VILNIAUS UNIVERSITETAS MATEMATIKOS IR INFORMATIKOS FAKULTETAS PROGRAMŲ SISTEMŲ KATEDRA

Selektyvios membranos modeliavimas amperometriniame biojutiklyje

*angl. pav.

Kursinis darbas

Atliko: 4 kurso 1 grupės studentas

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(parašas)

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Įvadas

<Tektas> *.

1. Mono-Layer Mono-Enzyme Model of Biosensor

Formules were taken from [BIKK09].

1.1. Mathematical Model

According to simplified scheme, the substrate (S) binds to the enzyme (E) and is converted to poduct (P),

$$S \stackrel{E}{\to} P \tag{1}$$

$$\frac{\delta S}{\delta t} = D_S \frac{\delta^2 S}{\delta x^2} - \frac{V_{max} S}{K_M + S},$$

$$\frac{\delta P}{\delta t} = D_P \frac{\delta^2 P}{\delta x^2} + \frac{V_{max} S}{K_M + S}, \quad x \in (0; d), \quad t > 0$$
(2)

1.2. Initial and Boundary Conditions

$$S(x,0) = 0, x \in [0;d),$$

$$S(d,0) = S_0,$$
(3)

$$P(x,0) = 0, x \in [0;d),$$

$$P(d,0) = P_0,$$
(4)

$$P(0;t) = 0, \ t > 0. \tag{5}$$

$$D_S \frac{\delta S}{\delta x} \bigg|_{x=0} = 0, t > 0. \tag{6}$$

$$\left. \frac{\delta S}{\delta x} \right|_{x=0} = 0, t > 0. \tag{7}$$

$$S(d,t) = S_0, t > 0, (8)$$

$$P(d,t) = P_0, t > 0. (9)$$

1.3. Characteeristics of the Biosensor Response

1.3.1. Response Time

$$T = \min_{i(t)>0} \left\{ t : \frac{t}{i(t)} \left| \frac{d_i(t)}{dt} \right| < \epsilon \right\}$$
 (10)

1.4. Numerical Approximation of Equations

1.4.1. Implicit Scheme

$$\frac{S_{i}^{j+1} - S_{i}^{j}}{\tau} = D \frac{S_{i+1}^{j+1} - 2S_{i}^{j+1} + S_{i-1}^{j+1}}{h^{2}} - \frac{V_{max}S_{i}^{j}}{K_{M} + S_{i}^{j}},
\frac{P_{i}^{j+1} - P_{i}^{j}}{\tau} = D \frac{P_{i+1}^{j+1} - 2P_{i}^{j+1} + P_{i-1}^{j+1}}{h^{2}} + \frac{V_{max}S_{i}^{j}}{K_{M} + S_{i}^{j}},
i = 1, ..., N - 1, j = 0, ..., M - 1.$$
(11)

1.4.2. Initial and Boundry Condtitions

$$S_i^0 = 0, i = 0, ..., N - 1,$$

 $S_N^0 = S_0,$ (12)

$$P_i^0 = 0, i = 0, ..., N - 1,$$

 $P_N^0 = P_0$ (13)

$$S_0^j = S_1^j, S_N^j = S_0, j = 1, ..., M,$$
 (14)

$$P_0^j = 0, P_N^j = P_0, j = 1, ..., M,$$
 (15)

1.4.3. Calculation Procedure

$$i(t_j) \approx t_j = n_e F D_p P_i^j / h, j = 1, ..., M.$$
 (16)

1.5. Validation of Numerical Solution

1.5.1. Inicial Conditions of Calculations

$$S_{0} = 1\mu M, P_{0} = 0\mu M,$$

$$D_{s} = 300\mu m^{2}/s, D_{p} = 300\mu m^{s}/s,$$

$$T = 50s, d_{1} = 10\mu m, d_{2} = 15\mu m, d_{3} = 100\mu m; d_{4} = 150\mu m,$$

$$\tau = 0.1s, h = 0.1\mu m,$$

$$V_{max} = 100\mu M/s, \epsilon = 0.05,$$

$$F = 961485, K_{M} = 100\mu M, n_{e} = 2.$$

$$(17)$$

1.5.2. Obtained Matrixes from the Given Implicit Scheme

Used: Finite Difference Solution, Implicit Scheme. Solved using Tridiagonal Matrix Algorythm.

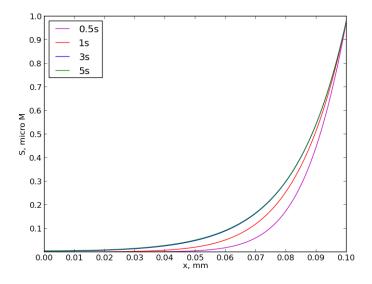
$$\begin{cases} S_{i+1}^{j+1} - S_{i}^{j+1} \left(1 + \frac{h^{2}}{D\tau} \right) = \frac{h^{2}}{D\tau} \left(\frac{V_{m} ax S_{i}^{j} \tau}{K_{M} + S_{i}^{j}} - S_{i}^{j} \right), when j = 0, \\ \dots, \\ S_{i-1}^{j+1} - S_{i}^{j+1} \left(2 + \frac{h^{2}}{D\tau} \right) + S_{i+1}^{j+1} = \frac{h^{2}}{D\tau} \left(\frac{V_{m} ax S_{i}^{j} \tau}{K_{M} + S_{i}^{j}} - S_{i}^{j} \right), when j = 1, \dots, M - 1, \\ \dots, \\ S_{0} + S_{i-1}^{j+1} - S_{i}^{j+1} \left(2 + \frac{h^{2}}{D\tau} \right) = \frac{h^{2}}{D\tau} \left(\frac{V_{m} ax S_{i}^{j} \tau}{K_{M} + S_{i}^{j}} - S_{i}^{j} \right), when j = M. \end{cases}$$

$$(18)$$

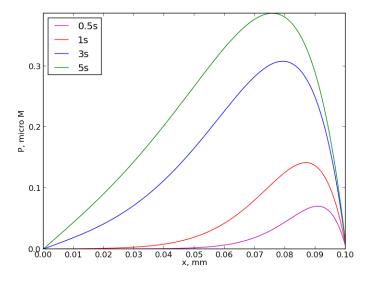
$$\begin{cases} P_{i+1}^{j+1} - P_{i}^{j+1} \left(2 + \frac{h^{2}}{D\tau} \right) = \frac{h^{2}}{D\tau} \left(-P_{i}^{j} - \frac{V_{m}axS_{i}^{j}\tau}{K_{M} + S_{i}^{j}} \right), when j = 0, \\ \dots, \\ P_{i-1}^{j+1} - P_{i}^{j+1} \left(2 + \frac{h^{2}}{D\tau} \right) + P_{i+1}^{j+1} = \frac{h^{2}}{D\tau} \left(\frac{V_{m}axS_{i}^{j}\tau}{K_{M} + S_{i}^{j}} - S_{i}^{j} \right), when j = 1, \dots, M - 1, \\ \dots, \\ P_{0} + P_{i-1}^{j+1} - P_{i}^{j+1} \left(2 + \frac{h^{2}}{D\tau} \right) = \frac{h^{2}}{D\tau} \left(\frac{V_{m}axS_{i}^{j}\tau}{K_{M} + S_{i}^{j}} - S_{i}^{j} \right), when j = M. \end{cases}$$

$$(19)$$

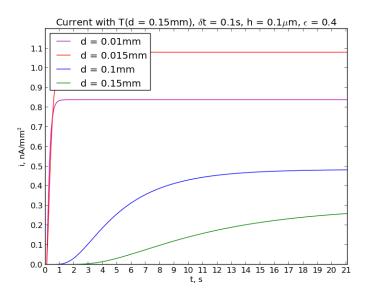
1.5.3. Results



1 pav.: Substrate



2 pav.: Product



3 pav.: Current

Literatūros sąrašas

[BIKK09] Romas Baronas, Feliksas Ivanauskas, IU IU Kulis, and Juozas Kulys. *Mathematical modeling of biosensors: an introduction for chemists and mathematicians*, volume 9. Springer, 2009.