

Biological Information Modeling

September 7, 2018

Abstract

Keywords:

Mathematics Subject Classification 2000:

1 Introduction

The describing of the biological problems :

1. the background information (by WX)
2. the expression of the biological experiments (by WX)

2 Mathematical Formulations

1. We use the following notations :

- (i) P_R stands by ...
- (ii) T_R stands by ...
- (iii) P_L stands by ...
- (iv) T_L stands by ...
- (v) a stands by ...
- (vi) b stands by ...
- (vii) c stands by ...

All these functions are positive.

The differential equations for the evolutions of genes :

$$\begin{aligned}\dot{P}_R &= a - cP_R, \\ \dot{T}_R &= a - b(T_R - P_R),\end{aligned}\tag{2.1}$$

where \dot{P}_R and \dot{T}_R are the derivatives of the two functions $P_R(t)$ and $T_R(t)$ with respect to t and the coefficients a, b and c are also functions of t .

2. The formulations for labeled genes from the labeling time t^* .

Since the labeling time is very small with respect to the evolution of genes, the coefficient $b = 0$ for labeling genes and $M_L(t^*) = 0$ at the labeling time. Hence solving (2.1), we obtain the formulation of the labeling genes:

$$\begin{aligned}P_L(t) &= \frac{a_t}{c_t}(1 - e^{-c_t t_L}), \\ T_L(t) &= a_t t_L.\end{aligned}\tag{2.2}$$

3 Fundamental Assumptions

In [1], it is assumed that

- (i) $a(t)$ is a ... in each time-periodic.
- (ii) $b(t)$ is a ... in each time-periodic.
- (iii) $c(t)$ is a ... in each time-periodic.

Hence solving (2.1) and (2.2), we obtain the formulations in [1] as follows :

$$\begin{aligned} a &= \dots \\ b &= \dots \\ c &= \dots \end{aligned}$$

While the problem is :

for two almost same genes, the correlations of a , b and c are very small.

4 Improvement of Models

To overcome this problem, we assume that the coefficients a , b and c are continuous functions and the variations of a , b and c are very small, i.e.,

$$\int_{t_0}^t |\dot{a}(s)|^2 + |\dot{b}(s)|^2 + |\dot{c}(s)|^2 ds$$

is small as possible. The reason is that the ... of each gene does not change very fast.

5 Improvement of Mathematical Formulations

Let $(P_R(t_0), T_R(t_0), P_L(t_0), T_L(t_0))$ and $(P_R(t_1), T_R(t_1), P_L(t_1), T_L(t_1))$ be obtained at two time points and t_L be the given labeling-time.

- (i) We insert several time points in $[t_0, t_1]$ labeled by $s_i = t_0 + \frac{i(t_1 - t_0)}{N}$, where N is an integer and $i = 0, \dots, N$. We label the approximation values of the functions a , b and c at each time periodic $[s_i, s_{i+1}]$ by a^i , b^i and c^i , $i = 0, \dots, N-1$ and the approximations values of the functions a , b and c at the time point $s_N = t_1$ by a^N , b^N and c^N .
- (ii) With $(P_L(t_0), T_L(t_0))$, $(P_L(t_1), T_L(t_1))$ and t_L , we obtain from (2.2) that

$$\begin{aligned} a^0 &= T_L(t_0)/t_L \\ a^N &= T_L(t_1)/t_L \\ \frac{c^0}{1 - e^{-c^0 t_L}} &= \frac{T_L(t_0)}{P_L(t_0)t_L} \\ \frac{c^N}{1 - e^{-c^N t_L}} &= \frac{T_L(t_1)}{P_L(t_1)t_L} \end{aligned}$$

- (iii) With the approximation coefficients a^i, b^i and c^i , we solve (2.1) in $[s_i, s_{i+1}]$ by the given initial value (P_R^i, T_R^i) to obtain the approximation value (P_R^{i+1}, T_R^{i+1}) for $i = 0, \dots, N-1$,

$$\begin{pmatrix} P_R^{i+1} \\ T_R^{i+1} \end{pmatrix} = \begin{pmatrix} P_R^i \\ T_R^i \end{pmatrix} + h \begin{pmatrix} a^i - c^i P_R^i \\ a^i - b^i(T_R^i - P_R^i) \end{pmatrix}$$

Here $h = (t_1 - t_0)/N$ and $(P_R^0, T_R^0) = (P_R(t_0), T_R(t_0))$.

After this process, we obtain an approximation value (P_R^N, T_R^N) to the final value $(P_R(t_1), T_R(t_1))$.

- (iv) The changing rate of the coefficients in the time periodic $[s_i, s_{i+1}]$ are approximated by

$$\begin{aligned} |\dot{a}(s)|^2 &\approx \left(\frac{a^{i+1} - a^i}{h} \right)^2, \\ |\dot{b}(s)|^2 &\approx \left(\frac{b^{i+1} - b^i}{h} \right)^2, \\ |\dot{c}(s)|^2 &\approx \left(\frac{c^{i+1} - c^i}{h} \right)^2, \end{aligned}$$

for $i = 0, \dots, N-1$. Hence the total change of the coefficients are approximated by

$$J(a, b, c) = \frac{1}{2} \sum_{i=0}^{N-1} \frac{(a^{i+1} - a^i)^2}{h} + \frac{(b^{i+1} - b^i)^2}{h} + \frac{(c^{i+1} - c^i)^2}{h},$$

which has an equivalent quadrature form

$$J(X) = \frac{1}{2} X^T A X + B^T X + C,$$

where $X = (a^1, \dots, a^{N-1}, b^0, \dots, b^N, c^1, \dots, c^{N-1})^T \in \mathbb{R}^{3N-1}$ and the matrix $A \in \mathbb{R}^{(3N-1) \times (3N-1)}$, the vector $B \in \mathbb{R}^{3N-1}$ and $C \in \mathbb{R}$ are given by

$$\begin{aligned} A &= \frac{N}{2(t_1 - t_0)} \begin{pmatrix} \bar{I}_{N-1} & & \\ & I_{N+1} & \\ & & \bar{I}_{N-1} \end{pmatrix}_{(3N-1) \times (3N-1)}, \\ B &= \frac{N}{t_1 - t_0} (-a^0, \quad 0_{(N-3)}, \quad -a^N, \quad 0_{(N+1)}, \quad -c^0, \quad 0_{(N-3)}, \quad -c^N)^T, \\ C &= \frac{N}{2(t_1 - t_0)} ((a^0)^2 + (a^N)^2 + (c^0)^2 + (c^N)^2), \end{aligned}$$

where

$$I_N = \begin{pmatrix} 1 & -1 & 0 & \cdots & 0 \\ -1 & 2 & -1 & \cdots & 0 \\ 0 & -1 & 2 & \cdots & 0 \\ 0 & \cdots & -1 & 2 & -1 \\ 0 & \cdots & 0 & -1 & 1 \end{pmatrix}_{N \times N},$$

and

$$\bar{I}_N = \begin{pmatrix} 2 & -1 & 0 & \cdots & 0 \\ -1 & 2 & -1 & \cdots & 0 \\ 0 & -1 & 2 & \cdots & 0 \\ 0 & \cdots & -1 & 2 & -1 \\ 0 & \cdots & 0 & -1 & 2 \end{pmatrix}_{N \times N}.$$

- (v) The coefficients a , b and c are solved by the following optimal problem with positive and nonlinear constraints, i.e.,

$$\begin{aligned} X &= \arg \min J(X) \\ \text{s.t. } X &> 0, \\ P_R^N(X) &= P_R(t_1), \\ T_R^N(X) &= T_R(t_1). \end{aligned}$$

6 Improvement of Solving Programs

- (i) Solving the differential equations with a higher-order method such as Runge's method.

$$\begin{aligned} \begin{pmatrix} K_1^i \\ L_1^i \end{pmatrix} &= \begin{pmatrix} a^i - c^i P_R^i \\ a^i - b^i (T_R^i - P_R^i) \end{pmatrix}, \\ \begin{pmatrix} K_2^i \\ L_2^i \end{pmatrix} &= \begin{pmatrix} a^i - c^i (P_R^i + h/2 K_1^i) \\ a^i - b^i ((T_R^i + h/2 L_1^i) - (P_R^i + h/2 K_1^i)) \end{pmatrix}, \\ \begin{pmatrix} K_3^i \\ L_3^i \end{pmatrix} &= \begin{pmatrix} a^i - c^i (P_R^i + h/2 K_2^i) \\ a^i - b^i ((T_R^i + h/2 L_2^i) - (P_R^i + h/2 K_2^i)) \end{pmatrix}, \\ \begin{pmatrix} K_4^i \\ L_4^i \end{pmatrix} &= \begin{pmatrix} a^i - c^i (P_R^i + h K_3^i) \\ a^i - b^i ((T_R^i + h L_3^i) - (P_R^i + h K_3^i)) \end{pmatrix}, \\ \begin{pmatrix} P_R^{i+1} \\ T_R^{i+1} \end{pmatrix} &= \begin{pmatrix} P_R^i \\ T_R^i \end{pmatrix} + \frac{h}{6} \left(\begin{pmatrix} K_1^i \\ L_1^i \end{pmatrix} + 2 \begin{pmatrix} K_2^i \\ L_2^i \end{pmatrix} + 2 \begin{pmatrix} K_3^i \\ L_3^i \end{pmatrix} + \begin{pmatrix} K_4^i \\ L_4^i \end{pmatrix} \right). \end{aligned}$$

- (ii) By introducing a penalty factor, the optimal problem is transformed into

$$\begin{aligned} \min J(X) + 0.5\alpha_1 (P_R^N(X) - P_R(t_1))^2 + 0.5\alpha_2 (T_R^N(X) - T_R(t_1))^2 \\ \text{s.t. } X > 0, \end{aligned}$$

where α_1, α_2 are positive large numbers. Then the gradient descent method is given by

$$\begin{aligned} X_{k+1} = X_k - \rho_k \left(\frac{\partial J}{\partial X}(X_k) + \alpha_1 (P_R^N(X_k) - P_R(t_1)) \frac{\partial P_R^N}{\partial X}(X_k) \right. \\ \left. + \alpha_2 (T_R^N(X_k) - T_R(t_1)) \frac{\partial T_R^N}{\partial X}(X_k) \right) \end{aligned}$$

where ρ_k is a small number such that $X_{k+1} > 0$ piecewise component. Of course

$$\frac{\partial J}{\partial X}(X_k) = AX_k + B.$$

Hence in the following, we present an approach for the approximations to $\frac{\partial P_R^N}{\partial X}(X_k)$ and $\frac{\partial T_R^N}{\partial X}(X_k)$.

(iii) Following from (2.1), in each $[s_i, s_{i+1}]$, one obtains that

$$\begin{aligned}
\frac{d}{dt} \left(\frac{\partial P_R}{\partial X} \right)_{3N-1} &= (\eta_i)_{3N-1} - P_R(\xi_i)_{3N-1} - c_i \left(\frac{\partial P_R}{\partial X} \right)_{3N-1} \\
&=: f_i \left(P_R, \left(\frac{\partial P_R}{\partial X} \right)_{3N-1} \right), \\
\frac{d}{dt} \left(\frac{\partial T_R}{\partial X} \right)_{3N-1} &= (\eta_i)_{3N-1} - (T_R - P_R)(\theta_i)_{3N-1} \\
&\quad - b_i \left(\left(\frac{\partial T_R}{\partial X} \right)_{3N-1} - \left(\frac{\partial P_R}{\partial X} \right)_{3N-1} \right) \\
&=: g_i \left(T_R - P_R, \left(\frac{\partial P_R}{\partial X} \right)_{3N-1}, \left(\frac{\partial T_R}{\partial X} \right)_{3N-1} \right),
\end{aligned} \tag{6.1}$$

where $\frac{\partial P_R}{\partial X}, \frac{\partial T_R}{\partial X}, \eta_i, \xi_i, \theta_i \in \mathbb{R}^{3N-1}$ are given by

$$\begin{aligned}
\eta_0 &= 0, \\
\eta_i &= (e_i, 0_{N+1}, 0_{N-1})^T, i = 1, \dots, N-1, \\
\xi_0 &= 0, \\
\xi_i &= (0_{N-1}, 0_{N+1}, e_i)^T, i = 1, \dots, N-1, \\
\theta_i &= (0_{N-1}, \bar{e}_{i+1}, 0_{N-1})^T, i = 0, \dots, N-1.
\end{aligned}$$

Here $\frac{\partial P_R}{\partial X}(0) = 0$, $\frac{\partial T_R}{\partial X}(0) = 0$, $e_i = (0, \dots, 0, 1, 0, \dots, 0)^T \in \mathbb{R}^{N-1}$, $i = 1, \dots, N-1$, and $\bar{e}_i = (0, \dots, 0, 1, 0, \dots, 0)^T \in \mathbb{R}^{N+1}$, $i = 1, \dots, N+1$.

(iv) Runge's method for (6.1) from s_i with $G_i^P \approx \frac{\partial P_R}{\partial X}(s_i)$ and $G_i^T \approx \frac{\partial T_R}{\partial X}(s_i)$ to s_{i+1} with $G_{i+1}^P \approx \frac{\partial P_R}{\partial X}(s_{i+1})$ and $G_{i+1}^T \approx \frac{\partial T_R}{\partial X}(s_{i+1})$ is given by :

$$\begin{aligned}
\begin{pmatrix} D_1^i \\ E_1^i \end{pmatrix} &= \begin{pmatrix} f_i(P_R^i, G_i^P) \\ g_i(T_R^i - P_R^i, G_i^P, G_i^T) \end{pmatrix}, \\
\begin{pmatrix} D_2^i \\ E_2^i \end{pmatrix} &= \begin{pmatrix} f_i(P_R^i + h/2K_1^i, G_i^P + h/2D_1^i) \\ g_i((T_R^i + h/2L_1^i) - (P_R^i + h/2K_1^i), G_i^P + h/2D_1^i, G_i^T + h/2E_1^i) \end{pmatrix}, \\
\begin{pmatrix} D_3^i \\ E_3^i \end{pmatrix} &= \begin{pmatrix} f_i(P_R^i + h/2K_2^i, G_i^P + h/2D_2^i) \\ g_i((T_R^i + h/2L_2^i) - (P_R^i + h/2K_2^i), G_i^P + h/2D_2^i, G_i^T + h/2E_2^i) \end{pmatrix}, \\
\begin{pmatrix} D_4^i \\ E_4^i \end{pmatrix} &= \begin{pmatrix} f_i(P_R^i + hK_3^i, G_i^P + hD_3^i) \\ g_i((T_R^i + hL_3^i) - (P_R^i + hK_3^i), G_i^P + hD_3^i, G_i^T + hE_3^i) \end{pmatrix}, \\
\begin{pmatrix} G_{i+1}^P \\ G_{i+1}^T \end{pmatrix} &= \begin{pmatrix} G_i^P \\ G_i^T \end{pmatrix} + \frac{h}{6} \left(\begin{pmatrix} D_1^i \\ E_1^i \end{pmatrix} + 2 \begin{pmatrix} D_2^i \\ E_2^i \end{pmatrix} + 2 \begin{pmatrix} D_3^i \\ E_3^i \end{pmatrix} + \begin{pmatrix} D_4^i \\ E_4^i \end{pmatrix} \right).
\end{aligned}$$

(v) Implementation.

(a) Given initial values $(P_R(t_0), T_R(t_0), P_L(t_0), T_L(t_0))$, $(P_R(t_1), T_R(t_1), P_L(t_1), T_L(t_1))$ and t_L .

- (b) Given an integer $N \geq 1$, $\alpha_1, \alpha_2 \gg 1$, $\rho \in (0, 1)$ and $\epsilon \ll 1$.
- (c) Calculate a^0, a^N, c^0, c^N .
- (d) Set $k = 0$ and give an initial guess X_0 for the other coefficients.
- (e) Calculate P_R^N and T_R^N by Runge's method for (2.1).
- (f) Calculate G_N^P and G_N^T by Runge's method for (6.1)
- (g) Calculate the gradient

$$g_k = \frac{\partial J}{\partial X}(X_k) + \alpha_1(P_R^N(X_k) - P_R(t_1)) \frac{\partial P_R^N}{\partial X}(X_k) + \alpha_2(T_R^N(X_k) - T_R(t_1)) \frac{\partial T_R^N}{\partial X}(X_k)$$

- (h) If $\max |g_k| < \epsilon$, the program is stopped. Otherwise update the guessing value by

$$X_{k+1} = X_k - \rho \frac{\min\{X_k\}}{\max |g_k|} g_k,$$

set $k = k + 1$ and go to Step (e).

References

- [1] INSPEcT