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R. TENNO, K. TIISMA, T. PAALME and R. VILU

STOCHASTIC CONTROL OF GROWTH IN FED-BATCH CULTURE OF BACTERIA

(Presented by E. Lippmaa)

Using the stoichiometric equation of growth and Monod type kinetic equation, a deterministic model of growth of microorganisms in fed-batch culture was built. A stochastic model of the process was developed on the basis of the deterministic description. A stochastic control algorithm which is using filtered data was developed.

1. Introduction

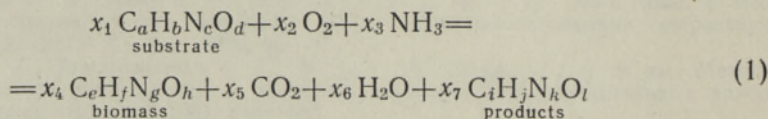
Interest to the computer-aided fermentations could be traced back to the sixties [1]. The first persons to point out the possibility of computer monitoring of indirectly measured parameters were Yamashita, Hoshi, and Inagaki in 1969 [2]. The idea has been implemented in several papers [3-9], where methodological and experimental problems of computer monitoring of fermentations, using indirectly measured parameters, have been investigated. Much less attention has been paid to the problems of parameter identification, using Kalman filter or other filtering schemes [10-11]. Realizations of on-line control algorithms, published, could be divided into three classes: a) control through monitoring and regulation of a certain parameter (RQ, etc) at specific value, which ensures the achievement of the formulated aim of the control [4, 12, 13], b) control on the basis of a deterministic kinetic process model [7, 14-17], and c) recently a stochastic control algorithm of fermentation processes was published [18].

A stochastic control algorithm which is using filtered data for partly observable fed-batch cultivation process of microorganisms was built in the present paper.

2. Results and discussion

2.1. A deterministic model of growth of microorganisms

Aerobic growth of microorganisms is described by a stoichiometric equation



in which x_1, \dots, x_7 are stoichiometric coefficients (not necessarily constant). Use of ammonia as nitrogen source does not restrict the generality of the equation. In eq. (1) we have taken into account only the four major elements comprising more than 95% of biomass — carbon, hydrogen, nitrogen and oxygen. Minor elements — phosphorus, sulphur, potassium, etc. as well as variability of stoichiometric coefficients will be taken into

account as uncontrolled disturbances in stochastic model of the growth process.

Stoichiometric coefficients obey elemental balance equations

$$\mathbf{A}\mathbf{X}=\mathbf{0}, \quad (2)$$

where

$$\mathbf{A}=\begin{bmatrix} a & 0 & 0 & -e & -1 & 0 & -i \\ b & 0 & 3 & -f & 0 & -2 & -j \\ c & 0 & 1 & -g & 0 & 0 & -k \\ d & 2 & 0 & -h & -2 & -1 & -l \end{bmatrix}, \quad \mathbf{X}=\begin{bmatrix} x_1 \\ \vdots \\ x_7 \end{bmatrix}.$$

It follows from eq. (1) that

$$\dot{Q}_1/x_1=\dot{Q}_2/x_2=\dot{Q}_3/x_3=\dot{Q}_4/x_4=\dot{Q}_5/x_5=\dot{Q}_6/x_6=\dot{Q}_7/x_7, \quad (3)$$

where $\dot{Q}_i=dQ_i(t)/dt$, $i=1, 2, \dots, 7$ is production or consumption rate of the corresponding components: Q_1 — growth substrate, Q_2 — oxygen, Q_3 — ammonia, Q_4 — biomass, Q_5 — carbon dioxide, Q_6 — water, and Q_7 — products, synthesized during the cultivation. It should be noted that all the parameters used throughout this work except elements of matrix \mathbf{A} and constants c_0, c_1, \dots, c_6 , together with $\sigma_1, \sigma_2, r_2, r_3$, introduced below, are functions of time. However, designation of the functional dependence is removed from the equations, and preserved only in the cases where emphasizing of the functional dependence is considered especially important.

The system of equations (2) has $(7\text{-rank } \mathbf{A}) > 3$ nontrivial solutions. Let us assume $\text{rank } \mathbf{A}=4$, which is correct in most cases of growth. Let us define vectors

$$\mathbf{Z}^T=(x_4, x_6, x_7), \quad \mathbf{Y}^T=(x_1, x_2, x_3, x_5)$$

and matrices \mathbf{B} and \mathbf{C} according to equation

$$\mathbf{A}\mathbf{X}=\mathbf{B}\mathbf{Y}+\mathbf{C}\mathbf{Z}.$$

If matrix \mathbf{B} is positively defined, then fundamental solution of eq. (2) is determined by

$$\mathbf{Y}=-\mathbf{B}^{-1}\mathbf{C}\mathbf{Z}$$

or

$$x_i=-\sum_{j=4, 6, 7} q_{ij}x_j, \quad i=1, 2, 3, 5, \quad (4)$$

where

$$\mathbf{B}^{-1}\mathbf{C}=(q_{ij}).$$

From eq. (3) and (4) it follows that

$$\dot{Q}_i=\mu_i\dot{Q}_4, \quad i=1, 2, 3, 5, \quad (5)$$

where

$$\mu_i=-[q_{i4}+q_{i6}\dot{Q}_6/\dot{Q}_4+q_{i7}\dot{Q}_7/\dot{Q}_4]^{-1}. \quad (6)$$

Biomass Q_4 , products Q_7 and water Q_6 formed (cf. eq. (1)) change in the fermentor according to (bounded) exponential law

$$\dot{Q}_4=FQ_4, \quad \dot{Q}_6=c_4Q_4, \quad \dot{Q}_7=c_5Q_4, \quad (7)$$

where

$$F=c_0Q/[V+c_1Q+c_2Q^2/V+c_3Q_7] \quad (8)$$

is a Monod (type) function taking into account growth inhibition by substrate and by products, and Q/V is concentration of unutilized growth substrate in fermentor

$$Q=\alpha-Q_1,$$

α is added (controlled) and Q_1 — utilized amount of growth substrate, V is the current culture volume

$$V = c_6 \alpha,$$

and c_0, \dots, c_6 — constants. In our case (see also [49]):

$$\begin{aligned} c_0 &= 40 \text{ mM}^{-1}\text{h}^{-1}, & c_1 &= 100 \text{ mM}^{-1}, & c_2 &= 10 \text{ mM}^{-1}, \\ c_3 &= 100 \text{ mM}^{-1}, & c_4 &= 0.4 \text{ h}^{-1}, & c_5 &= 0.1 \text{ h}^{-1}. \end{aligned}$$

It is derived from equations (6), (7) that

$$\mu_i = -[q_{i4} + ((c_4 q_{i6} + c_5 q_{i7})/F)]^{-1}, \quad i = 1, 2, 3, 5. \quad (9)$$

Eq. (5) and (9) together with eq. (7) and (8) form a system of nonlinear equations the solution of which is $(\mu_i, F)(t, V, Q_1, Q_4)$. Through integration by parts of (5):

$$\begin{aligned} Q_1 &= \tilde{Q}_1 - \tilde{\mu}_1 \tilde{Q}_4 - \int_0^t Q_4 \dot{\mu}_1 d\tau + \mu_1 Q_4, \\ Q_4 &= \tilde{Q}_4 - \tilde{Q}_5 / \mu_5 + \int_0^t [\dot{\mu}_5 / (\mu_5)^2] Q_5 d\tau + Q_5 / \mu_5, \end{aligned} \quad (10)$$

where « \sim » denotes the initial conditions, and « \cdot », as usually, are time derivatives, we obtain extended system (5), (7)–(10) the solution of which is $(\mu_i, F)(t, V, Q_5(\cdot))$. These functions will be used below in constructing the stochastic model of the growth process and in non-linear filtering of the data. It should be noted here that it is known from the experimental investigations that the functions $\mu_1(\cdot)$ and $\mu_5(\cdot)$ for a particular organism do not vary considerably. Usually the range for $\mu_1 = 2 \dots 3$ and for $\mu_5 = 0.2 \dots 0.5$. This fact simplifies essentially the calculation of the functions.

In the cases when F and μ_i could be considered constants, at least approximately, the calculations described by eq. (5), (7)–(10) are essentially simplified.

2.2. Stochastic model of the growth of microorganisms

Deterministic model does not take into account uncontrolled factors (peculiarities of physiological state of the cells depending on the characteristics of inoculum, deviations from homogeneity in fermentor, etc) which cause batch-to-batch variations of the growth process [10]. Beside these 'intrinsic' to the growth process stochastic factors, the errors of the measurements should be also taken into account. These considerations lead to the construction of the stochastic model of the growth. The simplest stochastic model of the growth process is obtained in the case when $\{Q_5(t)\}$ is an observable process. Let us define the following model for the growth biomass:

$$dQ_4 = F(t, V, Q_5) Q_4 dt + \sigma_4 dw_4, \quad (11)$$

for the amount of consumed growth substrate:

$$dQ_1 = \mu_1(t, V, Q_5) dQ_4 + \sigma_1 dw_1, \quad (12)$$

and for the measurement procedures:

$$d\xi_i = \mu_i(t, V, Q_5) dQ_4 + r_i dw_i, \quad i = 2, 3. \quad (13)$$

Here $\mathbf{w} = (w_1, w_4, w_2, w_3)^T$ and (\mathbf{w}_i) is a standard Wiener process, σ_i , r_i — constants which characterize the noise of the process of substrate consumption and biomass growth, respectively, and r_i , $i = 2, 3$ — mean square deviations of the measurements errors.

If the process $\{Q_5(t)\}$ is observed with low noise (low observation error), the real values of Q_5 could be replaced in equations (8)–(10) by observed values ξ_5 , and an additional equation is added to the system (13)

$$d\xi_5 = \mu_i(t, V, \xi_5) dQ_4 + r_5 d\omega_5. \quad (14)$$

Proposed replacement does not complicate the filtering problem. However, if $\{Q_5(t)\}$ is measured with high level of noise (great observation error), then the filtering problem is much more complicated, it has infinite dimensions.

2.3. Filtering

From the analysis of the stochastic model of the growth process, equations (11)–(14), it follows that for each t conditional mean $\mathbf{m}_t = M(\mathbf{Q}_t / \xi_{[0,t]})$ and conditional covariation $\mathbf{H}_t = \text{Cov}(\mathbf{Q}_t / \xi_{[0,t]})$ of the unobserved vector $\mathbf{Q}_t = (Q_4(t), Q_1(t))^T$ are calculated from equations

$$d\mathbf{m} = \mathbf{a}\mathbf{m} dt + \mathbf{G}(\mathbf{B}\mathbf{B}^T + \mathbf{R})^{-1}(d\xi - \mathbf{A}\mathbf{m} dt), \quad (15)$$

$$\mathbf{H} = \mathbf{a}\mathbf{H} + \mathbf{H}\mathbf{a}^T + \mathbf{b}\mathbf{b}^T - \mathbf{G}(\mathbf{B}\mathbf{B}^T + \mathbf{R})^{-1}\mathbf{G}^T,$$

where

$$\mathbf{a} = F \begin{bmatrix} 1 & 0 \\ \mu_1 & 0 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} \sigma_4 & 0 \\ \mu_1 \sigma_4 & \sigma_1 \end{bmatrix}, \quad \mathbf{A} = F \begin{bmatrix} \mu \\ 0 \end{bmatrix}^T,$$

$$\mathbf{B} = (\sigma_4 \mu, 0), \quad \mathbf{G} = \mathbf{b}\mathbf{B}^T + \mathbf{H}\mathbf{A}^T,$$

$$\mu = (\mu_i), \quad \xi = (\xi_i), \quad i=2, 3, 5 \quad \text{and}$$

$$\mathbf{R} = (r_i^2) \quad \text{— diagonal matrix.}$$

If one of the parameters ξ_2 or ξ_3 is not measured in the experiment, the calculations could be carried out according to equations (15) after excluding the appropriate coordinates of vectors μ and ξ , and also the appropriate row and line from matrix \mathbf{R} .

2.4. Control algorithm

In many important, from the practical point of view, cases of cultivation of microorganisms, the aim of the control is to maximize the amount of biomass synthesized from the beginning of cultivation. It follows from eq. (8) that the formulated aim is approximately achieved if the amount of unutilized growth substance increases slowly according to equation

$$Q(t) = \{V_t / c_2 [V_t + c_3 Q_7(t)]\}^{1/2},$$

i. e. if the control is given by an approximate nonlinear equation

$$\alpha(t) \approx Q_1(t) + \{(V_t / c_2) [V_t + c_3 c_5 \int_0^t Q_4(\tau) d\tau]\}^{1/2}. \quad (16)$$

In the framework of the stochastic approach the simplest aim of the control is maximization of a mean value of the amount of biomass $M\{Q_4(t) / \xi_{[0,t]}\}$ synthesized from the beginning of cultivation. The aim is approximately achieved by the control:

$$\alpha(t) = m_1(t) + \{(V_t / c_2) [V_t + c_3 c_5 \int_0^t m_4(\tau) d\tau]\}^{1/2}, \quad (17)$$

where $\mathbf{m}_t = (m_4(t), m_1(t))^T$ is the solution of filtering eq. (15). It could be shown that the proposed (separated) control (17) is admissible in the sense that the system of eq. (14) has unique strong solution [20], and there exists the feedback control from the observations.

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*Academy on Sciences of the Estonian SSR,
Institute of Cybernetics*

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*Academy on Sciences of the Estonian SSR,
Institute of Chemical Physics and Biophysics*

R. TENNO, K. TIISMA, T. PAALME, R. VILU

BAKTERITE KASVU STOHHAŠTILISE JUHTIMISE ALGORITM NENDE KULTIVEERIMISEKS FED-BATCH-REŽIIMIS

Kasutades kasvu stöhhiomeetristil võrrandit ning Monod' laadi kineetika võrrandit on loodud deterministlik mudel, mis kirjeldab bakterikultuuri kasvu *fed-batch*-režiimis. Kasvu stohhastiline mudel on loodud deterministliku mudeli alusel. Väljaarendatud stohhastilise juhtimise algoritm *fed-batch*-kultiveerimisrežiimi jaoks põhineb deterministlikul juhtimisrežiimil, kus tundmatud suurused on asendatud filtreeritud kaudsete mõõtmistulemustega.

P. ТЕННО, К. ТИИСМА, Т. ПААЛМЕ, Р. ВИЛУ

АЛГОРИТМ СТОХАСТИЧЕСКОГО УПРАВЛЕНИЯ РОСТОМ БАКТЕРИИ В РЕЖИМЕ ПЕРИОДИЧЕСКОГО КУЛЬТИВИРОВАНИЯ С ПОДПИТКОЙ

С использованием стехиометрического уравнения роста и уравнения Моно построена детерминистическая модель роста бактериальной культуры в режиме периодического культивирования с подпиткой и положена в основу соответствующей стохастической модели. Алгоритм стохастического управления реализует детерминистический закон управления путем замены неизвестных параметров отфильтрованными данными косвенных измерений.