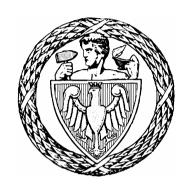
# Institute of Control and Computation Engineering Warsaw University of Technology





# Predictive controllers integrated with economic optimization tolerating actuator faults: application to a nonlinear plant

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#### Plan of presentation

- 1. Introduction
- 2. Control system
  - 2.1. Predictive algorithms
  - 2.2. Economic optimization problem
  - 2.3. Predictive algorithms integrated with economic optimization
- 3. Actuator faults handling
- 4. Example control system
- 5. Simulation experiments
  - 5.1. The same number of manipulated and controlled variables
  - 5.2. The system with additional manipulated variable
- 6. Summary

#### Introduction

- *Algorithms*: Model Predictive Control algorithms integrated with Economic Optimization (MPCEO)
- Aim: Continuation of the control system operation till the failure is fixed
- Assumption: Methods of detection and isolation of actuator faults are available, measurement of the actuator output, in particular
- Remark: Output constraints are often important for safety and economic effectiveness of the process

#### The idea of the predictive control

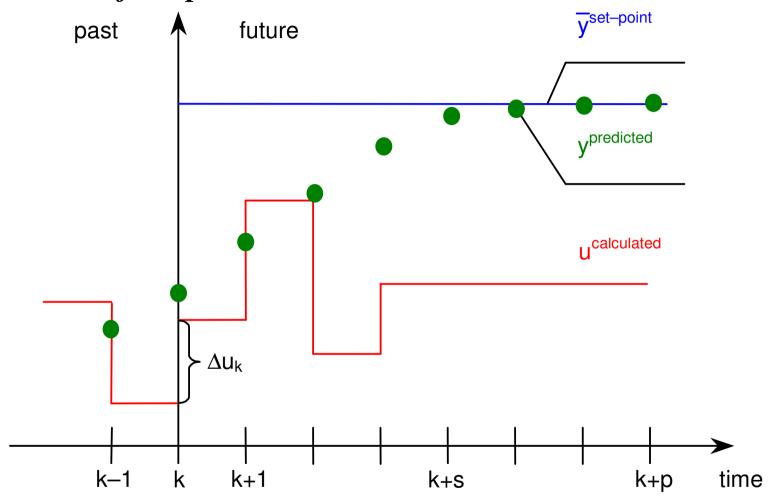


Fig. 1. Idea of predictive control; p – prediction horizon, s – control horizon,  $\Delta u_k$  – control signal change at current iteration

#### Numerical predictive control algorithms

Following problem is solved at each iteration:

$$\min_{\Delta u} \left\{ J_{MPC} = \sum_{j=1}^{n_y} \sum_{i=1}^p \kappa_j \cdot \left( \overline{y}_k^j - y_{k+i|k}^j \right)^2 + \sum_{j=1}^{n_u} \sum_{i=0}^{s-1} \lambda_j \cdot \left( \Delta u_{k+i|k}^j \right)^2 \right\}$$

subject to the constraints:

$$\Delta u_{min} \leq \Delta u \leq \Delta u_{max}$$

$$u_{min} \leq u \leq u_{max}$$

$$y_{min} \leq y \leq y_{max}$$

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subject to the constraints:

$$\Delta u_{min} \leq \Delta u \leq \Delta u_{max}, u_{min} \leq u \leq u_{max}, y_{min} \leq y \leq y_{max},$$

- In a nonlinear case, in order to avoid problems connected with general nonlinear optimization, effective algorithms with model linearization and quadratic optimization are used
- A few such algorithms are available, so the algorithm most suitable for a given nonlinear plant can be selected and a compromise between control performance and computation demand can be achieved

#### Economic optimization problem

$$\min_{\overline{y}} J_E(\overline{y}, \overline{u})$$

subject to

$$\overline{u}_{\min} \le \overline{u} \le \overline{u}_{\max}$$

$$\overline{y}_{\min} + \overline{r}_{\min} \le \overline{y} \le \overline{y}_{\max} - \overline{r}_{\max}$$

$$\overline{y} = F(\overline{u}, \widetilde{w})$$

 $F: \mathfrak{R}^{n_u} \times \mathfrak{R}^{n_w} \to \mathfrak{R}^{n_y}$  is a steady-state plant model (u - inputs, y - outputs, w - disturbances)

• Precise nonlinear steady–state plant model

#### Classical multilayer control system structure

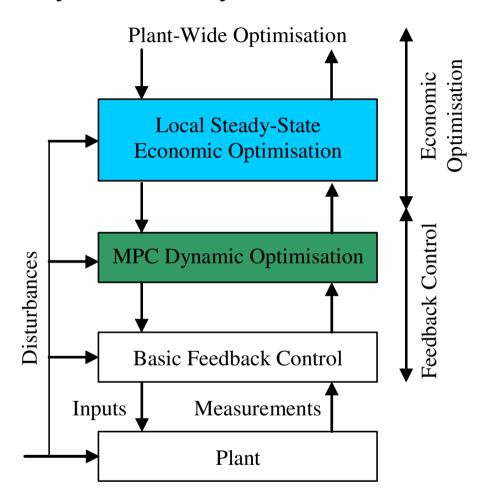


Fig. 2. Hierarchical control system structure with MPC advanced control layer

### MPC Integrated with Economic Optimization (MPCEO)

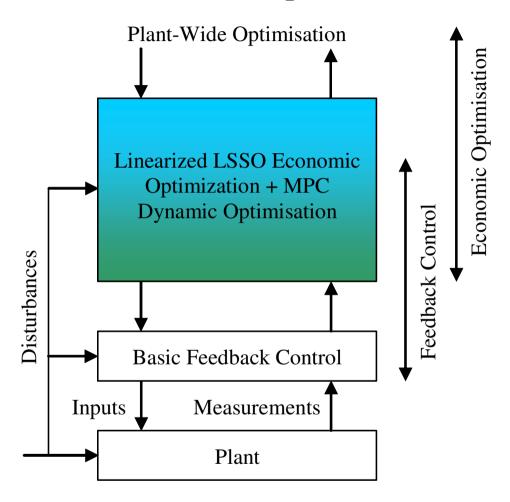


Fig. 3. Control system structure with MPC integrated with economic optimisation

#### MPC Integrated with Economic Optimization (MPCEO)

- One of the methods to cope with disturbances changing quickly comparing to the dynamics of the control plant
- The steady–state control plant model is linearized
- Only one quadratic optimization problem must be solved at each iteration
- The control system structure is simplified
- The economic optimization is performed more often than in the classic hierarchical approach

#### MPC Integrated with Economic Optimization (MPCEO)

$$\min_{\Delta u, \bar{y}} J_{MPC}(\bar{y}, \Delta u) + \gamma \cdot J_{E}(\bar{y}, \bar{u})$$

subject to

$$\Delta u_{min} \leq \Delta u \leq \Delta u_{max}$$

constraints from economic  $u_{min} \le u \le u_{max}$ , optimization problem

economic optimization performance function

$$\overline{u}_{\min} \leq \overline{u} \leq \overline{u}_{\max}$$

$$\overline{y}_{\min} + \overline{r}_{\min} \leq \overline{y} \leq \overline{y}_{\max} - \overline{r}_{\max}$$

$$\overline{y} = F(u(k-1), \widetilde{w}) + H(k)(\overline{u} - u(k-1))$$

linearization of the steady-state nonlinear model F

#### Actuator faults handling

• A set of equality constraints is added to the algorithm after fault detection

$$\Delta u_{k+i|k}^{m} = 0, i = 0, ..., s-1$$

m – number of control signal affected by the failure

- Application is relatively easy
- Elimination of some decision variables

#### Actuator faults handling

• Equality constraint is added

$$\overline{u}^m = u_{bl}^m$$

m – number of control signal affected by the failure  $u_{bl}^{m}$  – output of the actuator

- Easy application
- The steady-state model is in practice modified
- Measurement of the actuator output is often available

#### Control plant (evaporator system\*)

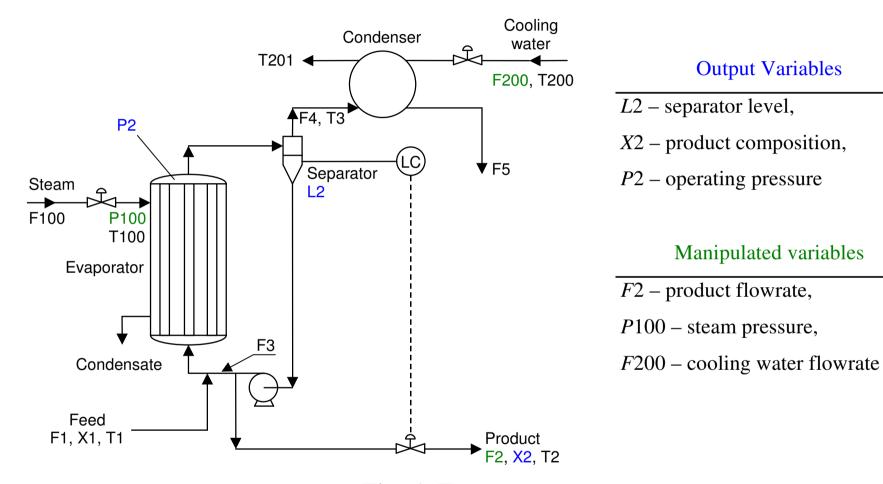


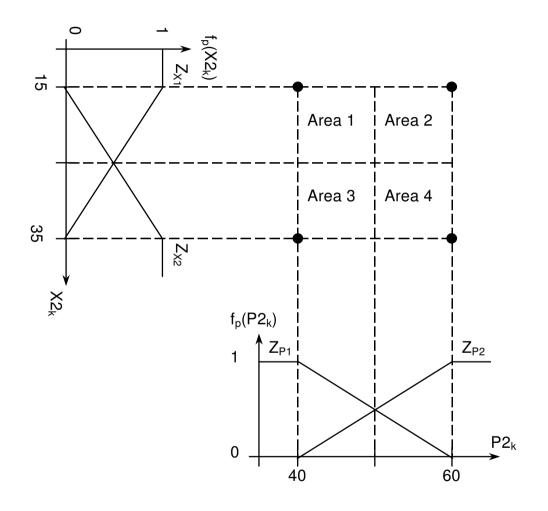
Fig. 4. Evaporator system

\* R.B. Newell, P.L. Lee: Applied process control – a case study; Prentice Hall, 1989

#### The MPCEO algorithm

- The manipulated variables are: steam pressure P100 and cooling water flow F200
- The controlled variables are: product composition *X*2 and pressure in the evaporator *P*2
- Measured disturbance *F*1 (feed flow)
- Based on the fuzzy DMC predictive algorithm,

## The MPCEO algorithm



Parameters:

$$\kappa_{\text{P2}} = \kappa_{\text{X2}} = 1,$$
 $\lambda_{\text{P100}} = \lambda_{\text{F200}} = \lambda_{F3} = 0.1$ 
 $p = 100, s = 10$ 

Fig. 5. Membership functions of the fuzzy MPCEO controller

#### The MPCEO algorithm

• Economic performance index (cost of production)

$$J_E = c_1 \cdot \overline{P}100 - c_2 \cdot \overline{F}2$$

• Constraints put on manipulated variables:

$$0 \text{ kPa} \le P100 \le 400 \text{ kPa}, 0 \text{ kg/min} \le F200 \le 400 \text{ kg/min},$$

• The product should fulfill purity criteria:

$$25 \% \le X2$$

- The appropriate soft constraints were put on the predicted *X*2 composition values
- The constraint put on  $\overline{X}$ 2 set—point was as follows

$$\overline{X}2_{\min} + \overline{r}_{\min}^{X2} \le \overline{X}2$$

$$\overline{X}2_{\min} = X2_{\min} = 25\%, \quad \overline{r}_{\min}^{X2} = 1\%$$

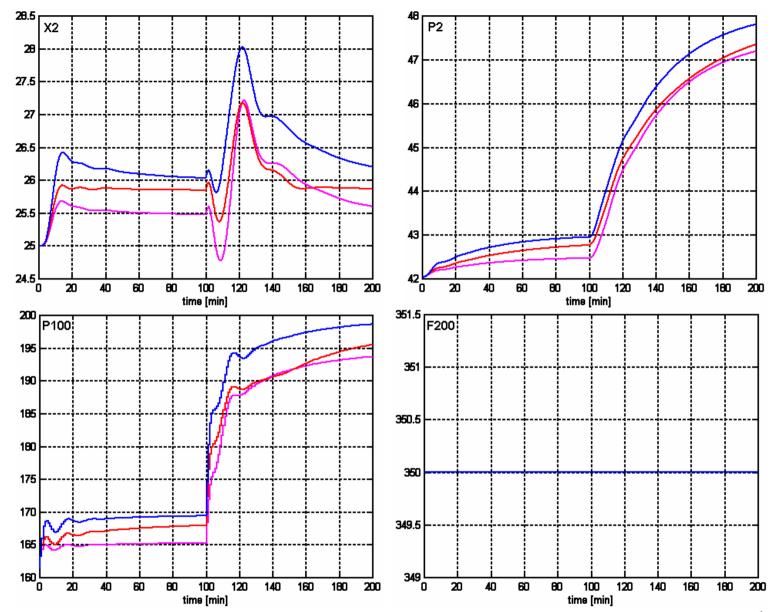


Fig. 6. Responses of the control system to a step decrease of F1 disturbance in the  $100^{th}$  minute; F200 actuator fault: not taken into consideration at all,

taken into consideration, additionally the equality constraint put on *X*2 set-point was added; above: output signals *X*2 and *P*2, below: control signals *P*100 and *F*200

#### P100 actuator blockade

- Optimizing procedure returned the message that there is no admissible solution
- Why there is no solution found?
- In order to answer this question steady—state characteristics should be analyzed

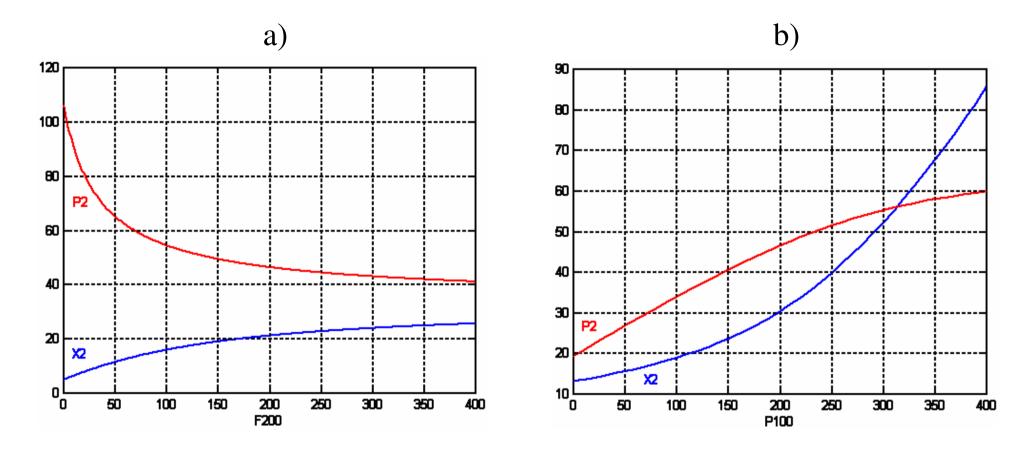


Fig. 7. Steady–state characteristics a) X2(F200) i P2(F200), b) X2(P100) i P2(P100), of the plant with blocked actuator of the manipulated variable a) P100, b) F200

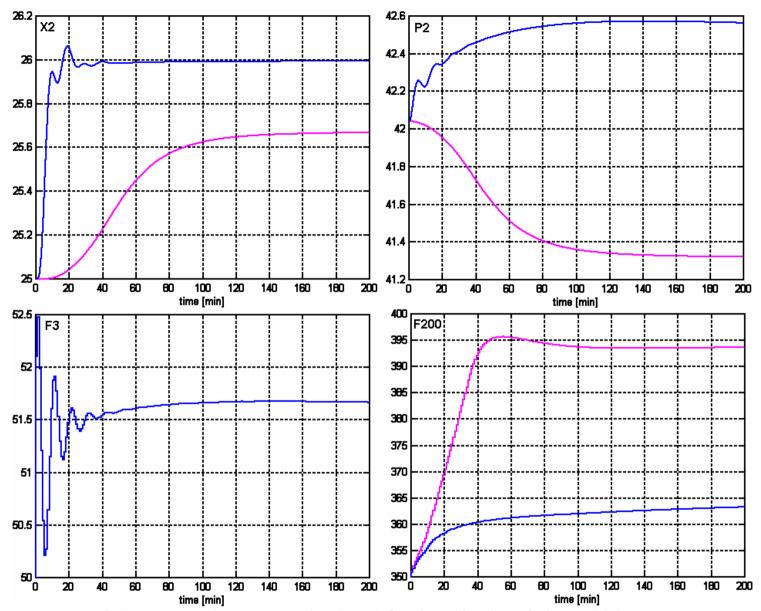


Fig. 8. Responses of the control system obtained for blockade of the *P*100 actuator: not taken into consideration, taken into consideration in the control system with additional manipulated variable above: output signals *X*2 and *P*2, below: control signals *F*3 and *F*200

#### **Summary**

- Effective and relatively little complicated method of actuator fault toleration in control systems with MPCEO algorithms and output constraints
- The method: equality constraints added to the optimization problem solved by the algorithm
- The method can be used in the MPCEO algorithm with either linear or nonlinear dynamic control plant model
- Further improvement of control system operation can be obtained, when the additional manipulated variable is available