



Norwegian University of  
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# Near-optimal control of batch processes

- by tracking of approximated sufficient conditions of optimality

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# What are we trying to do?

- Seek linear combination of measurements:  $c = Sy$
- Select the best matrix  $S$  using systematic procedure
- Allow separation of problem into:
  - Economic optimization
  - Control structure design via selecting  $S$
  - Implementation using for example PI controllers

# Implementation strategies

- Online methods
  - Measurements used to update model (or state estimate)
  - Requires online optimization
  - High computational load

- Offline methods
  - Measurements used directly in output feedback structure
  - Requires offline analysis
  - Low computational load

# Outline

- Goal: systematic control structure design
- Offline school of thought
- Remaining talk has three main parts:
  - a) Method development:  $c(t) = S(t)y(t)$
  - b) Linear example with stable plant
  - c) Nonlinear batch distillation example

# Mayer problem

$$\min_{u(t)} J(x(t_f))$$

$$\dot{x} = f(x, u, d), \quad x(0) = x_0.$$

$$y = h(x)$$

# Necessary and sufficient conditions for optimal control

$$H(t) = \lambda^T f(x, u, d), \quad \dot{\lambda} = -\partial H / \partial x$$

$$H_u = 0$$

$$\lambda(t_f) = \partial J / \partial x(t_f)$$

$$\delta^2 J = \int_0^{t_f} \begin{bmatrix} \delta x^T & \delta u^T \end{bmatrix} \begin{bmatrix} H_{xx} & H_{xu} \\ H_{ux} & H_{uu} \end{bmatrix} \begin{bmatrix} \delta x \\ \delta u \end{bmatrix} dt.$$

# Matching the Hamiltonian Matrix

$$\delta^2 J = \int_0^{t_f} \begin{bmatrix} \delta x^T & \delta u^T \end{bmatrix} \begin{bmatrix} H_{xx} & H_{xu} \\ H_{ux} & H_{uu} \end{bmatrix} \begin{bmatrix} \delta x \\ \delta u \end{bmatrix} dt.$$

- Form linear combinations:  $c = Sy = SCx$
- New approximation:

$$\delta^2 J \approx \frac{1}{2} \int_0^{t_f} (S(t)C(t)\delta x(t))^T Q_c (S(t)C(t)\delta x(t)) + \delta u^T H_{uu} \delta u dt$$

- Obviously, we must select  $R = H_{uu}$
- Also

$$(SC(t))^T Q_c(t) (SC(t)) = H_{xx}(t).$$

# Scaling matrix $Q_c$

- Variables “c” not known a priori
- Assume c’s independent
- Assume worst-case change in c to occur for all c’s at same d
- Scaling suggested:

$$Q_c(t) = \left( \text{diag} \left( S Q_y^{-1}(t) e \right) \right)^{-1}.$$

$$\Rightarrow (SC(t))^T \left( \text{diag}(S Q_y^{-1} e) \right)^{-1} (SC(t)) = H_{xx}(t).$$



# A linear-quadratic problem

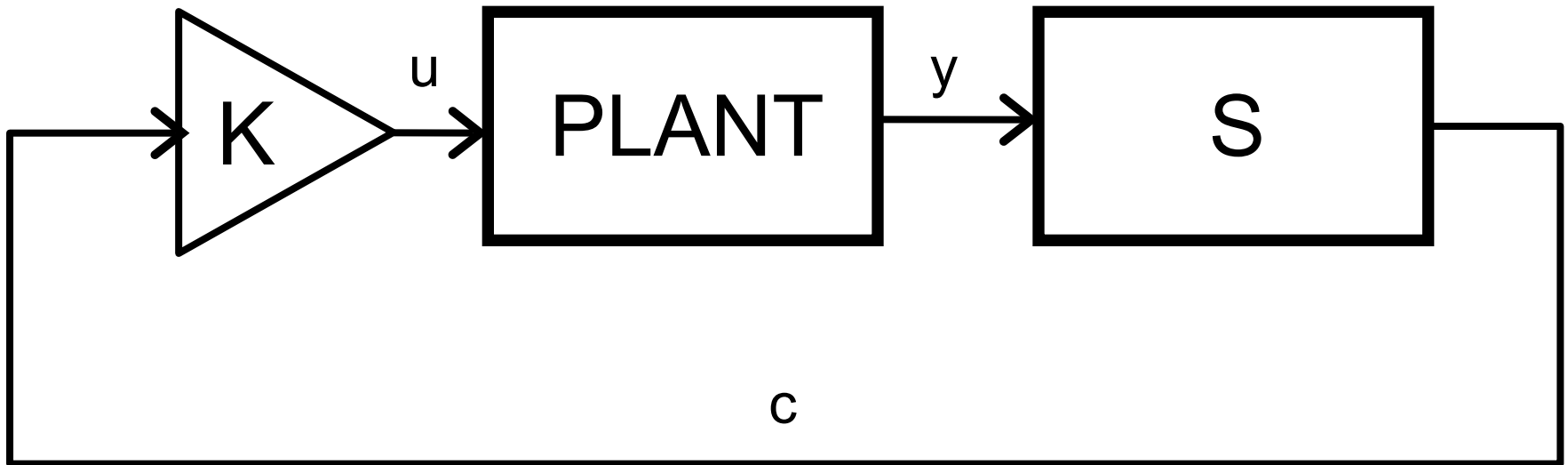
$$J = \frac{1}{2} \int_0^{t_f} \|x\|_2^2 + \|u\|_2^2 dt$$

$$\dot{x} = Ax + Bu,$$

$$y = Cx$$

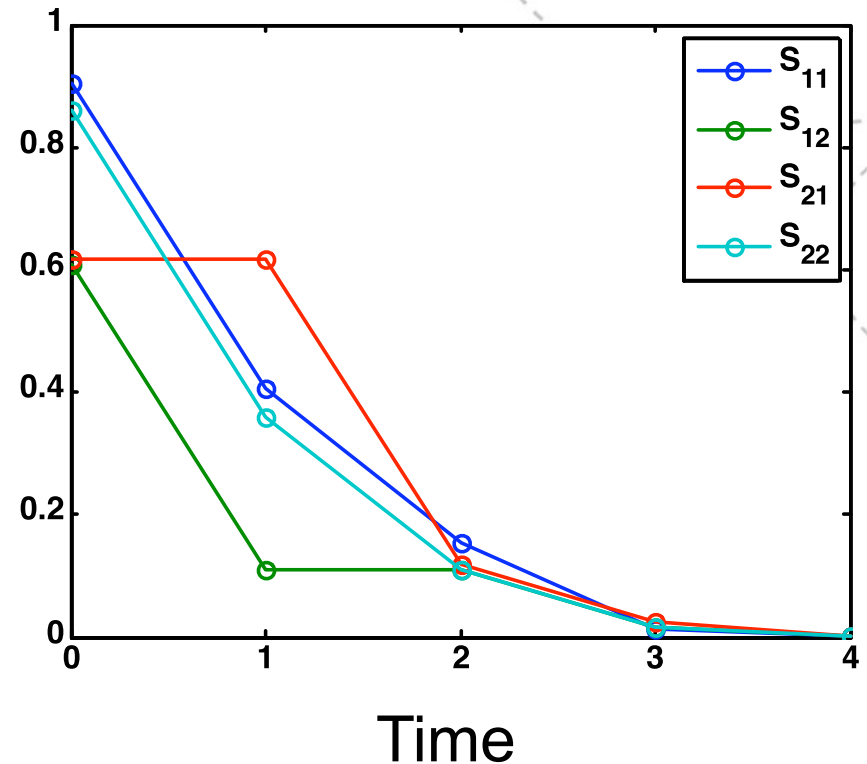
$$A = \begin{bmatrix} -3 & 1 & 1 \\ 0 & -2 & 1 \\ 0 & 0 & -3 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}, C_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

# Block diagram: output feedback

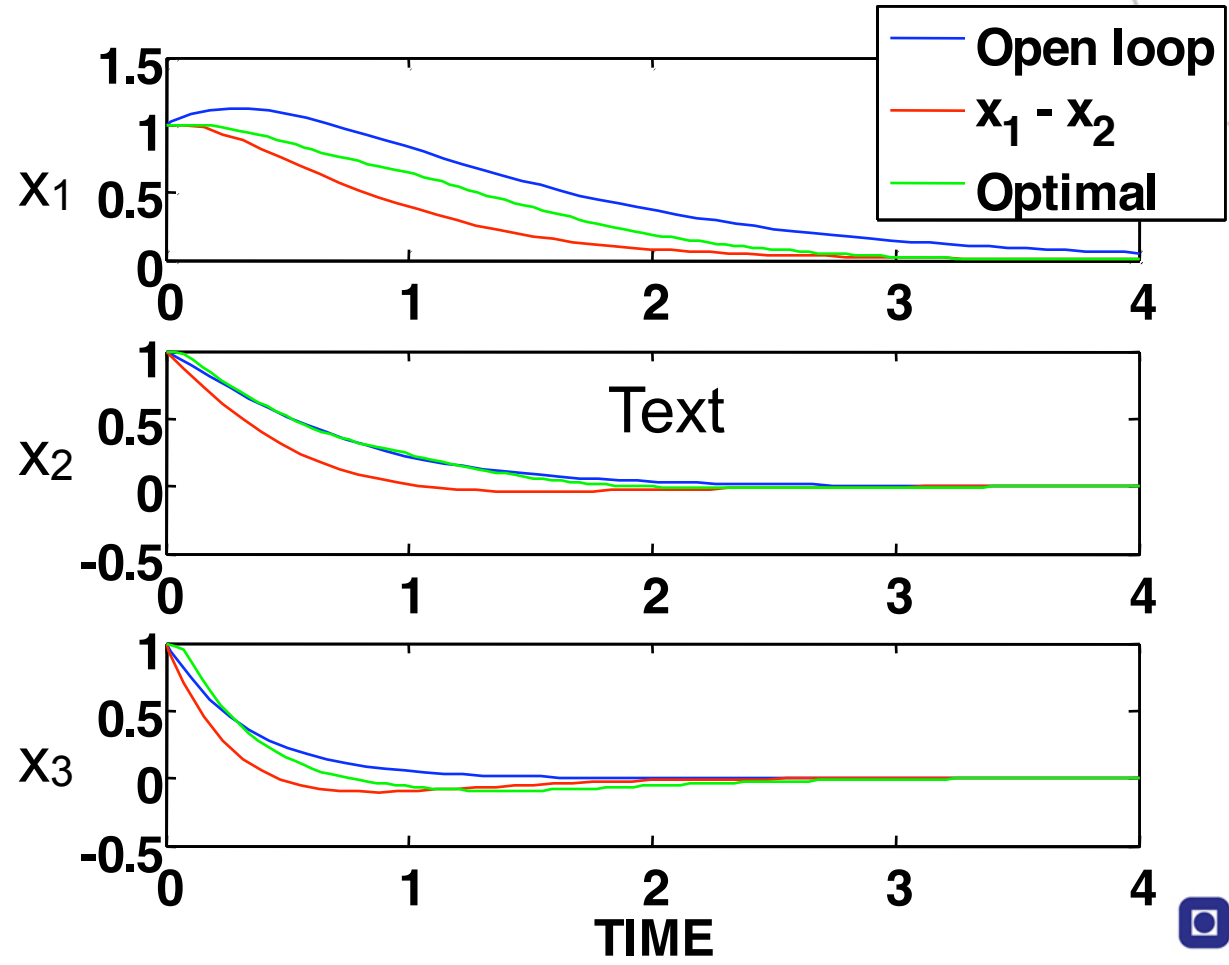


# Control strategy

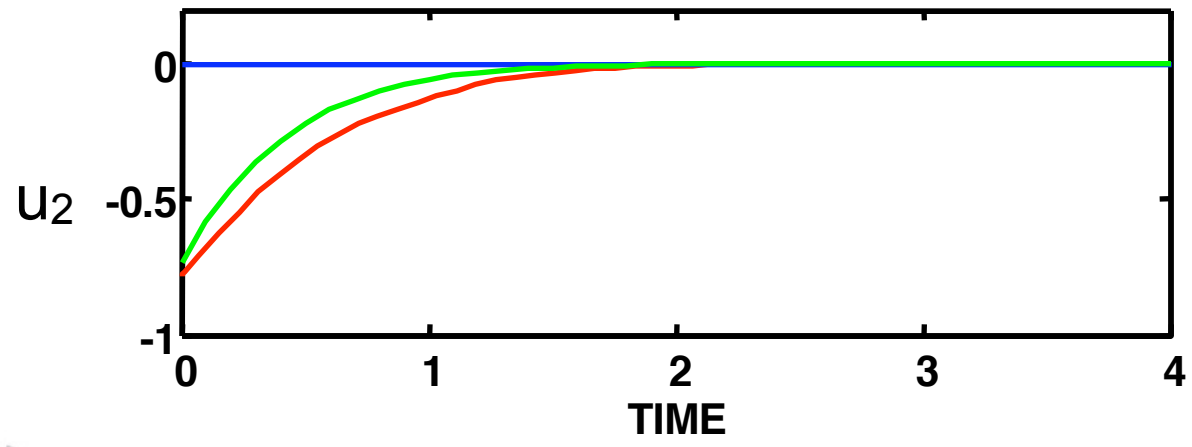
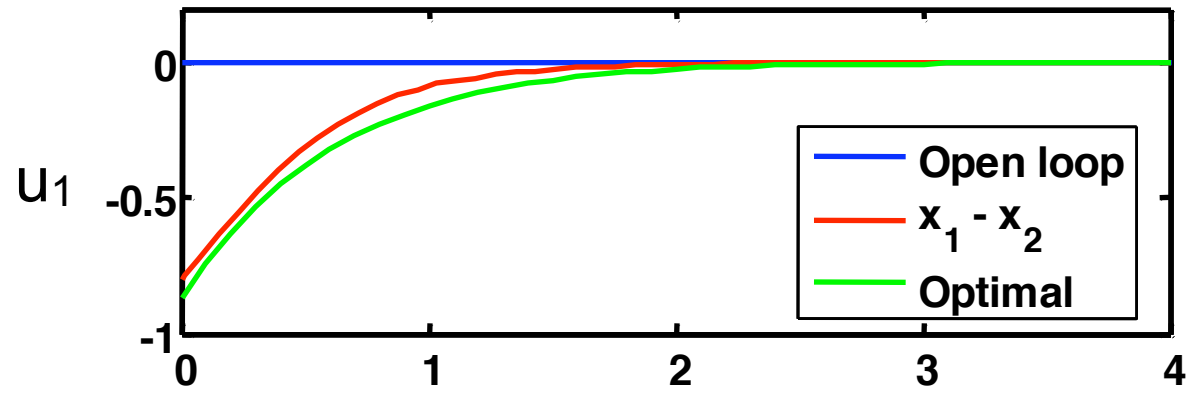
- $S$  computed for times  $\{0, 1, 2, 3, 4\}$
- Linear interpolation used between known points
- Static feedback with pure gain
- Tuned for  $x_0 = [1, 1, 1]$
- Look at responses for  $[1, 1, 1]$



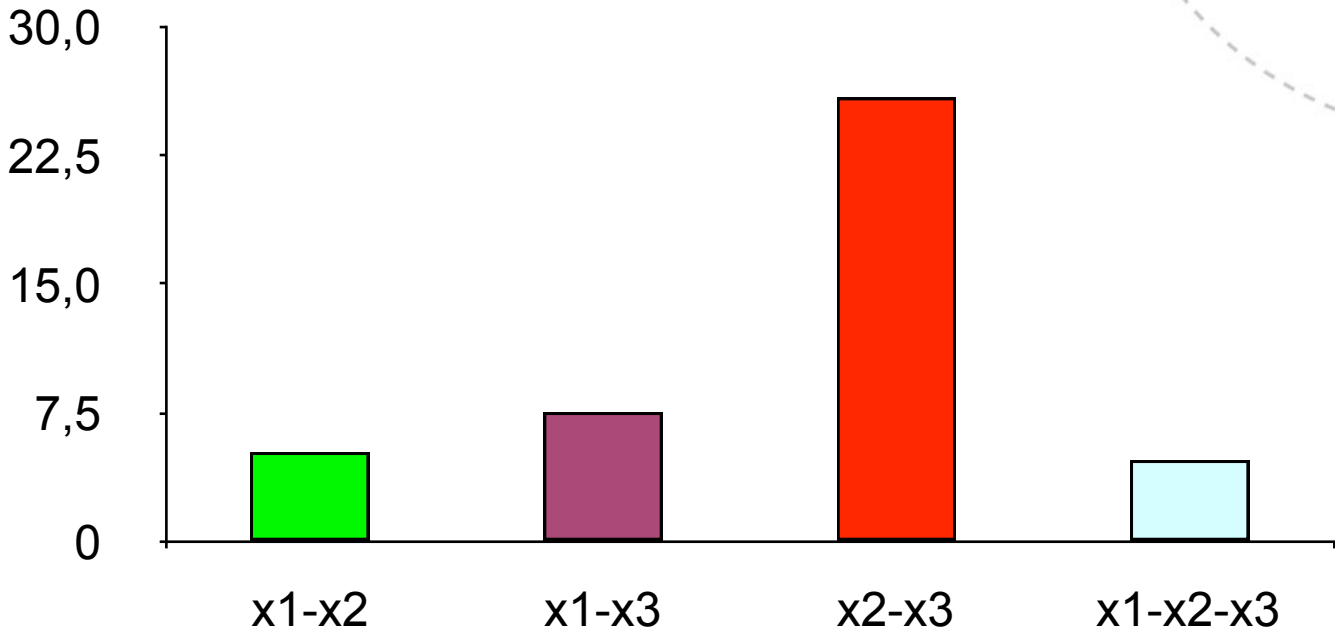
# State trajectories



# Input trajectories

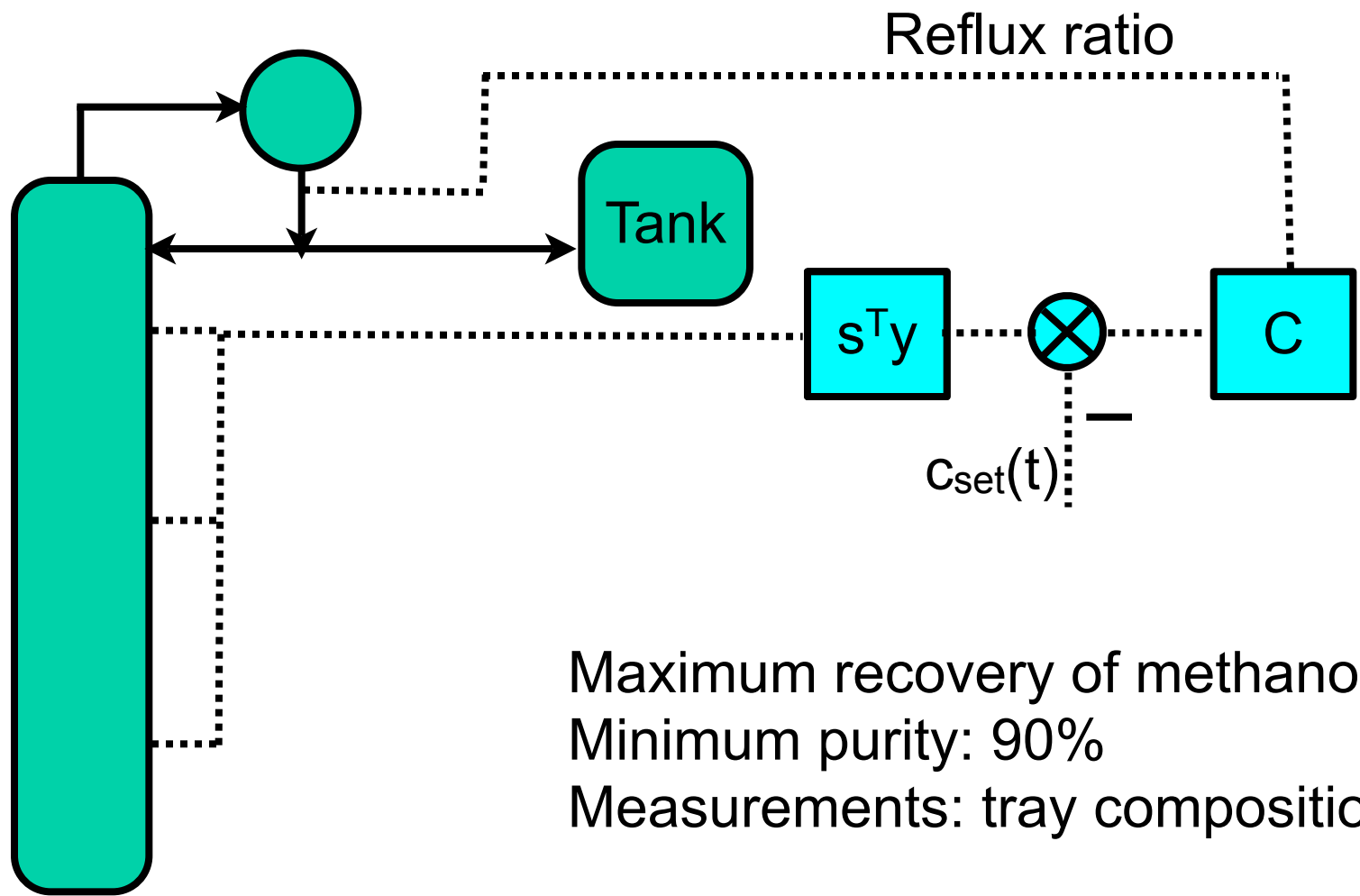


# Average losses (%)



Can improve loss by better controller tuning

# Batch distillation



Maximum recovery of methanol  
Minimum purity: 90%  
Measurements: tray compositions

MeOH/Allyl alcohol

# Computations

- Optimize for 49 vol%, 50 vol% and 51 vol% methanol to find  $Q_y$  ([www.gpops.org](http://www.gpops.org))
- Compute  $H_{xx}$
- Solve

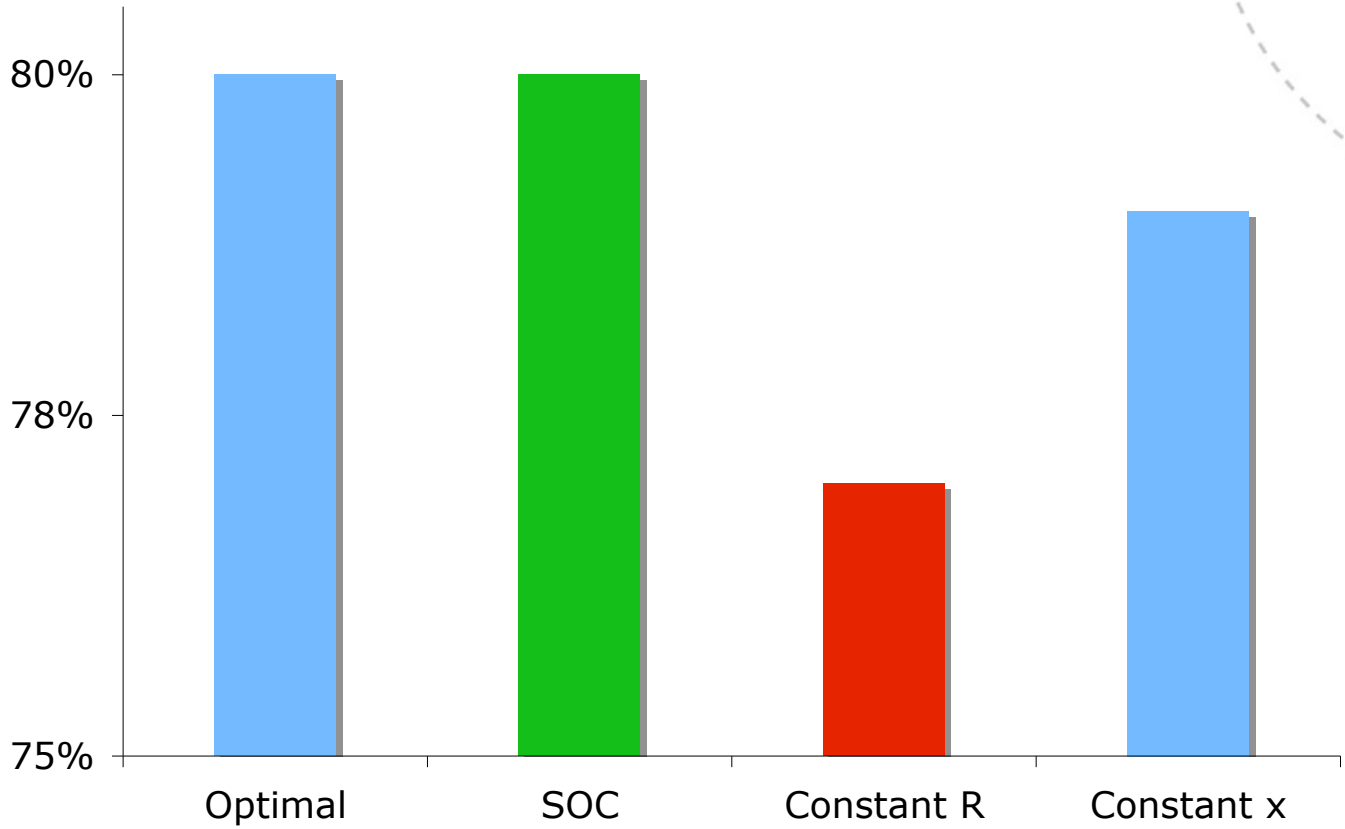
$$C^T S^T \text{diag}^{-1}(S Q_y^{-1} e) S C = H_{xx}$$

using direct search optimization for every hour

- Verify by simulations in closed loop



### Methanol Recovery



- Algorithm for control structure design
- Applicable to both linear and nonlinear systems
- A step on the way to automatic control structure design

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