbances ($\mathbf{c} = J_u = 0$)
- Problem: Usually no measurement of gradient



$$CV_1 = Hy$$

Nullspace method for H (Alstad):

$$HF=0 \text{ where } F=dy_{opt}/dd$$

$$J_u(u, d) = \underbrace{J_u(u_{opt}(d), d)}_{=0} + J_{uu} \cdot (u - u_{opt})$$

$$u - u_{opt} = (HG^y)^{-1}(c - c_{opt})$$

$$\text{Here: } c - c_{opt} = \Delta c - \Delta c_{opt}$$

where we have introduced deviation variables around a nominal optimal point (c^*, d^*) (where $c^* = c_{opt}(d^*)$)

Assume perfect control of c (no noise): $\Delta c = 0$

$$\text{Optimal change: } \Delta c_{opt} = H \Delta y_{opt} = HF \Delta d$$

$$\text{Gives: } J_u = -J_{uu}(HG^y)^{-1}HF \Delta d$$

$\Rightarrow HF = 0$ gives $J_u = 0$ for any disturbance Δd

“Exact local method”

$$\min_H \left\| \underbrace{J_{uu}^{1/2} (HG^y)^{-1}}_{\text{“Minimize” in Maximum gain rule (maximize } S_1 G J_{uu}^{-1/2}, G=HG^y)} \underbrace{H [FW_d \quad W_{ny}]}_{\text{“Scaling” } S_1} \right\|_2$$

“=0” in nullspace method (no noise)

Analytical solution:

$$H = G^y T (Y Y^T)^{-1} \text{ where } Y = [FW_d \quad W_{ny}]$$

- No measurement error: HF=0 (nullspace method)
- With measurement error: Minimize GF^c
- Maximum gain rule

Example. Nullspace Method for Marathon runner

u = power, d = slope [degrees]

y_1 = hr [beat/min], y_2 = v [m/s]

$$F = dy_{\text{opt}}/dd = [0.25 \quad -0.2]'$$

$$H = [h_1 \quad h_2]$$

$$\mathbf{HF} = \mathbf{0} \rightarrow h_1 f_1 + h_2 f_2 = 0.25 h_1 - 0.2 h_2 = 0$$

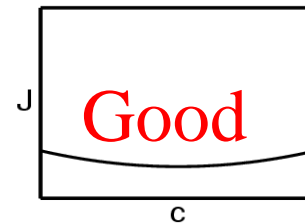
$$\text{Choose } h_1 = 1 \rightarrow h_2 = 0.25/0.2 = 1.25$$

Conclusion: $\mathbf{c} = \mathbf{hr} + 1.25 \mathbf{v}$

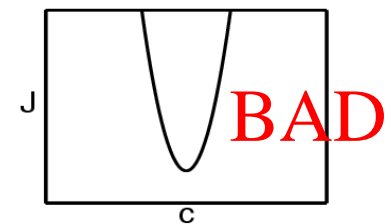
Control $\mathbf{c} = \mathbf{constant}$ \rightarrow hr increases when v decreases (OK uphill!)

In practice: What variable $c=Hy$ should we control? (for self-optimizing control)

1. **The *optimal value* of c should be *insensitive* to disturbances**
 - Small $HF = dc_{opt}/dd$
2. **c should be easy to measure and control**
3. **The *value* of c should be *sensitive* to the inputs (“maximum gain rule”)**
 - Large $G = HG^y = dc/du$
 - Equivalent: Want flat optimum



(b) Flat optimum: Implementation easy



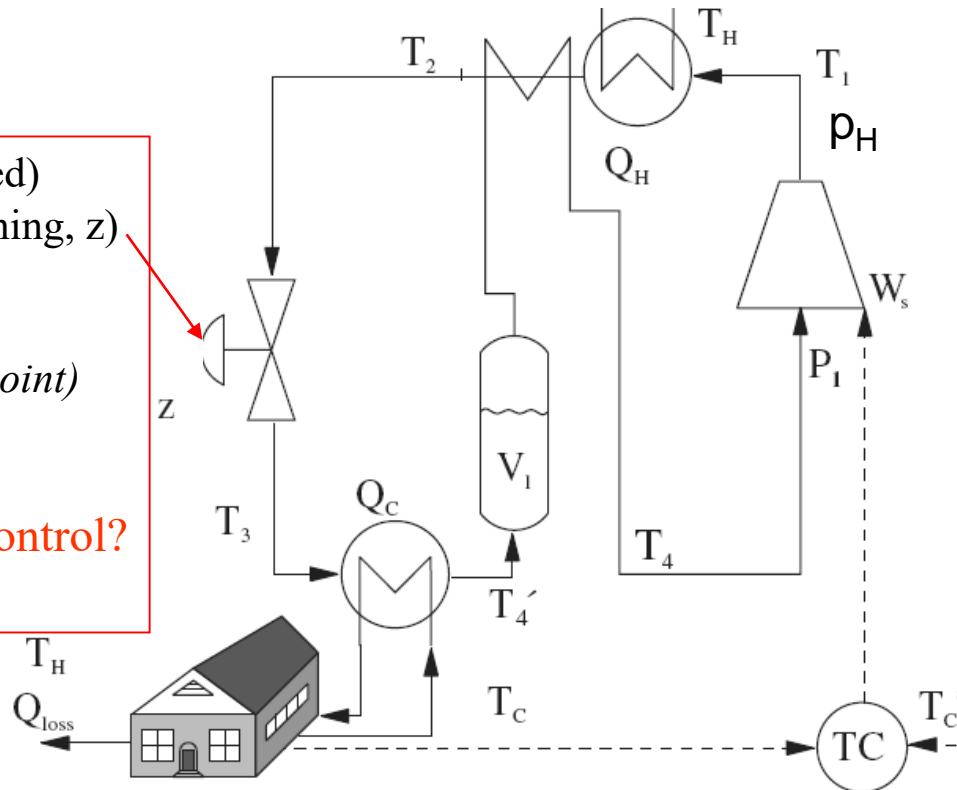
(c) Sharp optimum: Sensitive to implementation errors

Note: Must also find optimal setpoint for $c=CV_1$

Example: CO2 refrigeration cycle

$J = W_s$ (work supplied)
 DOF = u (valve opening, z)
 Main disturbances:
 $d_1 = T_H$
 $d_2 = T_{Cs}$ (setpoint)
 $d_3 = UA_{loss}$

What should we control?



CO2 refrigeration cycle

Step 1. One (remaining) degree of freedom ($u=z$)

Step 2. Objective function. $J = W_s$ (compressor work)

Step 3. Optimize operation for disturbances ($d_1=T_C$, $d_2=T_H$, $d_3=UA$)

- Optimum always unconstrained

Step 4. Implementation of optimal operation

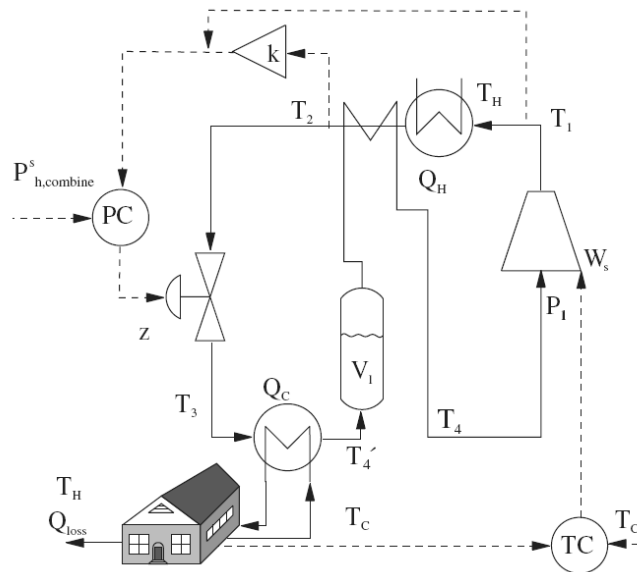
- No good single measurements (all give large losses):
 - p_h, T_h, z, \dots
- Nullspace method: Need to combine $n_u+n_d=1+3=4$ measurements to have zero disturbance loss
- Simpler: Try combining two measurements. Exact local method:
 - $c = h_1 p_h + h_2 T_h = p_h + k T_h; \quad k = -8.53 \text{ bar/K}$
- Nonlinear evaluation of loss: OK!

CO2 cycle: Maximum gain rule

Linear "maximum gain" analysis of controlled variables for CO₂ case

Variable (y)	Nom.	G = $\frac{\Delta y}{\Delta u}$	$\Delta y_{opt}(d_i)$			Δy_{opt}	n	Span y Δy_{opt} + n	G' = $\frac{ G }{span y}$
			d ₁ (T _H)	d ₂ (T _C)	d ₃ (UA _{loss})				
P _h /T ₂ ' (bar °C ⁻¹)	0.32	-0.291	0.140	-0.047	0.093	0.174	0.0033	0.177	0.25
P _h (bar)	97.61	-78.85	48.3	-15.5	31.0	59.4	1.0	60.4	1.31
T ₂ ' (°C)	35.5	36.7	16.27	-2.93	7.64	18.21	1	19.2	1.91
T ₂ ' - T _H (°C)	3.62	24	4.10	-1.92	5.00	6.75	1.5	8.25	2.91
z	0.34	1	0.15	-0.04	0.18	0.24	0.05	0.29	3.45
V ₁ (m ³)	0.07	0.03	-0.02	0.005	-0.03	0.006	0.001	0.007	4.77
T ₂ (°C)	25.5	60.14	8.37	0.90	3.18	9.00	1	10.0	6.02
P _{h,combine} (bar)	97.61	-592.0	-23.1	-23.1	3.91	33.0	9.53	42.5	13.9
m _{gco} (kg)	4.83	-11.18	0.151	-0.136	0.119	0.235	0.44	0.675	16.55

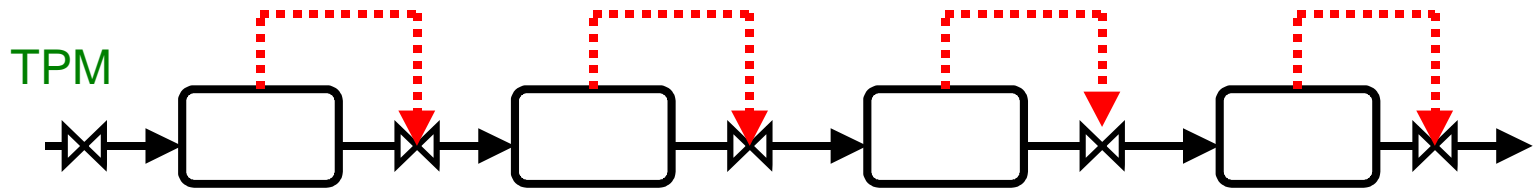
Nullspace method: $c = p_{h,combine} = h_1 p_h + h_2 T_2 = p_h + k T_2$; $k = -8.53$ bar/K



Step 4. Where set production rate?

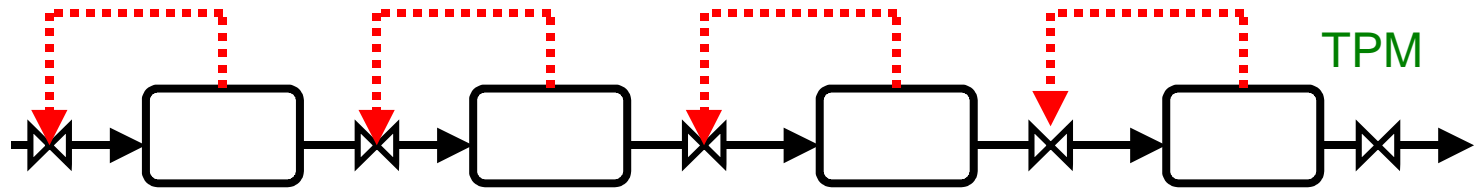
- Where locale the **TPM (throughput manipulator)?**
 - The ”gas pedal” of the process
- Very important!
- Determines structure of remaining inventory (level) control system
- Set production rate at (dynamic) bottleneck
- Link between **Top-down** and **Bottom-up** parts
- **NOTE: TPM location is a dynamic issue.**
 - Link to economics is to improve control of active constraints (reduce backoff)

Production rate set at inlet :
Inventory control in direction of flow*

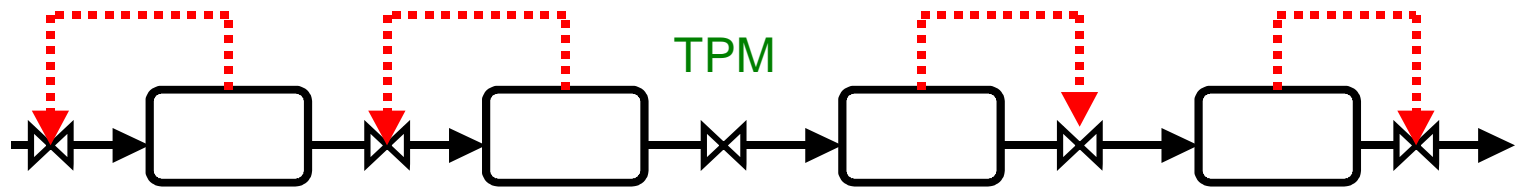


* Required to get “local-consistent” inventory control

Production rate set at outlet: Inventory control opposite flow

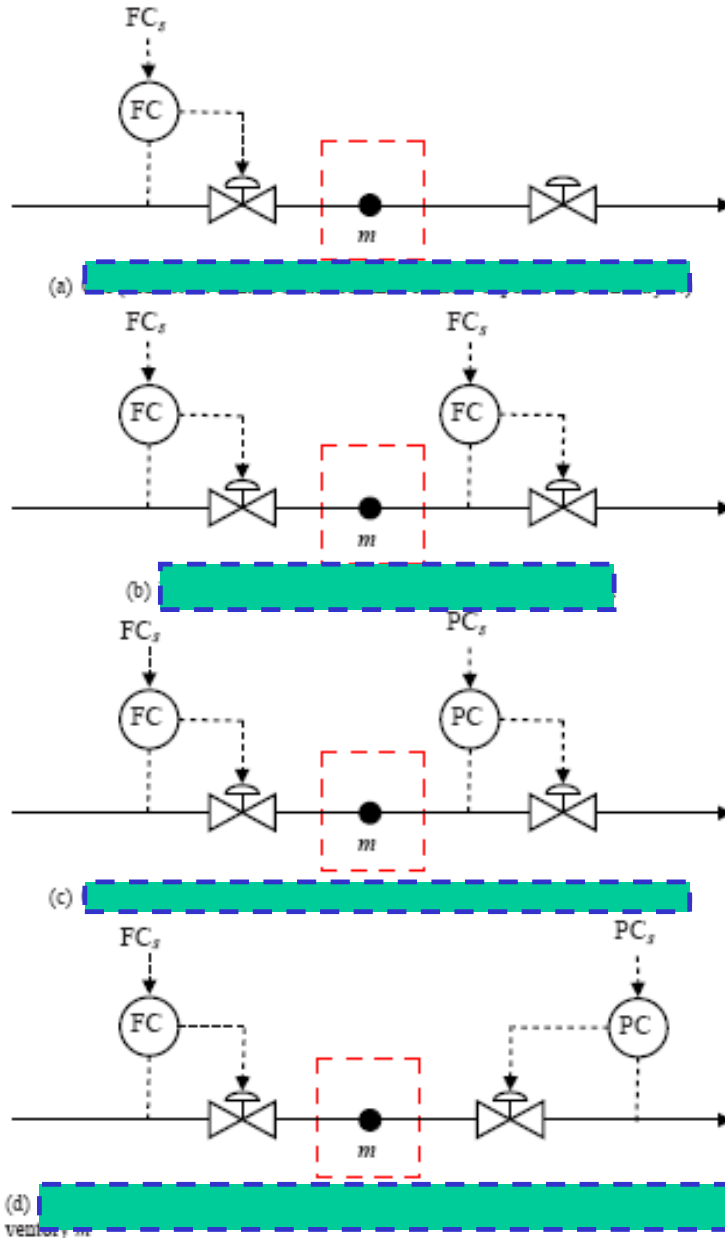


Production rate set inside process



General: “Need radiating inventory control around TPM” (Georgakis)

CONSISTENT?



Flow rate: $q = C_v f(z) \sqrt{\frac{p_1 - p_2}{\rho}}$ [m³/s]

Academic process control community fish pond

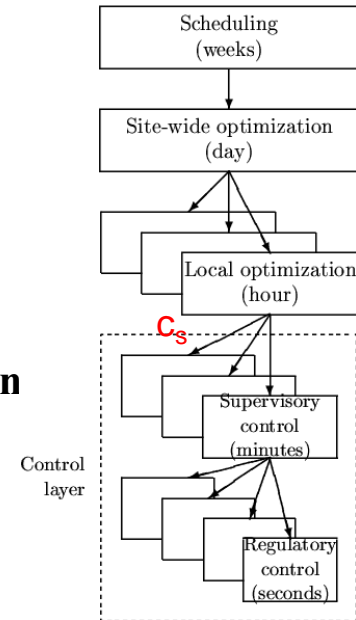
Optimal centralized Solution (EMPC)



Conclusion:

Systematic procedure for plantwide control

- **Start “top-down” with economics:**
 - **Step 1:** Define operational objectives and identify degrees of freedom
 - **Step 2:** Optimize steady-state operation.
 - **Step 3A:** Identify active constraints = primary CVs **c**.
 - **Step 3B:** Remaining unconstrained DOFs: Self-optimizing CVs **c**.
 - **Step 4:** Where to set the throughput (usually: feed)
- **Regulatory control I: Decide on how to move mass through the plan**
 - **Step 5A:** Propose “local-consistent” inventory (level) control structure.
- **Regulatory control II: “Bottom-up” stabilization of the plant**
 - **Step 5B:** Control variables to stop “drift” (sensitive temperatures, pressures,)
 - Pair variables to avoid interaction and saturation
- **Finally: make link between “top-down” and “bottom up”.**
 - **Step 6:** “Advanced/supervisory control” system (MPC):
 - CVs: Active constraints and self-optimizing economic variables +
 - look after variables in layer below (e.g., avoid saturation)
 - MVs: Setpoints to regulatory control layer.
 - Coordinates within units and possibly between units



<http://www.nt.ntnu.no/users/skoge/plantwide>

Summary and references

- The following paper summarizes the procedure:
 - S. Skogestad, "Control structure design for complete chemical plants", *Computers and Chemical Engineering*, **28** (1-2), 219-234 (2004).
- There are many approaches to plantwide control as discussed in the following review paper:
 - T. Larsson and S. Skogestad, "Plantwide control: A review and a new design procedure" *Modeling, Identification and Control*, **21**, 209-240 (2000).
- The following paper updates the procedure:
 - S. Skogestad, "Economic plantwide control", Book chapter in V. Kariwala and V.P. Rangaiah (Eds), *Plant-Wide Control: Recent Developments and Applications*, Wiley (2012).
- Another paper:
 - S. Skogestad "Plantwide control: the search for the self-optimizing control structure", *J. Proc. Control*, 10, 487-507 (2000).
- More information:

<http://www.nt.ntnu.no/users/skoge/plantwide>