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OPEN A van der Waals-like Transition **Between Normal and Cancerous Phases in Cell Populations Dynamics of Colorectal Cancer**

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Based on a deterministic continuous model of cell populations dynamics in the colonic crypt and in colorectal cancer, we propose four combinations of feedback mechanisms in the differentiations from stem cells (SCs) to transit cells (TCs) and then to differentiated cells (DCs), the four combinations include the double linear (LL), the linear and saturating (LS), the saturating and linear (SL), and the double saturating (SS) feedbacks, respectively. The relative fluctuations of the population of SCs, TCs, and DCs around equilibrium states with four feedback mechanisms are studied by using the Langevin method. With the increasing of net growth rate of TCs, it is found that the Fano factors of TCs and DCs go to a peak in a transient phase, and then increase again to infinity in the cases of LS and SS feedbacks. The "up-down-up" characteristic on the Fano factor (like the van der Waals loop) demonstrates that there exists a transient phase between the normal and cancerous phases, our novel findings suggest that the mathematical model with LS or SS feedback might be better to elucidate the dynamics of a normal and abnormal (cancerous) phases.

Mathematical modeling used as an indispensable complementary tool has played an important role in understanding the biological processes and phenomena which cannot be explained directly through the experimental studies due to the limitations of the biological experimental techniques¹. One such well-known example is to illustrate the cell population homeostasis in the intestinal crypt and the developmental processes of the colorectal cancer (CRC)^{2,3}. It is widely believed that earlier the CRC is diagnosed, the longer the patient survives⁴. Given the opinion that the origins of CRC lie in crypts of the intestine, illustrating the mechanisms involved in the dynamics of cells in colonic crypts is essential to understand the development of CRC. Single colorectal crypt can be mainly subdivided into three compartments which contain the stem cells (SCs), transit cells (TCs) and differentiated cells (DCs) separately from the bottom to the top of the crypt. In a normal crypt, SCs with unlimited proliferative capacity can regulate the tissue renewal rate to maintain homeostasis, update the crypt totally every 4-5 days through generating the TCs which only undergo 4-6 times of division^{5,6}, the TCs differentiate to DCs which exfoliate from the top of the crypt and form the luminal surface of gut. Johnston et al. proposed a deterministic compartmental continuous model^{2,3}, which can elucidate the mechanisms involved in normal intestinal crypt homeostasis dynamics and the gradual stages of CRC, and the combinations of feedback mechanisms in their model were assumed in the differentiations from stem cells (SCs) to transit cells (TCs) and then to differentiated cells (DCs).

On the one hand, although the linear feedback mechanism can be used to illustrate the homeostasis in a normal crypt, it cannot explain the uncontrolled exponential growth of cell populations in an abnormal (cancerous) colonic crypt unless coincidentally mutations occur. Some researchers pointed out that the other factors, such as microenvironmental influences and abnormal regulation of signal pathway, can also induce the uncontrolled growth and the tumor initiation⁷⁻¹². Thus, the saturating feedback mechanism, which was used to illustrate the uncontrolled cell population growth when the net growth rates of SCs or TCs are over a critical threshold value, may be better than the linear feedback in the continuous model^{2,3}. Moreover, the feedback mechanism in the

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Figure 1. Schematic diagram of the three compartments (SCs, TCs, and DCs) in a single crypt. The SCs differentiate into TCs which in turn differentiate into DCs, where the both differentiation rates α_2 and β_2 are controlled by feedbacks. SCs and TCs can also renew with rates α_3 and β_3 , respectively, and die with rates α_1 and β_1 , respectively. DCs can only die or be removed out of the compartment with rate γ .

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differentiation from SCs to TCs might not be same as that in the differentiation from TCs to DCs, the feedback in the differentiation from SCs to TCs maybe a linear or saturating, and that in the differentiation from TCs to DCs maybe a saturating or linear. Now a question is raised: which feedback mechanism is the better to elucidate the physiological properties in a normal and abnormal (cancerous) colonic crypt?

On the other hand, various experimental evidences suggested that the number of SCs in a normal colonic crypt is often very limited¹³⁻¹⁸. For instance, Potten and Loeffler concluded that different experimental techniques or assumptions got apparently different SCs numbers, the range of them is approximately 30–40, less than 16 or 4–16¹³. More following experiments demonstrated that the number is much smaller and is about 4–6 in each colonic crypt by using different experimental techniques¹⁴⁻¹⁶. More recently, Moore and Lyle found that the SCs number was only between 1.8 and 3.5 in human duodenum by using a two-dimensional model based on experimental results¹⁷. Kozar *et al.* revealed that the numbers of actual functional stem cells in intestinal crypts (5–7 in each crypt) and adenomas (9 per gland) are small through continuous clonal labeling¹⁸. The large relative fluctuations of cell populations due to the small number of SCs cannot be neglected in the growth of the three types of cells in intestinal crypt. Recently, based on the deterministic continuous model with two linear feedbacks, Pei *et al.*¹⁹ studied relative fluctuations of cell populations with the LL feedbacks. Thus, another question is raised: what are the diversities of relative fluctuations of cell populations around equilibrium states under the different combinations of feedback mechanisms?

In this paper, based on the deterministic model, we firstly use a Hill function form to represent a general feedback mechanism (including the linear and the saturating), then there are four combinations of feedback mechanisms in the cell populations dynamics, i.e. the LL, LS, SL, SS feedbacks. Secondly, in order to study the relative fluctuations in the populations of SCs, TCs, and DCs, we derive the formulae of Fano factor, covariance and susceptibility of cell populations around equilibrium states by using the Langevin method²⁰. Our theoretical results show that the stationary populations of TCs and DCs exhibit an approximately threshold behavior as a function of the net growth rate of TCs for the four feedback mechanisms. The reproductions of TCs and DCs can be classified into three phases: normal, transient, and cancerous. It is interesting that there is a so-called van der Waals loop in the Fano factors of TCs and DCs populations under the LS and SS feedback mechanisms, which can be used to as evidence of a phase transition²¹⁻²⁴. Our results demonstrate that LS and SS feedback mechanisms are more appropriate to illustrate the dynamics of the CRC, and the transient (coexistent) phase may correspond to the transient state (i.e., adenomas) before the CRC initiation²⁵⁻²⁸. Thirdly, we compare our theoretical results with these obtained by Gillespie algorithm²⁹ (an accurately simulating). We end with the conclusions and discussions.

General model with four feedback mechanisms

A general deterministic continuous model, which is based on the three compartments assumption in a single crypt³⁰ and can illustrate the homeostasis of the three types of cells (SCs, TCs, and DCs) in a colonic crypt and tumor initiation of a CRC^{2,3}, is used to describe the cell populations dynamics in a colonic crypt. The schematic representation of these three compartments in a crypt is given by Fig. 1².

In order to make the system structurally stable, the differentiations of both SCs and TCs are regulated by two alternative forms of feedback which maintain homeostasis in the crypt, respectively. Here we use a Hill function form to represent two possible feedback mechanisms (linear or saturating):

$$\frac{k_i N_i}{1 + m_i N_i} = \begin{cases} k_i N_i, & \text{if } m_i = 0, \text{ the linear feedback (L),} \\ \frac{k_i N_i}{1 + m_i N_i}, & \text{if } m_i \neq 0, \text{ the saturating feedback (S),} \end{cases}$$

where i = 0 and 1, N_0 and N_1 are the population numbers of SCs and TCs, respectively, k_0 and k_1 represent the rates of population response to change, while m_0 and m_1 are feedback parameters. In the case of saturating feedback, there is a saturating value $\alpha_2 + k_0/m_0$ (or $\beta_2 + k_1/m_1$) of the differentiation rate of SCs (or TCs). Therefore, there are four possible combinations (listed in Table 1), that is, LL feedback ($m_0 = 0$ and $m_1 = 0$), LS feedback ($m_0 = 0$ and $m_1 = 0.01$), SL feedback ($m_0 = 0.1$ and $m_1 = 0$), and SS feedback ($m_0 = 0.1$ and $m_1 = 0.01$).

In the deterministic description, the population dynamics in the crypt can be given by following ordinary differentiated equations (ODEs)^{2,3}

Feedback Parameters	$m_1 = 0$	$m_1 = 0.01$
$m_0 = 0$	LL feedback	LS feedback
$m_0 = 0.1$	SL feedback	SS feedback

 Table 1. Combinations of feedback mechanisms.

$$\frac{dN_0}{dt} = (\alpha_3 - \alpha_1)N_0 - \left(\alpha_2 + \frac{k_0 N_0}{1 + m_0 N_0}\right)N_0 \tag{1}$$

$$\frac{dN_1}{dt} = (\beta_3 - \beta_1)N_1 - \left(\beta_2 + \frac{k_1N_1}{1 + m_1N_1}\right)N_1 + \left(\alpha_2 + \frac{k_0N_0}{1 + m_0N_0}\right)N_0$$
(2)

$$\frac{dN_2}{dt} = \left(\beta_2 + \frac{k_1 N_1}{1 + m_1 N_1}\right) N_1 - \gamma N_2 \tag{3}$$

which give the steady state

$$N_0^S = \frac{\alpha}{k_0 - m_0 \alpha} \left(0 < \alpha < \frac{k_0}{m_0} \right) \tag{4}$$

$$N_1^S = \frac{\beta + \widetilde{\beta}}{2(k_1 - m_1\beta)} \left(-1 < \beta < \frac{k_1}{m_1} \right)$$
(5)

$$N_{2}^{S} = \frac{N_{1}^{S}}{\gamma} \left(\beta_{2} + \frac{k_{1} N_{1}^{S}}{1 + m_{1} N_{1}^{S}} \right)$$
(6)

with

$$\widetilde{\beta} = m_1 \psi + \sqrt{\left(\beta - m_1 \psi\right)^2 + 4\psi k_1} \tag{7}$$

$$\psi = \left(\alpha_2 + \frac{k_0 N_0^S}{1 + m_0 N_0^S}\right) N_0^S = (\alpha_3 - \alpha_1) N_0^S$$
(8)

and N_2 is the number of DCs, and $\alpha = \alpha_3 - \alpha_2 - \alpha_1$ and $\beta = \beta_3 - \beta_2 - \beta_1$, which denote the net (per capita) growth rates of SCs and TCs, respectively. Combining that $\alpha_1 + \alpha_2 + \alpha_3 = 1$ and $\beta_1 + \beta_2 + \beta_3 = 1$, the net growth rates (α and β) can be determined by the linear stability analysis: $0 < \alpha < k_0/m_0$ and $-1 < \beta < k_1/m_1$. So the k_0/m_0 and k_1/m_1 are the critical threshold values which discriminate the non-cancerous and cancerous states. When one of the net growth rates goes beyond its own threshold value, the cell population grows uncontrolled. In this paper, the values of the parameters $\beta_2 = 0.3$, $\gamma = 0.323$, $k_0 = 0.1$ and $k_1 = 0.01$ are determined^{2,3}. All the four situations can satisfy the healthy physiological case that the number of SCs is small and the number of each type of cells (SCs, TCs, and DCs) can get a stationary value under different initial conditions of cell population growth and make the system maintains homeostasis.

In the stochastic description, the dynamics of the three compartments are given by the following stochastic equations:

$$\frac{dN_0}{dt} = (\alpha_3 - \alpha_1)N_0 - \left(\alpha_2 + \frac{k_0 N_0}{1 + m_0 N_0}\right)N_0 + \xi_0(t)$$
(9)

$$\frac{dN_1}{dt} = (\beta_3 - \beta_1)N_1 - \left(\beta_2 + \frac{k_1N_1}{1 + m_1N_1}\right)N_1 + \left(\alpha_2 + \frac{k_0N_0}{1 + m_0N_0}\right)N_0 + \xi_1(t)$$
(10)

$$\frac{dN_2}{dt} = \left(\beta_2 + \frac{k_1 N_1}{1 + m_1 N_1}\right) N_1 - \gamma N_2 + \xi_2(t)$$
(11)

where $\xi_i(t)$ (*i*=0, 1, 2) are random variables. Here, the statistical properties of the random variables $\xi_i(t)$ around the steady state are derived by using two assumptions²⁰ (see Part 1 of Supplementary Information). All the mean



Figure 2. Time course of SCs, TCs, and DCs with different feedbacks and initial conditions. Initial conditions: upper row (**a**-**d**) $N_0(0) = 1.0$, $N_1(0) = 1.0$, and $N_2(0) = 1.0$; middle row (**e**-**h**) $N_0(0) = 20.0$, $N_1(0) = 30.0$, and $N_2(0) = 1.0$; bottom row (**i**-**l**) $N_0(0) = 35.0$, $N_1(0) = 50.0$, and $N_2(0) = 80.0$. The left parameter values are $\alpha = 0.4$, $\beta = 0.2$, and $\alpha_2 = 0.3$. All the parameters are measured in hr⁻¹.

values of $\xi_i(t)$ are zero, but the autocorrelation functions of $\xi_i(t)$ relate with the system's intrinsic characters (the parameters and the steady values of the system) and can be described by following equations

$$\langle \xi_0(t)\xi_0(t') \rangle = 2(\alpha + \alpha_2)N_0^{S}\delta(t - t')$$
(12)

$$\langle \xi_1(t)\xi_1(t')\rangle = 2\left(\beta_2 + \frac{k_1 N_1^S}{1+m_1 N_1^S}\right) N_1^S \delta(t-t')$$
(13)

$$\left\langle \xi_2(t)\xi_2(t')\right\rangle = 2\gamma N_2^S \delta(t-t') \tag{14}$$

and the cross-correlations between $\xi_i(t)$ take the forms

$$\left< \xi_0(t)\xi_1(t') \right> = -\left(\alpha_2 + \frac{k_0 N_0^S}{1 + m_0 N_0^S} \right) N_0^S \delta(t - t')$$
(15)

$$\left< \xi_1(t)\xi_2(t') \right> = -\left(\beta_2 + \frac{k_1 N_1^S}{1 + m_1 N_1^S}\right) N_1^S \delta(t - t')$$
(16)

$$\left\langle \xi_0(t)\xi_2(t')\right\rangle = 0\tag{17}$$

From the above equations (12)–(17), it is found that the autocorrelation of $\xi_0(t)$ is determined by the SCs's net growth rate α , differentiation rate α_2 , and its steady state value N_0^S , while the autocorrelation of $\xi_1(t)$, the cross-correlation between $\xi_0(t)$ and $\xi_1(t)$, and the cross-correlation between $\xi_1(t)$ and $\xi_2(t)$ have connection with the feedback mechanism on SCs or TCs. The autocorrelation of $\xi_2(t)$ is dependent of DCs's steady state value and the removal rate. There is no cross-correlation between $\xi_0(t)$ and $\xi_2(t)$.

Results

Relative fluctuation formulae of cell populations around equilibrium states. Equations (9)–(11) can be linearized around the steady state (N_0^S, N_1^S, N_2^S) with $x_0 = N_0 - N_0^S$, $x_1 = N_1 - N_1^S$ and $x_2 = N_2 - N_2^S$. By virtue of equations (4)–(8), equations (9)–(11) can be rewritten as

$$\frac{dx_0}{dt} = \left(-\alpha + \frac{m_0 \alpha^2}{k_0}\right) x_0 + \xi_0(t) \tag{18}$$

$$\frac{dx_1}{dt} = (\beta - \phi)x_1 + \left(\alpha_2 + 2\alpha - \frac{m_0\alpha^2}{k_0}\right)x_0 + \xi_1(t)$$
(19)

$$\frac{dx_2}{dt} = (\beta_2 + \phi)x_1 - \gamma x_2 + \xi_2(t)$$
(20)

where

$$\phi = \frac{k_1(\beta + \tilde{\beta})(4k_1 + m_1\tilde{\beta} - 3m_1\beta)}{(2k_1 - m_1\beta + m_1\tilde{\beta})^2}$$
(21)

The corresponding Fokker-Planck equation of the equations (18)-(20) is^{31,32}

$$\frac{\partial}{\partial t}P(x_0, x_1, x_2, t) = \left\{ \frac{\partial}{\partial x_0} \left[\left(\alpha - \frac{m_0 \alpha^2}{k_0} \right) x_0 \right] - \frac{\partial}{\partial x_1} \left[(\beta - \phi) x_1 + \left(\alpha_2 + 2\alpha - \frac{m_0 \alpha^2}{k_0} \right) x_0 \right] - \frac{\partial}{\partial x_2} \left[-\gamma x_2 + (\beta_2 + \phi) x_1 \right] + \frac{\alpha (\alpha + \alpha_2)}{k_0 - m_0 \alpha} \left(\frac{\partial^2}{\partial x_0^2} - \frac{\partial}{\partial x_0} \frac{\partial}{\partial x_1} \right) + \gamma N_2^S \left[\frac{\partial^2}{\partial x_1^2} + \frac{\partial^2}{\partial x_2^2} - \frac{\partial}{\partial x_1} \frac{\partial}{\partial x_2} \right] P(x_0, x_1, x_2, t)$$
(22)

According to the definition of the Fano factor (the ratio of the variance to the steady state value), it is used to characterize the relative size of fluctuation and can be calculated through the Fokker-Planck equation. So, we can get the formulae of Fano factors of SCs, TCs, and DCs, respectively

$$F_0 = \frac{\sigma_{00}}{N_0^S} = \frac{k_0(\alpha + \alpha_2)}{\alpha(k_0 - m_0\alpha)}$$
(23)

$$F_{1} = \frac{\sigma_{11}}{N_{1}^{S}} = \frac{2(k_{1} - m_{1}\beta)}{\beta + \tilde{\beta}}\sigma_{11}$$
(24)

$$F_2 = \frac{\sigma_{22}}{N_2^S} = 1 + \frac{\beta_2 + \phi}{\gamma N_2^S} \sigma_{12}$$
(25)

where the variances (σ_{00} , σ_{11} , σ_{22}) of three compartments and the covariance (σ_{01} , σ_{02} , σ_{12}), which can describe the correlations between fluctuations of the cell population, are given in Part 2 of Supplementary Information.

In addition, the susceptibility (which measures the sensitivity to variation of a parameter $z^{33,34}$) can be calculated by using its definition form

$$\chi_{i}(z) = \left| \frac{\partial \ln \langle M_{i} \rangle_{st}}{\partial \ln z} \right| = \left| \frac{z}{\langle M_{i} \rangle_{st}} \cdot \frac{\partial \langle M_{i} \rangle_{st}}{\partial z} \right|$$
(26)

where $\langle M_i \rangle_{st} = N_i^s (i=0, 1, 2)$, which means the steady value of the system. Here, we consider the effects of the net growth rates, so $z = \alpha$, β . According to this definition, the susceptibilities to the two net growth rates are derived in Part 3 of Supplementary Information.

Comparisons the diversities with four feedbacks due to the net growth rates. Although all the four feedbacks can maintain the homeostasis of the population in the colonic crypt (Fig. 2), the mutational processes, steady state values, and Fano factors are quite different under different feedbacks due to the net growth rates (α and β). Next, the differences on the mutational processes, steady state values, and the diversities of the fluctuation effects on Fano factors due to the net growth rates (α and β) of SCs and TCs in a colonic crypt with four feedback mechanisms are discussed in the range of physiologically reasonable parameter values^{2,3}.



Figure 3. Time course of SCs, TCs, and DCs in a sequence of mutations with different feedbacks. The initial parameters are taken to be $N_0(0) = 1.0$, $N_1(0) = 1.0$, $N_2(0) = 1.0$, and $\alpha = 0.286$, $\beta = 0.432$, $\alpha_2 = 0.3$, $\beta_2 = 0.3$. The mutations cause, successively, $\beta = 0.512$, $\alpha = 0.5$, $\beta = 0.697^3$. The last mutation is $\alpha = 1.1$ in the upper row, and $\beta = 1.1$ in the bottom row. All the parameters are measured in hr⁻¹.

A. Mutational processes and tumor initiation. According to equations (1)–(3), the mutational processes with four cases of feedbacks are illustrated in Fig. 3. Here, we consider the mutations which can increase the net growth rate α or β , and the mutations can give a selective advantage to the cell^{2,3}. The same initial parameter set $N_0(0) = 1.0$, $N_1(0) = 1.0$, $N_2(0) = 1.0$, and $\alpha = 0.286$, $\beta = 0.432$, $\alpha_2 = 0.3$, $\beta_2 = 0.3$ was chosen with the four cases of feedbacks. As time goes on, the processes go with the same sequence of mutations in all the four feedbacks except the last mutation in which $\alpha = 1.1$, $\beta = 0.697$ (upper row of Fig. 3) or $\alpha = 0.5$, $\beta = 1.1$ (bottom row of Fig. 3). Above two situations denote that the last mutation can cause an extreme large selective advantage, i.e. the net growth rate α is beyond the threshold value k_0/m_0 or β is beyond the threshold value k_1/m_1 . It is obvious that the total population of the crypt increases gradually with four feedbacks, which can illustrate a gradual stages with the mutational processes. However, there always have limited steady state values and the dynamics of the colonic crypt is controllable with LL feedback (first column of Fig. 3). When a saturating feedback is included, such as LS (second column of Fig. 3), SL (third column of Fig. 3), and SS (fourth column of Fig. 3) feedbacks, there can result in an uncontrolled growth of the cell population with the net growth rate beyond the threshold value of the saturating feedback (Fig. 3(c,d,f,h)). Thus, the model with one or two saturating feedbacks is appropriate to illustrate the processes of tumor initiation.

B. Steady states. From equations (4)–(6) and above results, the steady state values of SCs, TCs, and DCs are dependent on the systemic parameters. Here, we study the effects of one single net growth rate (α or β) on the steady state values, and the cell population as a function of α or β with four feedbacks is shown by Fig. 4. For a given value of β , with the increasing of α , the stationary population of SCs is small and linearly increased with LL and LS feedbacks (Fig. 4(a,b)), but it is linearly increased first, then to a uncontrolled infinite value later with SL and SS feedbacks (Fig. 4(a,-d)). At the same time, the cell population of TCs or DCs increases nonlinearly with all the four feedbacks (Fig. 4(a,-d)), and is controllable with LL and LS feedbacks (Fig. 4(a,b)), but uncontrollable with SL and SS feedbacks (Fig. 4(c,d)). It can conclude that, when the net growth rate α of SCs is close to the saturating threshold ($k_0/m_0 = 1$) with SL and SS feedback, the population of the crypt is rapidly increased and the tumor is more likely to be initiated. According to equation (4), when the feedback on the SCs is linear ($m_0=0$), i.e. LL and LS feedbacks, the stationary population of SCs is α/k_0 which is small because of $k_0 = 0.1$ and is proportional to $\alpha(0 \le \alpha \le 1)$; when the feedback on the SCs is saturating ($m_0=0.1$), i.e. SL and SS feedbacks, the stationary population of SCs is $\frac{\alpha}{k_0-m_0\alpha}$ which becomes uncontrollable with α close to the saturating threshold ($k_0/m_0=1$).

For a given value of α , with the increasing of β , the stationary population of SCs is a constant (see equation (4)) and small (Fig. 4(e–h)), the stationary populations of TCs and DCs exhibit an approximately threshold behavior, and the threshold is around $\beta = 0$ with all the four feedbacks (Fig. 4(e–h)). However, when the net growth rate β is larger than 1 (the threshold value), the whole population of the crypt is still controlled and limited with LL and SL feedbacks (Fig. 4(e,g)). Above results show that only when feedback on TCs is saturating can the uncontrolled increasing of cell populations or the cancerous phase be described, e.g. the LS or SS feedback, the population of the crypt increases uncontrolled and the cancerous phase are shown by Fig. 4(f,h). In the cases of LS or SS feedback, the development of the colonic crypt can be classified into three phases: normal ($\beta \ll 0$), transient ($\beta \sim 0$), and cancerous ($\beta \gg 0$)¹⁹, which can also correspond to three phases of tumor evolution: breakthrough, expansion and invasion³⁵.



Figure 4. Stationary population as a function of net growth rates α (the upper row) and β (the bottom row). (a-d) $\alpha_2 = 0.2$ and $\beta = 0.432$. (e-h) $\alpha = 0.2$ and $\alpha_2 = 0.3$. All the parameters are measured in hr⁻¹.



Figure 5. Diversities of fluctuation effects on Fano factors due to net growth rates with four different feedbacks. (a-c) $\alpha_2 = 0.2$, $\beta = 0.432$. (d,e) $\alpha = 0.2$, $\alpha_2 = 0.3$. All the parameters are measured in hr⁻¹.

C. A van der Waals-like Loop on Fano Factors. Above theoretical results indicate that the number of SCs in a crypt is always small, which is consistent with the experimental data in normal colonic crypt, except the SL and SS feedbacks with larger net growth rate α (Fig. 4(c,d)). Due to the small number of SCs, the relative high fluctuations of populations which can be characterized by the Fano factor are studied around the steady state by Langevin method. The Fano factors of SCs, TCs, and DCs with four feedbacks as a function of α are illustrated by Fig. 5(a–c) sequentially. With a fixed net growth rate β (β =0.432), it is obvious that the three Fano factors (F_0 , F_1 , F_2) with SL and SS feedbacks are quite different with LL and LS feedbacks. With the increasing of α , the Fano factor of SCs goes down rapidly for small values of α with four feedbacks, then reaches 1 for large values of α with LL and LS feedbacks, but goes up quickly again to infinity with SL and SS feedbacks when α is close to 1 (the threshold value k_0/m_0) (Fig. 5(a)); the Fano factors of TCs and DCs decrease linearly and slowly with LL and LS feedbacks but increase nonlinearly and quickly to infinity with SL and SS feedbacks (Fig. 5(b,c)). These effects of the net growth rate of SCs on the relative fluctuation around the steady state can be understood directly. When the net growth rate α is small, the numbers of SCs is very small (Fig. 4(a–d)) and the noise is relatively large on the SCs. When α increases, the cell population of SCs will increase and the relative intrinsic fluctuation goes down (Fig. 5(a)). Due to considering the effects of the net growth rate α , which puts effects directly on the SCs, the type



Figure 6. Diversities of fluctuation effects on Fano factors of TCs and DCs due to β with different α . $\alpha_2 = 0.3$. All the parameters are measured in hr⁻¹.

of feedback on SCs determines the variation tendency of Fano factors (F_0 , F_1 , F_2) (Fig. 5(a–c)). When α is close to the threshold value 1 with SL and SS feedbacks, the numbers of SCs, TCs, and DCs increase uncontrolled to infinity (Fig. 4(c,d)), then the Fano factors also increase to infinity (Fig. 5(a–c)).

Fano factors of TCs and DCs with four feedbacks as a function of β with determined α are shown by Fig. 5(d,e), while the Fano factor of SCs which has no relation with β (see equation (23)) is a constant. With the increasing of β , the Fano factors of both TCs and DCs with four feedbacks increase first, reach a maximum, and then decrease to a constant with LL and SL feedbacks but again increase uncontrolled to infinity with LS and SS feedbacks (Fig. 5(d,e)). When $\beta < 0$, the variation tendency of Fano factors is controlled by the type of feedback on SCs and on TCs when $\beta > 0$. The peak of Fano factors of TCs or DCs occurs in the transient phase $(\beta \sim 0)$ (Fig. 5(d,e)), which means that there is a significant change of relative fluctuations caused by an abruptly enhanced cell population of TCs or DCs. When β is close to the threshold value $(k_1/m_1 = 1)$ with the LS or SS feedback, the numbers of TCs and DCs increase uncontrolled to infinity (Fig. 4(f,h)) and the Fano factors also go up to infinity (Fig. 5(d,e)). This "up-down-up" characteristic on the Fano factors is similar to the van der Waals loop (Fig. 5(d,e)) which corresponds to the 'PV-isotherm' in the van der Waals theory of phase transition. This loop gives a continuous transition from liquid to vapor or from vapor to liquid, and it can demonstrate that there is a metastable state (superheated liquid or supercooled vapor) during this process. Thus, this van der Waals-like loop demonstrate that there is a transient phase between the normal and cancerous phases²¹⁻²⁴ in this paper. It was demonstrated that there is a transient benign state (e.g., adenomas)^{25–28} or an abrupt increase of TCs^{36-38} before CRC initiation or expansion phase of other tumors initiation³⁵.

Figure 6 shows the diversities of fluctuation effects on Fano factors of TCs and DCs due to β with different α under LS and SS feedbacks respectively. For the LS feedback, with the increasing of α , the intensity of relative fluctuations goes down in the normal ($\beta \ll 0$) and transient ($\beta \sim 0$) phases and is almost same in the cancerous phase ($\beta \gg 0$ or $\beta > 0.5$) (Fig. 6(a,c)). This is because that the Fano factors of TCs and DCs (F_1 and F_2) is decreasing with the increasing of α under the LS feedback (Fig. 5(b,c)), but when $\beta > 0.5$, i.e. in the cancerous phase, the β is the dominate factor, and the intensity of relative fluctuations is almost same. For the SS feedback, with the increasing of α , the behavior of Fano factors is different. When α is small ($\alpha < 0.5$), the intensity of relative fluctuations goes down with the increasing of α in the normal and transient phases and vice vesa (Fig. 6(b,d)). In the cancerous phase, the Fano factors F_1 and F_2 go to the infinity with different α . From Fig. 5(b,c), it is obvious that F_1 and F_2 increase nonlinearly and quickly with the increasing of α except for small α under the SS feedback. Except these differences between LS and SS feedbacks, there are some common points with these two feedbacks. There always exists a van der Waals loop with small α with the increasing of β (Fig. 6), which means there is a transient phase during the expansion of the tumor. However, when α is large, the tumor will go directly from normal phase to cancerous phase, just like the sublimation from liquid to vapor at high temperature.

In addition, the effects of the net growth rate α or β on the covariance and susceptibilities are compared and provided in Part 4 and 5 of Supplementary Information.



Figure 7. Fano factors as a function of α (upper row) and β (bottom row) obtained by Langevin method²⁰ (lines) and Gillespie algorithm²⁹ (symbols) with $\Omega = 200$. (a-d) $\alpha_2 = 0.4$, $\beta = 0.432$. (e-h) $\alpha = 0.2$, $\alpha_2 = 0.3$. All the parameters are measured in hr⁻¹.





Comparisons the results with different methods. The aim of this paper is mainly to compare the effects of the relative fluctuations of the cell population in a colonic crypt around the steady state with four feedbacks. There are mainly three methods to study the fluctuations of biochemical reaction systems. The first one is to solve the probability distribution for all the molecular components through the chemical master equation³⁹, which can be solved explicitly only in rare cases. In most of cases, the probability distribution is given out through computer simulation methods, such as the Gillespie algorithm²⁹. The second technique is to use the linear noise approximation (LNA) to approximately solve the master equation which is simplified to a linear Fokker-Planck equation by using van Kampen's expansion³⁹. The third approach is the Langevin method²⁰ which is mainly used in this paper. This method is more intuitive than the Langevin theory (the second method) and these two methods are equivalent at steady state²⁰. Here, we consider the additive noises as the relative fluctuation sources of the cell population in a colonic crypt (see equations (9)–(11)).

Comparing the Langevin method of Swain²⁰ with the usual Langevin theory or LNA³⁹, the statistical properties of random variables, such as mean square value, self-correlation function and mutual correlation function, are dependent of its own systemic parameters and the steady state values, and can be derived through two assumptions (i.e. the one-step biochemical processes and the small fluctuation at the steady state) in the Langevin method²⁰, whereas the usual Langevin theory²⁰ only artificially gives the statistical properties of intrinsic noises.

Comparing the results of the Langevin method²⁰ with those of Gillespie algorithm (an accurate simulating)²⁹, the theoretical and simulation results of the Fano factors as a function of α and β are compared with the same and different volumes of parameter Ω in Figs 7 and 8 with four feedbacks, respectively. It is obvious that the two kinds of results are consistent with each other in most region of parameters α and β except about $0 \le \alpha < 0.1$ and $-1 \le \beta < -0.4$. Because the number of the populations of SCs is very small with low values of net growth rates of α and β with four feedbacks (see Fig. 4), the relative intrinsic fluctuations will be very high, which is against the second assumption of the Langevin method (the small fluctuation at the steady state), and then the divergence between the results of the two approaches is remarkable. However, the results of the Fano factors around

the transient phase (the peaks) are consistent with each other by using of the two methods (Figs 7 and 8). So, our theoretical results obtained by the Langevin method are valid with large enough number of populations, i.e. $0.1 < \alpha < 1$ and $-0.4 < \beta < 1$.

Conclusions and Discussions

In this paper, we start from a general model of the colonic crypt cell population dynamics which contains four combinations of feedback mechanisms, i.e. LL, LS, SL and SS. Based on this general deterministic continuous model, the relative fluctuations between the populations of SCs, TCs, and DCs are considered, and the statistical properties of fluctuations are derived from the system's own systemic parameters and steady state values by using the two assumptions of Langevin method²⁰.

The formulae of Fano factors, covariances, and susceptibilities are derived with the general model. Although the dynamics of cell populations of SCs, TCs, and DCs can reach a steady state with the four feedbacks, which is reasonable in the range of physiologically parameter values, the steady state values, and the effects of relative fluctuations on Fano factors, covariances, and susceptibilities obtained by using of Langevin method²⁰ are different as a function of the net growth rate α (or β) of SCs (or TCs). Our theoretical results show that the stationary populations of TCs and DCs exhibit an approximately threshold behavior as a function of the net growth rate of TCs for the four feedback mechanisms. The reproductions of TCs and DCs can be classified into three phases: normal, transient, and cancerous. The LL feedback which cannot produce the uncontrolled growth is not suitable for the cancerous phase with increasing of the net growth rate of SCs or TCs. It is interesting that there is a so-called van der Waals loop in the Fano factors of TCs and DCs populations under the LS and SS feedback mechanisms, which can be considered as an evidence of phase transition, and this transient phase may be corresponded to the adenomas state before CRC tumor initiation²⁵⁻²⁸ or expansion phase in other tumors initiation³⁵.

This three-compartment model discussed in this paper is a general model in normal tissue differentiation or the development of other tumors initiation. Thus, it is important to know which one is the best of these four feedback mechanisms for modeling the dynamics of a normal and abnormal (CRC) colonic crypt or other tumors initiation? Due to the difficulty of identifying cancer stem cells (CSCs), some researchers believed that the quantity of CSCs is small in tumors, just like the number of SCs in the normal tissues. A recent research¹⁸ reports that CSCs serve only to replace stem cell loss, with rare clonal victors driving gland repopulation and tumor growth, and the number of CSCs per crypt is five to seven in CRC. Our results show that the LL and LS feedbacks are better than the other two, which always can control the number of SCs in a small value. The LL feedback cannot illustrate the uncontrolled cell populations growth unless a coincidentally irreversible genetic mutation happens to remove the one of the two linear feedbacks. However, a recent report revealed that chronic myeloid leukemia can be cured by decreasing the proliferating rate of TCs without killing CSCs⁴⁰, which means the cancerous phase can be reversed to the normal phase with controlling the net growth rate of TCs. Therefore, the LS feedback is better than LL feedback. On the other hand, some experiments demonstrate that the CSCs number can be a substantial proportion of the tumor⁴¹⁻⁴³, such as in melanoma⁴¹, some human acute myeloid leukemia (AML)⁴² or in more aggressive type or late stage of cancer, e.g. breast cancer⁴³. Then these may use distinct mechanisms from other solid malignancies⁴⁴, and the SL and SS feedbacks may be more appropriate to illustrate this situation, because the SCs number can be large and any proportion when the net growth rate α increases. However, above theoretical results and the other results^{2,3,18,36–38} showed that the net growth rate of TCs plays an important role in the tumor initiation and metastasis. Thus, we suggest that the feedback on the TCs is probable a saturating feedback in a real colonic crypt and in colorectal cancer or in other tumors. In a conclusion, the saturating feedback on the TCs is necessary and the feedback on the SCs may be linear or saturating depending on the stages or types of the tumor.

References

- 1. Gatenby, R. A. & Maini, P. K. Mathematical oncology: cancer summed up. Nature 421, 321 (2003).
- Johnston, M. D., Edwards, C. M., Bodmer, W. F., Maini, P. K. & Chapman, S. J. Mathematical modeling of cell population dynamics in the colonic crypt and in colorectal cancer. *Proc. Natl. Acad. Sci. USA.* 104, 4008–4013 (2007).
- 3. Johnston, M. D., Edwards, C. M., Bodmer, W. F., Maini, P. K. & Chapman, S. J. Examples of mathematical modeling: tales from the crypt. Cell Cycle 6, 2106–2112 (2007).
- Merlos-Suarez, A. et al. The intestinal stem cell signature identifies colorectal cancer stem cells and predicts disease relapse. Cell Stem Cell 8, 511–524 (2011).
- 5. Clayton, E. et al. A single type of progenitor cell maintains normal epidermis. Nature 446, 185–189 (2007).
- 6. Watt, F. M. & Jensen, K. B. Epidermal stem cell diversity and quiescence. EMBO Mol. Med. 1, 260-267 (2009).
- 7. Medema, J. P. & Vermeulen, L. Microenvironmental regulation of stem cells in intestinal homeostasis and cancer. *Nature* 474, 318–326 (2011).
- 8. Crosnier, C., Stamataki, D. & Lewis, J. Organizing cell renewal in the intestine: stem cells, signals and combinatorial control. *Nature Rev. Genet.* 7, 349–359 (2006).
- 9. Hardwick, J. C., Kodach, L. L., Offerhaus, G. J. & van den Brink, G. R. Bone morphogenetic protein signalling in colorectal cancer. *Nature Rev. Cancer.* **8**, 806–812 (2008).
- 10. Reya, T. & Clevers, H. Wnt signalling in stem cells and cancer. Nature 434, 843-850 (2005).
- 11. Kim, K. A. et al. Mitogenic influence of human R-spondin1 on the intestinal epithelium. Science 309, 1256–1259 (2005).
- 12. Hanahan, D. & Weinberg, R. A. Hallmarks of cancer: the next generation. Cell 144, 646–674 (2011).
- Potten, C. S. & Loeffler, M. Stem cells: attributes, cycles, spirals, pitfalls and uncertainties. Lessons for and from the crypt. Development 110, 1001–1020 (1990).
- 14. Bjerknes, M. & Cheng, H. Clonal analysis of mouse intestinal epithelial progenitors. Gastroenterology 116, 7–14 (1999).
- 15. Marshman, E., Booth, C. & Potten, C. S. The intestinal epithelial stem cell. *BioEssays* 24, 91–98 (2002).
- 16. Wang, Z., Matsudaira, P. & Gong, Z. STORM: a general model to determine the number and adaptive changes of epithelial stem cells in teleost, murine and human intestinal tracts. *PLoS ONE* **5**, e14063 (2010).
- Moore, N. & Lyle, S. Quiescent, slow-cycling stem cell population in cancer: a review of the evidence and discussion of significance. J. Oncol. 2011, 396076 (2011).
- Kozar, S. et al. Continuous clonal labeling reveals small numbers of functional stem cells in intestinal crypts and adenomas. Cell Stem Cell 13, 626–633 (2013).

- 19. Pei, Q. M. et al. Fluctuations of cell population in a colonic crypt. Phys. Rev. E. 89, 032715 (2014).
- Swain, P. S. Efficient attenuation of stochasticity in gene expression through post-transcriptional control. J. Mol. Biol. 344, 965–976 (2004).
 - 21. Alder, B. J. & Wainwright, T. E. Phase transition in elastic disks. Phys. Rev. 127, 359–361 (1962).
 - Broughton, J. Q., Gilmer, G. H. & Weeks, J. D. Molecular-dynamics study of melting in two dimensions. Inverse-twelfth-power interaction. *Phys. Rev. B.* 25, 4651–4669 (1982).
 - 23. Evans, D. J. Melting in soft disk systems. Phys. Lett. A. 88, 48–50 (1982).
 - 24. Alonso, J. J. van der Waals loops and the melting transition in two dimensions. Phys. Rev. E. 59, 2659-2663 (1999).
 - 25. Radtke, F. & Clevers, H. Self-renewal and cancer of the gut: two sides of a coin. Science 307, 1904–1909 (2005).
 - 26. Fearon, E. R. & Vogelstein, B. A genetic model for colorectal tumorigenesis. Cell 61, 759-767 (1990).
 - 27. Huang, J. et al. APC mutations in colorectal tumors with mismatch repair deficiency. Proc. Natl. Acad. Sci. USA. 93, 9049-9054 (1996).
 - Summers, R. M. et al. Computed tomographic virtual colonoscopy computer-aided polyp detection in a screening population. Gastroenterology 129, 1832–1844 (2005).
- 29. Gillespie, D. T. Exact stochastic simulation of coupled chemical reactions. J. Phys. Chem. 81, 2340-2361 (1977).
- Tomlinson, I. P. & Bodmer, W. F. Failure of programmed cell death and differentiation as causes of tumors: some simple mathematical models. Proc. Natl. Acad. Sci. USA. 92, 11130–11134 (1995).
- 31. Jia, Y. & Li, J. R. Steady-state analysis of a bistable system with additive and multiplicative noises. Phys. Rev. E. 53, 5786–5792 (1996).
- 32. Jia, Y. & Li, J. R. Reentrance phenomena in a bistable kinetic model driven by correlated noise. Phys. Rev. Lett. 78, 994-997 (1997).
- Jia, Y., Liu, W. H., Li, A. B., Yang, L. J. & Zhan, X. Intrinsic noise in post-transcriptional gene regulation by small non-coding RNA. Biophys. Chem. 143, 60–69 (2009).
- Scott, M., Ingalls, B. & Kaern, M. Estimations of intrinsic and extrinsic noise in models of nonlinear genetic networks. *Chaos* 16, 026107 (2006).
- 35. Vogelstein, B. & Kinzler, K. W. The path to cancer-three strikes and you're out. N. Engl. J. Med. 373, 1895–1898 (2015).
- 36. Dieter, S. M. *et al.* Distinct types of tumor-initiating cells form human colon cancer tumors and metastases. *Cell Stem Cell* **9**, 357–365 (2011).
- 37. Botchkina, G. Colon cancer stem cells-from basic to clinical application. Cancer Lett. 338, 127-140 (2013).
- 38. Boman, B. M., Fields, J. Z., Cavanaugh, K. L., Guetter, A. & Runquist, O. A. How dysregulated colonic crypt dynamics cause stem cell overpopulation and initiate colon cancer. *Cancer Res.* **68**, 3304–3313 (2008).
- 39. van Kampen, N. G. Stochastic Processes in Physics and Chemistry, 2nd ed. (Elsevier, Amsterdam, 1997).
- Lenaerts, T., Pacheco, J. M., Traulsen, A. & Dingli, D. Tyrosine kinase inhibitor therapy can cure chronic myeloid leukemia without hitting leukemic stem cells. *Haematologica* 95, 900–907 (2010).
- 41. Quintana, E. et al. Efficient tumour formation by single human melanoma cells. Nature 456, 593-598 (2008).
- 42. Kelly, P. N., Dakic, A., Adams, J. M., Nutt, S. L. & Strasser, A. Tumor growth need not be driven by rare cancer stem cells. Science 317, 337 (2007).
- Brooks, M. D., Burness, M. L. & Wicha, M. S. Therapeutic Implications of Cellular Heterogeneity and Plasticity in Breast Cancer. Cell Stem Cell 17, 260–271 (2015).
- 44. Visvader, J. E. & Lindeman, G. J. Cancer stem cells: current status and evolving complexities. Cell Stem Cell 10, 717–728 (2012).

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Author Contributions

K.Q. and Y.J. designed the project, K.Q., L.W., J.S., A.Y., P.H., D.S., J.B.K. and X.Z. performed the calculations, K.Q. and Y.J. analyzed the results. K.Q. and Y.J. wrote the manuscript.

Additional Information

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