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Productivity optimization of cultures of *Saccharomyces cerevisiae* using a simple robust control strategy

Laurent Dewasme, Frederic Renard, Alain Vande Wouwer
Automatic Control Laboratory
University of Mons

16/03/2010

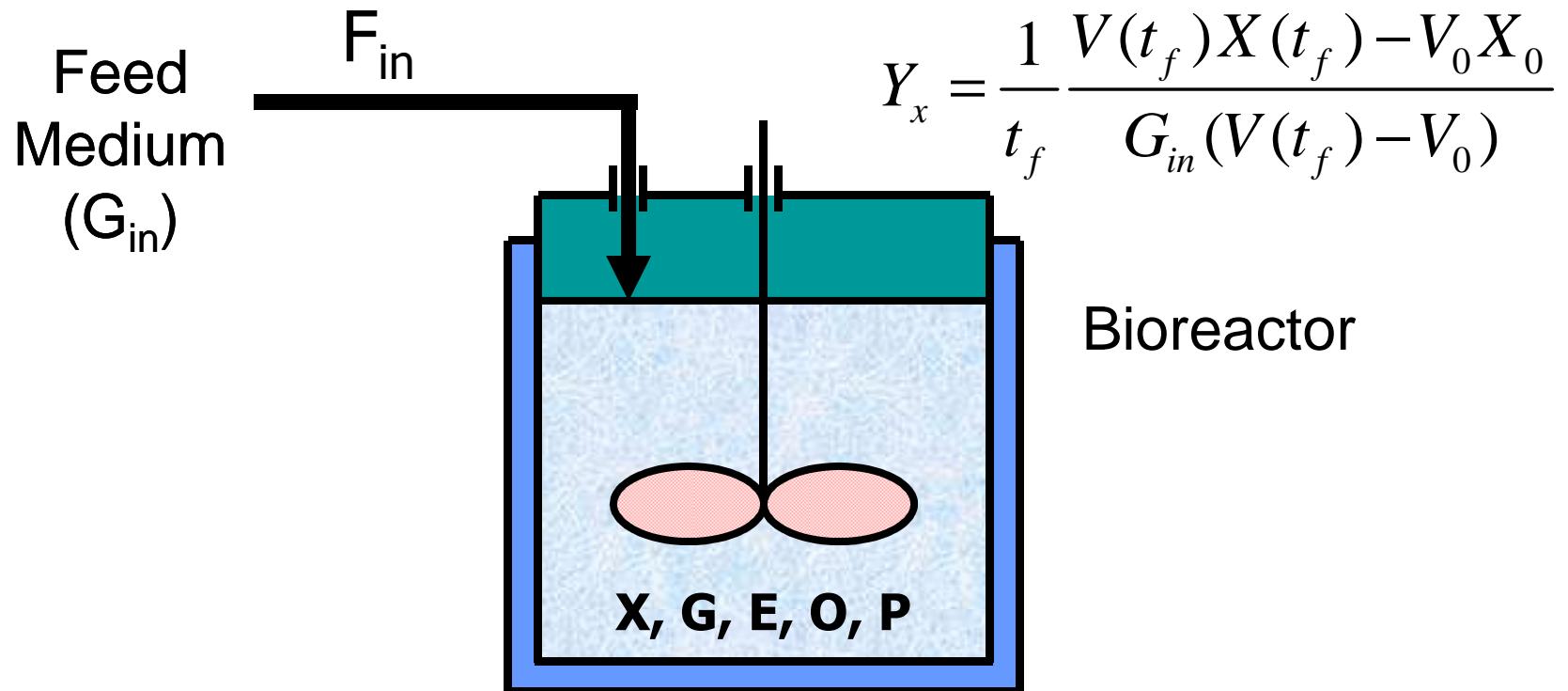
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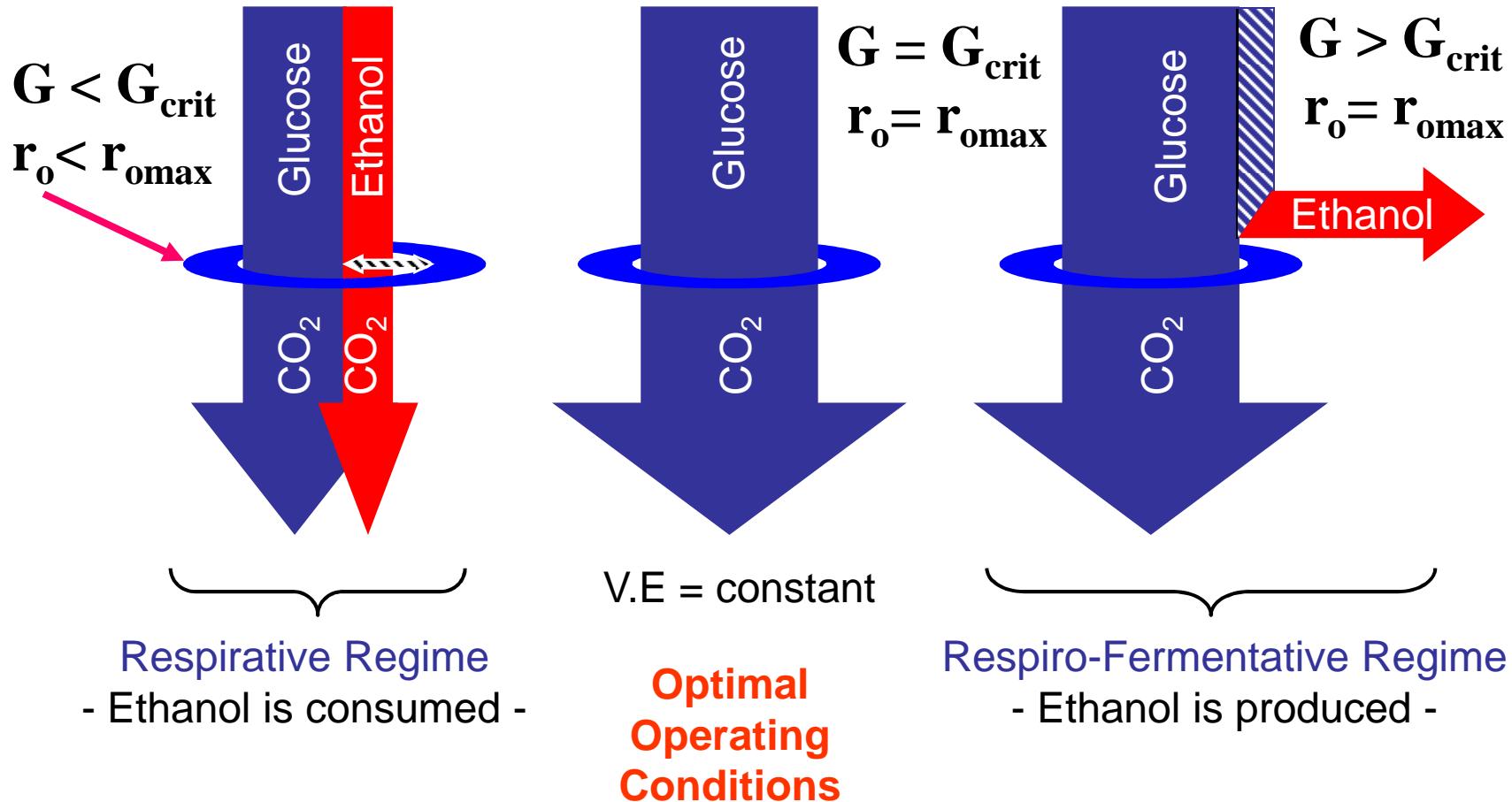
Introduction

- ❑ Process : yeast fed-batch culture
- ❑ Aim : maximize biomass productivity



Introduction

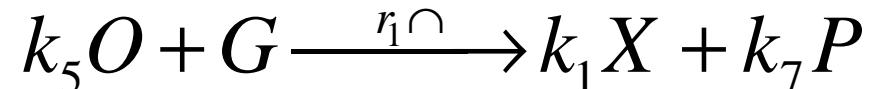
☐ Yeast limited respiratory capacity



Introduction

□ Reaction scheme

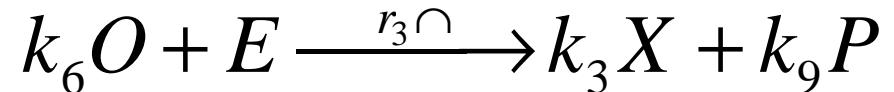
- Glucose oxidation:



- Glucose fermentation:



- Ethanol oxidation:



$$\begin{bmatrix} \dot{X} \\ \dot{G} \\ \dot{E} \\ \dot{O} \\ \dot{P} \end{bmatrix} = \begin{bmatrix} k_1 & k_2 & k_3 \\ -1 & -1 & 0 \\ 0 & k_4 & -1 \\ -k_5 & 0 & k_6 \\ k_7 & k_8 & k_9 \end{bmatrix} \cdot \begin{bmatrix} r_1 \cdot X \\ r_2 \cdot X \\ r_3 \cdot X \end{bmatrix} - D \cdot \begin{bmatrix} X \\ G \\ E \\ O \\ P \end{bmatrix} + \begin{bmatrix} 0 \\ G_{in} \cdot D \\ 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ OTR \\ -CTR \end{bmatrix}$$

(r₅ = 0)

SSSS : respiration fermentative régime

Outline

→ Model linearization

- Feed-Oxygen transfer
- Feed-Ethanol transfer

- Simple design of two adaptive robust controllers
- Comparison of the two control strategies
- Experimental results
- Conclusions

Model linearization

Feed-Oxygen Model

□ Main assumption:

$$\begin{cases} \frac{d(VG)}{dt} \approx 0 \\ q_s = r_1 + r_2 \\ q_o = k_5 r_1 + k_6 r_3 \end{cases} \rightarrow X = \frac{F_{in} G_{in}}{q_s V}$$

and $\frac{dO}{dt} = -\left(\frac{q_o}{q_s} G_{in} + O\right) \frac{F_{in}}{V} + k_L a (O_{sat} - O)$

→ Defining a nominal trajectory around which the system is linearized: $F_{in}^*(t)$, $V^*(t)$ so that $O^*(t) = O_{ref}$

$$\frac{q_o^*}{q_s^*} \approx k_5 \quad (\text{considering only respiration on glucose so that } r_2 \text{ and } r_3 \text{ are equal to 0})$$

Model linearization

Feed-Oxygen Model

- Considering $O^*(t) = O_{ref}$:

$$D^* = \frac{F_{in}^*}{V^*} = \frac{OTR^*}{k_5 G_{in} + O^*}$$

which is constant.

If the variation of V is neglected, a Taylor series expansion limited to the first order around the nominal trajectory allows to write:

$$\frac{d\delta O}{dt} = -\left(\frac{k_5 G_{in} + O^*}{V^*}\right)\delta F_{in} - (k_L a + D^*)\delta O \quad (*)$$

$$\delta O = O - O^*$$

$$\delta F_{in} = F_{in} - F_{in}^*$$

and, assuming that D is negligible in comparison with $k_L a$, (*) becomes:

$$\frac{d\delta O}{dt} = -\left(\frac{k_5 G_{in} + O^*}{V^*}\right)\delta F_{in} - k_L a \delta O \quad (5)$$



Model linearization

Feed-Oxygen Model

- As the principal variation of the volume is due to the added feed-rate:

$$\frac{dV}{dt} = F_{in}^*$$

the solution of this differential equation is:

$$F_{in}^*(t) = D^* V_0 \exp(D^* t)$$

Putting all together

$$\frac{d(O - O^*)}{dt} = -\left(\frac{k_5 G_{in} + O^*}{V^*}\right)[F_{in} - D^* V_0 \exp(D^* t)] - k_L a (O - O^*)$$

whose Laplace transform is given by:

$$O(p) = -\frac{\frac{1}{k_L a} \frac{k_5 G_{in} + O^*}{V^*}}{1 + \frac{1}{k_L a} p} \left[F_{in}(p) - \frac{D^* V_0}{p - D^*} \right]$$



Model linearization

Feed-Oxygen Model

- Finally, the discrete-time linear model is given by:

$$O(k) = \frac{b}{1-a} q^{-1} [F_{in}(k) - d_i(k)]$$

$$d_i(k) = \frac{c}{1-\gamma} q^{-1} \delta(k)$$

- An equivalent Feed-ethanol model can be obtained considering another optimal trajectory $E^*(t)=E_{ref}$ and assuming again that:

$$\frac{d(VG)}{dt} \approx 0 \Leftrightarrow r_2 XV = F_{in}G_{in} - r_1 XV$$

Simple manipulations lead to:

$$E(k) = \frac{b}{1-a} q^{-1} [F_{in}(k) - d_i(k)]$$

$$d_i(k) = \frac{c}{1-\gamma} q^{-1} \delta(k)$$

Model linearization

Feed-Ethanol and feed-oxygen models

Parameters	Oxygen model model:	Ethanol model	
	Both regimes $\delta(k)$	Respiro-fermentative	Respirative
a	$\exp(-k_L a T_s)$	1	1
b	$\frac{1}{1 - \frac{k_5 G_{in}}{k_L q} V^*} + O^*(1-a)$	$T_s \frac{k_4 G_{in} - E^*}{V^*}$	$T_s \frac{\frac{k_5}{k_6} G_{in} - E^*}{V^*}$
c	$V_o \alpha_i(k) \gamma$	$\frac{k_4 r_1^*}{k_4 G_{in} - E^*} V_0 X_0$	$\frac{k_4 \frac{r_o^*}{k_6} V_0 X_0}{\frac{k_5}{k_6} G_{in} - E^*}$
$F_{in}(k)$	$- \exp(D^* T_s)$ $+ \exp(D^* T_s)$	$b q^{-1}$ $1 - a q^{-1}$	$O(k) \text{ or } E(k)$ $\exp(\mu T_s)$ $\exp(\mu T_s)$

Outline

- Model linearization
- ➔ Simple design of an adaptive robust controller
 - RST controller
 - Adaptation scheme
 - Some simulation results
 - Model improvements
- Comparison of the two control strategies
- Experimental results
- Conclusions

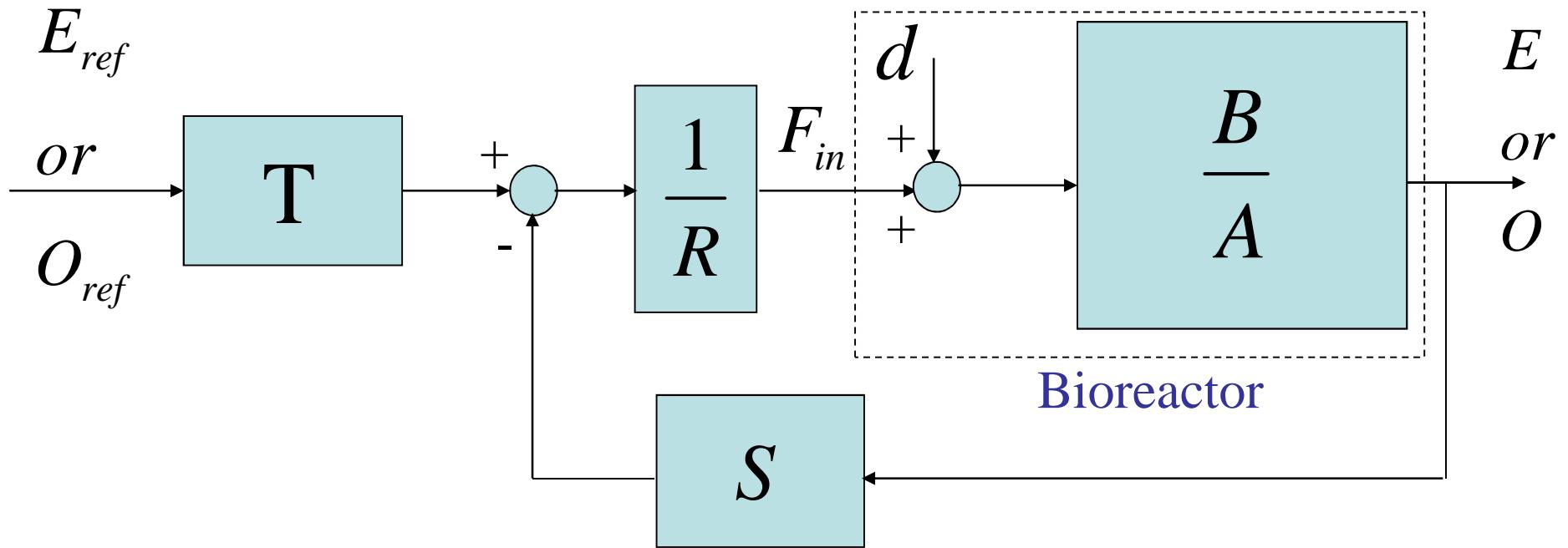
Simple design of an adaptive robust controller

- Exponential disturbance.
- Variation of the kinetic parameter:

$$\gamma = \exp(\mu T_s) \text{ or } \exp(D^* T_s)$$

- Variation of the gain b according to the operating regime.
- The controller must be able to:
 - Reject an exponential disturbance (Internal model principle).
 - Robustify the system against uncertainties on γ and b .

RST controller



- Controller:

$$R(q^{-1})F_{in}(k) = -S(q^{-1})E(k) + T(q^{-1})E_{ref}(k)$$

Simple design of an adaptive robust controller

- Polynomials R,S,T computed by a pole-placement procedure.
- Unstable pole in the polynomial $D = 1 - \gamma q^{-1}$
→ must be included in the R polynomial.

- Closed loop:

$$E = \frac{BT}{AR + BS} E_{ref} + \frac{BR}{AR + BS} d$$

- Reference model: $A_m D_m R' + B_m S_m$

$$H_m = \frac{B_m(q^{-1})A_m(1)}{A_0 A_m + B_m(1)A_m(q^{-1})}$$

Robustness

Tracking behaviour
Tracking behaviour

Simple design of an adaptive robust controller

- 1) Choose the tracking dynamics by choosing the A_m zeros.
- 2) Tune the robust behaviour by choosing the A_0 zeros.
- 3) Solve the diophantine equation:

$$A D R' + BS = A_0 A_m$$

so as to obtain the R' and S polynomials.

- 4) T is given by:
$$T = A_0 \frac{A_m(1)}{B(1)}$$

Adaptation scheme

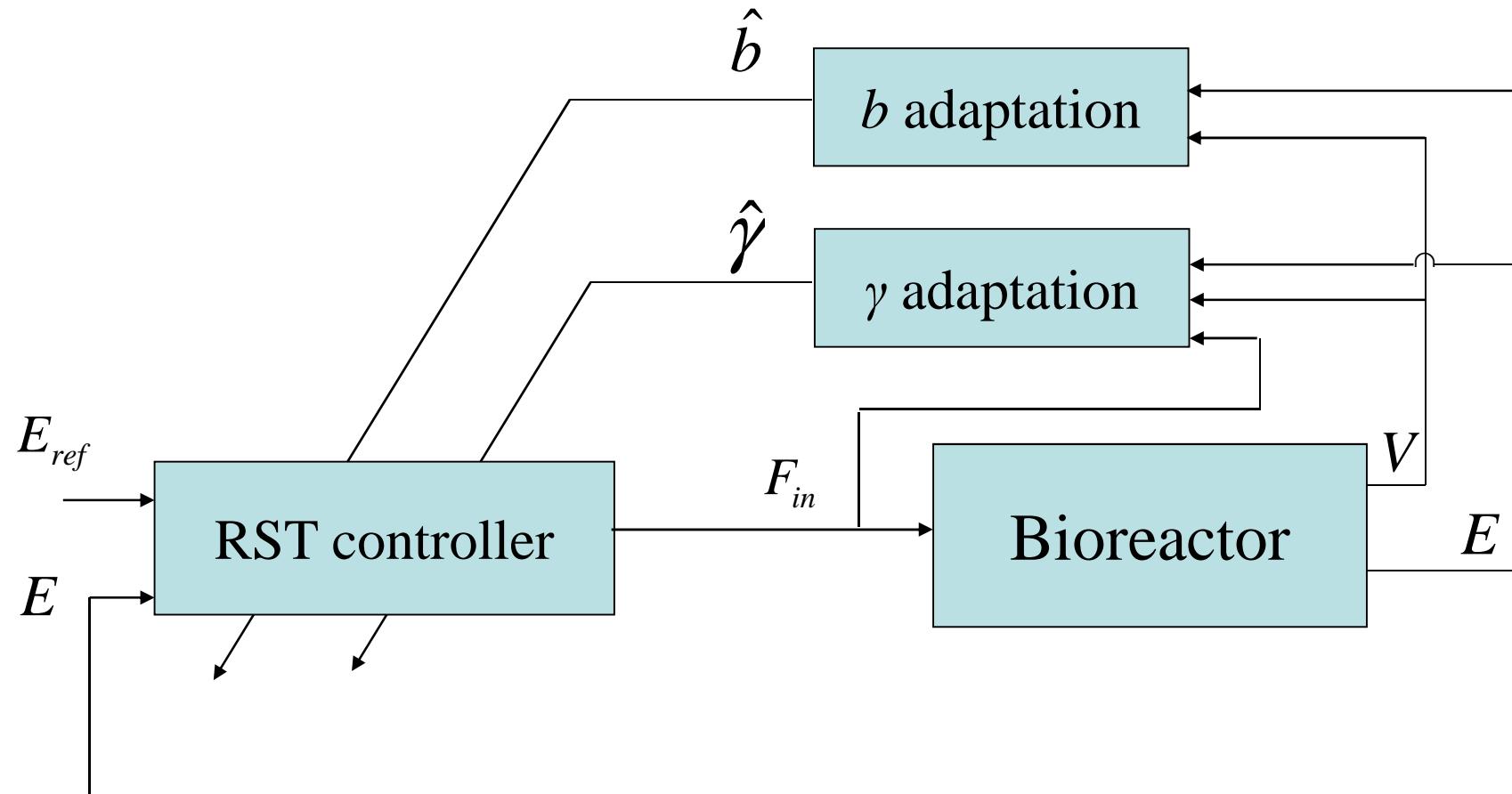
- Sometimes, μ varies during the culture and must be adapted (RLS algorithm):

$$\begin{aligned}\hat{d}(k) &= Ay(k) - Bu(k) \\ D(q^{-1})\hat{d}(k) &= 0\end{aligned}\qquad \longrightarrow \qquad \begin{aligned}\hat{d}(k) &= \hat{d}(k-1) \gamma \\ \gamma &= \exp(\mu T_s)\end{aligned}$$

- The gain b is updated with the volume V and the ethanol concentration E measurements:

$$b(k) = \frac{k_4 G_{in} - E(k)}{V(k)} T_s$$

Adaptation scheme



Case study: ethanol regulation

□ Initial and operating conditions:

- $X_0 = 0.4 \text{ g/l}$
- $E_0 = 0.5 \text{ g/l}$
- $G_0 = 0.0125 \text{ g/l}$
- $V_0 = 5 \text{ l}$
- $G_{in} = 500 \text{ g/l}$
- $E_{in} = 0.1 \text{ g/l}$
- $E_{ref} = 1 \text{ g/l}$

□ Controller parameters:

- $a_1 = 0.7$
- $b_0 = (k_4 G_{in} - E_0) / V_0$

Controller design

1) $A_m = 1 - 0.9 q^{-1}$

2) $A_0 = 1 - 0.7 q^{-1}$

3) Solving the diophantine equation:

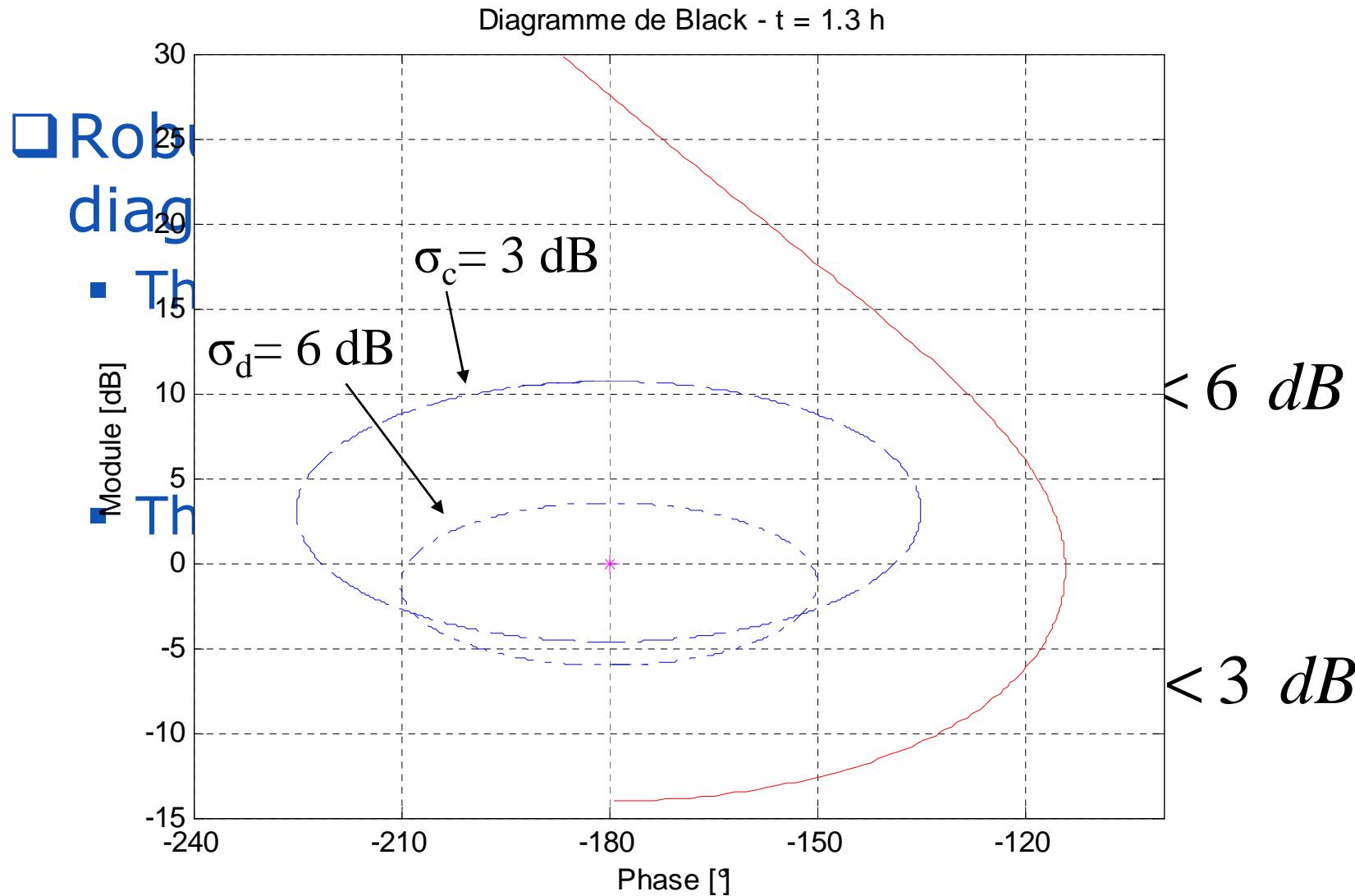
$$A \ D \ R' + BS = A_0 A_m$$

We obtain $R' = 1$

$$S = s_0(\gamma, b) + s_1(\gamma, b) q^{-1}$$

4) $T = (1 - 0.7 q^{-1}) A_m(1)/b$

Controller design



Controller design

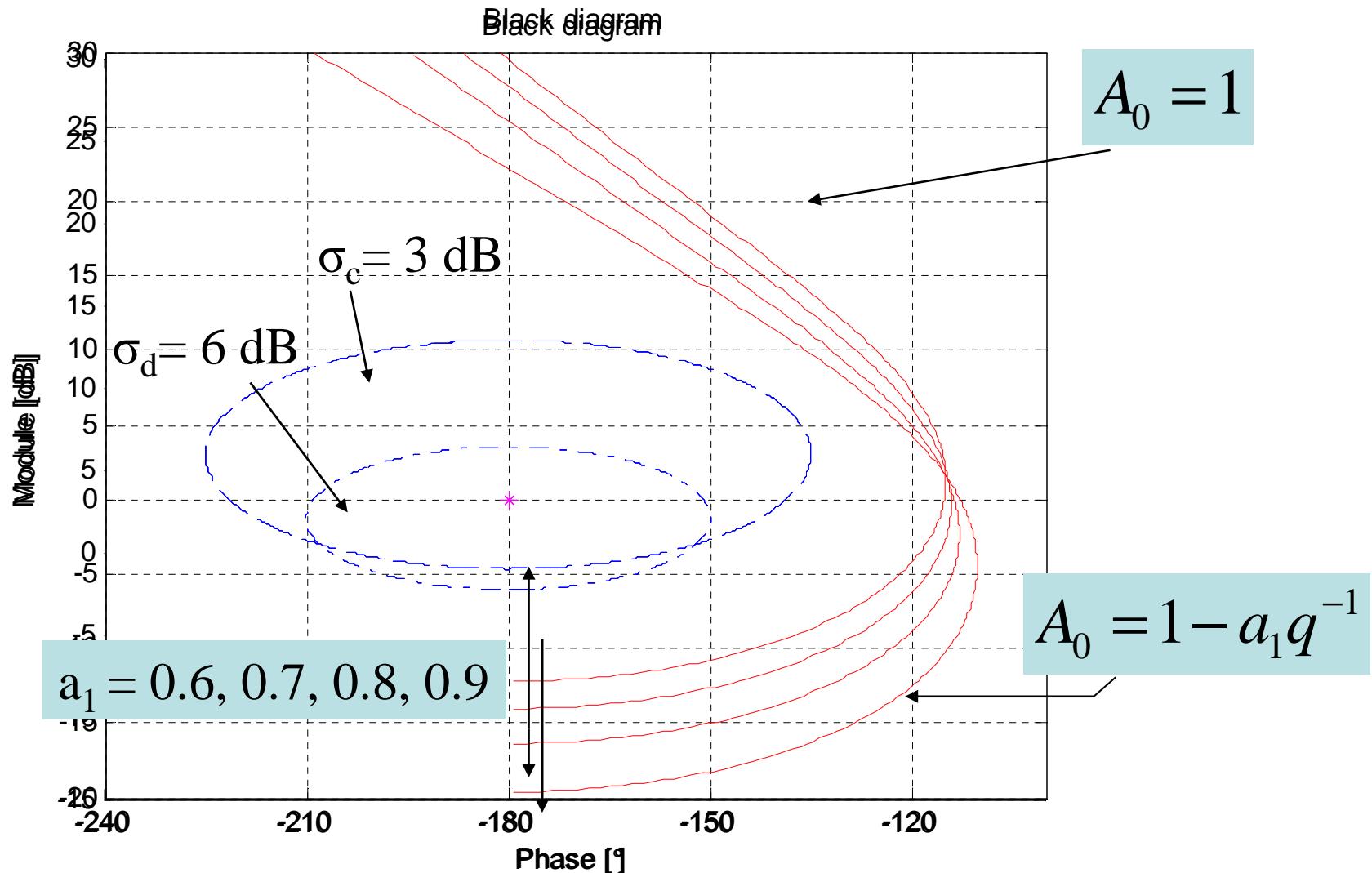
□ Robustness insured by the a_1 parameter:

$$a_1 = 0.7$$

□ Good disturbance rejection when γ is well adapted:

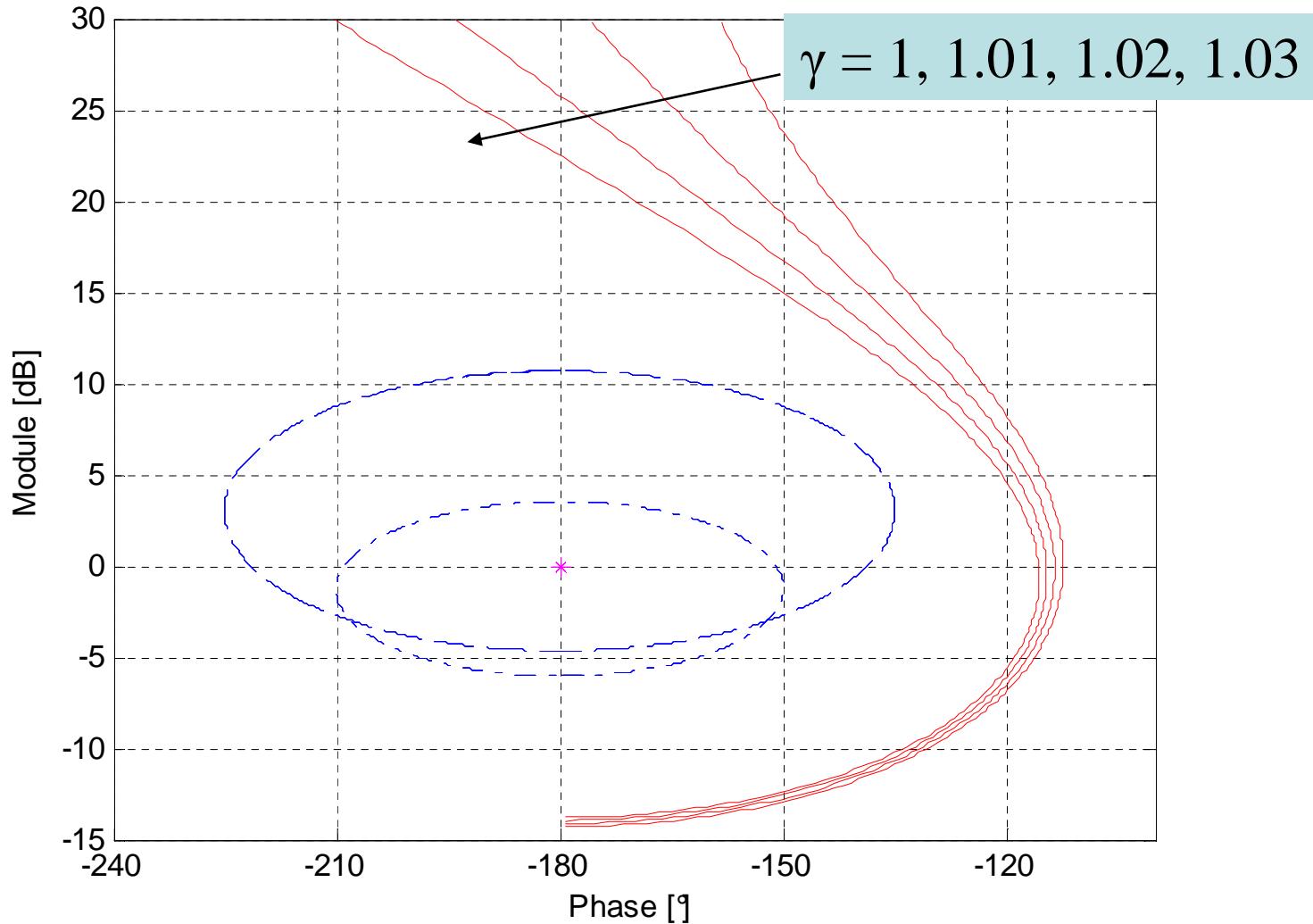
$$\gamma \approx 1.023$$

Robustness analysis



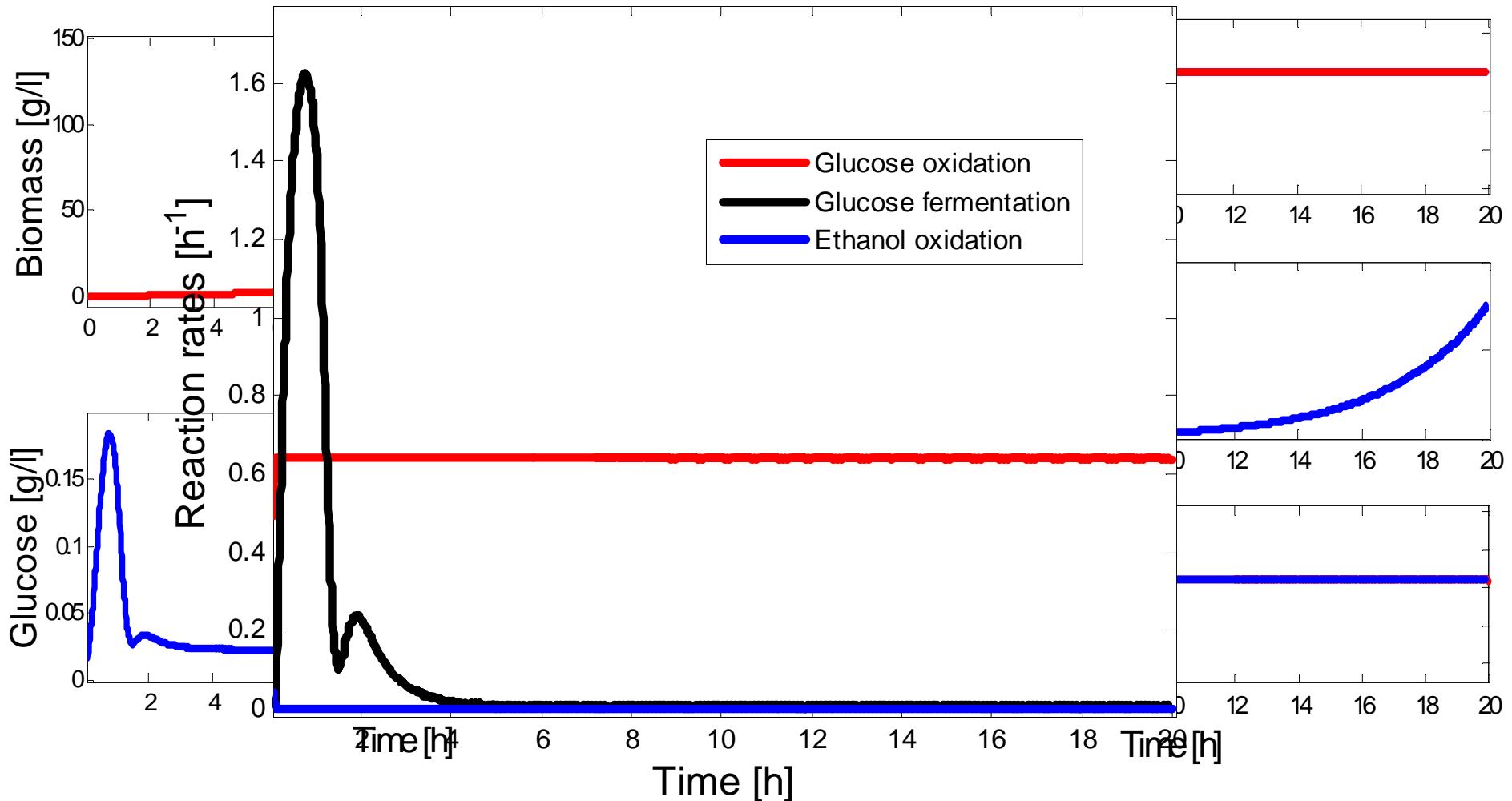
Controller design

Black diagram



Simulation results

Reaction rates

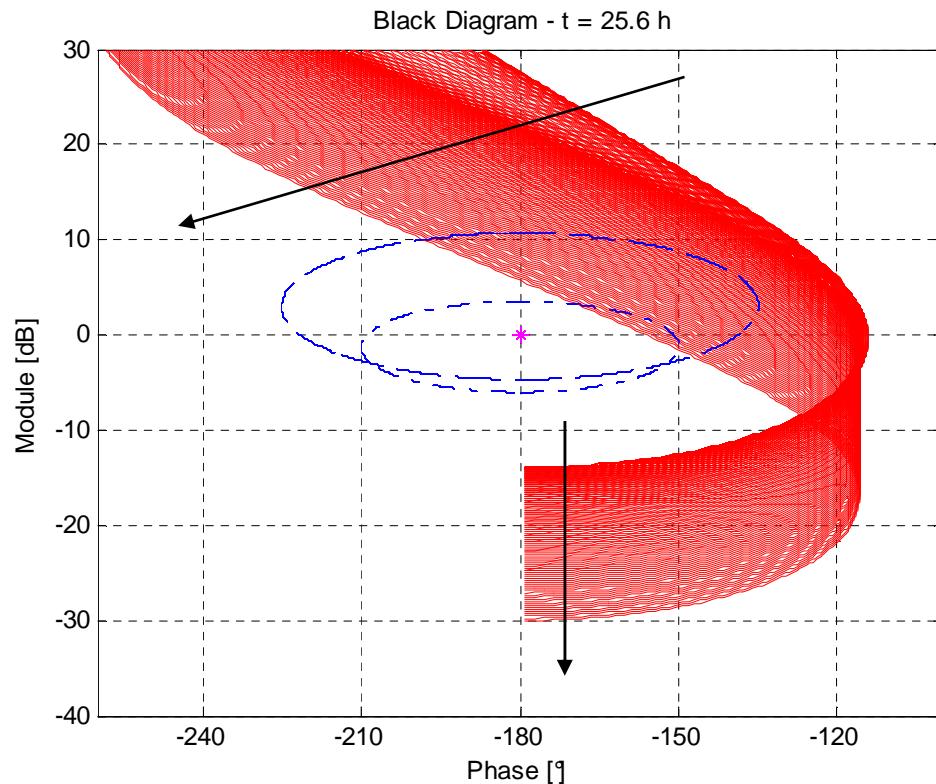


Adaptation

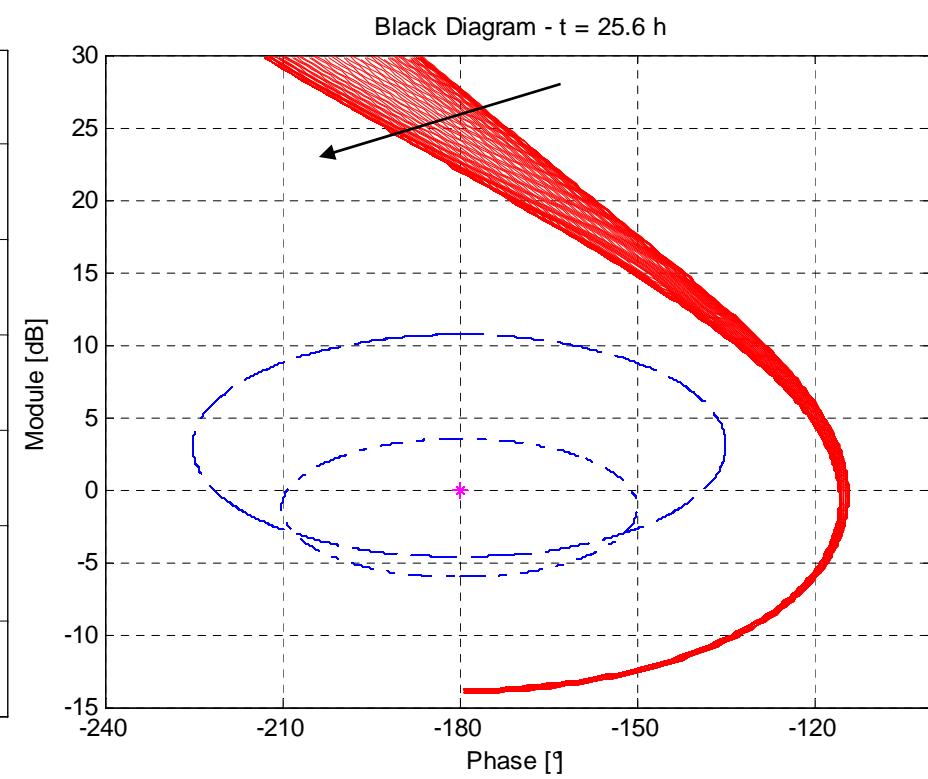
If b is not adapted the gain margin increases but the phase margin decreases when V grows

Adaptation

b is constant

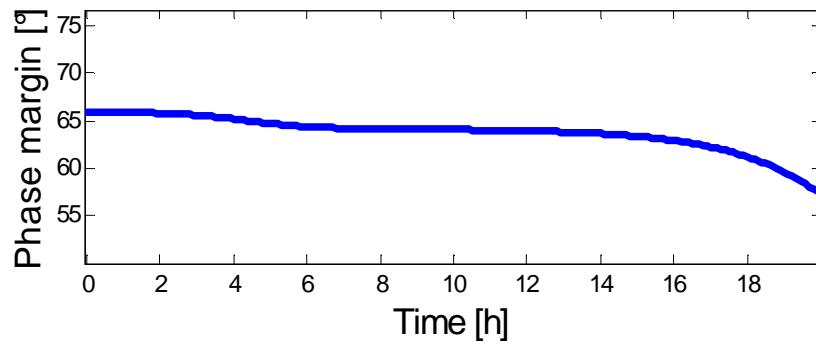
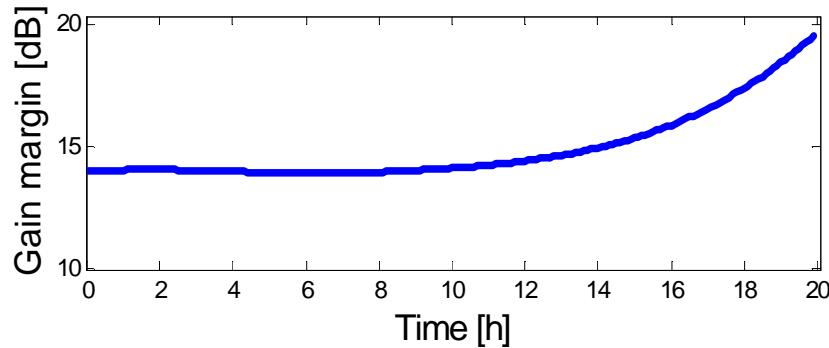


b is adapted

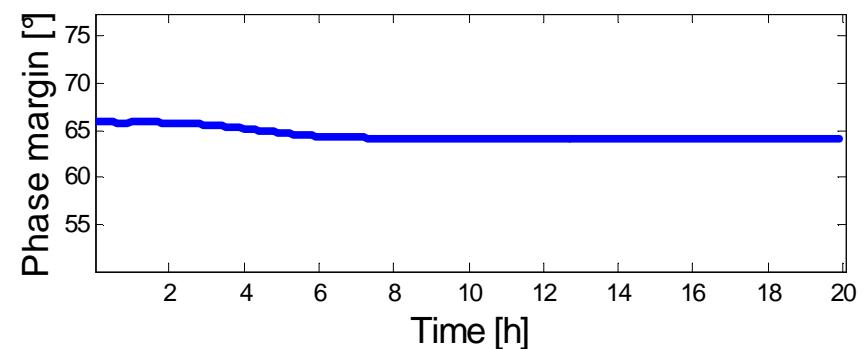
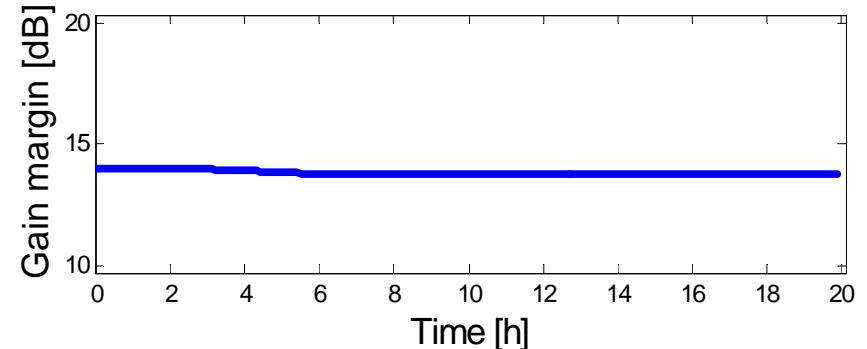


Adaptation

b is constant



b is adapted



Model improvements

- ❑ Ethanol response to feed variations is delayed (≈ 12 min).
- ❑ Delay = 2 sampling periods.
- ❑ Ethanol probe dynamics generally neglected (1-3min).
- ❑ New model:

$$H(q^{-1}) = \frac{b q^{-3} \left[(T_s + T_{mes}(\nu - 1)) + (T_{mes} - \nu(T_s + T_{mes})) q^{-1} \right]}{(1 - q^{-1})(1 - \nu q^{-1})}$$

$$\nu = e^{\left(\frac{T_s}{T_{mes}} \right)}$$

Outline

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Simulations

□ Initial and operating conditions:

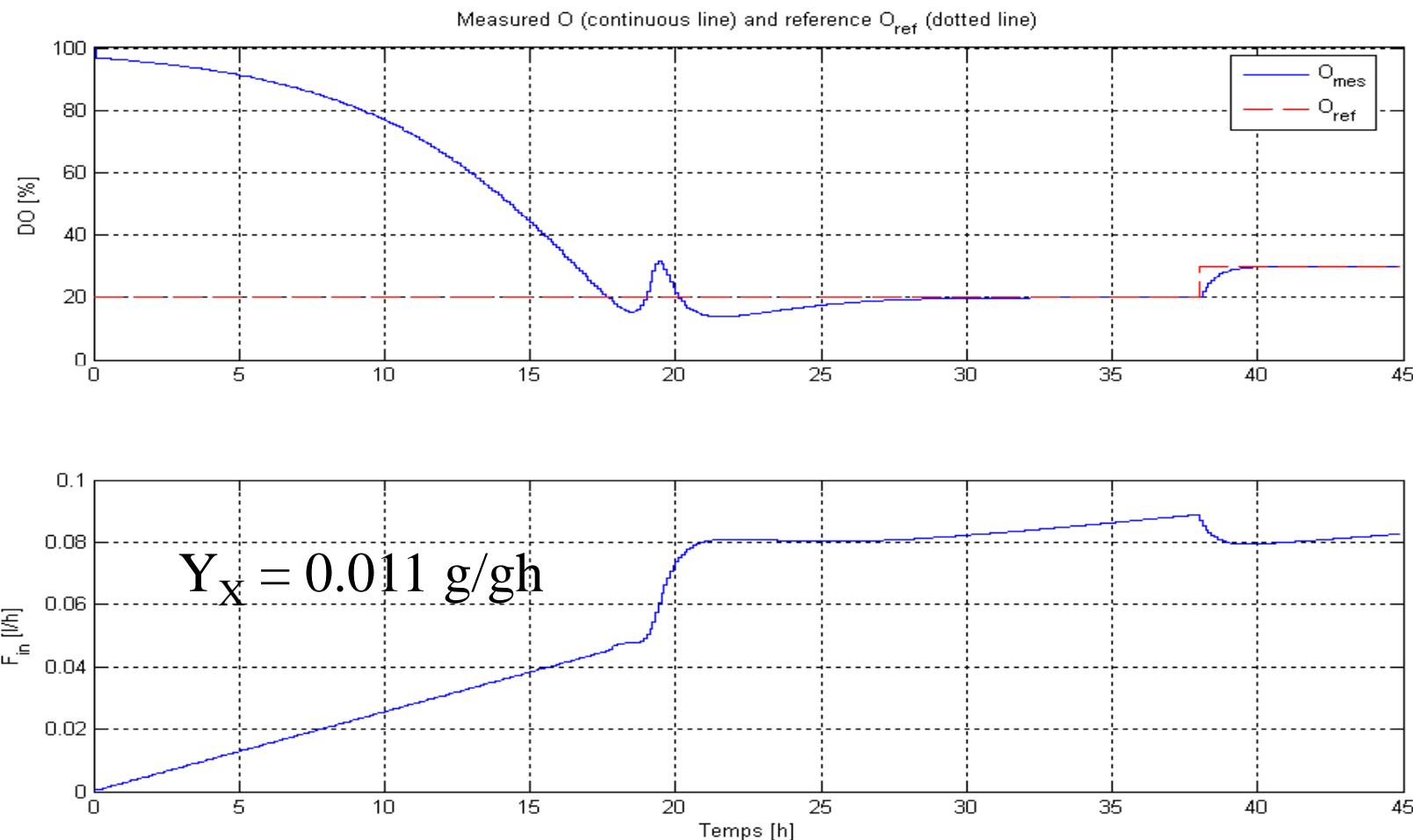
- $X_0 = 0.4 \text{ g/l}$
- $E_0 = 0.9 \text{ g/l}$
- $G_0 = 0.012 \text{ g/l}$
- $O_0 = 100\%$
- $V_0 = 6.8 \text{ l}$
- $G_{in} = 350 \text{ g/l}$
- $E_{ref} = 1 \text{ g/l}$ or $O_{ref} = 20\%$

□ Controller parameters:

- $A_0 = 1 - 0.7q^{-1}$ for Oxygen regulation
- $A_0 = (1 - 0.9q^{-1})(1 - 0.95q^{-1})$ for Ethanol regulation
- $b_0 = (k_4 G_{in} - E_0)/V_0$

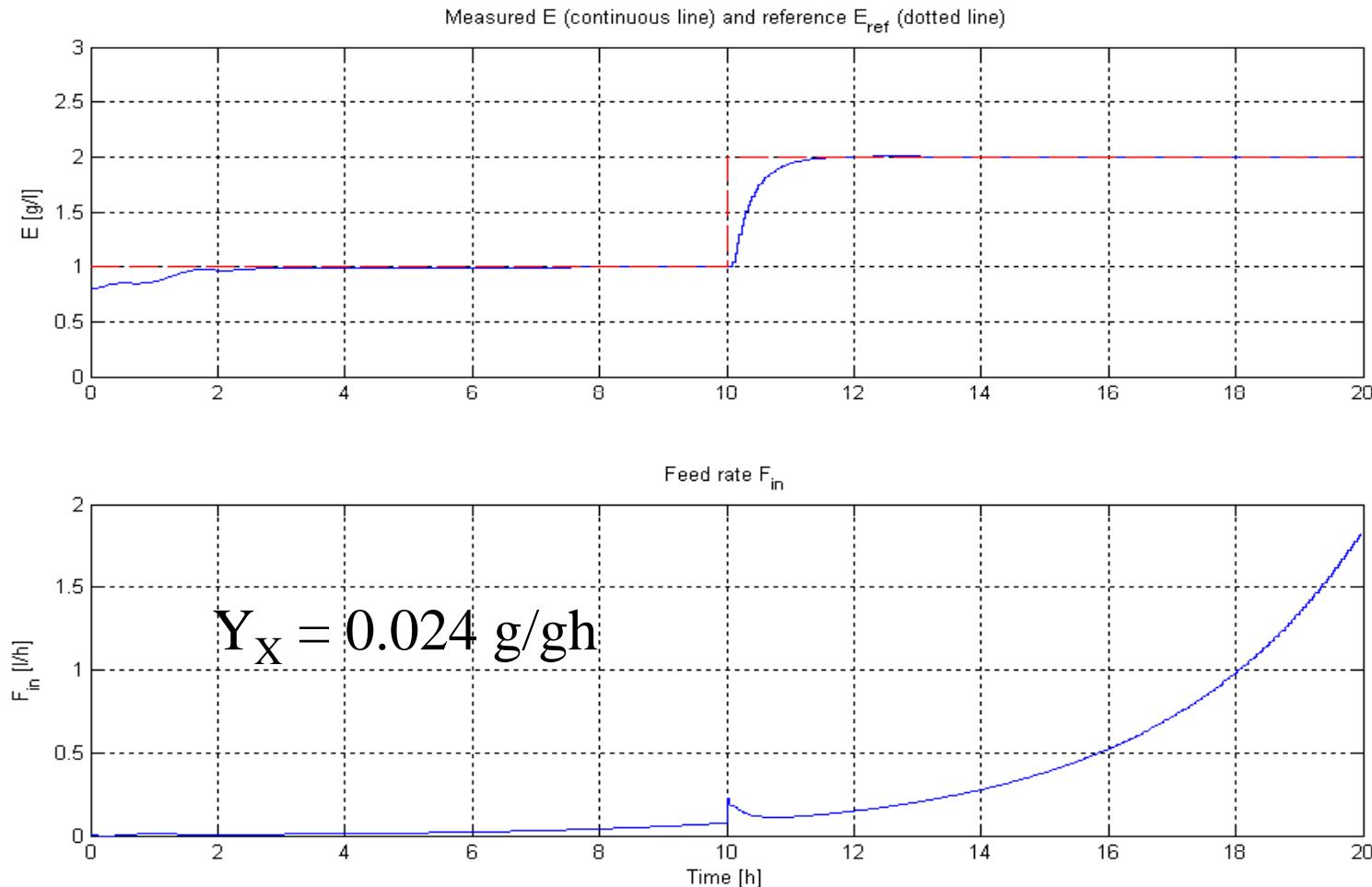
Simulation results

PO2-Feed regulation

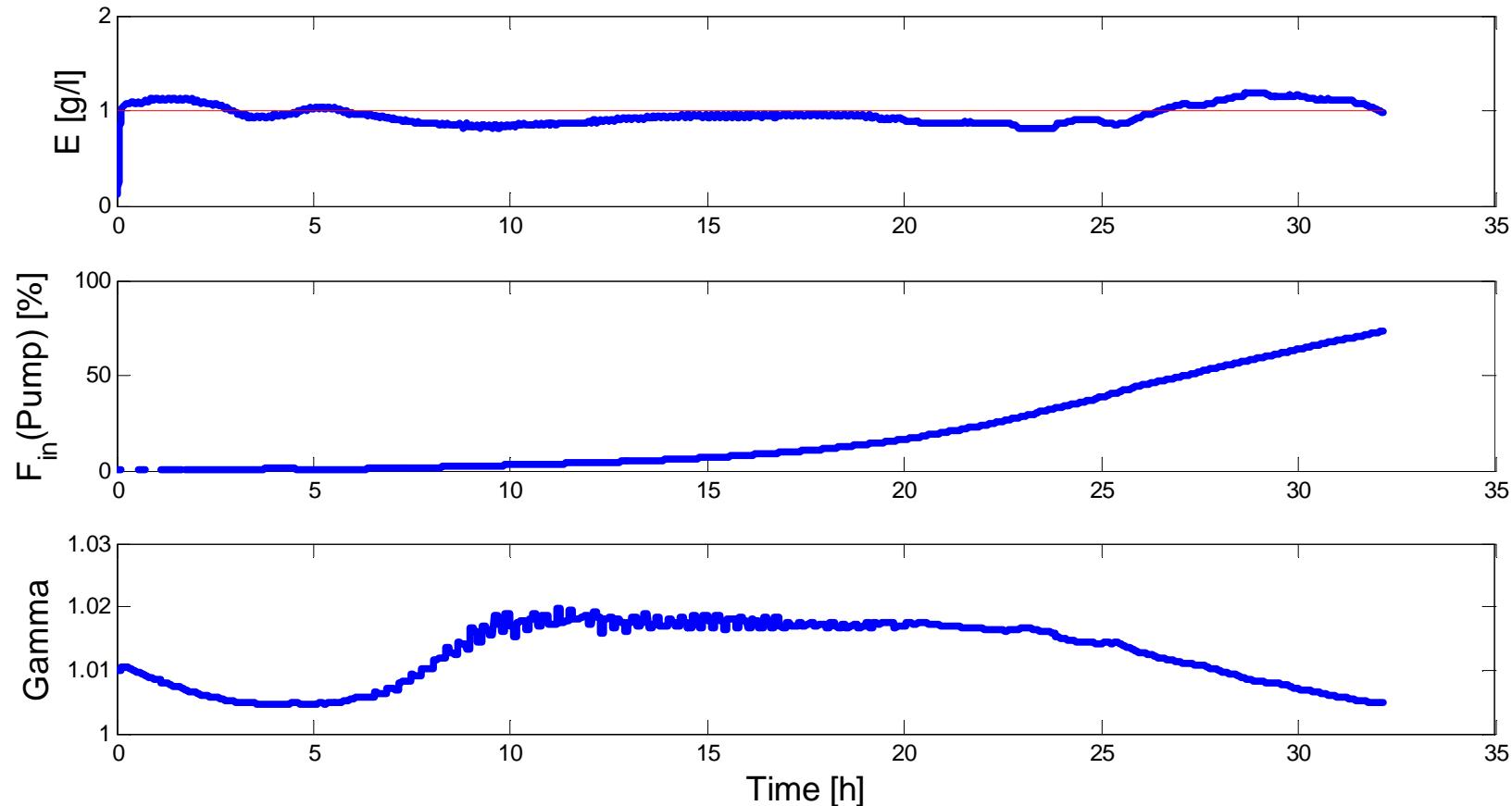


Simulation results

Ethanol-Feed regulation



Experimental results



Conclusions

- 2 simple linear models with the same structure.
- Linear framework is well-adapted to stability and robustness analysis.
- Only one measured concentration: O or E.
- Improvements of the model based on experimental observations.
- 40% improvement in productivity compared to conventional bioprocesses.

Some extensions

□ $E_{in} > E_{ref}$:

- Imposes the consumption of ethanol (respirative regime).

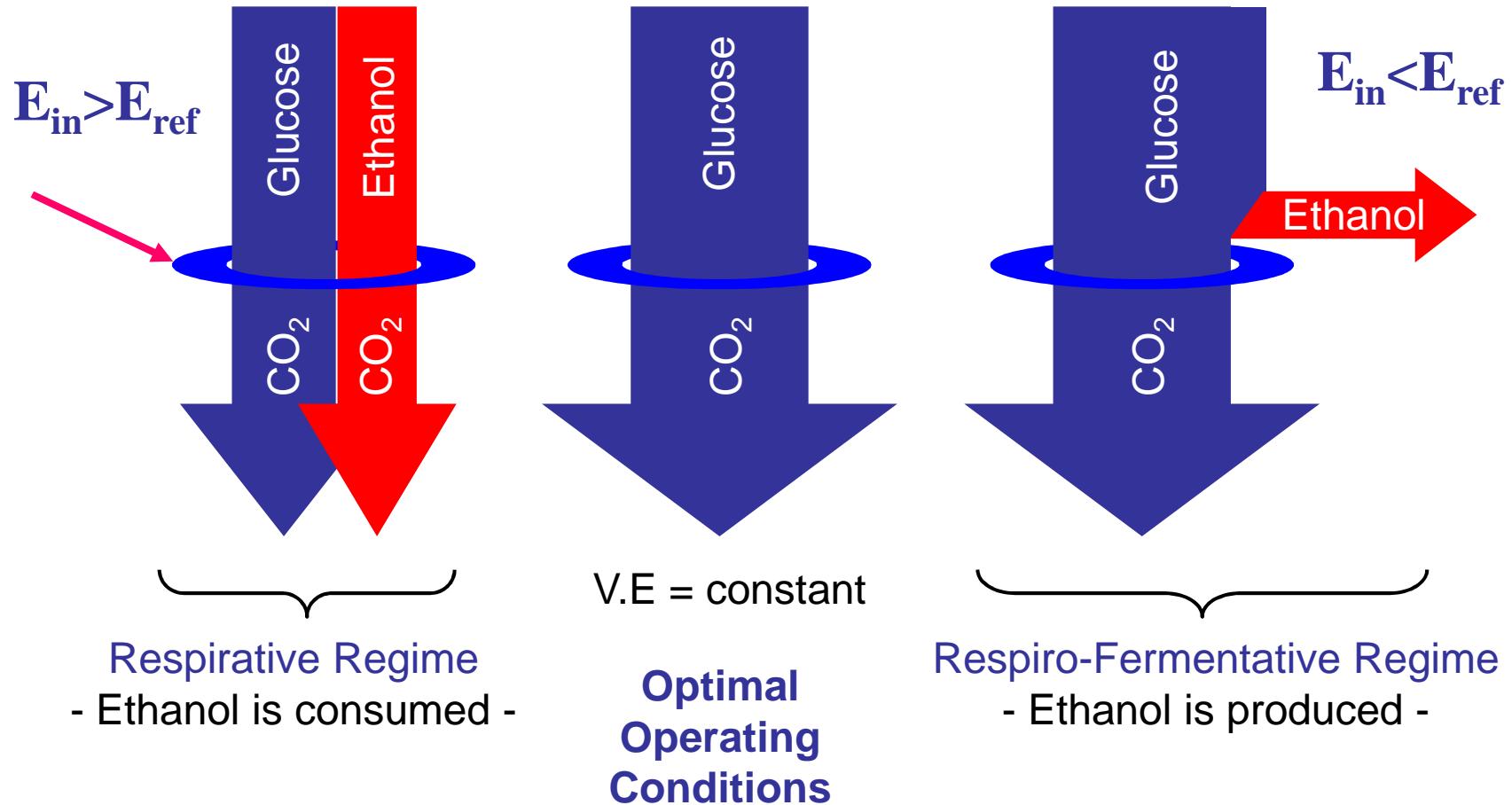
□ $E_{in} < E_{ref}$:

- Imposes the production of ethanol (respiro-fermentative).

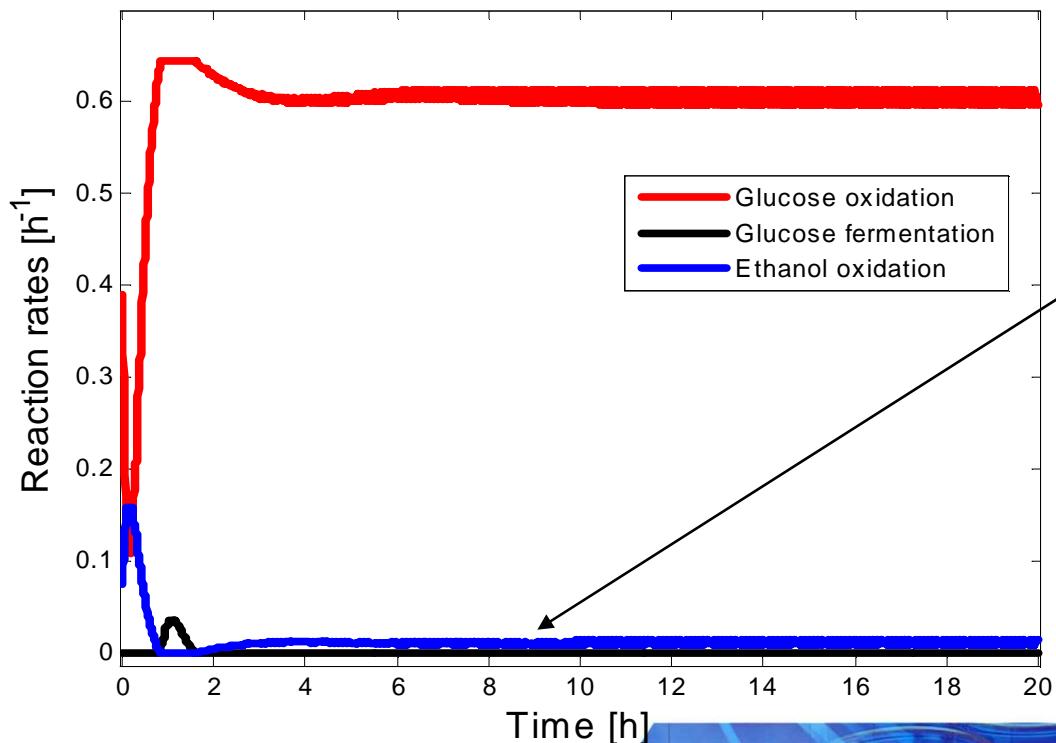
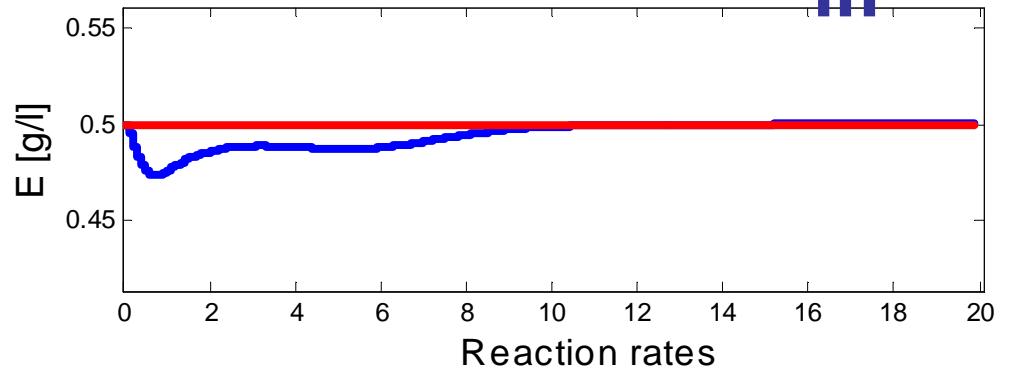
□ Interesting in some applications.

Biological interpretation

□ Yeasts' limited respiratory capacity



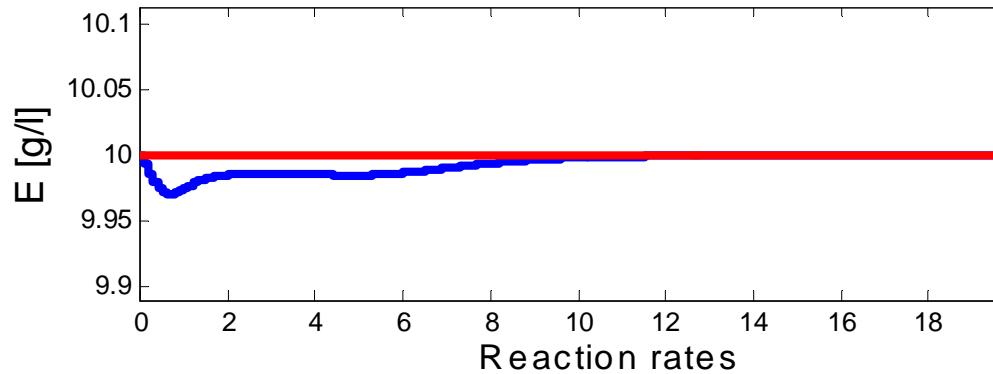
Forced respiratory regime : $E_{in} > E_{ref}$



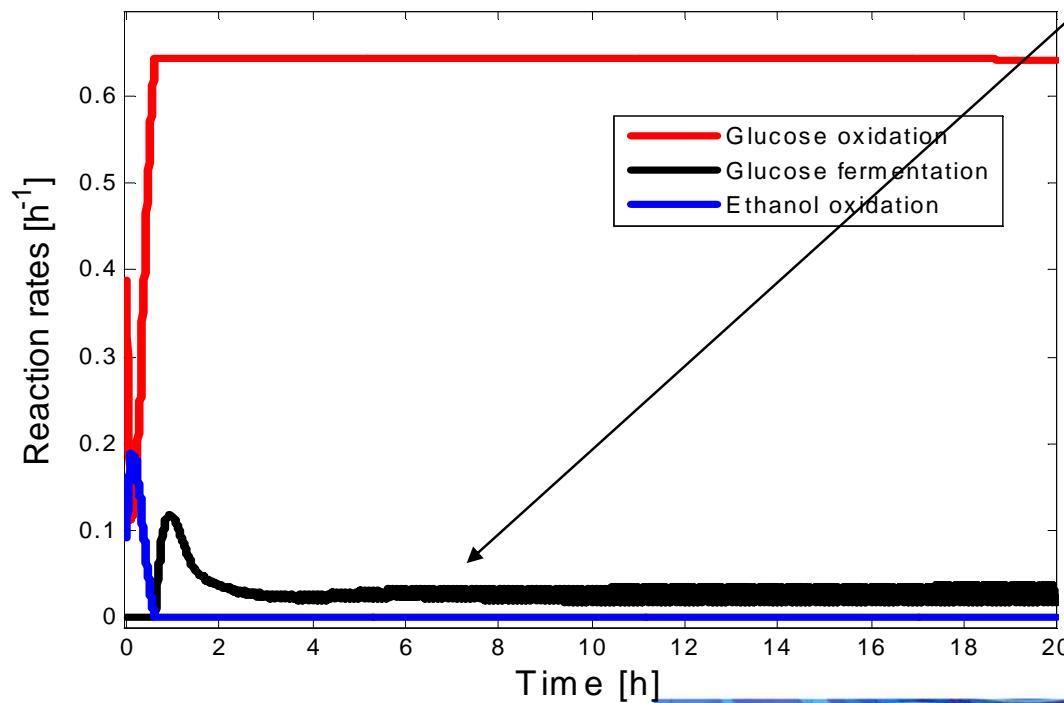
Ethanol consumption

Only in respiratory
regime

Forced respiro-fermentative regime : $E_{in} < E_{ref}$



Ethanol production



Only in respiro-
fermentative regime