Visual synchronization of two 3-variable Lotka–Volterra oscillators based on DNA strand displacement

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DNA strand displacement as a theoretical foundation is helpful in constructing reaction networks and DNA circuits. Research on chemical kinetics is significant to exploit the inherent potential property of biomolecular systems. In this study, we investigated two typical reactions and designed DNA strands with a fluorophore and dark quencher for reaction networks using a 3-variable Lotka–Volterra oscillator system as an example to show the convenience of and superiority for observation of dynamic trajectory using our design, and took advantage of the synchronization reaction module to synchronize two 3-variable Lotka–Volterra oscillators. The classical theory of chemical reaction networks can be used to represent biological processes for mathematical modeling. We used this method to simulate the nonlinear kinetics of a 3-variable Lotka–Volterra oscillator system.

1. Introduction

Nonlinear kinetics exists widely in mass chemical reaction equations. Chemical reactions with DNA strand displacement (DSD) also have the property of being nonlinear and cascading. We can use the property of cascading to construct complex chemical reaction networks (CRNs) as well as to compile chemical equations into circuit systems, for example, in pulse counter and incremeneter state machines.1–3

With the aid of DNA strand displacement, massive DNA circuits, including digital and analog circuits, have been designed and validated by experiments.4–13 The outputs of the DNA circuit are represented by the concentrations of the DNA strands, but the detection of concentration is difficult or even impossible in some cases. For example, the structure of a DNA polymer is massive, complex, and unstable, such that detection of the concentration of the DNA polymer is impossible.14 If the DNA strand is designed with a fluorophore and dark quencher, observation of the output is convenient and fast. Luminous intensity is enhanced with an increase in the concentration of DNA strands.15 Thus, it is helpful to measure the concentration of DNA strands by fluorescence intensity. This property is particularly useful for digital circuits because these circuits are encoded with binary bits, where 2 possible values of each bit are “0” and “1”. If the luminous intensity is strong enough, the value of the output is “1”, otherwise the value of the output is “0”. For an analog circuit, the output is represented qualitatively by luminous intensity; for polymerization of a hybridization chain reaction,14 characterization of the hyperbranched chain reaction can be recognized by fluorescence intensity.

Synchronization is an important dynamical behavior and a general and ubiquitous phenomenon that has been investigated in a variety of fields such as complex networks, chemical reaction networks,16–18 secure communication, and genetic regulatory networks. Recently, it was found that synchronization is beneficial to DNA nanotechnology; for example, synchronization is applied in self-assembly,19 gene expression20 and DNA replication.21 Synchronized systems can enjoy common dynamical behavior and isochronous signal transition or transmission. Thus, synchronization has extensive application potential in digital and analog DNA circuits.

As compared to previous studies, we have proposed catalysis, degradation and synchronization reaction modules with a fluorophore and dark quencher, and identified different DNA strands with different colors of fluorophores, where the synchronization reaction module is a new reaction module. Then, we compiled these reaction modules into the 3-variable Lotka–Volterra oscillator and realized synchronization of two 3-variable Lotka–Volterra oscillator systems. From the simulations of these reaction modules, it was demonstrated that the DNA implementation of CRN can imitate the dynamical behavior of an ideal formal reaction, and there is a relationship between the reaction rates of DNA CRNs, intermediate CRNs and ideal formal CRNs.

We used visual DSD software to design the DNA strand and test the feasibility of the chemical reaction with DNA strand displacement, and utilized chemical reaction network (CRN) software to numerically simulate the dynamic behavior of the 3-variable Lotka–Volterra oscillator.
Fig. 1 DNA strand displacement with the fluorophore and quencher. The arrow in a DNA strand indicates the 3' end. The domain \( a_2 \) is 12 nucleotides, and the toehold domain is usually less than 10 nucleotides, so the toehold binding reaction is reversible.

2. Recent work

2.1. DNA strand displacement with fluorophore and dark quencher

We presented a reversible, bimolecular chemical reaction, as shown in Fig. 1, where \( A_1 \) and \( A_2 \) are reactants and \( A_3 \) and \( A_4 \) are products; \( k_1 \) denotes the forward reaction rate, and \( k_2 \) denotes the backward reaction rate. The reaction begins with an invader DNA strand \( X \) reacting with \( H \), the hairpin is open and the toehold binding reaction is reversible.

2.2. DNA strand displacement with hairpin structure

Hairpin is a special structure for single strand DNA (ssDNA), which can protect some domains. As shown in Fig. 2, domains \( a_2 \), \( b_1 \), and \( b_2 \) are protected before the hairpin is open. When an invader DNA strand \( X \) reacts with \( H \), the hairpin is open and then, domains \( a_2 \), \( b_1 \), and \( b_2 \) can be used for further reactions.

3. Results

3.1. CRN modules

In DNA strand displacement as a type of chemical reaction, there is a corresponding relationship between DNA CRNs and the ideal formal CRNs, where there are intermediate CRNs in the transition between DNAs and ideal formal CRNs. In order to endure CRNs modularity, we propose three types of reaction modules by DNA CRNs and ideal CRNs, where \( [X]_0 \), \( [XM]_0 \) and \( [X]_1 \) represent the initial concentration of species \( X \), \( X_M \) and \( X_1 \) in DNA CRNs, intermediate CRNs and ideal CRNs, respectively, and satisfied \( [X]_0 = [XM]_0 = [X_1]_0 \). \( [Y]_0 \), \( [YM]_0 \) and \( [Y_1]_0 \) represent the initial concentration of species \( Y \), \( Y_M \) and \( Y_1 \) in DNA CRNs, intermediate CRNs and ideal CRNs, respectively, and satisfied \( [Y]_0 = [YM]_0 = [Y_1]_0 \).

In the catalysis reaction module (1), as shown in Fig. 3(b), \( X'_{YM} \) is produced by \( X_M \) in (5), and the increase in \( X'_M \) will catalyze (6), while the producer \( X_M \) can catalyze the production of \( X'_M \). According to (1), \( d[X]/dt = -q_m X_m[X]_0 \) and \( d[XM]/dt = -\theta_m [XM]_0 \) by (5); if DSD (1) and (2) can approximate to the intermediate (3), \( q_i = \theta_i/C_m \) is obtained. According to (3), \( d[X]/dt = -q_m X_m[X]_0 \), while \( d[XM]/dt = -\theta_m [XM]_0 \) based on (6). If DSD (3) and (4) can approximate to the intermediate (6), \( q_m = \theta_m/C_m \), \( d[X]/dt = -\theta_m [X]_0 \) and \( d[XM]/dt = -q_m [XM]_0 \), according to (4) and (6); if (4) is equivalent to (6), \( q_i = \theta_i/C_m \). Similarly, we can obtain \( k_1 = \theta_b \) based on (5), (6) and (7). Initial concentrations of the auxiliary species \( A_i \), \( C_i \), \( P_j \), \( Sp_i \) and \( Sq_j \) are set to \( C_m \), where \( [X_0] \ll C_m \) and \( [X]_0, [Y]_0, [YM]_0 \) are the initial concentrations of \( X \) and \( Y \). As shown in Fig. 3(c), the evolutions of \( X \) in DNA CRNs, intermediate CRNs and ideal CRNs are approximately coincident. This shows that DNA CRNs, intermediate CRNs and ideal formal CRNs in the catalysis reaction module (1) are equivalent, but there are nuances between DNA CRNs, intermediate CRNs and ideal formal CRNs.

In catalysis reaction module (2), as shown in Fig. 4(b), \( Y'_M \) is produced by \( X_M \) and \( Y_M \) in (13), and the increase in \( Y'_M \) will catalyze (14), while the producer \( Y_M \) can catalyze the production of \( Y'_M \) and consumption of \( X_M \). According to ref. 1, we have, \( q_i = \theta_i \). Similar to catalysis reaction module (1), \( q_i = \theta_i/C_m \), \( q_m = \theta_m/C_m \), and \( k_2 = \theta_b \). Initial concentrations of the auxiliary species \( A_i \), \( B_i \), \( H_i \), \( P_j \), \( Sp_i \) and \( Sq_j \) are set to \( C_m \), where \( [X_0] \ll C_m \). As shown in Fig. 4(c), the evolutions of \( Y \) in DNA CRNs, intermediate CRNs and ideal CRNs are approximately coincident. This indicates that DNA CRNs, intermediate CRNs and ideal CRNs in the catalysis reaction module (2) are equivalent, but there are nuances between DNA CRNs, intermediate CRNs and ideal CRNs.

In the synchronization reaction module, it is assumed that \( X \) and \( M_a \) attain instant equilibrium through (18), with \( [X]_0/[M_a]_0 = q_m/q_i \) when \( X_M \) and \( M_a \) attain instant equilibrium through (24), we obtain \( [X_M]_0/[M_a]_0 = \theta_m/C_m \). According to (19) and (25), we have \( d[M_a]/dt = -C_m q_s [M_a]_0 \) as a result, \( q_s = \theta_s C_m \). Taking (25) and (28) into account, we obtain \( d[Y]_0/dt = \theta_s [X]_0 \), and \( d[Y_1]/dt = k_b [X]_0 \), respectively; if intermediate reaction (25) can be approximated to the ideal formal reaction (28), \( \theta_s [M_a]_0 = k_b [X]_0 \). Because \( [X]_0/[M_a]_0 = q_m/q_i \), when \( X \) and \( M_a \) attain instant equilibrium and \( [X]_0 = [X]_0/[M_a]_0 \), we can obtain \( k_b = q_i \theta_b/q_m + q_i = q_i \theta_b C_m q_i + q_i \). The initial concentrations of the auxiliary species \( N_{ai} \), \( N_{bi} \), \( N_{ci} \), \( N_{di} \), \( N_{bi} \), \( N_{ei} \) and \( N_f \) are set to \( C_m \), where \( [X]_0 \ll C_m \). Fig. 6(c) shows that \( X \) and \( Y \) can share the same dynamic behavior in DNA CRNs, intermediate CRNs and ideal formal 

Fig. 2 DNA strand displacement with a hairpin structure. The toehold \( b_1 \) is as long as the other toeholds. The hairpin cannot open spontaneously as a result of the dissociation of \( b_1 - b_1' \) hybridization at the end of the branch migration, the loop portion \( a_2 - b_1 - b_2 \) of the hairpin is protected if it does not open.
CRNs, but their synchronous values are different. The reasons for this phenomenon are as follows:

I. In DNA CRNs

$$[X]_0 + [Y]_0 \rightarrow [X]_i + [Y]_i + [M_{A_i}]_i + [M_{B_i}]_i + [M_{C_i}]_i + [M_{D_i}]_i$$  

When DNA CRNs approach reaction equilibrium, the following can be obtained:

$$[X]_i = [Y]_i$$

$$[M_{A_i}]_i = [M_{C_i}]_i$$

$$[M_{B_i}]_i = [M_{D_i}]_i = 0$$

Considering (30) and (31), we can obtain the synchronous value:
When intermediate CRNs approach reaction equilibrium, we can obtain

\[ [X_M]_t = [Y_M]_t = \frac{(X_M)_0 + (Y_M)_0}{2(q_i + q_m)} \]  (35)

II. In intermediate CRNs

\[ [X_M]_0 + [Y_M]_0 = [X_M]_t + [Y_M]_t + [M_a]_t + [M_c]_t \]  (34)

III. In ideal CRNs
Fig. 5  Degradation reaction module $Y \rightarrow \emptyset$: DNA implementation of formal species $Y$ and degradation with auxiliary species $F_i$; in (16), species $Y$ reacts with species $F_i$ producing inert waste, where the initial concentration of auxiliary species $F_i$ is set to $C_m$.

Fig. 6  Synchronization reaction module. (a) A list of DNA reactions; (b) exhibits the DNA CRNs and appropriate ideal formal CRNs, where intermediate CRNs is the transition between DNAs and ideal formal CRNs; (c) represents the process of synchronization between $X$ and $Y$ in DNA CRNs, intermediate CRNs and ideal formal CRNs, respectively, where $C_m = 10^4 \text{nM}$, $[X_0] = [X_d_0] = 4 \text{nM}$, and $[Y_0] = [Y_d_0] = 8 \text{nM}$; $q_{s_i} = 10^{-6} \text{nM s}^{-1}$, $q_{s_i} = 3 \times 10^{-4} \text{nM s}^{-1}$ and $q_{m} = 10^{-4} \text{nM s}^{-1}$. In (18), $X$ react with $N_a$, producing $M_a$, reversely; the producer $M_a$ reacts with $N_{d_i}$ to produce $M_{d_i}$ in (19), and $Y$ is produced by $M_{d_i}$ and $N_{c_i}$, which is equivalent to $Y_i$ being produced by $X_i$ in ideal CRNs (28). DSD implementation of ideal reaction (29) is similar to (28).
\[ [X_1]_0 + [Y_1]_0 = [X_1]_0 + [Y_1]_0 \] (36)

Eqn (36) is satisfied when \( X_1 \) and \( Y_1 \) approach instant equilibrium through (27) and (28)

\[ [X_1]_0 = [Y_1]_0 \] (37)

Therefore, the synchronous value of \( X_1 \) and \( Y_1 \) are obtained as follows:

\[ [X_1]_0 = [Y_1]_0 = \frac{[X_1]_0 + [Y_1]_0}{2} \] (38)

### 3.2. 3-Variable Lotka–Volterra oscillator

The Lotka–Volterra model is one of the earliest sound mathematical principles \(^{27-29}\) to describe population dynamics, and the Lotka–Volterra oscillator is known to be an ecological food chain oscillator. A 2-variable Lotka–Volterra oscillator can be described by ideal formal CRNs as follows:\(^1\)

\[
\begin{align*}
X_1 & \xrightarrow{r_1} 2X_1 \\
X_1 + X_2 & \xrightarrow{r_2} 2X_2 \\
X_2 & \xrightarrow{r_3} \phi
\end{align*}
\] (39a)(39b)(39c)

where \( X_1 \) and \( X_2 \) are reactants; \( r_1, r_2, \) and \( r_3 \) represent the reaction rate of reaction (39a–c), respectively. Reaction (39a) leads to an increase in \( X_1 \), and \( X_1 \) can catalyze reaction (39a).

The product of \( X_1 \) will increase the product \( X_2 \), and \( X_2 \) can catalyze reaction (39b). When the concentration of \( X_2 \) increases to a certain degree, an increase in \( X_2 \) will lead to consumption of \( X_1 \) and the decrease in \( X_1 \) will slow down reaction (39b), which can lead to a decrease in \( X_2 \). Therefore, in the cycle, the concentrations of \( X_1 \) and \( X_2 \) will fluctuate over time.

According to the basic Lotka–Volterra oscillator, a 3-variable Lotka–Volterra oscillator model can be described by ideal CRNs as follows:

\[
\begin{align*}
X_1 & \xrightarrow{r_1} 2X_1 \\
X_1 + X_2 & \xrightarrow{r_2} 2X_2 \\
X_1 + X_3 & \xrightarrow{r_3} 2X_3 \\
X_2 & \xrightarrow{r_4} \phi \\
X_3 & \xrightarrow{r_5} \phi
\end{align*}
\] (40a)(40b)(40c)(40d)(40e)

CRNs (40) can be represented by the following set of ODEs:

\[
\begin{align*}
\dot{X}_1 &= r_1X_1 - r_2X_1X_2 - r_3X_1X_3 = F_1(X_1) \\
\dot{X}_2 &= r_2X_1X_2 - r_4X_2 = F_2(X_2) \\
\dot{X}_3 &= r_3X_1X_3 - r_5X_3 = F_3(X_3)
\end{align*}
\] (41)(42)(43)

Fig. 8(a) shows the evolution of Lyapunov exponents \( \lambda_i \) (\( i = 1,2,3 \)) of ideal formal CRNs (40) with the chemical reaction rate parameter, \( r_1 \). It is known that the kinetic characteristics of the system can be characterized by the Lyapunov exponent, if exist in the range in which one Lyapunov exponent is positive and another Lyapunov exponent is negative, indicating that the system is at a chaotic state. Furthermore \( \lambda_1 < 0, \lambda_2 > 0 \) and \( \lambda_3 > 0 \) when the parameter is taken as \( r_1 = 5.5/300 \text{ s}^{-1} \), and the phase map of the state variables \( X_1, X_2, \) and \( X_3 \) is shown in Fig. 8(b).

### 3.3. Synchronization of two 3-variable Lotka–Volterra oscillator systems

Considering another 3-variable Lotka–Volterra oscillator system

\[
\begin{align*}
Y_1 & \xrightarrow{r_1} 2Y_1 \\
Y_1 + Y_2 & \xrightarrow{r_2} 2Y_2 \\
Y_1 + Y_3 & \xrightarrow{r_3} 2Y_3
\end{align*}
\] (44a)(44b)(44c)
where the initial concentrations of the above two systems are different. In this study, we used four types of fluorophores as reporters: ssDNA $X_1$ and $Y_1$ are identified by the TYE563 fluorophore (blue ball) with $\lambda_{ex} = 549$ nm, $\lambda_{em} = 563$ nm; ssDNA $X_2$ and $Y_2$ are identified by the ROX fluorophore (red ball) with $\lambda_{ex} = 588$ nm, $\lambda_{em} = 608$ nm; ssDNA $X_3$ and $Y_3$ are identified by FAM fluorophore (green ball) (45e) with $\lambda_{ex} = 495$ nm and $\lambda_{em} = 520$ nm; and the dark quencher is Lowa Black RQ.11

In order to realize internal synchronization between two chemical reaction networks, we added coupling terms as follows:

\[
\begin{align*}
X_1 & \xrightarrow{r_1} 2X_1 \\
X_1 + X_2 & \xrightarrow{r_2} 2X_2 \\
X_1 + X_3 & \xrightarrow{r_3} 2X_3 \\
X_2 & \xrightarrow{r_4} \phi \\
X_3 & \xrightarrow{r_5} \phi \\
X_1 & \xrightarrow{r_6} Y_1 \\
X_2 & \xrightarrow{r_7} Y_2 \\
X_3 & \xrightarrow{r_8} Y_3 \\
Y_1 & \xrightarrow{r_9} 2Y_1 \\
Y_1 + Y_2 & \xrightarrow{r_{10}} 2Y_2 \\
Y_1 + Y_3 & \xrightarrow{r_{11}} 2Y_3
\end{align*}
\]

ODEs of CRNs (45) and (46) can be described as follows:

\[
\begin{align*}
\dot{X}_1 &= r_1 X_1 - r_2 X_1 X_2 - r_3 X_1 X_3 + r_6 \\
&= F_1(X_1) + r_6 (Y_1 - X_1) \\
\dot{X}_2 &= r_2 X_1 X_2 + r_6 Y_2 - r_4 X_2 - r_6 X_2 \\
&= F_2(X_2) + r_6 (Y_2 - X_2) \\
\dot{X}_3 &= r_3 X_1 X_3 + r_6 Y_1 - r_5 X_3 - r_6 X_3 \\
&= F_3(X_3) + r_6 (Y_3 - X_3)
\end{align*}
\]

\[
\begin{align*}
\dot{Y}_1 &= r_1 Y_1 - r_2 Y_1 Y_2 - r_3 Y_1 Y_3 + r_6 (X_1 - Y_1) \\
&= G_1(Y_1) + r_6 (X_1 - Y_1) \\
\dot{Y}_2 &= r_2 Y_1 Y_2 + r_6 X_2 - r_4 Y_2 - r_6 Y_2 \\
&= G_2(Y_2) + r_6 (X_2 - Y_2) \\
\dot{Y}_3 &= r_3 Y_1 Y_3 + r_6 X_3 - r_5 Y_3 - r_6 Y_3 \\
&= G_3(Y_3) + r_6 (X_3 - Y_3)
\end{align*}
\]

The error state was defined as $e_i = y_i - x_i$ ($i = 1, 2, 3$), and the dynamics of error can be described as follows:

\[
\begin{align*}
\frac{de_1}{dt} &= G_1(Y_1) - F_1(X_1) - 2r_6 e_1 \\
\frac{de_2}{dt} &= G_2(Y_2) - F_2(X_2) - 2r_6 e_2 \\
\frac{de_3}{dt} &= G_3(Y_3) - F_3(X_3) - 2r_6 e_3
\end{align*}
\]
In simulations, the two 3-variable Lotka–Volterra oscillator systems with different initial values, and the evolution of the maximum Lyapunov exponent of eqn (53) with the chemical reaction rate \( r_6 \) is given in Fig. 9. When the maximal Lyapunov exponent \( \lambda_m \) is negative, the synchronization of the two 3-variable Lotka–Volterra oscillator systems can be obtained.\(^a\)

CRNs (45) and (46) can be described by the abovementioned DNA reaction modules as follows:

\[
\begin{align*}
S_1 + S_p &\xrightarrow{qs \omega} X_2 + Sq_1 \\
X_1 + A_2 &\xrightarrow{q_2 \omega} T_2 + B_2 \\
X_3 + T_2 &\xrightarrow{q_2} \text{waste} + C_2 \\
C_2 + H_2 &\xrightarrow{q_2 \omega} \text{waste} + 2X'_1 \\
X'_1 + P_2 &\xrightarrow{q_2 \omega} S_2 \\
S_2 + S_p &\xrightarrow{q_2 \omega} X_3 + Sq_2 \\
X_2 + F_1 &\xrightarrow{q_2 \omega} E_1 \\
X_3 + F_2 &\xrightarrow{q_2 \omega} E_2 \\
Y_1 + A'_3 &\xrightarrow{q_2 \omega} T_3 + B'_3 \\
T'_3 + C'_3 &\xrightarrow{q_2 \omega} 2Y'_1 + \text{waste} \\
Y'_1 + P'_3 &\xrightarrow{q_2 \omega} S'_1 \\
S'_1 + S_p &\xrightarrow{q_2 \omega} Y_1 + Sq'_1 \\
Y_1 + A_3 &\xrightarrow{q_2 \omega} T_3 + B_3 \\
Y_2 + T_3 &\xrightarrow{q_2 \omega} \text{waste} + C_3 \\
C_3 + H_1 &\xrightarrow{q_2 \omega} \text{waste} + 2Y'_2 \\
Y'_2 + P_3 &\xrightarrow{q_2 \omega} S_3 \\
S_3 + S_p &\xrightarrow{q_2 \omega} Y_2 + Sq_3 \\
Y_1 + A_4 &\xrightarrow{q_2 \omega} T_4 + B_4 \\
Y_3 + T_4 &\xrightarrow{q_2 \omega} \text{waste} + C_4 \\
C_4 + H_4 &\xrightarrow{q_2 \omega} \text{waste} + 2Y'_4 \\
Y'_4 + P_4 &\xrightarrow{q_2 \omega} S_4 \\
S_4 + S_p &\xrightarrow{q_2 \omega} Y_4 + Sq_4 \\
Y_2 + F_3 &\xrightarrow{q_2 \omega} E_3 \\
Y_3 + F_4 &\xrightarrow{q_2 \omega} E_4 \\
X_1 + U_1 &\xrightarrow{q_2 \omega} V_1 + U'_1
\end{align*}
\]

\( r_1 = 5.5/300 \text{ s}^{-1}, \quad q_1 = r_2 C_m, \quad C_m = 2 \times 10^{-5} \text{ M} \)

\( r_2 = 10^6 \text{ M} \text{ s}^{-1}, \quad q_2 = 10^3 \text{ M} \text{ s}^{-1}, \quad q_m = 10^6 \text{ M} \text{ s}^{-1} \)

\( r_3 = 1.008 \times 10^6 \text{ M} \text{ s}^{-1}, \quad r_3 = r_5 C_m, \quad q_3 = 10^3 \text{ M} \text{ s}^{-1} \)

\( r_4 = 4/300 \text{ s}^{-1}, \quad q_4 = 10^3 \text{ M} \text{ s}^{-1} \)

\( r_5 = 4/300 \text{ s}^{-1}, \quad q_5 = 10^3 \text{ M} \text{ s}^{-1} \)

\( r_6 = 0.8/300 \text{ s}^{-1} \)

\( q_6 = 2 \times 10^3 \text{ M} \text{ s}^{-1} \)

\( q_7 = r_6 C_m \)

\( q_8 = r_6 C_m \)

\( q_9 = 2 \times 10^2 \text{ M} \text{ s}^{-1} \)

\( q_10 = r_6(q_m + q_9)(q_2 C_m) \)

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**Table 1** Parameter Values of the Lotka–Volterra Oscillator

| Parameter | Value | Parameter | Value | Parameter | Value |
|-----------|-------|-----------|-------|-----------|-------|
| \( r_1 \) | 5.5/300 s\(^{-1}\) | \( q_1 \) | \( r_2 C_m \) | \( C_m \) | \( 2 \times 10^{-5} \text{ M} \) |
| \( r_2 \) | 10^6 M s\(^{-1}\) | \( q_2 \) | \( 10^3 \text{ M} \text{ s}^{-1} \) | \( q_m \) | \( 10^6 \text{ M} \text{ s}^{-1} \) |
| \( r_3 \) | 1.008 \( \times 10^6 \text{ M} \text{ s}^{-1} \) | \( r_3 \) | \( r_5 C_m \) | \( q_3 \) | \( 10^3 \text{ M} \text{ s}^{-1} \) |
| \( r_4 \) | 4/300 s\(^{-1}\) | \( q_4 \) | \( 10^3 \text{ M} \text{ s}^{-1} \) | \( q_4 \) | \( r_6 C_m \) |
| \( r_5 \) | 4/300 s\(^{-1}\) | \( q_5 \) | \( 10^3 \text{ M} \text{ s}^{-1} \) | \( q_5 \) | \( 2 \times 10^3 \text{ M} \text{ s}^{-1} \) |
| \( r_6 \) | 0.8/300 s\(^{-1}\) | \( q_6 \) | \( r_6 C_m \) | \( q_6 \) | \( r_6(q_m + q_9)(q_2 C_m) \) |
Visual DSD software can be used to realize the approximations of the CRNs (45) and (46). The DSD implementation of the catalysis reactions (54) and (59) are shown in Fig. 3. The DSD implementation of the catalysis reactions (55), (56), (60) and (61) are shown in Fig. 4. The reactions (57), (58), (62) and (63) are degradation reactions. The DSD implementation of reactions (57), (58), (62) and (63) are shown in Fig. 5. Reactions (64)–(66) are used as the synchronization module. The DSD implementation of reactions (64)–(66) are shown in Fig. 6.

**4. Simulation results**

The parameters of the 3-variable Lotka–Volterra oscillator are given in Table 1, and $[X_1]_0$, $[X_2]_0$, $[X_3]_0$, $[Y_1]_0$, $[Y_2]_0$ and $[Y_3]_0$ represent the initial values of species $X_1$, $X_2$, $X_3$, $Y_1$, $Y_2$ and $Y_3$, respectively. $[X_1]_0 = 20 \text{nM}$, $[X_2]_0 = 12 \text{nM}$, $[X_3]_0 = 8 \text{nM}$, $[Y_1]_0 = 1 \text{nM}$, $[Y_2]_0 = 50 \text{nM}$, and $[Y_3]_0 = 15 \text{nM}$, where the values of chemical reaction rates $r_i (i = 1, \ldots, 5)$ are referenced.

The evolution of the 3-variable Lotka–Volterra oscillator based on ideal CRNs and the 3-variable Lotka–Volterra oscillator based on DNA CRNs are shown in Fig. 7 and 8. When the three coupling terms are added to the system, the evolution of the 3-variable Lotka–Volterra oscillator based on DNA CRNs and the 3-variable Lotka–Volterra oscillator based on ideal CRNs are given in Fig. 9. According to Fig. 10, it can be found that the error between $Y_i$ and $X_i$ $(i = 1, 2, 3)$ approaches zero, which indicates that synchronization among $X_i$ and $Y_i$ with different initial concentrations is realized. It is clear that errors between $X_i$ and $Y_i$ in the ideal CRNs approach zero, which is similar to errors in DNA CRNs, as shown in Fig. 10 and 11.

In CRNs, luminous intensity relates to the concentration of DNA strands. We can observe the process of synchronization of the two 3-variable Lotka–Volterra oscillator systems through luminous intensity. The luminous intensity of $X_i$ and $Y_i$ are different without adding coupling terms to the systems. The luminous intensity of $X_i$ and $Y_i$ reach unanimity when there is synchronization among $X_i$ and $Y_i$. 

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**Fig. 10** Process of synchronization of two 3-variable Lotka–Volterra oscillator systems based on DNA and ideal CRNs.

**Fig. 11** Evolution of errors between two 3-variable Lotka–Volterra oscillator systems based on DNA and ideal CRNs.
5. Conclusion

We proposed a 3-variable Lotka–Volterra oscillator system as an example and realized synchronization of two 3-variable Lotka–Volterra oscillator systems with different initial concentrations. For more convenient observation and more accurate detection of synchronization, DNA strands were designed with fluorophores and quenchers. We demonstrated the contributions of fluorophores and quenchers using the CRNs modules, and analyzed the differences between DNA CRNs and ideal formal CRNs. We used Visual DSD software as a design tool for the DNA strands and CRN language to simulate the synchronization of two 3-variable Lotka–Volterra oscillator systems, one each in DNA CRNs and ideal CRNs. Numerical simulation demonstrates that the method of synchronization in this article is effective for ideal CRNs as well as DSD implementation CRNs.

Conflicts of interest

There are no conflicts to declare.

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