CELLULAR GROWTH AND BERWALD CURVATURES

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1. Introductory biological background

Genome= the total collection of genetic material in the chromosomes of an organism.

 $\mathbf{DNA}=$ a double stranded molecule made out of two complementary strands of equal length containing four bases: adenine (A), guanine (G), cytosine (C) and thynine (T), held together (quite strongly) by hydrogen bands.

 \mathbf{mRNA} = the twin sister of a single stranded *DNA*: it keeps the same A, C, G bases while T (thynine) is replaced by U (uracil).

Protein= a sequence of amino acids (20 types).

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DNA Structure



2. Central Dogma of Molecular Biology

What is a gene?

The unit of hereditary information that occupies a fixed position (locus) on chromosomes (*Encyclopedia Britannica*).

A gene is a segment of DNA together with a transformation to a segment (s) of mRNA (A. Carbone and M. Gromov, [CG]).

-alternative splicing in eukariotic cells, i.e. the same segment of **DNA** may lead to the production of different **mRNA**s

-overlapping genes, i.e. different segments of mRNA are produced from overlapping segments of DNA.

 $DNA \xrightarrow{F} mRNA \xrightarrow{G} protein$

gene = $(G \circ F)^{-1}$ (protein)







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3. Differential equations



$$\frac{dr}{dt} = f(p) - V \cdot r \qquad \frac{dp}{dt} = L \cdot r - U \cdot p,$$

where the variables r = r(t), p = p(t) are functions of time, and:

n	The number of genes in the genome.
r	mRNA concentrations, <i>n</i> -dimensional vector-valued functions of t .
p	Protein concentrations, n -dimensional vector-valued functions of t .
f(p)	Transcription functions, n -dimensional vector polynomials on p .
L	Translations constants, $n \times n$ nondegenerate diagonal matrix.
V	Degradations rates of $mRNA$'s, $n \times n$ nondegenerate diagonal matrix.
U	Degradations rates of Proteins, $n \times n$ nondegenerate diagonal matrix.

• mRNA production: $\frac{d^2r}{dt^2} + 2G(r, \dot{r}) = 0,$ • Protein production: $\frac{d^2p}{dt^2} + 2H(p, \dot{p}) = 0.$ Finslerian example: $F^2 = \frac{1}{2}e^{2(\alpha x^1 + \beta x^2)} \left[\left(\frac{dx^1}{dt}\right)^2 + \left(\frac{dx^2}{dt}\right)^2 \right]$ (Volterra-Hamilton stars)

system).

4. Michaelis-Menten kinetics

- Enzyme
- \bullet Substrate
- Biochemical reactions

association : two proteins combine together to form a complex

disassociation : a substrate splits in two reaction products: an enzyme and a product.

$$S + E \underset{k_{-1}}{\overset{k_1}{\longleftrightarrow}} SE \overset{k_2}{\longrightarrow} P + E_1$$

S : substrate; E : enzyme (catalyst); P: product,

 k_1, k_{-1}, k_2 constant rate parameters.

• Law of mass action

$$\frac{ds}{dt} = -k_1 es + k_{-1} c, \qquad \frac{de}{dt} = -k_1 es + (k_{-1} + k_2) c, \\
\frac{dc}{dt} = k_1 es - (k_{-1} + k_2) c, \qquad \frac{dp}{dt} = k_2 c.$$

- s, e, c, p: concentrations of the reactants S, E, SE, P.
- Phase plane analysis.
- Hopf bifurcation and limit cycles.

5. SIBIL: a software for simulations of biological systems

The Holy Grail of Molecular Biology: construct faithful mathematical model of living cell (genes, proteins, protein complexes, mRNA, etc.).

• System architecture.



• Circadian rhythm in Drosophila.



6. Differential system in Drosophila

$$\frac{dPer_m}{dt} = S_1 \cdot \frac{\left(\frac{CC_p}{P_1}\right)^p}{1 + \left(\frac{PT_n}{M_1}\right)^m + \left(\frac{CC_n}{P_1}\right)^p} - D_1 \cdot \frac{Per_m}{L_1 + Per_n} - d \cdot Per_m$$
$$\frac{dPer_c}{dt} = S_2 \cdot Per_m - A_1 \cdot Per_c \cdot Tim_c + A_2 \cdot PT_c - D_2 \cdot Dbt_c \cdot \frac{Per_c}{L_2 + Per_c} - d \cdot Per_c$$

$$\frac{dTim_m}{dt} = S_3 \cdot \frac{\left(\frac{CC_n}{Q_1}\right)^q}{1 + \left(\frac{PT_n}{N_1}\right)^n + \left(\frac{CC_n}{Q_1}\right)^q} - D_3 \cdot \frac{Tim_m}{L_3 + Tim_m} - d \cdot Tim_m$$

$$\frac{dTim_c}{dt} = S_4 \cdot Tim_m - A_1 \cdot Per_c \cdot Tim_c + A_2 \cdot PT_c - D_4 \cdot \frac{Tim_c}{L_4 + Tim_c} - d \cdot Tim_c$$

$$\frac{dPT_c}{dt} = A_1 \cdot Per_c \cdot Tim_c - A_2 \cdot PT_c - V_1 \frac{PT_c}{K_1 + PT_c} + V_2 \frac{PT_n}{K_2 + PT_n}$$

$$- D_5 \frac{PT_c}{L_4 + Tim_c} - d \cdot PT_c$$

$$\frac{dPT_n}{dt} = V_1 \frac{PT_c}{K_1 + PT_c} - V_2 \frac{PT_n}{K_2 + PT_n} - D_6 \frac{PT_n}{L_6 + PT_n} - d \cdot PT_n$$
$$\frac{dClk_n}{dt} = S_5 \cdot \frac{\left(\frac{PT_n}{R_1}\right)^r}{1 + \left(\frac{CC_c}{O_1}\right)^\circ + \left(\frac{PT_n}{R_1}\right)^r} - D_7 \cdot \frac{Clk_n}{L_7 + Clk_n} - d \cdot Clk_n$$

$$\begin{aligned} \frac{dCC_c}{dt} &= A_3 \cdot Clk_c \cdot Cyc_c - A_4 \cdot CC_c - V_3 \frac{CC_c}{K_3 + CC_c} + V_4 \frac{CC_n}{K_4 + CC_n} \\ &\quad - D_9 \frac{CC_c}{L_9 + CC_c} - d \cdot CC_c \end{aligned}$$
$$\begin{aligned} \frac{dCC_n}{dt} &= V_3 \frac{CC_c}{K_3 + CC_c} - V_4 \frac{CC_n}{K_4 + CC_n} - D_{10} \frac{CC_n}{L_{10} + CC_n} - d \cdot CC_n \\ \frac{dClk_c}{dt} &= S_6 \cdot Clk_m - A_3 \cdot Clk_c \cdot Cyc_c + A_4 \cdot CC_c - D_8 \cdot \frac{Clk_c}{L_8 + Clk_c} - d \cdot Clk_c, \end{aligned}$$

where we have 10 variables of time

 $\begin{array}{ll} Tim_m, Per_m, Clk_m, & \text{mRNAs} \\ Tim_c, Per_c, Clk_c, & \text{proteins translated from mRNAs} \\ PT_c, PT_n, CC_c, CC_n, & \text{protein complexes} \\ Dbt_c, Cyc_c, & \text{catalysts, (i.e. constants)} \\ \text{and 56 constant parameters } S_1, \dots, S_6; D_1, \dots, D_{10}; A_1, \dots, A_4; B_1, B_2, B_3; C_1, C_2, \\ C_3; M_1, N_1, O_1, P_1, Q_1, R_1, K_1, \dots, K_4; L_1, \dots, L_{10}; V_1, \dots, V_4; o, p, q, r, m, n, d. \end{array}$

7. A simple example: Tyson's Model for the Cell Division Cycle

• Cell cycle



• Early embryonic cell cycle

$$\left| \stackrel{S}{\rightarrow} \right| \stackrel{M}{\rightarrow} \left| \stackrel{S}{\rightarrow} \right| \stackrel{M}{\rightarrow} \left| \rightarrow \dots \right|$$

• Maturation promoting factor (MPF) peaks abruptly at metaphase (M).

 $cyclin + cdc \rightarrow MPF.$

• The cell cycle regulatory pathway.



• Differential system of cell cycle regulatory pathway.

$$\begin{aligned} \frac{dC_2}{dt} &= k_6 \cdot M - k_8 \cdot P \cdot C_2 + k_9 \cdot CP \\ \frac{dCP}{dt} &= -k_3 \cdot CP \cdot Y + k_8 \cdot P \cdot C_2 - k_9 \cdot CP \\ \frac{dpM}{dt} &= k_3 \cdot CP \cdot Y - pM \cdot F(M) + k_5 \cdot P \cdot M \\ \frac{dM}{dt} &= pM \cdot F(M) - k_5 \cdot P \cdot M - k_6 \cdot M \\ \frac{dY}{dt} &= k_1 \cdot aa - k_2 \cdot Y - k_3 \cdot CP \cdot Y \\ \frac{dYP}{dt} &= k_6 \cdot M - k_7 \cdot YP, \end{aligned}$$

where $F(M) = k + k_4 \frac{M}{CT^2}$. • Tyson model.

$$\frac{du}{dt} = (v - u)(k + k_4 u^2) - k_6 u$$
$$\frac{dv}{dt} = k_1 - k_6 u,$$

u and v are the relative concentration of active MPF and total cyclin minus degraded cyclin relative to total cdc2, respectively.

Parameter ranges:

 $k := 0.018 \text{ min}^{-1}$ rate constant for the dephosphorylation of cdc2,

 $k_1 := 0.015 \text{ min}^{-1}$ rate constant of cyclin synthesis,

 k_4 : = 10 - 1000 min⁻¹ (adjustable) constant describing the auto catalytic activation of MPF by the dephosphorylation of cdc2,

 $k_6 := 0.1 - 10 \text{ min}^{-1}$ (adjustable) constant describing breakdown of the active cdc2-cyclin complex.

8. The KCC Theory

Find the basic invariants of the SODE

$$\begin{aligned} \frac{d^2 x^i}{dt^2} + g^i(x, \dot{x}, t) &= 0, \ i \in \{1, 2, \dots, n\}, \\ x(t_0) &= x_0, \dot{x}(t_0) = \dot{x}_0, \ (x, \dot{x}, t) \in \Omega \subset \mathbb{R}^n \times \mathbb{R}^n \times \mathbb{R}^1 \end{aligned}$$

under the nonsingular coordinate transformation

$$\bar{x}^i = f^i(x^1, \dots, x^n), \ i \in \{1, 2, \dots, n\},\ \bar{t} = t.$$

• KCC-covariant differential

$$\frac{D\xi^i}{dt} = \frac{d\xi^i}{dt} + \frac{1}{2}g^i_{;r}\xi^r,$$

";" partial differentiation with respect to \dot{x} .

• First KCC- invariant

$$\frac{D\dot{x}^i}{dt} = \frac{1}{2}g^i_{;r}\dot{x}^r - g^i = \varepsilon^i$$

Lemma. The functions $g^i = g^i(x, \dot{x}, t)$ are 2 homogeneous in \dot{x} if and only if $\varepsilon^i = 0$.

Variation of the trajectories into nearby one

$$\bar{x}^i(t) = x^i(t) + \xi^i(t)\eta.$$

The variational equations

$$\frac{d^2\xi^i}{dt^2} + g^i_{\;;\,r} \; \frac{d\xi^r}{dt} + g^i_{\;,\,r}\xi^r = 0,$$

"," indicates partial differentiation with respect to x^r .

$$\frac{D^2\xi^i}{dt^2} = P_r^i\xi^r,$$

where

$$P_{j}^{i} = -g^{i}_{,j} - \frac{1}{2}g^{r}g^{i}_{;r;j} + \frac{1}{2}\dot{x}^{r}g^{i}_{,r;j} + \frac{1}{4}g^{i}_{;r}g^{r}_{;j} + \frac{1}{2}\frac{\partial g^{i}_{;j}}{\partial t}$$

The Second KCC-invariant of the system or deviation curvature tensor.

The third, fourth and fifth invariants are:

$$\begin{split} R^{i}_{jk} &= \frac{1}{3} \left(P^{i}_{j;k} - P^{i}_{k;j} \right), \\ B^{i}_{jk\ell} &= R^{i}_{jk;\ell}, \\ D^{i}_{jk\ell} &= g^{i}_{j;k;\ell}. \end{split}$$

A basic result of the KCC-theory is the following

Theorem A. Two SODE's on Ω can be locally transformed, one into the other, if and only if their five KCC-invariants ε^i , P^i_j , R^i_{jk} , $B_j{}^i{}_{k\ell}$,

 $D_j{}^i{}_{k\ell}$ are equivalent tensors. In particular, there are local coordinates (\bar{x}) for which $g^i(\bar{x}, \dot{x}, t) = 0$ if and only if all five KCC-tensors vanish.

9. Stability of trajectories

• Linear stability: the Liapunov stability of the steady states.

Theorem B. If a first order system has steady state (x_0, y_0) , it is stable iff the real part of the eigen values of its Jacobian at (x_0, y_0) are strict negative. If the eigen values have only imaginary part, then (x_0, y_0) is neutrally stable, otherwise it is unstable.

• Jacobi stability: the Liapunov stability of the whole trajectories, or transient states.

Theorem C. The trajectories of the SODE are JACOBI STABLE iff the real parts of the eigenvalues of the deviation curvature tensor (P_j^i) are strict negative everywhere in Ω , otherwise it is JACOBI UNSTABLE.

The eigenstructure of the deviation curvature tensor P_i^j is an alternative to the classical Floquet theory, with the eigenvalues of P_i^j replacing the Floquet exponents.

10. The one dimensional case

Let us consider the FODE

$$\dot{x}^1 = F(x^1, x^2), \ \dot{x}^2 = G(x^1, x^2)$$

and let us assume that we can eliminate one variable so that we obtain the SODE

$$\dot{x} = y, \ \dot{y} = -g^1(x, y),$$

with a steady-state given by $(x_0, 0)$.

The variational equation

$$\ddot{\xi} + (g^1_{;1})_0 \ \dot{\xi} + (g^1_{,1})_0 \ \xi = 0,$$

The coefficients are evaluated at some fixed reference trajectory $(x_0(t), y_0(t))$ and are functions of the parameter along this curve, t.

• Second invariant: $P_1^1 = -g_{,1}^1 - \frac{1}{2}g^1g_{;1;1}^1 + \frac{1}{2}yg_{,1;1}^1 + \frac{1}{4}g_{;1}^1g_{;1}^1$.

Consequence.

- a. The trajectories of SODE are Jacobi stable in Ω if and only if $P_1^1 < 0$ everywhere in Ω . This is equivalent to periodic deviation for the FODE.
- b. The trajectories are Jacobi unstable in Ω if and only if $P_1^1 \ge 0$ everywhere in Ω . This is equivalent to aperiodic deviation for the FODE.
 - Linear stability analysis of the steady states.
 - Jacobi stability analysis of the whole trajectories.

11. Jacobi stability for cell cycle

• Tyson model (FODE)

$$\frac{du}{dt} = (v - u)(k + k_4 u^2) - k_6 u$$
$$\frac{dv}{dt} = k_1 - k_6 u,$$

u and v are the relative concentration of active MPF and total cyclin minus degraded cyclin relative to total cdc2.

• The associated SODE

$$\ddot{x} + g(x, y) = 0,$$

where x = v, y = dx/dt and

(4.2)
$$g = Ay^{3} + (Bx + C)y^{2} + (Dx + E)y + Fx + G$$

The constants A, B, C, D, E, F, G can be expressed by means of the parameters k_1, k_4, \ldots

The embrionic cell development can be described by the Jacobi stability analysis of production of the total cyclin (relative to the total cdc2).

The second invariant for the Tyson model:

$$\begin{split} P_1^1 &= -\frac{3}{4} A^2 y^4 + (-A B x - A C) y^3 + (-\frac{3}{2} E A - \frac{3}{2} D A x) y^2 \\ &+ (-\frac{1}{2} D - 3 F A x - 3 G A) y + (\frac{1}{4} D^2 - F B) x^2 \\ &+ (-G B + \frac{1}{2} D E - F C) x - F - G C + \frac{1}{4} E^2 \end{split}$$

Remark.

The variational equation, $D^2\xi^1/dt^2 = P_1^1 \xi^1$, indicates that the deviation will be *periodic*, as for the *simple harmonic oscillator*, if $P_1^1 < 0$, and *aperiodic* otherwise, as for the *wave-guide equation* (similar to an harmonic oscillator with reverse sign), where the trajectories diverge.

• The final expression for P_1^1

$$\begin{split} P_1^1 &= -\frac{3}{4} \frac{k_4^2 y^4}{k_6^4} + (-\frac{k_4^2 x}{k_6^3} + 3 \frac{k4^2 k_1}{k_6^4}) y^3 \\ &+ (3 \frac{k_4^2 k_1 x}{k_6^3} - \frac{1}{4} \frac{6 k_4 k_6^3 + 6 k_4 k k_6^2 + 18 k_4^2 k_1^2}{k_6^4}) y^2 \\ &+ (-\frac{1}{4} \frac{(12 k_6 k4^2 k_1^2 + 12 k_4 k_6^3 k) x}{k_6^4} \\ &- \frac{1}{4} \frac{-16 k_4 k_1 k_6^3 - 12 k_4^2 k_1^3 - 12 k_4 k_1 k k_6^2}{k_6^4}) y - k_4 k x^2 \\ &- \frac{1}{4} \frac{(-4 k_6 k_4^2 k_1^3 - 12 k_4 k_6^3 k_1 k) x}{k_6^4} \\ &- \frac{1}{4} \frac{2 k_6^5 k + 3 k_4^2 k_1^4 + 10 k_6^3 k_4 k_1^2 - k^2 k_6^4 + 6 k_4 k_1^2 k k_6^2 - k_6^6}{k_6^4} \end{split}$$

Numerical results I.



Figure 2. This graph shows $P_1^1(x_0, y_0)$ at the steady-states for $k_6 \in [0.1, 10]$.



Figure 3. This graph shows $P_1^1(x_0, y_0)$ at the steady-states for k_6 varying within region A. The deviation vector is aperiodic for most of the range. Note that it becomes periodic for k_6 values close to the B range.



Figure 4. This graph shows $P_1^1(x_0, y_0)$ at the steady-states for k_6 varying within region B. The deviation vector is periodic for all of the range.



Figure 5. This graph shows $P_1^1(x_0, y_0)$ at the steady-states for k_6 varying within region C. The deviation vector is aperiodic for most of the range.

Numerical results II.



Figure 7. This graph shows P_1^1 against x and y for $k_4=180$ and $k_6=0.1924781307$, the unique k_6 -value within region A such that $P_1^1(x_0, y_0) = 0$ at the steady-state (x = 0.09143007729, y = 0).



Figure 8. This graph shows P_1^1 against x and y for $k_4=180$ and $k_6=0.2911043274$, for which $P_1^1(x_0, y_0)$ (Figure 3) is most negative (associated with the limit cycle in the model), corresponding to the point ($x_0 = 0.08177456483$, $y_0 = 0$, $P_1^1 = -0.1316018785$) in this graph.



Figure 9. This graph shows P_1^1 against x and y for $k_4=180$ and $k_6=1.460381230$, for which $P_1^1(x_0, y_0)$ (Figure 3) has its second minimum point (associated with the excitable switch in the model), corresponding to the point ($x_0 = 0.4157874576$, $y_0 = 0$, $P_1^1 = -0.0540182526$) in this graph.



Figure 10. This graph shows P_1^1 against x and y for $k_4=180$ and $k_6=1.900932822$, the unique k_6 -value within region C such that $P_1^1(x_0, y_0) = 0$ at the steady-state (x = 0.5214518040, y = 0).

• Linear stability analysis.

• *Mode A-Linear Stability*: steady state of high MPF activity that corresponds to the metaphase arrest of unfertilized eggs.

• Mode B-Linear Unstability: autonomous oscillations (limit cycle) that corresponds to rapid cyclin gin early embrios.

• *Mode C-Linear Stability*: excitable steady state with low MPF activity that corresponds to the interphase arrest of resting cells.

	Region A		Region B	Region C		
Linear	stable	stable	unstable	stable	stable	
stability			+++			k6
Jacobi	0.1 +++	0.192 	0.2 	1.5 	1.9 ++++	10
stability	aperiodic	periodic	periodic	periodic	aperiodic	

• Jacobi stability analysis.

• Mode A: Jacobi stability for almost all values of k_6 , but become instable near the boundary with mode B.

The egg stops developing as usual and dies away.

• *Mode B*: Jacobi unstability with two distinct local minima, one corresponding to the limit cycle of the system (the deepest one) and the other corresponding to the cell excitable switch. The instability region is larger than initially estimated by linear stability analysis.

• Mode C: Jacobi stability for almost all values of k_6 , but become instable near the boundary with mode B.

The cell is perturbed out the basin of attraction and it dies away or it divides endlessly.

Remarks.

• Estimation of the eigenvalues of P_i^j at a certain point gives the behavior of the trajectories in a neighbourhood of that point.

• Jacobi stability analysis gives a more accurate picture of cell growth than the linear stability analysis alone.

Future work.

The process of cell development investigation and drug design.



Open problems.

1. Biological meaning of the geometry associated to a SODE (in the sense of Bryant or Foulon)?

2. What kind of mathematics do we need to get a fightfull picture of the cell that can be implemented as a software?

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