

## MODELING AND CONTROL OF A CONTINUOUS BIOREACTOR WITH CROSS-FLOW FILTRATION

Ying Zhao and Sigurd Skogestad\*

Department of Chemical Engineering, University of Trondheim-NTH  
 N-7034 Trondheim, Norway

\*Author to whom correspondence should be addressed, e-mail: skoge@kjemi.unst.no,  
 phone: +47 - 7359 - 4154, fax: +47 - 7359 - 4080

**Abstract.** This work is based on an industrial application of a continuous bioreactor with cross-flow filtration. In this paper the general features of this application are studied. The control objective for this biochemical process is to maintain cell concentration at a desired high concentration in order to maximize the production of cell mass. In this work a controller-independent controllability measure, the partial disturbance gain (PDG), is used for control structure selection and controllability analysis with respect to disturbance rejection, and we study the possibility of partial control of this bioreactor.

**Key Words.** Continuous bioreactor; controllability analysis; disturbance rejection; partial control.

### 1. INTRODUCTION

The studied biochemical process is the propionibacteria fermentation process. Dairy propionibacteria are commercially important in the production of "eyes" and typical flavors in Swiss-type cheeses as they ferment lactic acid with the production of propionic acid and carbon dioxide. Thus there is a strong interest in food industry to continuously obtain the propionibacteria with high concentration at the maximum production rate. To attain this, the development of efficient on-line computer control system is essential.

A schematic overview of this continuous fermentation process is shown in Fig. 1.

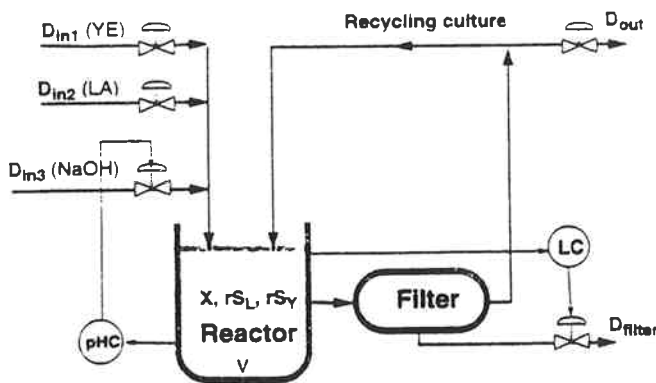


Fig. 1. A continuous bioreactor with cross-flow filtration.

The operation objective of this process is to get high concentration cultivation of propionibacteria in this bioreactor with cross-flow filtration. There are three feed streams into this continuous bioreactor: Yeast extract (YE) acts as nitrogen source for incorporation into cell mass synthesis ( $D_{in1}$ ), lactic acid (LA) acts as both energy source for cell growth and carbon source

for incorporation into cell mass synthesis ( $D_{in2}$ ), and the base solution ( $D_{in3}$ ) is used for compensating pH changes in the bioreactor. The reactor effluent is fed to a filter which is used for increasing the cell concentration by bleeding off some used media at flow rate  $D_{filter}$ , and a part of the remaining stream including cell mass, organic acids and media is recycled to the bioreactor.  $D_{out}$  denotes the product stream flow rate. All flow rates  $F$  are normalized with respect to the bioreactor volume  $V$ , i.e.,  $D = F/V$ .

In the paper simple frequency-dependent tools are used for control structure selection and controllability analysis with respect to disturbance rejection. In particular, by using a frequency-dependent tool: the partial disturbance gain (Skogestad and Wolff, 1992, and Zhao and Skogestad, 1994), the possibility of partial control of this continuous bioreactor is studied, which is the main consideration in this paper.

### 2. DYNAMIC MODEL OF PROCESS AND ITS CHARACTERISTICS

The following assumptions are made.

1. The concentrations of cell mass and substrates are uniform respectively in the whole fermenter-filter system.
2. The bioreactor volume is constant.

Material balances yield:

$$\frac{dX}{dt} = (\mu - D_{out}) X \quad (1)$$

$$\frac{dr_{S_L}}{dt} = -\frac{\mu X}{K_L} - (D_{filter} + D_{out}) r_{S_L} + D_{in2} \quad (2)$$

$$\frac{dr_{S_Y}}{dt} = -\frac{\mu X}{K_Y} - (D_{filter} + D_{out}) r_{S_Y} + D_{in1} \quad (3)$$

where

$X$  — Cell concentration

$r_{S_L} = S_L/S_{fL}$  — Normalized LA concentration with respect to its feed concentration  $S_{fL}$ .

$rS_Y = S_Y/S_{fY}$ — Normalized YE concentration with respect to its feed concentration  $S_{fY}$ .  
 $K_L = Y_{X/L}S_{fL}$ — Product of cell yield coefficient on LA and LA feed concentration.  
 $K_Y = Y_{X/Y}S_{fY}$ — Product of cell yield coefficient on YE and YE feed concentration.  
 $\mu$ — the specific cell growth rate.

In this fermentation process YE acts as growth-limiting substrate, and it is assumed that the Monod model is used for describing this fermentation process kinetics.

$$\mu = \mu_m \frac{rS_Y}{rK_s + rS_Y}$$

where  $\mu_m$  represents the maximum specific cell growth rate and  $rK_s = K_s/S_{fY}$  is the normalized saturation constant with respect to YE feed concentration.

### 2.1. Steady state behavior

From eqs. (1-3), the steady state values of cell and substrate concentrations can be calculated:

$$\frac{dX}{dt} = 0 \Rightarrow \mu = D_{out} \Rightarrow rS_Y = \frac{rK_s D_{out}}{\mu_m - D_{out}} \quad (4)$$

$$\frac{drS_L}{dt} = 0 \Rightarrow rS_L = \frac{D_{in2} - \frac{D_{out} X}{K_L}}{D_{filter} + D_{out}} \quad (5)$$

$$\frac{drS_Y}{dt} = 0 \Rightarrow X = \frac{D_{in1} - (D_{filter} + D_{out}) rS_Y}{D_{out}} K_Y \quad (6)$$

The nominal model parameter values and steady-state data are given in Table 1.

Table 1 Steady State Data

Flow rates [1/h]	$D_{in1}$	$D_{in2}$	$D_{filter}$	$D_{out}$
Concentrations	$X$	$rS_L$	$rS_Y$	
	29.43 [g/l]	0.017	0.000667	
parameters	$rK_s$	$K_L$	$K_Y$	$\mu_m$
	0.000172	0.6 [g/l]	16 [g/l]	0.141 [1/h]

(1.) The steady state relationship between the cell concentration ( $X$ ) and the cell production rate ( $D_{out}$ ) is shown in Fig. 2.

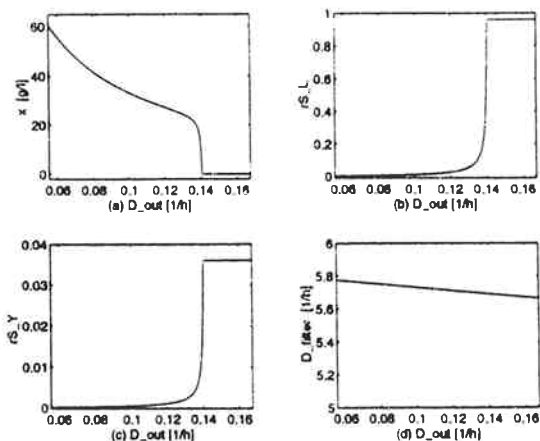


Fig. 2. Steady state values of  $X$ ,  $rS_L$  and  $rS_Y$  as functions of  $D_{out}$  with other flow rates at constant except adjusting  $D_{filter}$  to keep  $V$  constant

When  $D_{out}$  increases,  $X$  decreases and substrate concentrations  $rS_Y$  and  $rS_L$  increase. If  $D_{out}$  reaches to

a critical value  $D_{out,max}$  which can be calculated from eq. (6), the cell can not grow fast enough to keep up with its outflow, and the culture is washed out of the reactor and then  $rS_L$  and  $rS_Y$  rise to their maximum values respectively.

(2.) The steady state relationship between  $X$  and the growth-limiting substrate YE feed flow rate ( $D_{in1}$ ) is shown in Fig. 3.

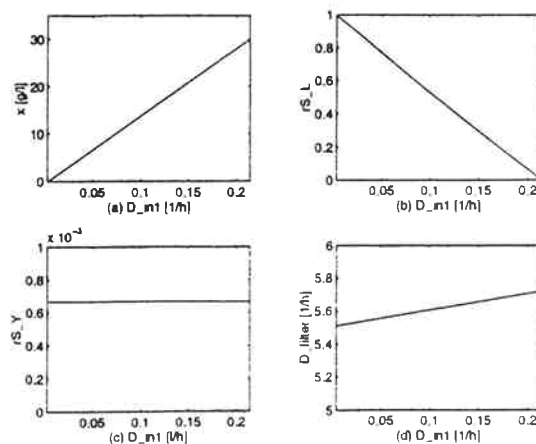


Fig. 3. Steady state values of  $X$ ,  $rS_L$  and  $rS_Y$  as functions of  $D_{in1}$  with other flow rates at constant except adjusting  $D_{filter}$  to keep  $V$  constant

In this case,  $X$  is proportional to  $D_{in1}$  while  $rS_Y$  does not vary with  $D_{in1}$ .  $rS_L$  decreases with increasing  $D_{in1}$  and then this gives an upper bound to  $D_{in1}$  (see Fig. 3b), i.e.,  $D_{in1}$  should be below  $D_{in1,max}$ , otherwise  $rS_L = 0$ . On the other hand,  $D_{in1}$  should be above a critical value  $D_{in1,min}$  (see Fig. 3a), otherwise at very low YE feed flow rate, a large fraction of cells may die from starvation since the limiting substrate is not being added fast enough to permit maintenance of cell growth.

In addition, the LA feed flow rate  $D_{in2}$  does not have much effect except on  $rS_L$ .

**In conclusion**, in this continuous bioreactor with multiple substrate streams the cell concentration is sensitive to changes in  $D_{out}$ , which is different from a general biochemical CSTR with single substrate stream where the cell concentration almost remains constant throughout most of the range of the dilution rates (except close to the  $D_{washout}$ ). The growth-limiting substrate concentration  $rS_Y$  only depends on the cell production rate  $D_{out}$  as in a general CSTR. In addition,  $X$  is also sensitive to changes in  $D_{in1}$  but rather insensitive to the changes in  $D_{in2}$ , and  $rS_L$  is sensitive to changes in feed flow rates  $D_{in1}$  and  $D_{in2}$  but almost remains constant throughout most of the range of  $D_{out}$ .

### 2.2. Dynamic behavior

The main control objective is to keep cell concentration  $X$  at a given high concentration. We have three output variables  $X$ ,  $rS_L$  and  $rS_Y$ , and five manipulated inputs as shown in Fig. 1. Since pH is already controlled by the base addition rate  $D_{in3}$  and the reactor volume is assumed constant such that

$D_{in1} + D_{in2} + D_{in3} = D_{filter} + D_{out}$ , three independent manipulated variables  $D_{in1}$ ,  $D_{in2}$  (or  $D_{filter}$ ) and  $D_{out}$  remain. Since large flow rates are preferred to control reactor volume (level), and this leaves two alternative manipulated inputs  $D_{in2}$  and  $D_{filter}$  for reactor volume control. However biotechnologically it is preferred to use  $D_{filter}$  because for a given organism the ratio  $D_{in1}/D_{in2}$  should be nearly constant.

In the following we will use a plant description of the form

$$y(s) = G(s)u(s) + G_d(s)d(s) \quad (7)$$

where  $G$  and  $G_d$  denote the process and disturbance model, and

$$y = \begin{bmatrix} X \\ rS_L \\ rS_Y \end{bmatrix}; \quad u = \begin{bmatrix} D_{in1} \\ D_{in2} \\ D_{out} \end{bmatrix}; \quad d = \begin{bmatrix} D_{in1d} \\ D_{in2d} \\ D_{outd} \\ K_Y \\ rK_s \\ \mu_m \end{bmatrix}$$

The variables are scaled as follows:

The allowed maximum output changes are

$$\bar{\Delta}X = 10\% \bar{X}; \quad \bar{\Delta}rS_L = 30\% r\bar{S}_L; \quad \bar{\Delta}rS_Y = 20\% r\bar{S}_Y$$

The allowed maximum input changes are 50% changes in  $D_{in1}$ ,  $D_{in2}$  and  $D_{out}$  respectively.

The expected maximum disturbance changes in inputs are 5% changes in  $D_{in1}$ ,  $D_{in2}$  and  $D_{out}$  respectively. Here we have also included disturbances in the model parameters  $K_Y$ ,  $rK_s$  and  $\mu_m$  which are sensitive to variations in the environment conditions such as temperature, pH and aeration rate etc. The expected maximum disturbance changes are

$$\bar{\Delta}K_Y = 10\% \bar{K}_Y; \quad \bar{\Delta}rK_s = 10\% r\bar{K}_s; \quad \bar{\Delta}\mu_m = 10\% \bar{\mu}_m$$

In the following the overbar ( $\bar{\cdot}$ ) used to denote steady-state values will be deleted to simplify notation.

The plant has no poles or transmission zeros in the RHP, and thus there are no fundamental problems related with instability, inverse responses or inherent bandwidth limitations.

The steady-state elements of the open-loop disturbance matrix  $G_d$  (appropriately scaled) are

$$G_d(0) = \begin{bmatrix} 0.51 & -0.01 & -0.55 & 1.00 & -0.02 & 0.09 \\ -9.31 & 9.31 & 0.84 & -18.29 & 0.34 & -1.68 \\ 0.00 & 0.00 & 1.22 & 0.00 & 0.50 & -2.44 \end{bmatrix}$$

The steady-state open-loop gain matrix in terms of scaled variables is

$$G(0) = \begin{bmatrix} 5.09 & -0.09 & -5.46 \\ -93.15 & 93.15 & 8.42 \\ 0.00 & 0.00 & 12.21 \end{bmatrix}$$

Their frequency-dependent plots are shown in Fig 4. We see that feedback control is necessary to reject disturbances; the output  $X$  is sensitive to  $D_{in1}$  and  $D_{out}$  but rather insensitive to  $D_{in2}$ ;  $D_{in1}$  and  $D_{in2}$  have large effect on  $rS_L$ ; at steady state  $rS_Y$  only depends on  $D_{out}$ .

### 3 × 3 control

The steady-state elements of relative gain array (RGA) and the closed loop disturbance gain (CLDG) which is the appropriate measure when we use decentralized control (Hovd and Skogestad, 1992) are given as following,

$$RGA(0) = \begin{bmatrix} 1.02 & -0.02 & 0.00 \\ -0.02 & 1.02 & 0.00 \\ 0.00 & 0.00 & 1.00 \end{bmatrix}$$

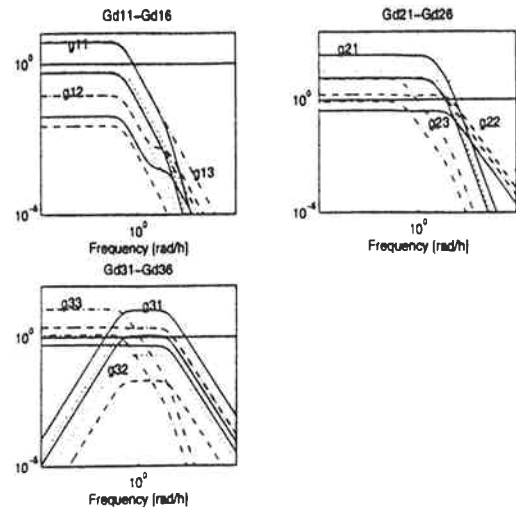


Fig. 4. Frequency-dependent plots of  $G$  and  $G_d$

$$CLDG(0) = \begin{bmatrix} 0.51 & 0.00 & 0.00 & 1.00 & 0.21 & -1.02 \\ 0.00 & 9.31 & 0.00 & 0.01 & 3.81 & -18.63 \\ 0.00 & 0.00 & 1.22 & 0.00 & 0.50 & -2.44 \end{bmatrix}$$

The CLDG show that for all three outputs, the most difficult disturbance to reject is changes in the parameter  $\mu_m$ . An objective is to control this plant by using the decentralized control, and we see that there is one possible set of pairings  $D_{in1}-X$ ,  $D_{in2}-rS_L$  and  $D_{out}-rS_Y$  corresponding to the positive steady-state RGA values and without input constraints problem as all  $|g_{ii}| > |[CLDG]_{ik}|$  for  $i = 1, 2, 3$  and  $k = 1, \dots, 6$ . However implementation of a  $3 \times 3$  control system of bioreactor is expensive and difficult. Therefore the main consideration in this paper is on whether one may achieve acceptable control performances of all three outputs by controlling only one or two of them, i.e., the plant is partially controlled.

## 3. CONTROLLABILITY ANALYSIS AND CONTROL STRUCTURE SELECTION

### 3.1. Measures for evaluating controllability

We now close only one loop ( $1 \times 1$  control) or two loops ( $2 \times 2$  control) and define  $y = [y_c \ y_u]^T$  and  $u = [u_c \ u_u]^T$ , where

$y_c$ — controlled outputs

$y_u$ — uncontrolled outputs

$u_c$ — manipulated inputs used to control  $y_c$

$u_u$ — unused inputs

The problem is to select the controlled outputs ( $y_c$ ), and the unused inputs ( $u_u$ ) such that :

1. Use of  $u_c$  to control  $y_c$  should yield satisfactory control performance.
2. With these control loops closed the uncontrolled outputs  $y_u$  should be relatively insensitive to disturbances.

To evaluate both these issues a controllability analysis of the alternative structures needs to be performed. A useful tool when considering issue 2 is the partial disturbance gain (PDG) which is the effect of a disturbance on uncontrolled outputs when the plant is under partial control.

For a square plant with only one output  $y_j$  perfectly controlled ( $1 \times 1$  control), the sensitivity of the uncontrolled outputs  $y_i$  to a disturbance  $d_k$  is derived (see appendix):

$$PDG_{1 \times 1} = \left( \frac{\partial y_i}{\partial d_k} \right)_{y_j, u_i \neq j} = [G_d]_{ik} - [G(\hat{G})^{-1}]_{ij}[G_d]_{jk} \quad (8)$$

where  $u_i \neq j$  are unused inputs,  $\hat{G}$  is the matrix consisting of only the diagonal elements of  $G$ , i.e.,  $\hat{G} = \text{diag}\{g_{ii}\}$ .

Next consider two outputs  $y_{l \neq i}$  perfectly controlled ( $2 \times 2$  control). Since we have a  $3 \times 3$  control system this is the same as having 1 input  $u_j$  "unused" and the sensitivity of the uncontrolled output  $y_i$  to a disturbance  $d_k$  is given by Skogestad and Wolff in 1992:

$$PDG_{2 \times 2} = \left( \frac{\partial y_i}{\partial d_k} \right)_{u_j, y_{l \neq i}} = [G^{-1}G_d]_{jk} / [G^{-1}]_{jj} \quad (9)$$

For simultaneous disturbances  $d_k, k = 1, 2, \dots$ , the worst overall effect may be evaluated by taking the sum of element magnitudes for each "pairing". This gives rise to a combined  $PDG$ -matrix, denoted  $CPDG$ , with elements

$$[CPDG]_{ij} = \sum_k |[PDG]_{ij}|$$

The objective of the controllability analysis in this paper is to find the best partial control scheme so that we can get the acceptable disturbance sensitivity for the uncontrolled outputs even under simultaneous disturbances. To attain this, we prefer pairings corresponding to  $CPDG$ -elements less than 1 at all frequencies.

### 3.2. Case study

#### $1 \times 1$ control

The effect of disturbances on the uncontrolled outputs are given by the  $PDG$ . The steady-state combined  $PDG$  ( $CPDG$ ) is listed in Table 2. From this and the frequency response of  $CPDG$  (not shown in the paper), the best partial control scheme is the  $D_{in1} \rightarrow rS_L$  - scheme No. 2. It yields  $CPDG = 1$  for  $X$  and  $CPDG = 4.16$  for  $rS_Y$ . Practical implementation of this partial control system is possible since on-line measurement of lactic acid concentration is easier than cell mass or yeast extract concentration (Yeast extract is a complex medium and its composition is chemically undefined and may vary as the fermentation process is in progress).

Table 2 The combined partial disturbance gain

No.	Controlled pairing		$CPDG$		
	$u_c$	$y_c$	$\Sigma_k \frac{\partial X}{\partial d_k}$	$\Sigma_k \frac{\partial rS_L}{\partial d_k}$	$\Sigma_k \frac{\partial rS_Y}{\partial d_k}$
1	$D_{in1}$	$X$		18.31	4.16
2	$D_{in1}$	$rS_L$	1.00		4.16
3	$D_{in1}$	$rS_Y$	$\infty$	$\infty$	
4	$D_{in2}$	$X$		2.20	4.16
5	$D_{in2}$	$rS_L$	2.14		4.16
6	$D_{in2}$	$rS_Y$	$\infty$	$\infty$	
7	$D_{out}$	$X$		36.43	6.09
8	$D_{out}$	$rS_L$	23.64		53.57
9	$D_{out}$	$rS_Y$	2.72	36.91	

Schemes No. 3 and 6 with  $rS_Y$  controlled by  $D_{in1}$  or  $D_{in2}$  are not feasible since disturbance sensitivities of uncontrolled outputs  $X$  and  $rS_L$  are infinite because  $rS_Y$  only depends on  $D_{out}$ . Schemes No. 1 and No. 7-9 are unacceptable with large sensitivities of uncontrolled outputs to disturbances.

### $2 \times 2$ control

In this case the effect of disturbances on the controlled output are evaluated by eq. (9) and the steady-state  $CPDG$  values are given in Table 3. The best  $2 \times 2$  control system is No. 6 with  $u_c = [D_{in1} D_{out}]^T$  and  $y_c = [rS_L rS_Y]^T$ . It gives almost acceptable sensitivity of the uncontrolled  $X$  to disturbances. However comparing with the case by  $1 \times 1$  control, we see that cell concentration  $X$  is more sensitive to disturbances than the scheme No. 2 in Table 2.

Table 3 The combined partial disturbance gain

No.	Controlled pairing		$CPDG$		
	$u_c$	$y_c$	$\Sigma_k \frac{\partial X}{\partial d_k}$	$\Sigma_k \frac{\partial rS_L}{\partial d_k}$	$\Sigma_k \frac{\partial rS_Y}{\partial d_k}$
1.	$D_{in1}$	$X$			4.16
2.	$D_{in1}$	$rS_L$		$\infty$	
3.	$D_{in1}$	$rS_Y$	$\infty$		
4.	$D_{in2}$	$X$			4.17
5.	$D_{in2}$	$rS_L$		31.2	
6.	$D_{in1}$	$rS_Y$	1.71		
7.	$D_{out}$	$X$			6.02
8.	$D_{out}$	$rS_L$		2765.8	
9.	$D_{out}$	$rS_Y$	2.68		

### 3.3. Dynamical simulation results

In this section we present nonlinear simulation results to compare with the controllability analysis results presented in the previous section. In all simulations simple PI controllers are used for the controlled outputs and we consider the step responses to the combined disturbances in parameters  $K_Y, rK_s$  and  $\mu_m$  as shown in Fig. 5(d).

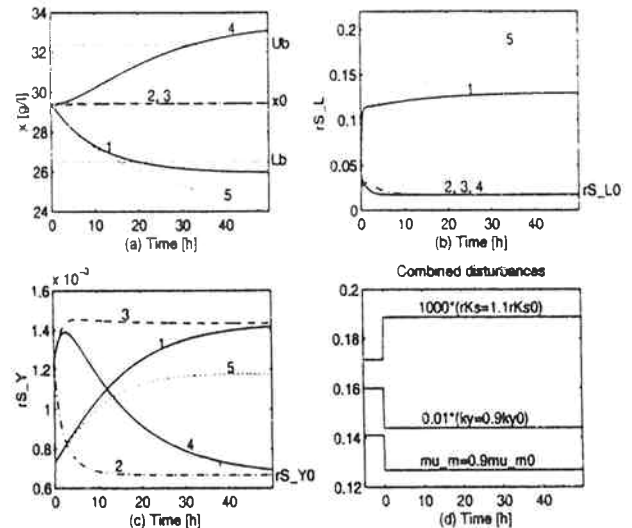


Fig. 5. Time responses for the combined step disturbances in parameters  $K_Y, rK_s$ , and  $\mu_m$ . 1: uncontrolled. 2:  $3 \times 3$  control. 3:  $1 \times 1$  control  $D_{in1} \rightarrow rS_L$ . 4:  $2 \times 2$  control. 5:  $1 \times 1$  control  $D_{in1} \rightarrow rS_Y$ .

From simulation results we find: The system can not be left uncontrolled, otherwise after long time (for example during the weekend) as

shown in Fig. 5 (curves 1) the cell concentration  $X$  will decrease and below its lower bound (Lb); and the substrate concentration, especially  $rS_L$ , will be much high and then result losing more substrate.

As shown in Fig. 5 (curves 2), if all outputs  $X$ ,  $rS_L$  and  $rS_Y$  were on-line measurable, then the  $3 \times 3$  control system ( $X$  is controlled by  $D_{in1}$ ,  $rS_L$  is controlled by  $D_{in2}$  and  $rS_Y$  is controlled by  $D_{out}$ ) could work well against the combined disturbances in parameters as all three outputs remain at their desired steady states without violating the input constraints.

With  $1 \times 1$  partial control scheme  $D_{in1} \rightarrow rS_L$  shown in Fig. 5 (curves 3),  $rS_L$  is perfectly controlled without violating the input constraint of  $D_{in1}$ ; the uncontrolled  $X$  almost goes back its desired value after several hours so that the main operation objective of this fermentation process has been achieved. The deviation of uncontrolled  $rS_Y$  from its desired value is large but tolerable.

With  $2 \times 2$  partial control system  $u_c = [D_{in1} D_{out}]^T$  and  $y_c = [rS_L rS_Y]^T$  shown in Fig. 5 (curves 4), we find no improvement compared to the  $1 \times 1$  scheme  $D_{in1} \rightarrow rS_L$  except the  $rS_L$  returns quicker to its desired value, and after long transition process  $rS_Y$  may be controlled at its desired value. However the uncontrolled  $X$  exceeds its upper bound (Ub).

The partial control scheme  $D_{in1}$  (or  $D_{in2}$ )  $\rightarrow rS_Y$  with  $D_{out}$  left in manual should be avoided, otherwise as shown in Fig 5 (curves 5) the uncontrolled  $X$  decreases quickly and eventually down to zero and then the reaction will stop:  $D_{in1}$  has to be continuously lowered in order to try to keep  $rS_Y$  constant, however, it is not possible to keep  $rS_Y$  constant in the long run while  $D_{out}$  is unchanged because  $rS_Y$  is independent of  $D_{in1}$  at steady state, and finally  $D_{in1}$  will reach 0 and results in a failure of the reactor.

#### 4. CONCLUSIONS

Based on both linear controllability analysis and simulation results, we suggest that to control this continuous bioreactor with cross-flow filtration is by using a  $1 \times 1$  partial control scheme ( $D_{in1} \rightarrow rS_L$ ) where the YE feed flow rate  $D_{in1}$  is used to control lactic acid concentration  $rS_L$  which furthermore is relatively easy to implement practically.

In this paper we have generalized the partial disturbance gain which was proposed by Skogestad in 1992 to the case that a plant is under  $1 \times 1$  partial control. A careful comparison shows that the simulation results are consistent with the linear controllability analysis. This indicates that the partial disturbance gain (PDG) is an effective tool for controllability analysis without considering the controller design.

**Acknowledgment.** Thanks to Ivar Storø for fruitful discussions.

#### REFERENCES

- Hovd, M. and S. Skogestad, 1992, "Simple Frequency-Dependent Tools for Control System Analysis, Structure Selection and Design", *Automatica*, **28**, 5, p. 989-996.
- Skogestad, S., and E. A. Wolff, 1992, "Controllability measures for disturbance rejection", *Proc. of IFAC Workshop on Interactions Between Process Design and Control*, London, UK.
- Zhao, Y. and S. Skogestad, 1994, "A comparison of various control schemes for continuous bioreactor", *Proc. of IFAC Symposium ADCHEM'94*, Kyoto, Japan.

#### APPENDIX

Derivation of PDG when only one output  $y_j$  of the plant is perfectly controlled with inputs  $u_{i \neq n}$  left in manual.

For the uncontrolled outputs  $y_{i \neq j}$ , set  $u_{i \neq n} = 0$  and the eq. (7) yields

$$y_i = g_{in} u_n + [G_d]_{ik} d_k \quad (10)$$

$$y_j = g_{jn} u_n + [G_d]_{jk} d_k \quad (11)$$

$$y_j = 0 \Rightarrow u_n = -(g_{jn})^{-1} [G_d]_{jk} d_k \quad (12)$$

By substituting eq. (12) into eq. (10) and thus

$$y_i = -g_{in} (g_{jn})^{-1} [G_d]_{jk} d_k + [G_d]_{ik} d_k \quad (13)$$

By taking the partial derivative of  $y_i$  w.r.t  $d_k$ , we derive

$$PDG_{1 \times 1} = \left( \frac{\partial y_i}{\partial d_k} \right)_{y_j, u_{i \neq n}} = [G_d]_{ik} - g_{in} (g_{jn})^{-1} [G_d]_{jk} \quad (14)$$

When  $j = n$ , we have

$$PDG_{1 \times 1} = \left( \frac{\partial y_i}{\partial d_k} \right)_{y_j, u_{i \neq j}} = [G_d]_{ik} - [G(\dot{G})^{-1}]_{ij} [G_d]_{jk} \quad (15)$$