

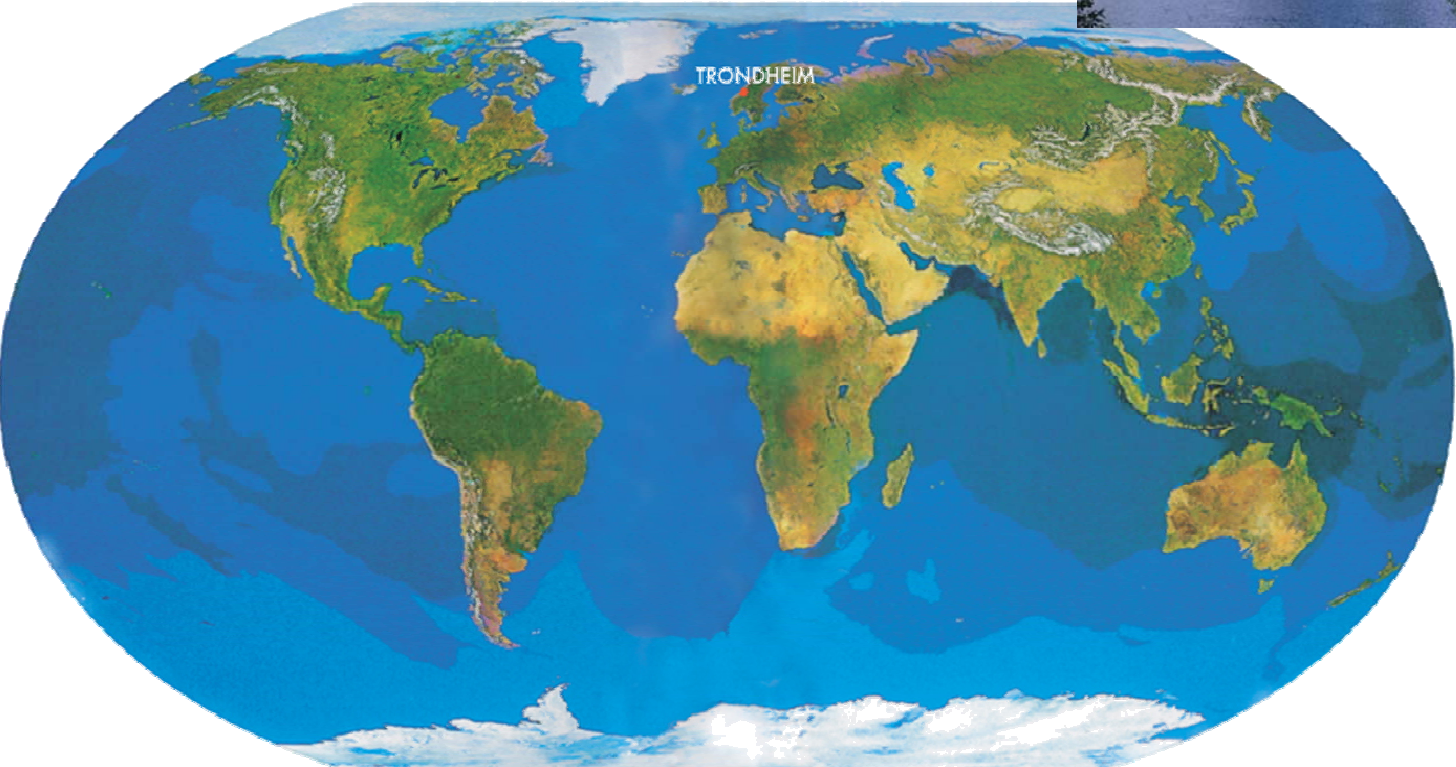
PLANTWIDE CONTROL

Sigurd Skogestad

Department of Chemical Engineering
Norwegian University of Science and Technology (NTNU)
Trondheim, Norway

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Trondheim, Norway



Arctic circle

North Sea

Trondheim

NORWAY

SWEDEN

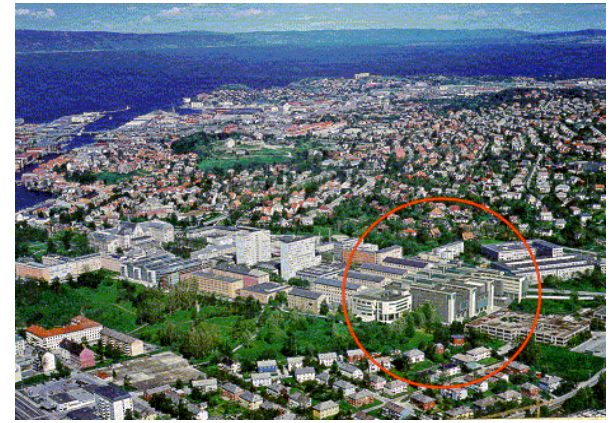
Oslo

DENMARK

GERMANY

UK





NTNU, Trondheim



Outline

- Control structure design (plantwide control)
- A procedure for control structure design
 - I Top Down
 - Step 1: Degrees of freedom
 - Step 2: Operational objectives (optimal operation)
 - Step 3: What to control ? (primary CV's) (self-optimizing control)
 - Step 4: Where set production rate?
 - II Bottom Up
 - Step 5: Regulatory control: What more to control (secondary CV's) ?
 - Step 6: Supervisory control
 - Step 7: Real-time optimization
- Case study: HDA
- ++ Simple PID tuning rules

Main message

- 1. Control for economics (Top-down steady-state arguments)
 - Primary controlled variables $c = y_1$
- 2. Control for stabilization (Bottom-up; regulatory PID control)
 - Secondary controlled variables y_2 (“inner cascade loops”)

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

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

- 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.

- Carl Nett (1989):

Minimize control system complexity subject to the achievement of accuracy specifications in the face of uncertainty.

“Plantwide control” = “Control structure design for complete chemical plant”

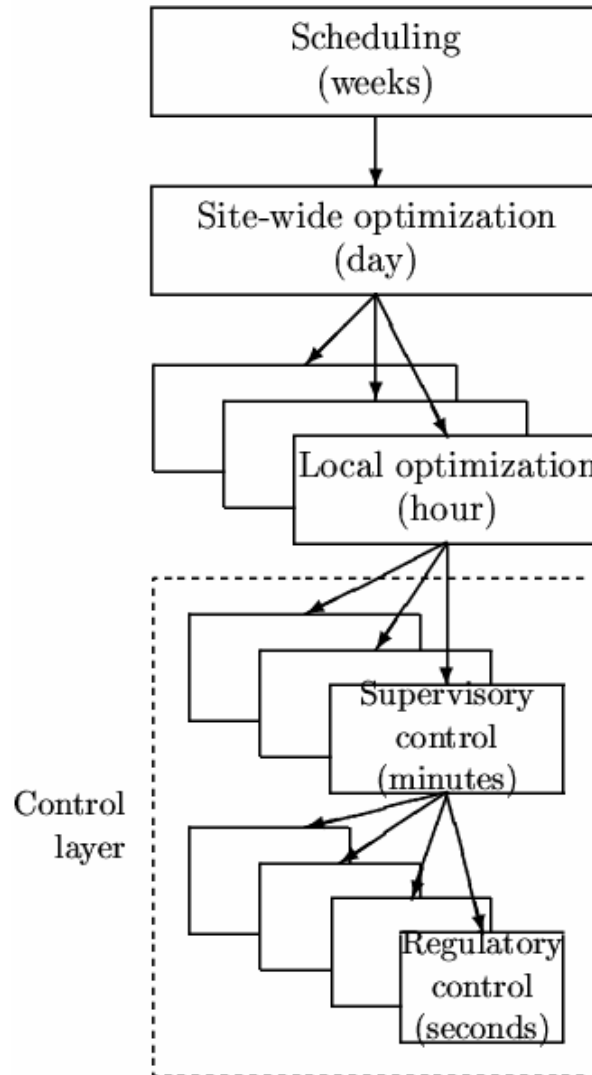
- *Not* the tuning and behavior of each control loop,
- But rather the *control philosophy* of the overall plant with emphasis on the *structural decisions*:
 - *Selection of controlled variables (“outputs”)*
 - *Selection of manipulated variables (“inputs”)*
 - *Selection of (extra) measurements*
 - *Selection of control **configuration*** (structure of overall controller that interconnects the controlled, manipulated and measured variables)
 - *Selection of controller type* (PID, decoupler, MPC, LQG etc.).

Main simplification: Hierarchical structure

RTO

MPC

PID



Need to define objectives and identify main issues for each layer

Regulatory control (seconds)

- *Purpose*: “Stabilize” the plant by controlling selected “secondary” variables (y_2) such that the plant does not drift too far away from its desired operation
- Use simple single-loop **PI(D) controllers**
- *Status*: Many loops poorly tuned
 - Most common setting: $K_c=1$, $\tau_I=1$ min (default)
 - Even wrong sign of gain K_c

Regulatory control.....

- *Trend*: Can do better! Carefully go through plant and retune important loops using standardized tuning procedure
- Exists many tuning rules, including Skogestad (SIMC) rules:
 - $K_c = (1/k) (\tau_1 / [\tau_c +])$ $\tau_I = \min (\tau_1, 4[\tau_c +])$, Typical: $\tau_c =$
 - “Probably the best simple PID tuning rules in the world” © Carlsberg
- *Outstanding structural issue*: What loops to close, that is, which variables (y_2) to control?

Supervisory control (minutes)

- *Purpose:* Keep primary controlled variables ($c=y_1$) at desired values, using as degrees of freedom the setpoints y_{2s} for the regulatory layer.
- *Status:* Many different “advanced” controllers, including feedforward, decouplers, overrides, cascades, selectors, Smith Predictors, etc.
- *Issues:*
 - Which variables to control may change due to change of “active constraints”
 - Interactions and “pairing”

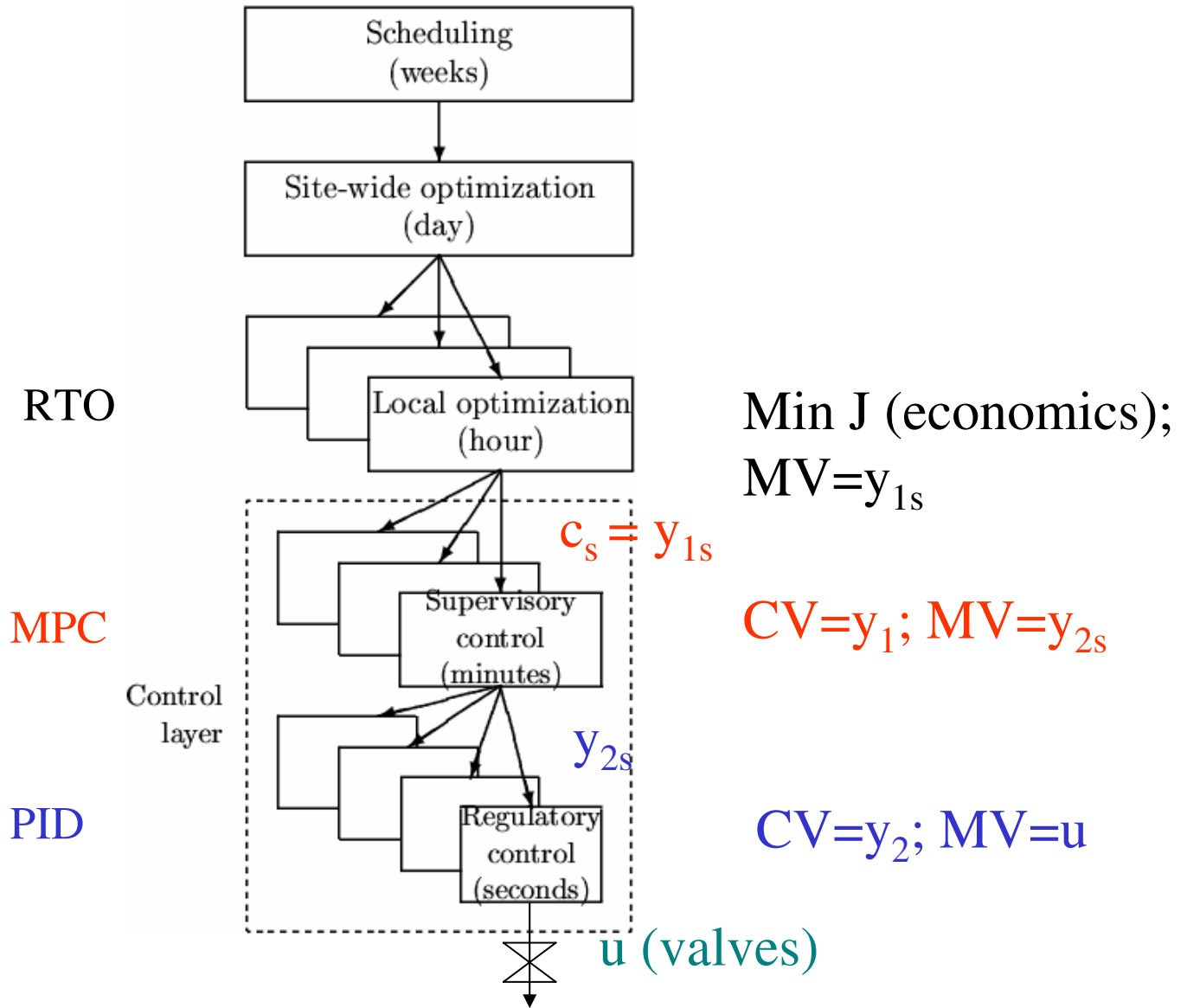
Supervisory control.....

- *Trend:* Model predictive control (**MPC**) used as unifying tool.
 - Linear multivariable models with input constraints
 - Tuning (modelling) is time-consuming and expensive
- *Issue:* When use MPC and when use simpler single-loop decentralized controllers ?
 - MPC is preferred if active constraints (“bottleneck”) change.
 - Avoids logic for reconfiguration of loops
- *Outstanding structural issue:*
 - What primary variables **$c=y_1$** to control?

Local optimization (hour)

- *Purpose:* Minimize cost function J and:
 - Identify active constraints
 - Recompute optimal setpoints y_{1s} for the controlled variables
- *Status:* Done manually by clever operators and engineers
- *Trend:* Real-time optimization (**RTO**) based on detailed nonlinear steady-state model
- *Issues:*
 - Optimization not reliable.
 - Need nonlinear steady-state model
 - Modelling is time-consuming and expensive

Objectives of layers: MV's and CV's



Stepwise procedure plantwide control

I. TOP-DOWN

Step 1. DEGREES OF FREEDOM

Step 2. OPERATIONAL OBJECTIVES

Step 3. WHAT TO CONTROL? (primary CV's $c=y_1$)

Step 4. PRODUCTION RATE

II. BOTTOM-UP (structure control system):

Step 5. REGULATORY CONTROL LAYER (PID)

“Stabilization”

What more to control? (secondary CV's y_2)

Step 6. SUPERVISORY CONTROL LAYER (MPC)

Decentralization

Step 7. OPTIMIZATION LAYER (RTO)

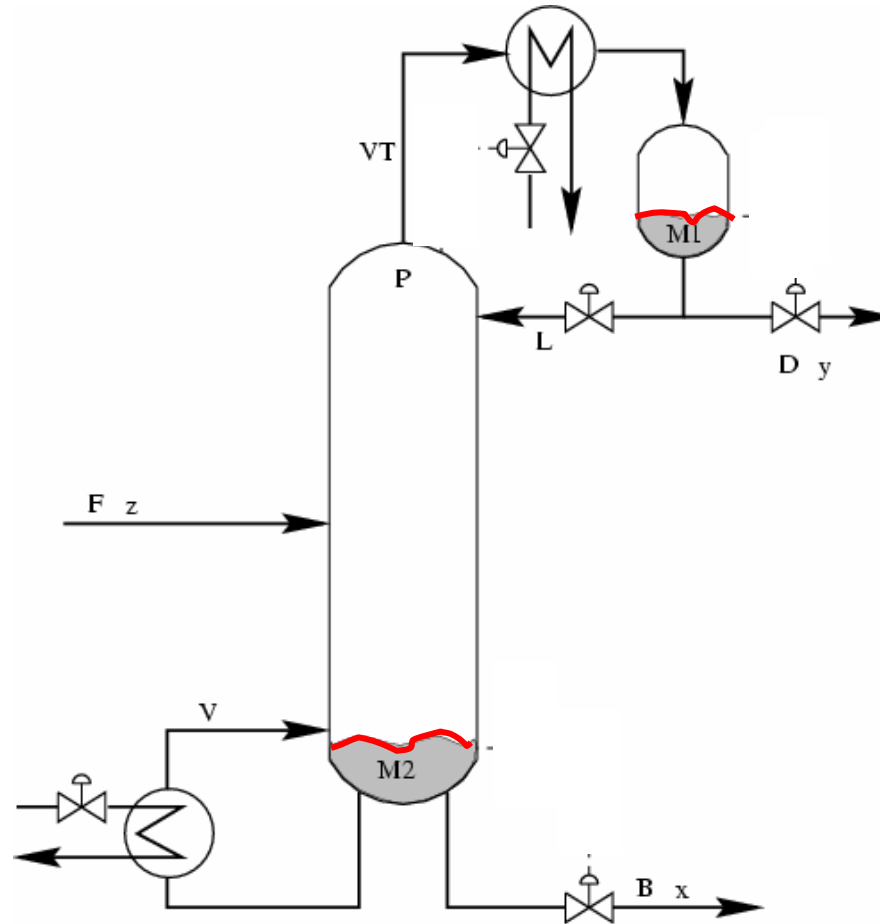
Can we do without it?

Steady-state degrees of freedom (N_{ss}): Typical number for some process units

- each external feedstream: 1 (feedrate)
- splitter: $n-1$ (split fractions) where n is the number of exit streams
- mixer: 0
- compressor, turbine, pump: 1 (work)
- adiabatic flash tank: 0^*
- liquid phase reactor: 1 (holdup-volume reactant)
- gas phase reactor: 0^*
- heat exchanger: 1 (duty or net area)
- distillation column excluding heat exchangers: 0^* + number of sidestreams
- **pressure*** : add 1DOF at each extra place you set pressure (using an extra valve, compressor or pump!). Could be for **adiabatic flash tank, gas phase reactor, distillation column**

* Pressure is normally assumed to be given by the surrounding process and is then not a degree of freedom

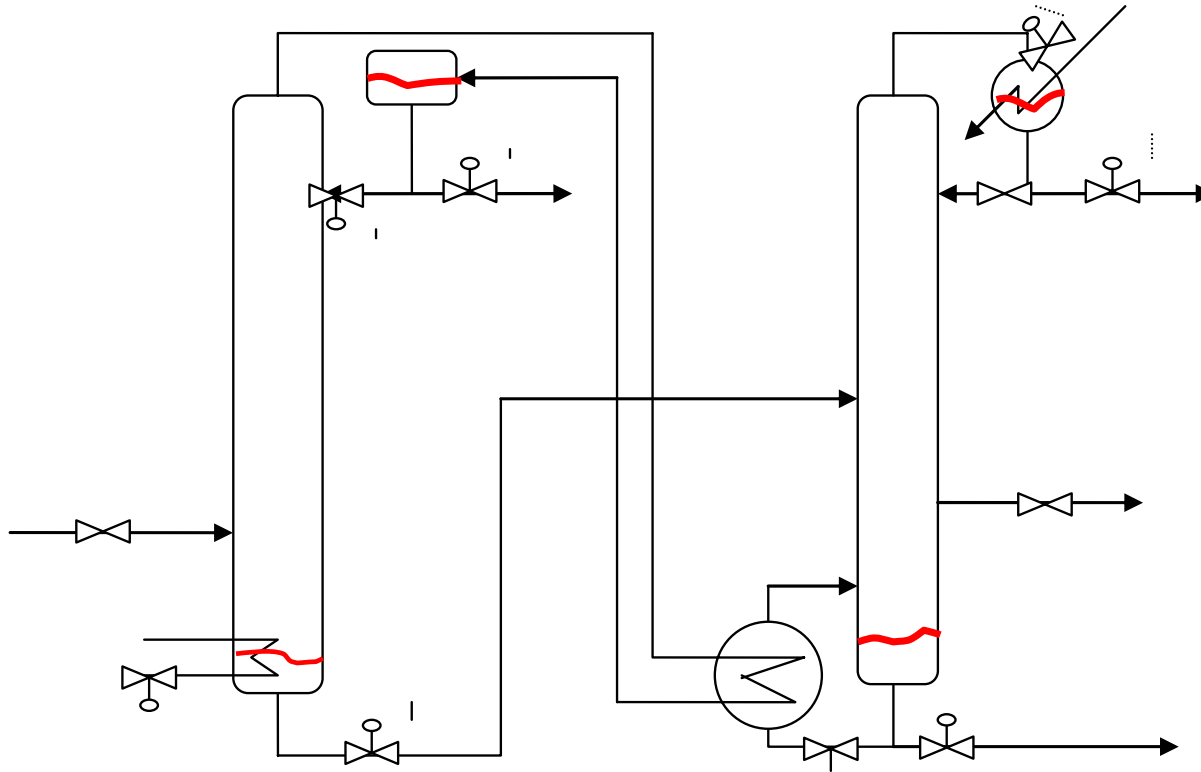
Distillation column with given feed and pressure



“Typical number”,

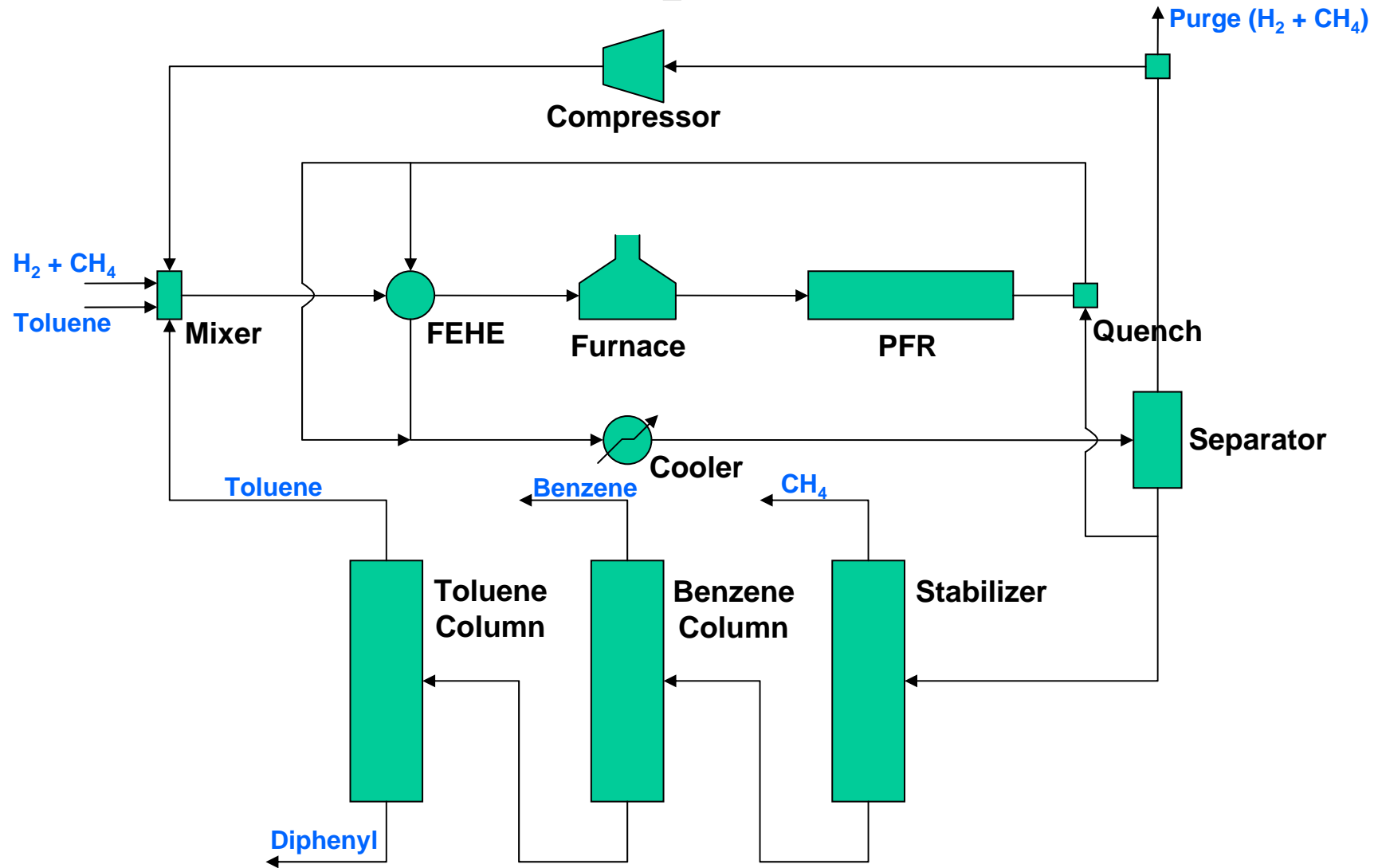
$$N_{ss} = 0 \text{ (distillation)} + 2 * 1 \text{ (heat exchangers)} = 2$$

Heat-integrated distillation process

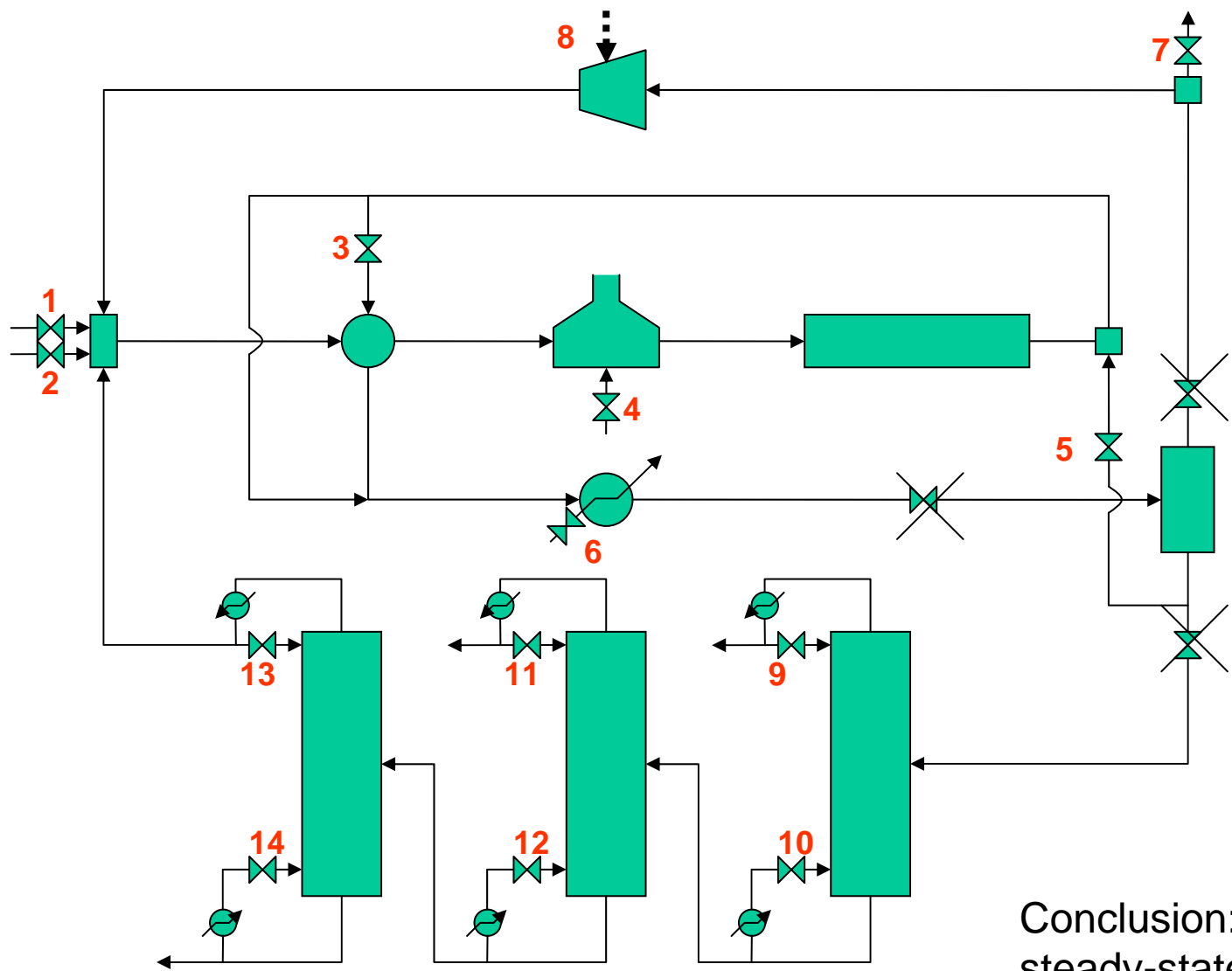


Typical number, $N_{ss} = 1$ (feed) + $2*0$ (columns) + $2*1$ (column pressures) + 1 (sidestream) + 3 (hex) = 7

HDA process



HDA process: steady-state degrees of freedom



feed: 1, 2
 hex: 3, 4, 6
 splitter: 5, 7
 compressor: 8
 distillation: rest

Conclusion: 14
 steady-state
 DOFs

Assume given column pressures

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Optimal operation (economics)

- What are we going to use our degrees of freedom for?
- Define scalar cost function $J(u_0, x, d)$
 - u_0 : degrees of freedom
 - d : disturbances
 - x : states (internal variables)

Typical cost function:

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

- Optimal operation for given d :

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

subject to:

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

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

Optimal operation

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

Two main cases (modes) depending on marked conditions:

1. Given feed

Amount of products is then usually indirectly given and $J = \text{cost energy}$.

Optimal operation is then usually *unconstrained*:

“maximize efficiency (energy)”

Control: Operate at optimal trade-off (not obvious how to do and what to control)

2. Feed free

Products usually much more valuable than feed + energy costs small.

Optimal operation is then usually *constrained*:

“maximize production”

Control: Operate at bottleneck (“obvious”)

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Implementation of optimal operation

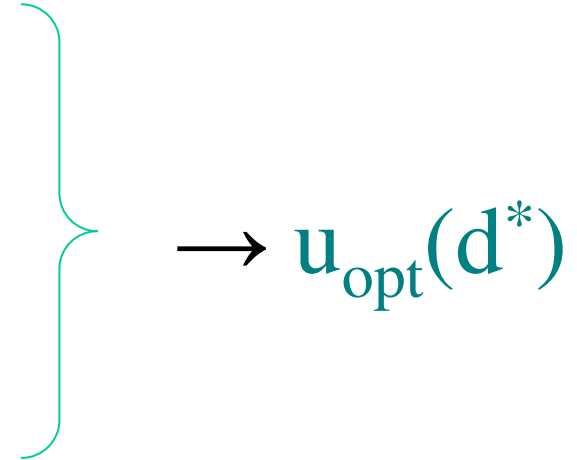
- Optimal operation for given d^* :

$$\min_{\mathbf{u}} J(\mathbf{u}, \mathbf{x}, \mathbf{d})$$

subject to:

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

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



Problem: Usually cannot keep \mathbf{u}_{opt} constant because disturbances \mathbf{d} change

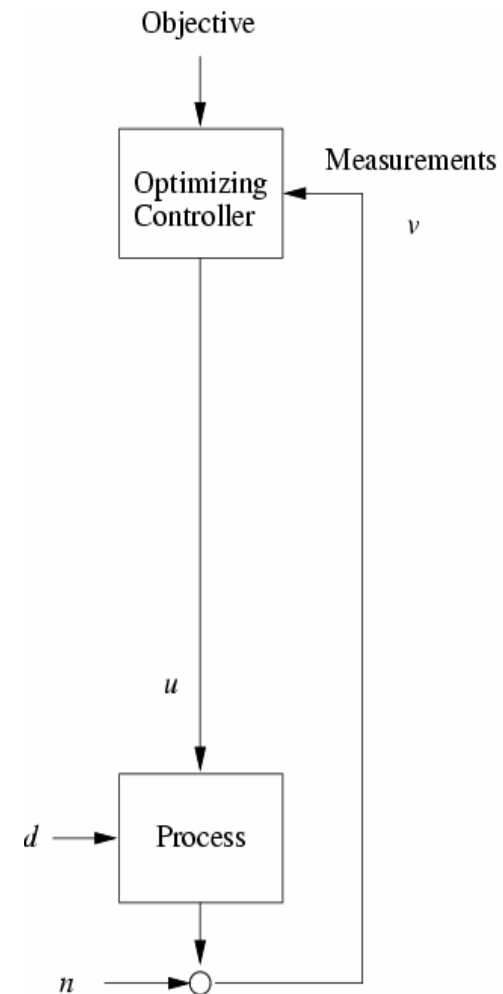
How should be adjust the degrees of freedom (\mathbf{u})?

Implementation of optimal operation (Cannot keep u_{0opt} constant)

”Obvious” solution: Optimizing control

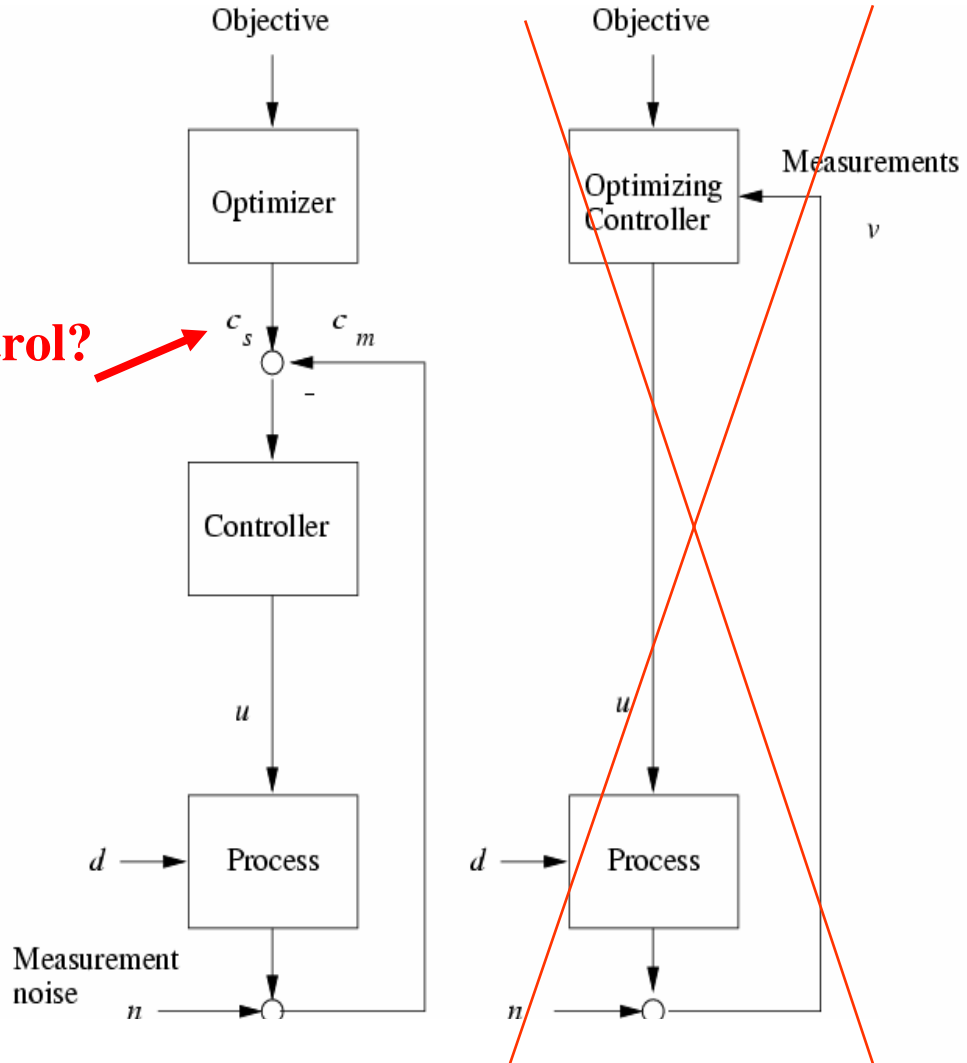
Estimate d from measurements
and recompute $u_{opt}(d)$

Problem: Too complicated
(requires detailed model *and*
description of uncertainty)



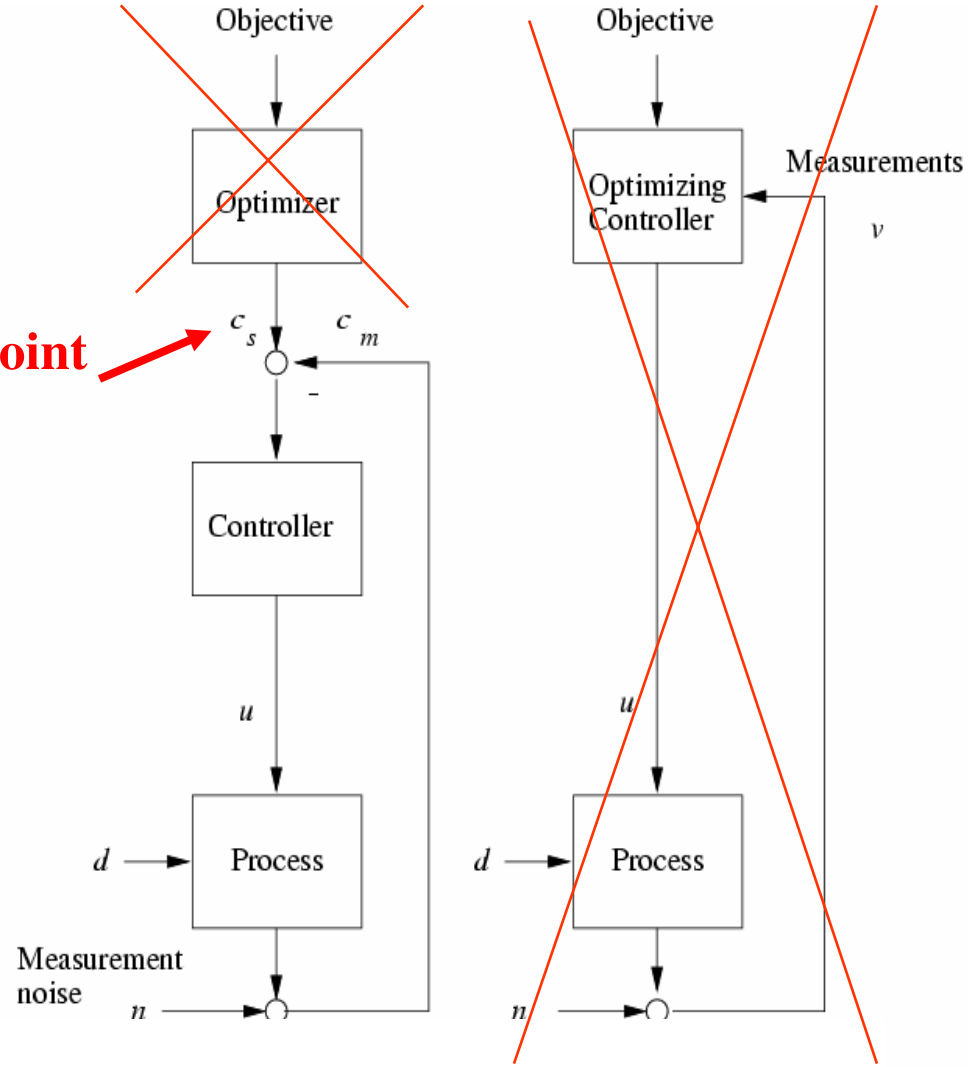
In practice: Hierarchical decomposition with separate layers

What should we control?



Self-optimizing control: When constant setpoints is OK

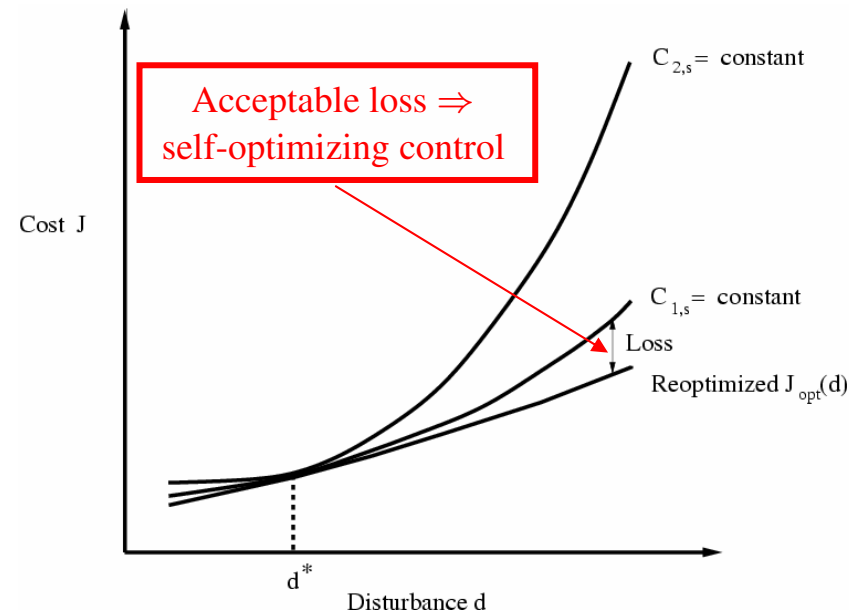
Constant setpoint



Unconstrained variables: Self-optimizing control

- Self-optimizing control:*
 Constant setpoints c_s give
 "near-optimal operation"
 (= acceptable loss L for expected
 disturbances d and
 implementation errors n)

$$L(d) = J(c_s + n, d) - J_{opt}(d)$$



What c 's should we control?

- Optimal solution is usually at constraints, that is, most of the degrees of freedom are used to satisfy “active constraints”, $g(u,d) = 0$
- **CONTROL ACTIVE CONSTRAINTS!**
 - $c_s =$ value of active constraint
 - Implementation of active constraints is usually simple.
- **WHAT MORE SHOULD WE CONTROL?**
 - Find “self-optimizing” variables c for remaining unconstrained degrees of freedom u .

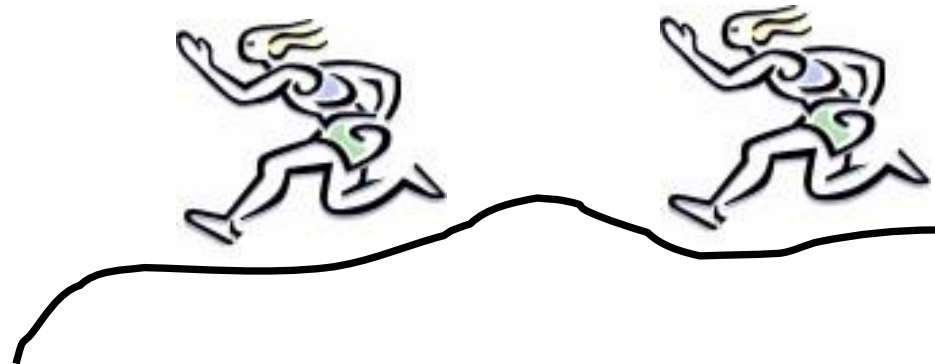
What should we control? – Sprinter

- Optimal operation of Sprinter (100 m), $J=T$
 - One input: "power/speed"
 - **Active constraint control:**
 - Maximum speed ("no thinking required")



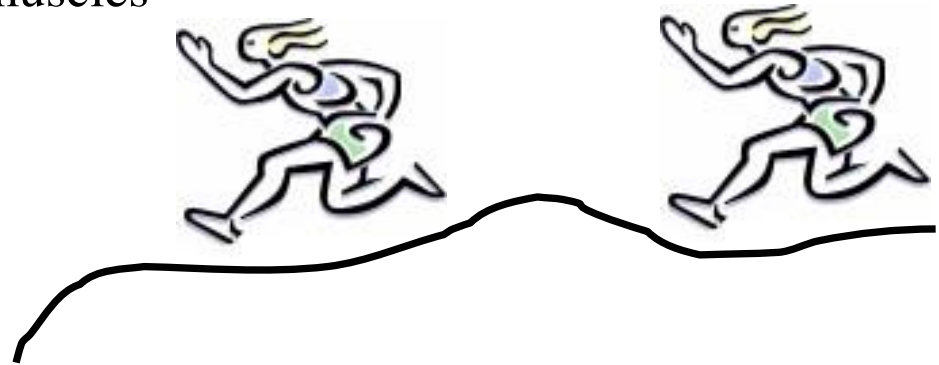
What should we control? – Marathon

- Optimal operation of Marathon runner, $J=T$
 - No active constraints
 - Any **self-optimizing** variable c (to control at constant setpoint)?



Self-optimizing Control – Marathon

- Optimal operation of Marathon runner, $J=T$
 - Any self-optimizing variable c (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



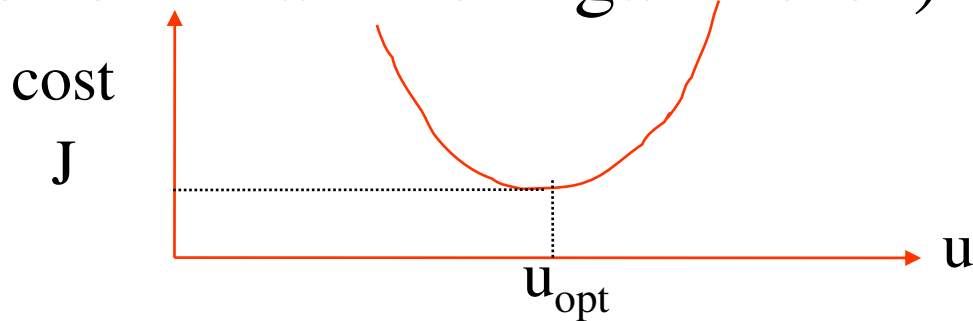
Unconstrained variables: What should we control?

- *Intuition*: “Dominant variables” (Shinnar)
- Is there any systematic procedure?

Candidate controlled variables

- We are looking for some “magic” variables **c** to control....
What properties do they have?
 - **Intuitively 1:** Should have small optimal range Δc_{opt}
 - since we are going to keep them constant!
 - **Intuitively 2:** Should have small “implementation error” n
 - **Intuitively 3:** Should be sensitive to inputs u (remaining unconstrained degrees of freedom), that is, the gain G_0 from u to c should be large
 - G_0 : (unscaled) gain from u to c
 - large gain gives flat optimum in c
 - Charlie Moore (1980’s): Maximize minimum singular value when selecting temperature locations for distillation
- } **span(c)**
- Will show shortly: Can combine everything into the “maximum gain rule”:
 - **Maximize scaled gain $G = G_0 / \text{span}(c)$**

Mathematic local analysis (Proof of “maximum gain rule”)



$$\text{Loss} = J(u, d) - J_{\text{opt}}(d) = \underbrace{\left(\frac{\partial J}{\partial u}\right)_{\text{opt}}^T}_{=0} (u - u_{\text{opt}}(d))$$

$$\cdot + \frac{1}{2} (u - u_{\text{opt}}(d))^T \underbrace{\left(\frac{\partial^2 J}{\partial u^2}\right)_{\text{opt}}}_{=J_{uu}} (u - u_{\text{opt}}(d)) + \dots$$

$$\cdot \approx \frac{1}{2} (c - c_{\text{opt}}(d))^T G_0^{-T} J_{uu} G_0^{-1} (c - c_{\text{opt}}(d))$$

$$\cdot \leq \frac{1}{2} G^{-T} J_{uu} G^{-1}$$

where $G_0 =$ unscaled gain ($\Delta c = G_0 \Delta u$)

$G =$ scaled gain $= \frac{|G_0|}{\text{span}(c)}$;

$\text{span}(c) =$ max. expected $c - c_{\text{opt}}(d)$

Minimum singular value of scaled gain

*Maximum gain rule (Skogestad and Postlethwaite, 1996):
Look for variables that maximize the scaled gain $\underline{\sigma}(G)$
(minimum singular value of the appropriately scaled
steady-state gain matrix G from u to c)*

$$\text{Loss} \approx \frac{\bar{\sigma}(J_{uu})}{2} \cdot \frac{1}{\underline{\sigma}(G)^2}; \quad G = \frac{G_0}{\text{span}(c)}$$

$\underline{\sigma}(G)$ is called the Morari Resiliency index (MRI) by Luyben

Detailed proof: I.J. Halvorsen, S. Skogestad, J.C. Morud and V. Alstad,

"Optimal selection of controlled variables", *Ind. Eng. Chem. Res.*, **42** (14), 3273-3284 (2003).

Improved minimum singular value rule for ill-conditioned plants

$$\text{Maximize } \underline{\sigma}(G J_{uu}^{-1/2})$$

G: Scaled gain matrix (as before)

J_{uu} : Hessian for effect of u's on cost

Problem: J_{uu} can be difficult to obtain

Improved rule has been used successfully for distillation

Unconstrained degrees of freedom:

Maximum gain rule for scalar system

$$\text{Loss} \approx \frac{|J_{uu}|}{2} \cdot \frac{1}{|G|^2}$$

Scaled steady-state gain from u to c :

$$|G| = \frac{|G_0|}{|c_{\text{opt}}| + |n|} = \frac{|G_0|}{\text{optimal span } c}$$

Summary unconstrained degrees of freedom:

Looking for “**magic**” variables to keep at constant setpoints.

How can we find them systematically?

Candidates

$c = y_i$: Single measurements, e.g. pressure, temperature, composition

$c = \frac{y_i}{y_j}$: Combinations of measurements (e.g flow ratios)

A. Start with: Maximum gain (minimum singular value) rule:

(Scaled) gain $\underline{\sigma}(G)$ from u to y should be large

B. Then: “Brute force evaluation” of most promising alternatives.

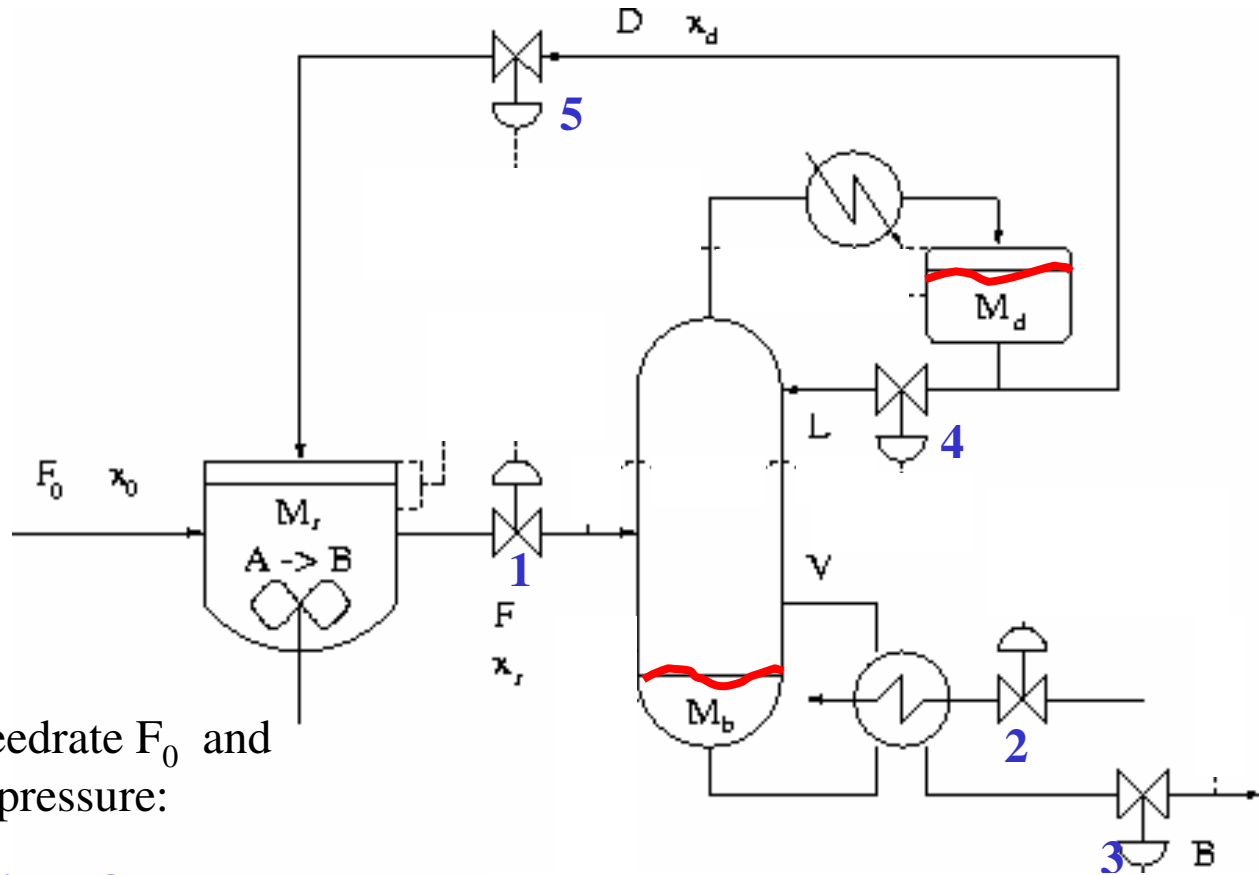
Evaluate loss when the candidate variables c are kept constant.

In particular, may be problem with **feasibility**

C. More general candidates: Find optimal linear combination (matrix H):

$$c = h_1 y_1 + h_2 y_2 + \dots + h_n y_n = Hy$$

EXAMPLE: Recycle plant (Luyben, Yu, etc.)



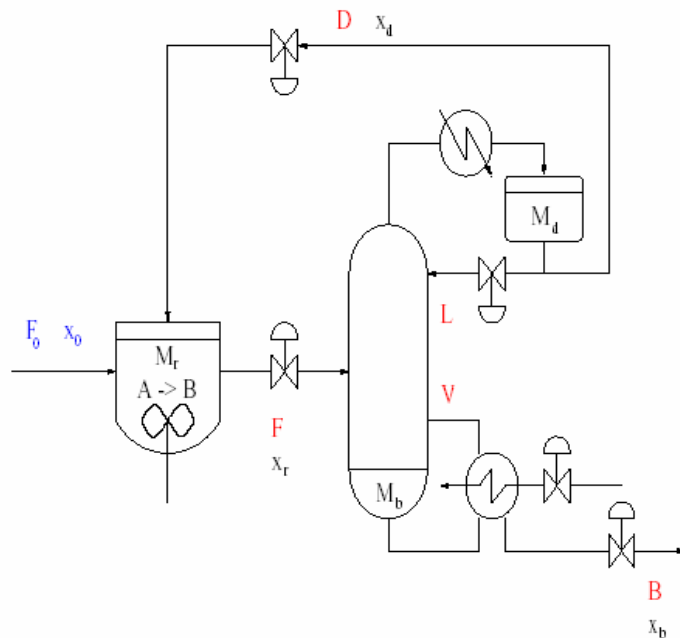
Given feedrate F_0 and column pressure:

Dynamic DOFs: $N_m = 5$

Column levels: $N_{0y} = 2$

Steady-state DOFs: $N_0 = 5 - 2 = 3$

Recycle plant: Optimal operation



Manipulated variables:

$$\mathbf{m}^T = [V \ L \ B \ D \ F]$$

Steady-state degrees of freedom: **3**

Minimize costs

$$J = V$$

Constraints:

$$x_b \leq 0.015 \frac{\text{molA}}{\text{mol}} \quad \text{active}$$

$$M_r \leq 2800 \text{ kmol} \quad \text{active}$$

$$V \leq 5000 \text{ kmol/h}$$

$$\text{Flows} \geq 0 \text{ kmol/h}$$

Disturbances:

$$\mathbf{d}^T = [F_0 \ x_0] = [460 \pm 92 \frac{\text{kmol}}{\text{h}} \ 0.90 \pm 0.05]$$

1 remaining unconstrained degree of freedom

A. Maximum gain rule: Steady-state gain

Rank	c	$ G(0) \cdot 10^3$
1	x_D	13.1
2	L/F	8.9
3	D/L	7.7
4	D/V	5.8
5	V/L	4.5
6	B/L	4.1
7	V/F	4.0
8	B/D	3.3
9	L	3.0
10	B/F	2.6
11	D	2.6
12	F/F_0	2.5
13	D/F	2.5
14	F	1.9
15	B/V	0
15	V	0
15	x_r	0
15	B	0

Conventional:
Looks good

Luyben rule:
Not promising
economically

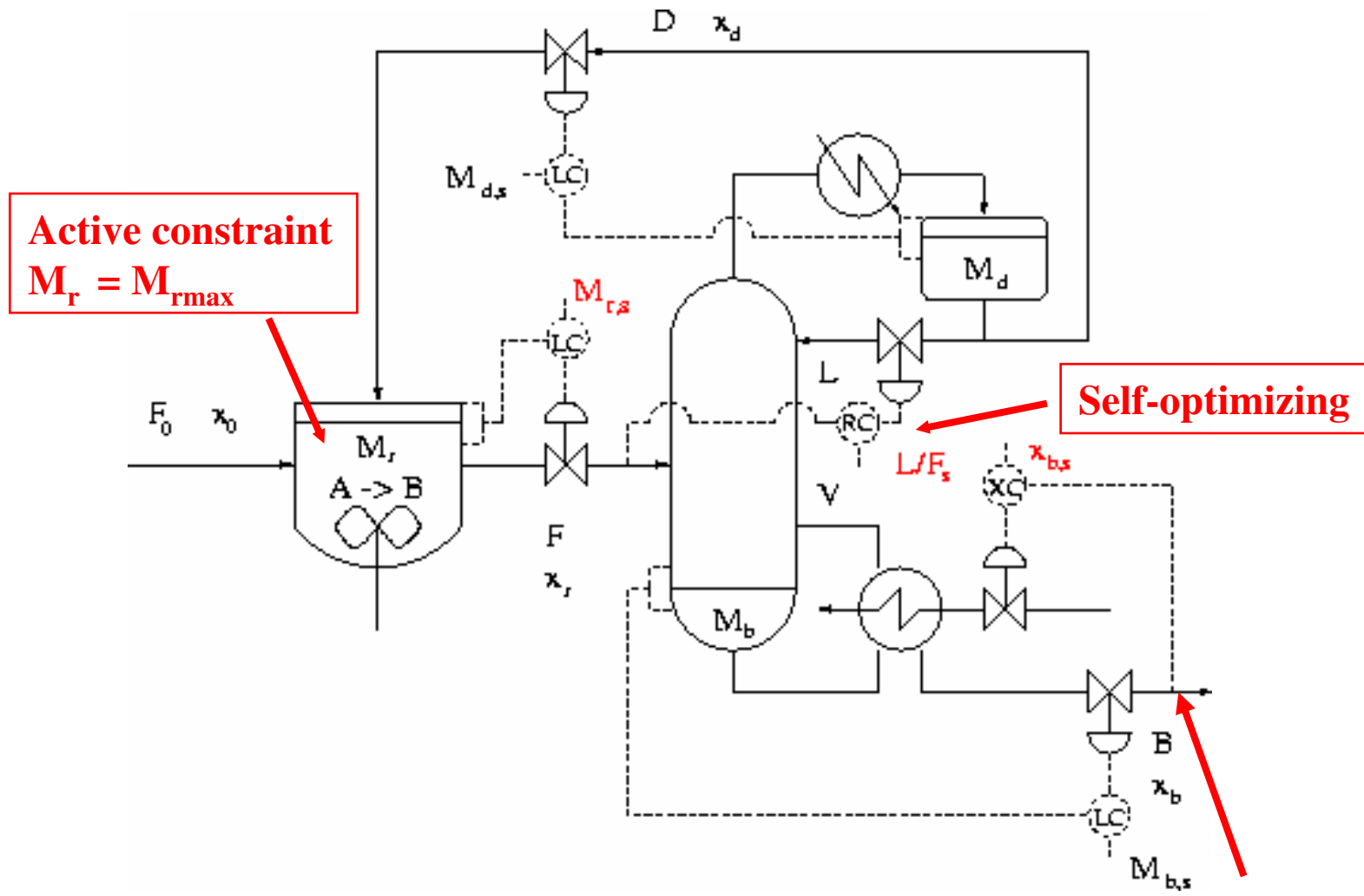
Gain from $u = L$ to c with active constraints (M_r and x_B) constant.
 Scaling: $\Delta c_i = \max_d |c_{opt,i}(d) - c_{opt,i}(d^*)|$ + implementation error.
 $d = F_0, z_F$.

How did we find the gains in the Table?

1. Find nominal optimum
2. Find (unscaled) gain G_0 from input to candidate outputs: $c = G_0 u$.
 - In this case only a single unconstrained input (DOF). Choose at $u=L$
 - Obtain gain G_0 numerically by making a small perturbation in $u=L$ while adjusting the other inputs such that the **active constraints are constant** (bottom composition fixed in this case)
3. Find the span for each candidate variable
 - For each disturbance d_i make a typical change and reoptimize to obtain the optimal ranges $c_{opt}(d_i)$
 - For each candidate output obtain (estimate) the control error (noise) n
 - $span(c) = \sum_i |c_{opt}(d_i)| + n$
4. Obtain the scaled gain, $G = G_0 / span(c)$


 IMPORTANT!

Conclusion: Control of recycle plant



L/F constant: Easier than “two-point” control

Assumption: Minimize energy (V)

Summary:

Procedure selection controlled variables

1. Define economics and operational constraints
2. Identify degrees of freedom and important disturbances
3. Optimize for various disturbances
4. Identify (and control) active constraints (off-line calculations)
 - May vary depending on operating region. **For each operating region do step 5:**
5. Identify “self-optimizing” controlled variables for remaining degrees of freedom
 1. (A) Identify promising (single) measurements from “maximize gain rule” (gain = minimum singular value)
 - (C) Possibly consider measurement combinations if no promising
 2. (B) “Brute force” evaluation of loss for promising alternatives
 - Necessary because “maximum gain rule” is local.
 - In particular: Look out for feasibility problems.
 3. Controllability evaluation for promising alternatives

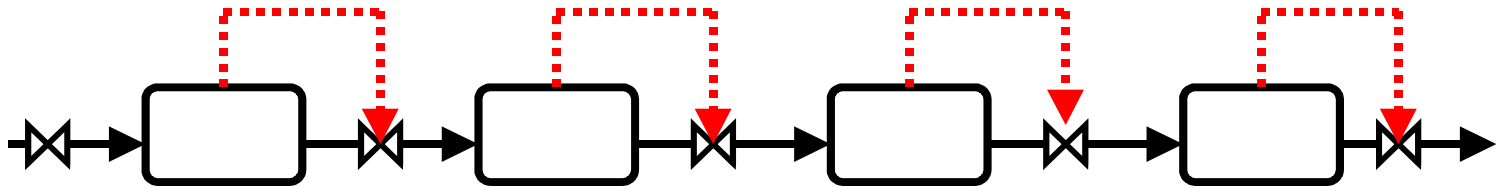
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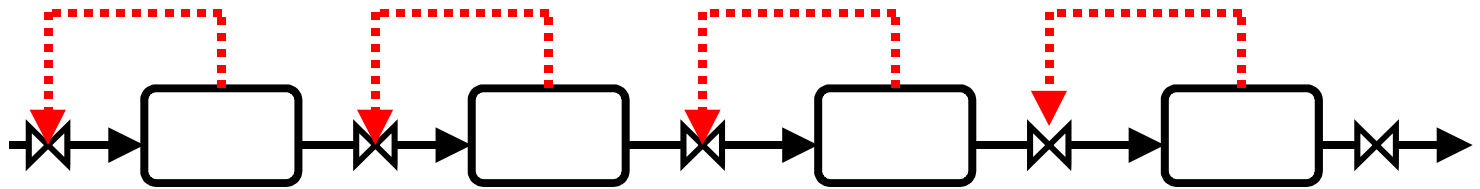
Step 4. Where set production rate?

- Very important!
- Determines structure of remaining inventory (level) control system
- Set production rate at (dynamic) bottleneck
- Link between **Top-down** and **Bottom-up** parts

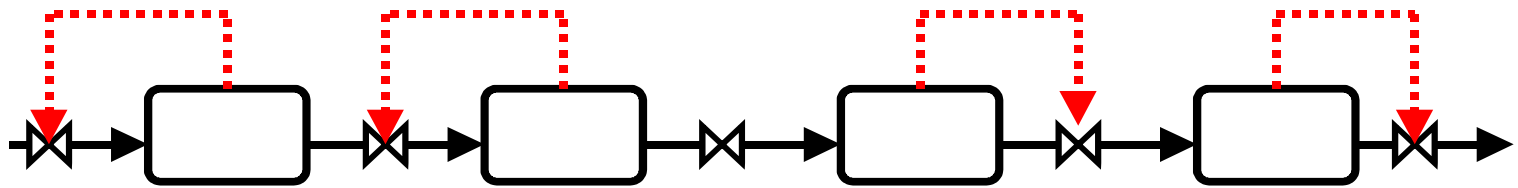
Production rate set at inlet :
Inventory control in direction of flow



Production rate set at outlet: Inventory control opposite flow



Production rate set inside process

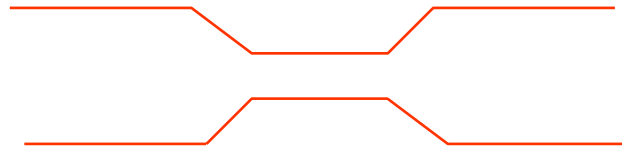


Where set the production rate?

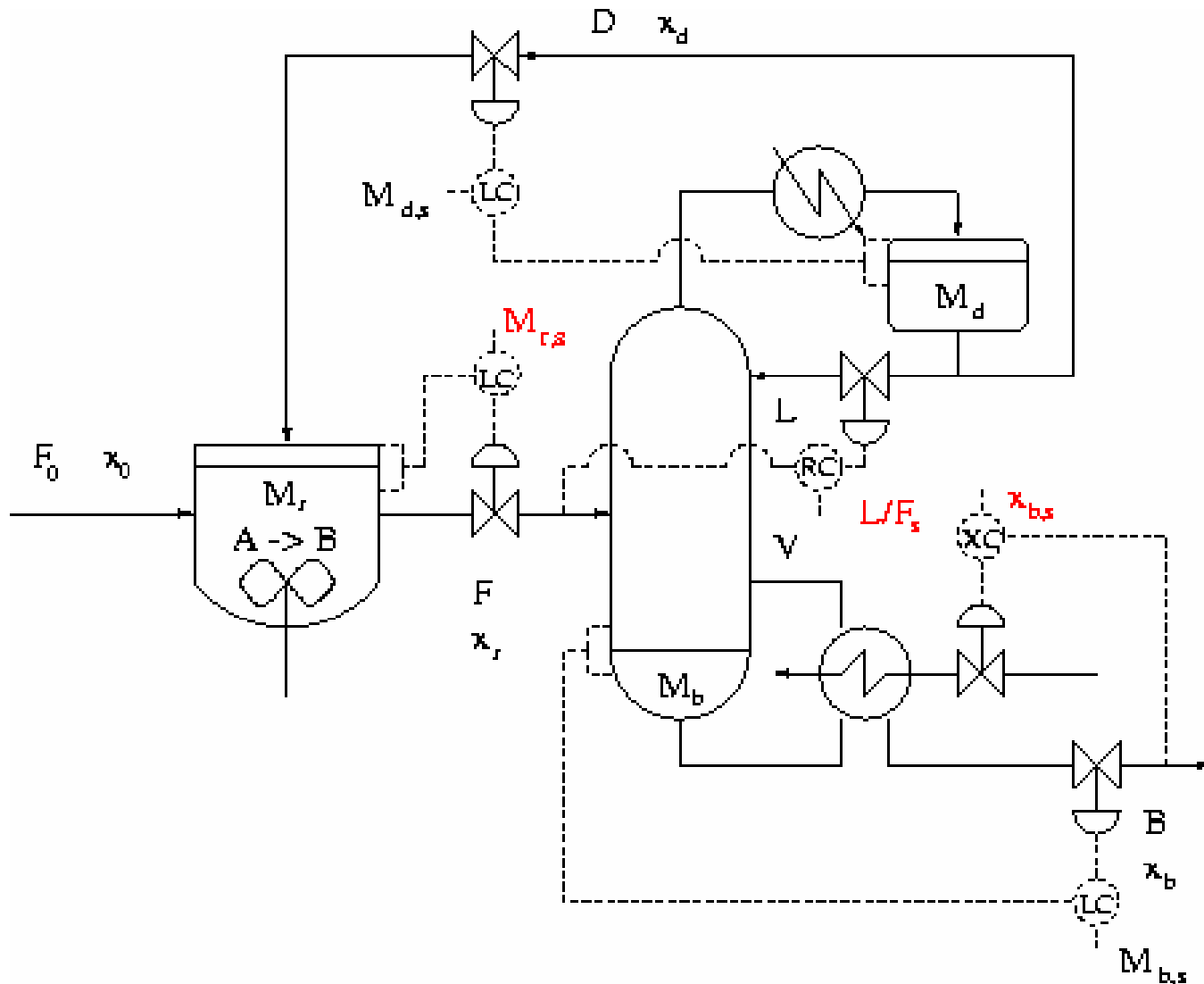
- Very important decision that determines the structure of the rest of the control system!
- May also have important economic implications

Often optimal: Set production rate at bottleneck!

- "A bottleneck is an extensive variable that prevents an increase in the overall feed rate to the plant"
- If feed is cheap and available: Optimal to set production rate at bottleneck
- If the flow for some time is not at its maximum through the bottleneck, then this loss can never be recovered.

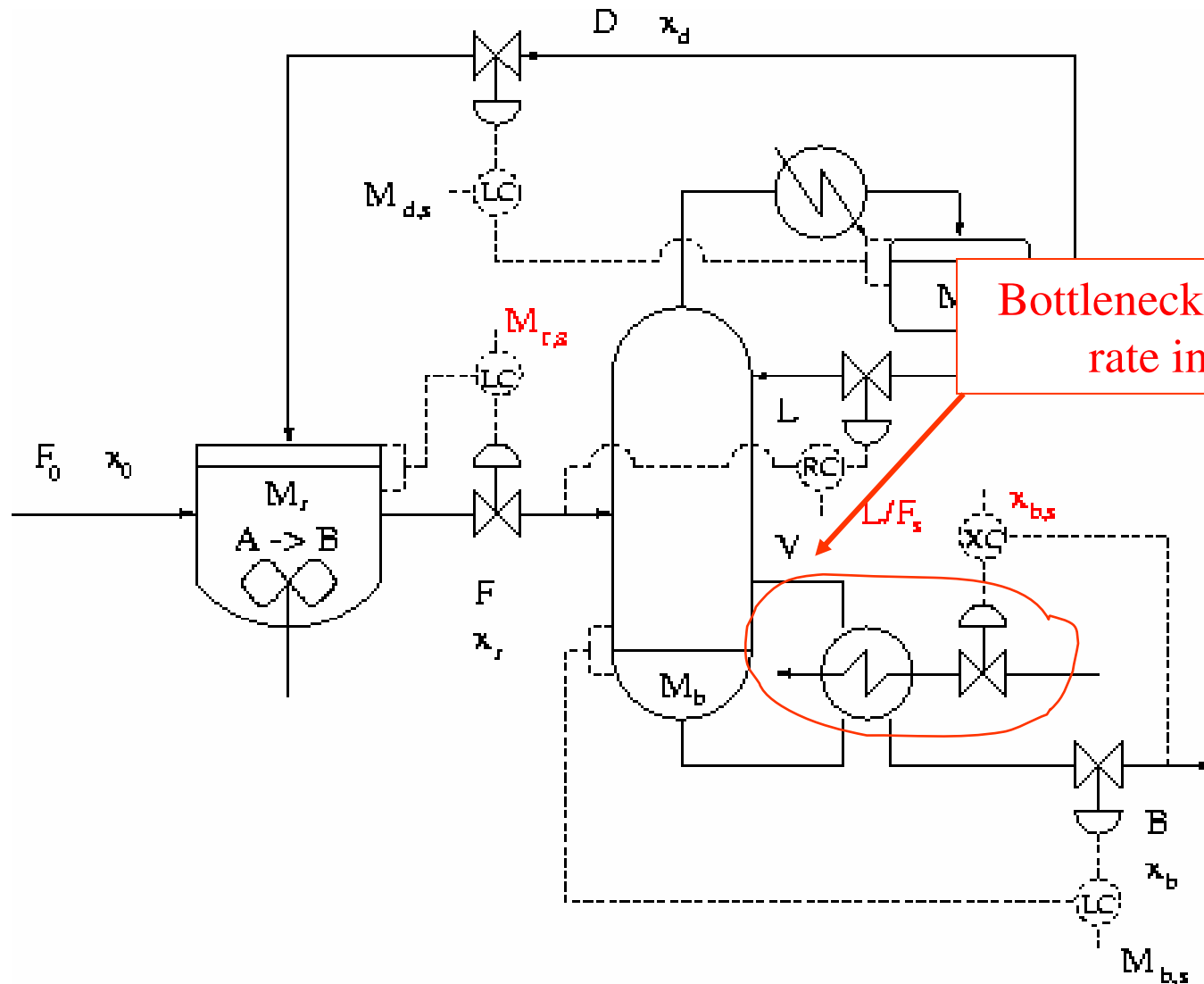


Reactor-recycle process: Given feedrate with production rate set at inlet



Reactor-recycle process:

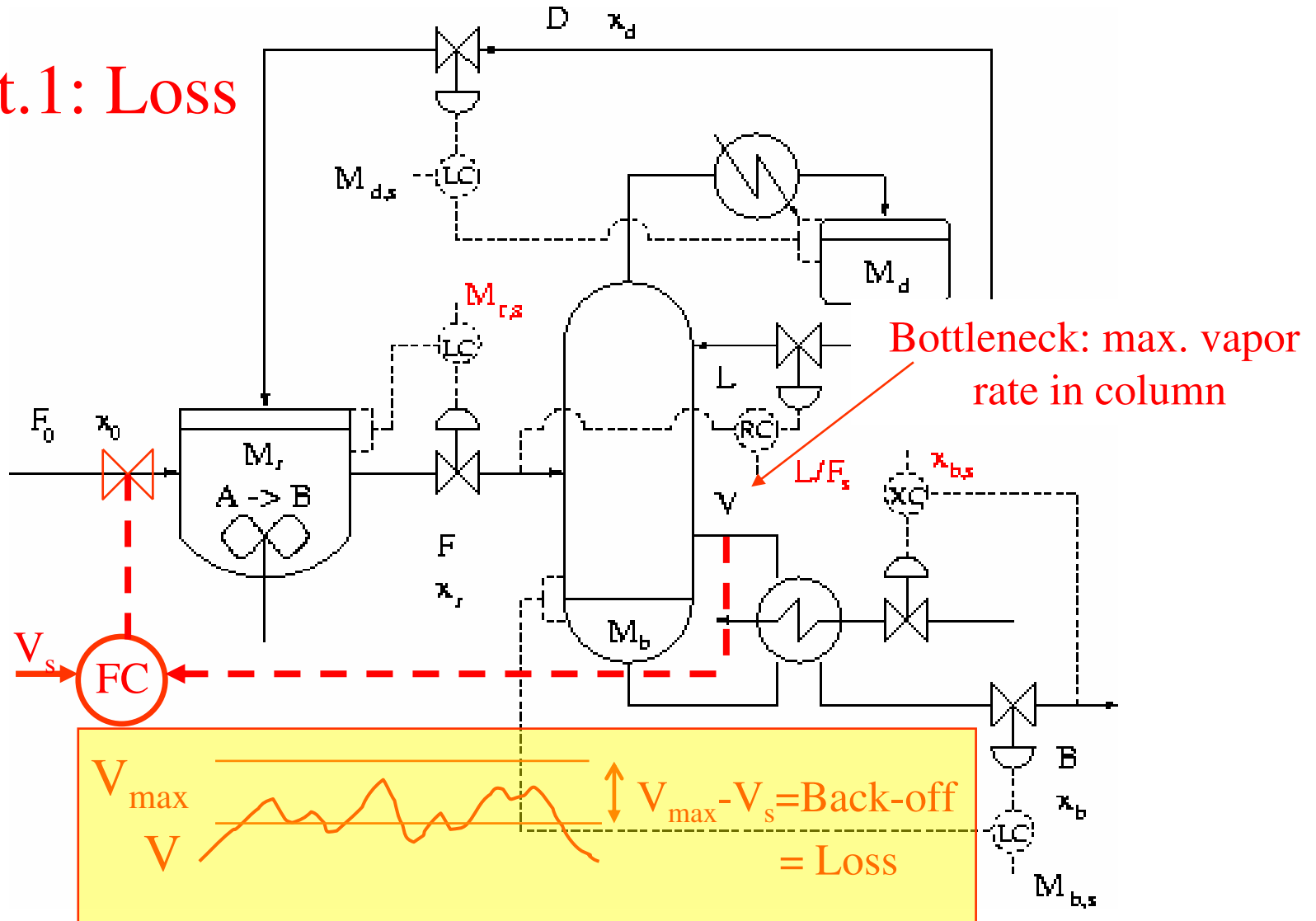
Want to maximize feedrate: reach **bottleneck** in column



Reactor-recycle process with production rate set at inlet

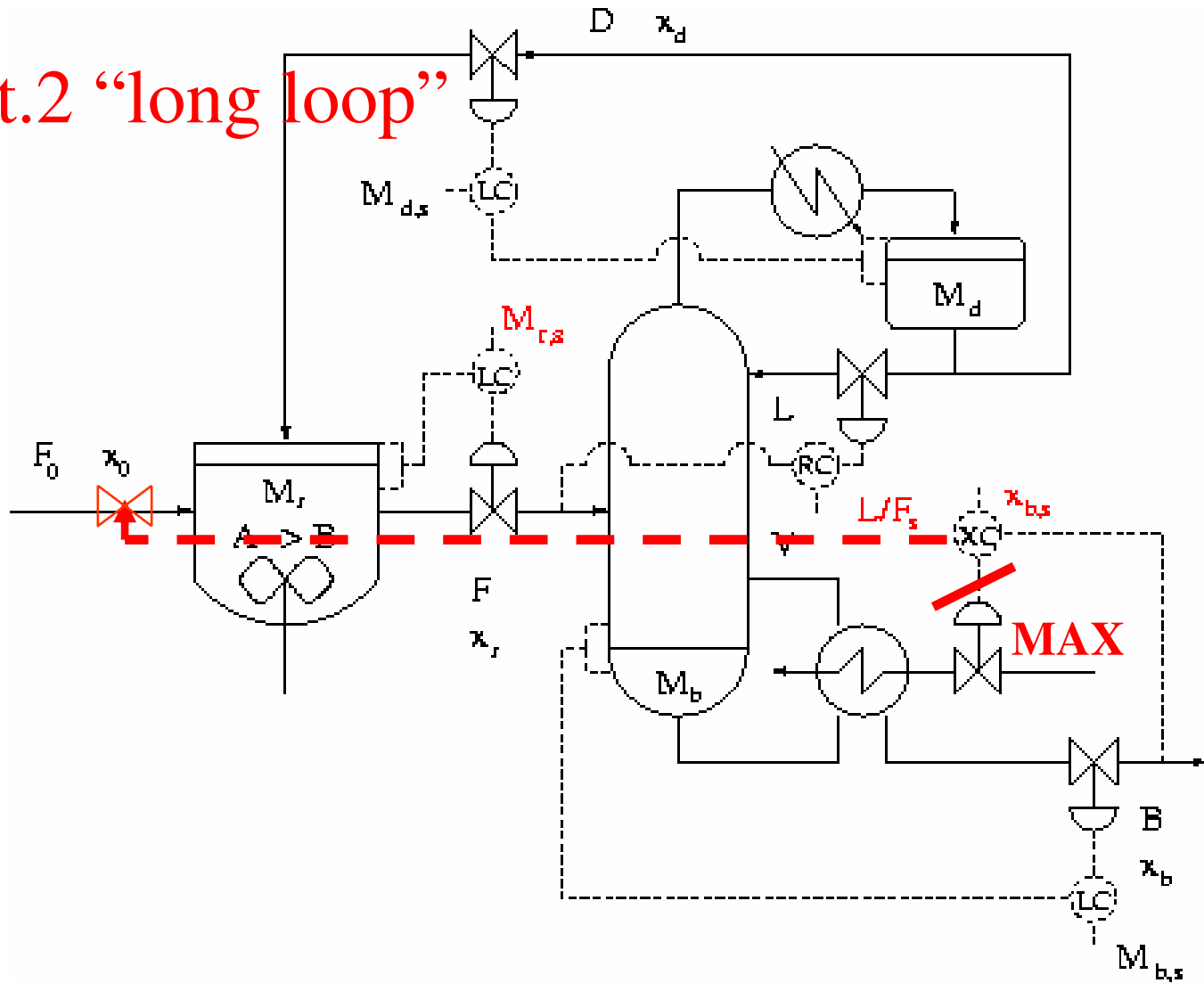
Want to maximize feedrate: reach bottleneck in column

Alt.1: Loss



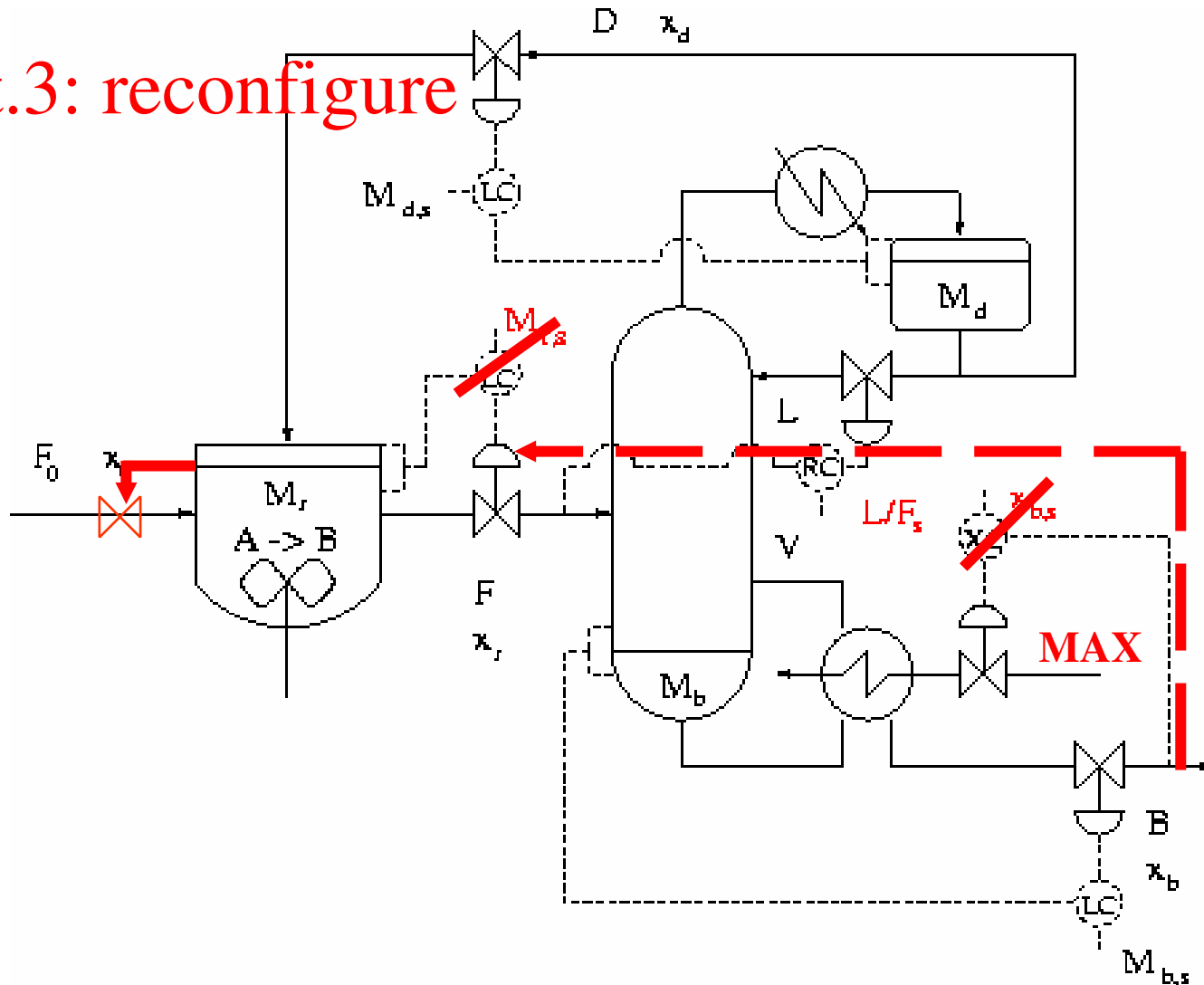
Reactor-recycle process with increased feedrate:
 Optimal: **Set production rate at bottleneck**

Alt.2 "long loop"



Reactor-recycle process with increased feedrate:
 Optimal: **Set production rate at bottleneck**

Alt.3: reconfigure



Alt.4: Multivariable control (MPC)

- Can reduce loss
- BUT: Is generally placed on top of the regulatory control system (including level loops), so it still important where the production rate is set!

Conclusion production rate manipulator

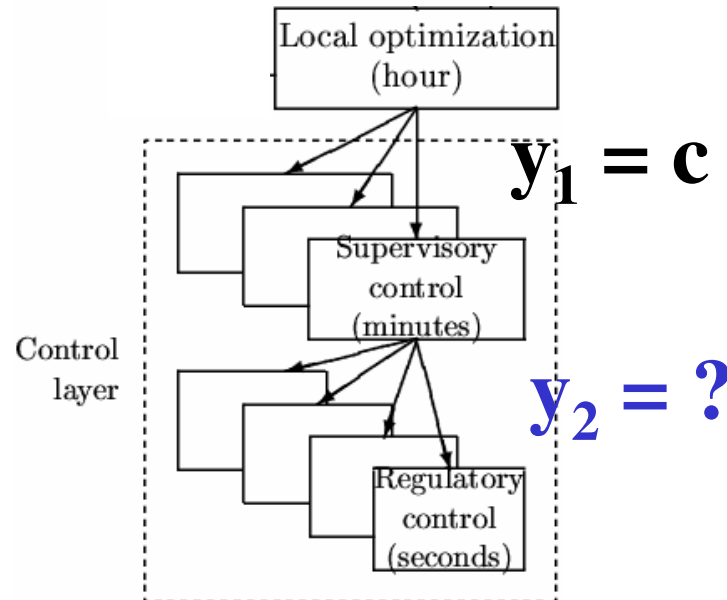
- Think carefully about where to place it!
- Difficult to undo later

Outline

- Control structure design (plantwide control)
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 - Step 3: What to control ? (self-optimizing control)
 - Step 4: Where set production rate?
 - II Bottom Up
 - Step 5: Regulatory control: What more to control ?
 - Step 6: Supervisory control
 - Step 7: Real-time optimization
- Case studies

Step 5. Regulatory control layer

- *Purpose*: “Stabilize” the plant using local SISO PID controllers
- Enable manual operation (by operators)
- Main structural issues:
 - **What more should we control?** (secondary cv's, y_2)
 - Pairing with manipulated variables (mv's u_2)



Objectives regulatory control layer

1. Allow for manual operation
2. Simple decentralized (local) PID controllers that can be tuned on-line
3. Take care of “fast” control
4. Track setpoint changes from the layer above
5. Local disturbance rejection
6. **Stabilization** (mathematical sense)
7. Avoid “drift” (due to disturbances) so system stays in “linear region”
 - “**stabilization**” (practical sense)
8. Allow for “slow” control in layer above (supervisory control)
9. Make control problem easy as seen from layer above

Implications for selection of y_2 :

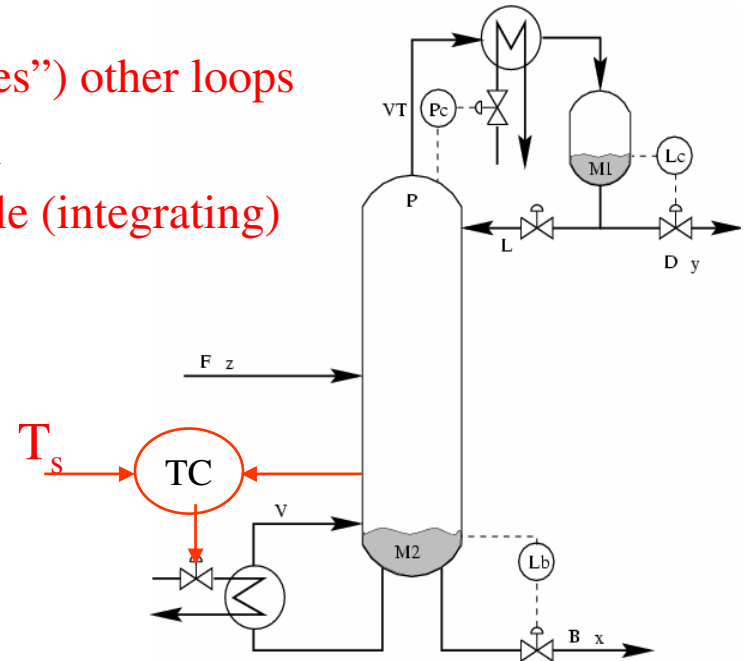
1. Control of y_2 “stabilizes the plant”
2. y_2 is easy to control (favorable dynamics)

Rules for selecting y_2 (and u_2 to be paired with y_2)

1. y_2 should be easy to measure
2. Control of y_2 stabilizes the plant
3. y_2 should have good controllability, that is, favorable dynamics for control
4. y_2 should be located “close” to a manipulated input (u_2) (follows from rule 3)
5. The (scaled) gain from u_2 to y_2 should be large
6. The effective delay from u_2 to y_2 should be small
7. Avoid using inputs u_2 that may saturate (should generally avoid saturation in lower layers)

Example: Distillation

- Primary controlled variable: $y_1 = c = x_D, x_B$ (compositions top, bottom)
- BUT: Delay in measurement of x + unreliable
- Regulatory control: For “stabilization” need control of (y_2):
 - Liquid level condenser (M_D)
 - Liquid level reboiler (M_B)
 - Pressure (p) **Disturbs (“destabilizes”) other loops**
 - Holdup of light component in column
(temperature profile) **Almost unstable (integrating)**



T-loop in bottom

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 - **Step 6: Supervisory control**
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Step 6. Supervisory control layer

- *Purpose*: Keep primary controlled outputs $c=y_1$ at optimal setpoints c_s
- Degrees of freedom: Setpoints y_{2s} in reg.control layer
- *Main structural issue*: **Decentralized or multivariable?**

Decentralized control (single-loop controllers)

Use for: Noninteracting process and no change in active constraints

- + Tuning may be done on-line
- + No or minimal model requirements
- + Easy to fix and change
- Need to determine pairing
- Performance loss compared to multivariable control
- Complicated logic required for reconfiguration when active constraints move

Multivariable control (with explicit constraint handling = MPC)

Use for: Interacting process and changes in active constraints

- + Easy handling of feedforward control
- + Easy handling of changing constraints
 - no need for logic
 - smooth transition
- Requires multivariable dynamic model
- Tuning may be difficult
- Less transparent
- “Everything goes down at the same time”

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Step 7. Optimization layer (RTO)

- *Purpose*: Identify active constraints and compute optimal setpoints (to be implemented by supervisory control layer)
- *Main structural issue*: Do we need RTO? (or is process self-optimizing)
- RTO not needed when
 - Can “easily” identify change in active constraints (operating region)
 - For each operating region there exists self-optimizing var

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- Conclusion / References

Conclusion

Procedure plantwide control:

- I. Top-down analysis** to identify degrees of freedom and primary controlled variables (look for self-optimizing variables)
- II. Bottom-up analysis** to determine secondary controlled variables and structure of control system (pairing).

More examples and case studies

- HDA process
- Cooling cycle
- Distillation (C3-splitter)
- Blending

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- ... + more.....

See home page of S. Skogestad:

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