

# Molecular Sensing and Computing Systems

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(Invited Paper)

**Abstract**—Advances in the field of synthetic biology have been key to demonstration of molecular computing systems in general and DNA in particular. This paper presents an overview of how continuous-time, discrete-time, and digital signal processing systems can be implemented using molecular reactions and DNA. In this paper, discrete-time systems refer to sampled signals with continuous signal amplitude. Signals that are sampled in discrete time steps with digital amplitude are referred to as digital signals. Delay elements in sampled signals are implemented using molecular reactions in the form of molecular transfer reactions. Completion of all phases of transfer reactions once corresponds to a computation cycle. These molecular systems can be implemented in a fully-synchronous, globally-synchronous locally-asynchronous or fully-asynchronous manner. The paper also presents molecular sensing systems where molecular reactions are used to implement analog-to-digital converters (ADCs) and digital-to-analog converters (DACs). Molecular implementations of digital logic systems are presented. A complete example of the addition of two molecules using digital implementation is described where the concentrations of two molecules are converted to digital by two 3-bit ADCs, and the 4-bit output of the digital adder is converted to analog by a 4-bit DAC. This system is demonstrated using both molecular reactions and DNA. A brief comparison of molecular and electronic systems is also presented.

**Index Terms**—Molecular systems, DNA, analog, digital, logic, signal processing, digital signal processing, analog-to-digital conversion, digital-to-analog conversion.

## I. INTRODUCTION

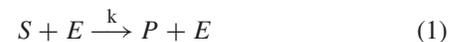
THE field of Molecular computing in general and DNA computing in particular have advanced remarkably in last 20-25 years. The progress in the broad field of synthetic biology continues to accelerate at a rate faster than Moore's law that refers to doubling in the number of devices on an integrated circuit (IC) chip every 18 months. A similar growth in synthetic biology is referred as Carlson's law [1], [2]. Early publications on molecular computing [3] and DNA computing [4] demonstrated the ability to compute using biological and chemical molecules, as an alternative to computing using silicon ICs [5]. Molecular computing has the potential to revolutionize monitoring concentrations or rates of change of concentrations of proteins and targeted drug delivery.

There is significant interest in synthesis of molecular circuits, *in vitro* or *in vivo*, to understand biological processes or

to re-engineer them by implementing particular computations [6]–[11]. This broad field, referred as synthetic biology, has seen remarkable progress since its inception in 2000. Efforts in synthetic biology have culminated in manipulation or even construction of a variety of molecular systems [12]–[15].

Molecular computing does not need to compete with traditional computing. Rather it is meant for activating or inhibiting pathways or monitoring proteins or delivering drugs at a very slow rate. The computation rates in molecular systems are typically 10–15 orders of magnitude slower than traditional computing. For example, monitoring a protein 4 times a day requires a sample period of 21,600 seconds compared to a clock period of 1 ns for a clock speed of 1 GHz. Fortunately, today's DNA circuits can meet these sample rate constraints for simple circuits. As the molecular computing technology evolves, it will be possible to realize molecular circuits that can implement computationally complex operations and/or at faster rates. Furthermore, massive parallelization is an important advantage of chemical and biological systems. For example it is known that more than 10M reactions per second can be performed in a human cell [16]. This power can be exploited for a dense data processing or storage in a cell-sized area.

A molecular system consists of a set of chemical reactions where reactants combine to form products. For example (1) shows a molecular reaction where  $S$  and  $E$  are reactants,  $P$  and  $E$  are products and  $k$  is the rate constant.



In this reaction, one molecule of  $E$  combines with one molecule of  $S$  to produce one molecule of  $P$  and one molecule of  $E$ . The dynamic behavior of reactant and product concentrations can be modeled using mass-action kinetic model. In this model the ordinary differential equations are used to model the concentration of each species as a function of time [17], [18]. For reaction (1) we have

$$\begin{aligned} \frac{dP}{dt} &= -\frac{dS}{dt} = kSE \\ \frac{dE}{dt} &= 0 \end{aligned} \quad (2)$$

The speed of reaction is proportional to the concentration of inputs and the rate constant.

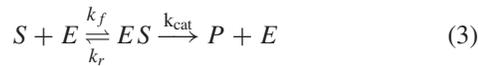
Whereas the mass-action kinetics model is a natural language for describing biochemical reactions, additional issues need to be considered for achieving more accurate modeling of biochemical systems by ODEs from mass-action model. For example if reaction (1) represents an enzymatic reaction with

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$E$ ,  $S$  and  $P$  are enzyme, substrate and product, respectively, the Michaelis-Menten kinetics provides a more accurate modeling of this reaction. Based on this model enzyme reactions are initiated by a binding interaction between the enzyme ( $E$ ) and the substrate ( $S$ ) to form a complex ( $ES$ ), which in turn is converted into a product ( $P$ ) and the enzyme. This can be represented by the following reactions:



where  $k_f$ ,  $k_r$ , and  $k_{cat}$  denote the rate constants [19] and the double arrows between  $S$  and  $ES$  represent the fact that enzyme-substrate binding is a reversible process. The ODEs for these reaction can be solved for different assumptions of the system to obtain the dynamic behavior of molecules. For example for a more probable situation, when the enzyme concentration is much less than the substrate concentration, the rate of product formation is given by

$$\frac{dP}{dt} = V_{max} \frac{S}{K_M + S} = k_{cat} E_0 \frac{S}{K_M + S}. \quad (4)$$

$K_M$  is Michaelis constant and defined as the substrate concentration at which the reaction rate is at half-maximum and  $E_0$  is the initial concentration of the enzyme.  $K_M$  and  $E_0$  are specific for each enzyme and can be obtained by experiment and are available for most of enzymes.

Different *in vivo* or *in vitro* mediums have been used for synthesizing a desired molecular system. For example, for *in vivo* case, promising approaches have used RNA interference (RNAi) and silencing (siRNA) to construct logic gates [20]–[22]. For the *in vitro* the DNA strand-displacement [23] is a well established medium for implementing and scaling up molecular systems. In this paper we use DNA strand-displacement as the medium for implementing and simulating our designs.

Computing or signal processing systems can either be analog or discrete-time. In *analog* processing, the input and output correspond to continuous-time signals. In *discrete-time* processing, the continuous-time signal is first sampled using a sampler, then processed in discrete time steps, and finally converted to a continuous-time signal if necessary by some form of interpolation. If the sampled signal in a discrete-time system is also discretized in amplitude, then it is referred to as a *digital* signal. A digital signal processing (DSP) system requires an analog-to-digital converter (ADC), processing of digital signals and finally a digital-to-analog conversion (DAC). Most information processing systems today store, process or transmit digital information. Discrete-time signal processing provides significantly higher accuracy than continuous-time since the delay elements can be realized with high-precision. In [24], it was recognized that the strength of a molecule was significantly degraded in an analog delay line with increase in the order of the system or the number of delays. In contrast, delay lines implemented in a discrete-time molecular or DNA system do not suffer from significant degradation. Digital processing provides even higher robustness and precise control in processing the signal in temporal or spectral domain than discrete-time

signals. We differentiate discrete-time as sampled in time but continuous in amplitude and digital as sampled in time and discretized in amplitude.

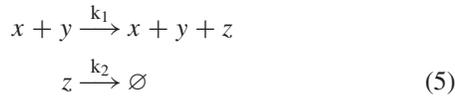
This paper presents synthesis of molecular computing systems that can be analog, discrete-time or digital. Analog and discrete-time processing of molecular systems have been published before. Synthesizing molecular and DNA reactions to implement continuous-time linear filters was first presented in [25]. Signal processing systems, implemented as either discrete-time or digital, contain delay elements. Delay elements transfer the molecules from their inputs to outputs without altering the concentration every computation cycle. Delay elements were first synthesized using molecular reactions in [26]. These systems can operate either in a fully-synchronous manner [27] using a two-phase clock, or in a locally-asynchronous globally-synchronous manner [26], [28], or in a fully-asynchronous manner [29] and [30]. The goal of this paper is two-fold. First, this paper presents a review of past work on continuous-time and discrete-time processing systems. Second, a new methodology to synthesize molecular ADCs and molecular DACs are presented. Molecular and DNA implementations of a complete digital processing system using ADC, digital computing and DAC are presented. These molecular designs can be scaled up with respect to their complexity. However, due to the resource limitation in living cells, they are more suitable for *in vitro* implementation.

This paper is organized as follows. In Section II, We provide a brief review of continuous-time systems and illustrate molecular implementation of a simple first-order analog filter. Discrete-time signal processing systems using fully-synchronous and locally-asynchronous globally-synchronous manner are reviewed in Section III. Molecular circuits for ADCs, digital logic, DACs, and a complete digital processing system that adds two numbers using 2 ADCs, one digital adder, and 1 DAC are described in Section IV. DNA implementation of the complete digital system and its simulation results are presented in Section V. A comparison of molecular and electronic systems is presented in Section VI. Finally Section VII provides discussion and concluding remarks.

## II. MOLECULAR CONTINUOUS-TIME SYSTEMS

Molecular implementations of continuous-time or analog systems have been described in many past publications [31]–[34]. Study of analog molecular systems is important since it has been proven that computations in living cells are mostly analog [31]–[33].

Analog computations can be implemented with chemical reaction networks (CRNs) efficiently with respect to the number of reactions and molecular species. For example, as presented for the first time in [33] and [34], implementing a molecular adder via analog computation is simple: we have two input concentrations to be added; both are transferred to the same molecular type by means of two independent reactions. In one application of an *in vivo* analog adder, two inputs may correspond to regulating the expressions of a common protein from two independent genetic promoters [33]. Analog multiplication can be simply implemented by two molecular reactions [35]:



From mass-action kinetics model we have

$$\frac{dz}{dt} = k_1xy - k_2z \quad (6)$$

where  $x, y$ , and  $z$  are molecular concentrations of their corresponding molecular types. In the steady-state  $\frac{dz}{dt} = 0$ , thus,  $z = \frac{k_1}{k_2}xy$ . The output  $z$  represents a scaled version of the product  $xy$ . Analog implementation of more complex functions such as square roots and logarithmic additions have been presented in [34]. Implementation of linear continuous-time systems with biochemical reactions has been presented in [25]. We briefly describe this method with an example. Each signal,  $u$ , is represented by the difference in concentration between two particular molecular types,  $u^+$  and  $u^-$ , where  $u^+$  and  $u^-$  are defined as:

$$u^+ = \begin{cases} u & \text{if } u > 0 \\ 0 & \text{otherwise} \end{cases} \quad (7)$$

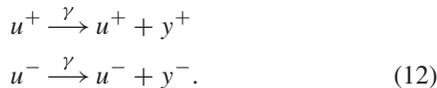
and

$$u^- = \begin{cases} |u| & \text{if } u < 0 \\ 0 & \text{otherwise.} \end{cases} \quad (8)$$

Any linear continuous-time system can be implemented using three building blocks: integrator, gain and summation. Using mass-action kinetics model, these blocks can be approximated by a minimal set of chemical reactions, referred as: catalysis, degradation, and annihilation reactions described by (9), (10), and (11), respectively.



where  $\gamma$  and  $\eta \in R^+$ . Reaction (9) is a concise representation of the following two reactions:



This notation is also adopted for other reactions with double superscripts. For each molecular type, an annihilation reaction is necessary to ensure a minimal representation of the molecule. For example, if  $y$  is used in a reaction network, the reaction  $y^+ + y^- \rightarrow \emptyset$  should be added.

*Integration:* Reactions (13) implement integration,  $y(t) = \int_0^t u(\tau)d\tau + y(0)$  with  $t \in R$ :



where  $\alpha \in R^+$ . For these reactions we have

$$\begin{aligned} \left. \begin{aligned} \frac{dy^+}{dt} &= \alpha u^+ \\ \frac{dy^-}{dt} &= \alpha u^- \end{aligned} \right\} \Rightarrow \frac{dy}{dt} &= \frac{dy^+}{dt} - \frac{dy^-}{dt} \\ &= \alpha u^+ - \alpha u^- = \alpha u \Rightarrow y(t) = \alpha \int_0^t u(\tau)d\tau + y(0). \end{aligned} \quad (14)$$

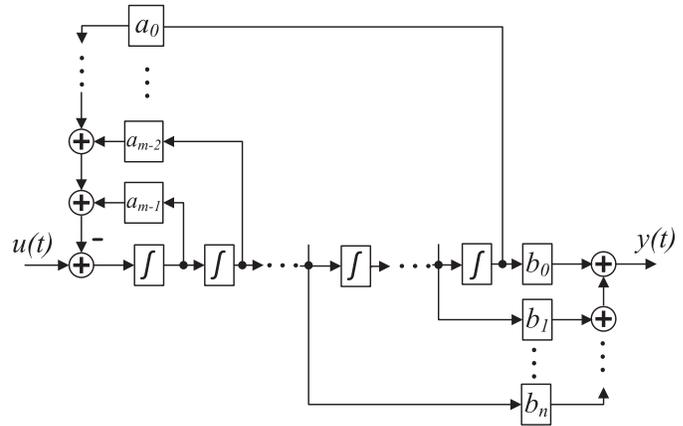


Fig. 1. Constructing linear I/O systems based on transfer function  $\frac{Y(s)}{U(s)} = \frac{B(s)}{A(s)}$ , using integration, gain, and summation blocks.

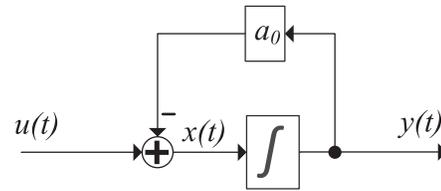


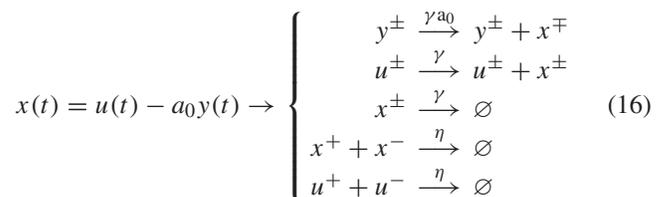
Fig. 2. A first order low-pass continuous-time filter.

*Gain and Summation:* The following chemical reactions output a linear combination of the input signals,  $u_i$ , with corresponding gain  $k_i$ .



where  $y$  represents the output,  $k_i, \gamma \in R^+$  for  $i \in 1, 2, \dots, n$ . In the special case  $n = 1$ , this chemical representation approximates the gain block,  $y = k_1u_1$  for  $k \geq 0$ . For  $n \geq 2$  this chemical representation approximates the summation block,  $y = \sum_{i=1}^n k_i u_i$  [25]. Suppose  $U(s)$  and  $Y(s)$  represent the Laplace transforms of input and output, respectively. Any linear I/O system with the transfer function  $\frac{Y(s)}{U(s)} = \frac{B(s)}{A(s)}$  can be approximated by using integration, gain, and summation blocks where  $B(s) = b_n s^n + b_{n-1} s^{n-1} + \dots + b_1 s + b_0$  and  $A(s) = s^m + a_{m-1} s^{m-1} + \dots + a_1 s + a_0$  and  $m \geq n$ . Figure 1 illustrates how  $\frac{Y(s)}{U(s)}$  can be constructed using these basic building blocks [36], [37].

A PI controller has been implemented in [25] using these blocks. Here, we illustrate an example molecular implementation of a first-order low-pass continuous-time filter, shown in Figure 2. The transfer function for this filter is  $\frac{1}{s+a_0}$ . It can be approximated by the following reactions:



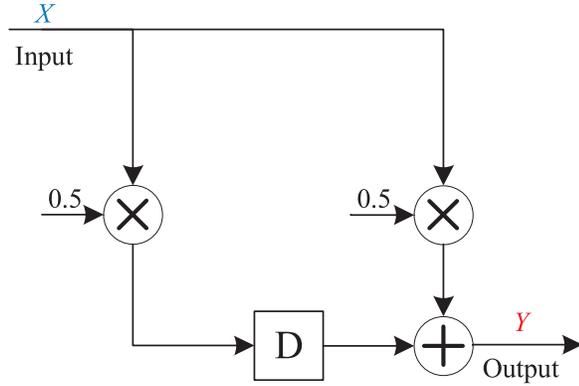


Fig. 3. Block diagram for the moving-average filter [26].

$$\frac{dy}{dt} = x(t) \rightarrow \begin{cases} x^\pm \xrightarrow{\gamma} x^\pm + y^\pm \\ y^+ + y^- \xrightarrow{\eta} \emptyset \end{cases} \quad (17)$$

### III. MOLECULAR DISCRETE-TIME SYSTEMS

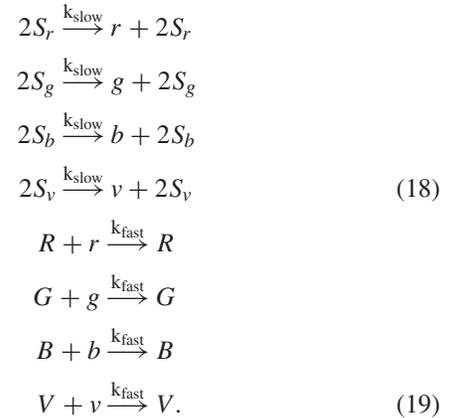
For discrete-time systems the corresponding computations start after the inputs are sampled at specific points in time. In these systems the timings of signal transfers need to be synchronized in order to avoid any interference in computations. The concept of a computational cycle in a molecular system is critical. Three different synchronization schemes have been proposed; these include: fully-synchronous, globally-synchronous locally-asynchronous, and fully-asynchronous. Fully synchronous systems are synchronized by a two-phase clock [27], [26]. In a globally-synchronous locally-asynchronous systems, three proteins, referred as Red ( $R$ ), Green ( $G$ ) and ( $Blue$ ) are introduced. The transfer of  $R$  to  $G$ ,  $G$  to  $B$  and  $B$  to  $R$  completes a computational cycle. The global  $RGB$  clock provides global synchronization [28], [26]. Fully-asynchronous systems do not make use of any global clock [29], [30]. Typically,  $RGB$  clocked systems are the fastest, while the fully-asynchronous systems are the slowest as these involve more phases of transfers. The protein transfer operation is a slow operation and is the bottleneck in molecular systems with respect to sample period. Although fully-synchronous systems require a two-phase clock, this clock is designed from a 4-phase protein transfer mechanism. This paper presents a brief review of the fully-synchronous and the  $RGB$  systems. Fully-asynchronous systems are not reviewed in this paper; however, the reader is referred to [29], [30].

All reactions in the discrete-time system are implemented using only two coarse rate categories for the reaction rate constants, i.e.,  $k_{fast}$  and  $k_{slow}$ . Given reactions with any such set of rates, the computation is correct. It does not matter how fast the fast reactions are or how slow the slow reactions are - only that all fast reactions fire relatively faster than slow reactions. We illustrate both schemes with a simple example, a moving-average filter. In fact, it is a first-order discrete-time low-pass filter. The circuit diagram for the filter is shown in Figure 3. It produces an output value that is one-half the current input value plus one-half the previous value. Given a time-varying input signal  $X$ , the output signal  $Y$  is a moving average, i.e.,

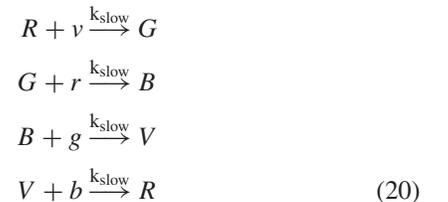
a smoother version of the input signal. Since there is no feedback in the system, it is called a *finite impulse response* (FIR) filter [38].

#### A. Fully-Synchronous Framework

In this framework a global clock signal synchronizes signal transfers in the system. For a molecular clock, reactions are chosen that produce sustained oscillations in terms of chemical concentrations. With such oscillations, a low concentration corresponds to a logical value of zero; a high concentration corresponds to a logical value of one. Techniques for generating chemical oscillations are well established in the literature. Classic examples include the Lotka-Volterra, the Brusselator, and the Arsenite-Iodate-Chlorite systems [39], [40]. Unfortunately, none of these schemes is quite suitable for synchronous sequential computation: the required clock signal should be symmetrical, with abrupt transitions between the phases. A new design was proposed in [26] and [27] for multi-phase chemical oscillator. For a 4-phase oscillator the phases can be represented by molecular types  $R, G, B, V$ . First consider the reactions



In reactions (18), the molecular types  $r, g, b, v$  are generated slowly and constantly, from source types  $S_r, S_g, S_b, S_v$ , whose concentrations do not change with the reactions. In reactions (19), the types  $R, G, B, V$  quickly consume the types  $r, g, b, v$ , respectively. Call  $R, G, B, V$  the phase signals and  $r, g, b, v$  the absence indicators. The latter are only present in the absence of the former. The reactions



transfer one phase signal to another, in the absence of the previous one. The essential aspect is that, within the  $R, G, B, V$  sequence, the full quantity of the preceding type is transferred to the current type before the transfer to the succeeding type begins. To achieve sustained oscillation, we introduce positive feedback. This is provided by the reactions

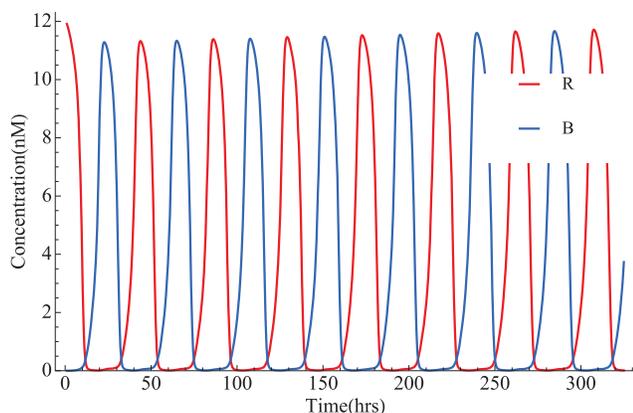
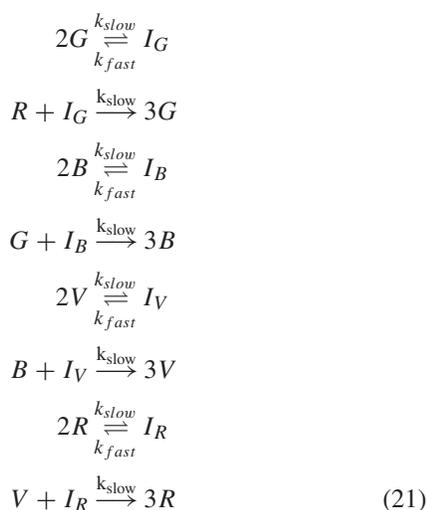


Fig. 4. simulation results for R and B phases of a four-phase oscillator [26].



Consider the first two reactions. Two molecules of  $G$  combine with one molecule of  $R$  to produce three molecules of  $G$ . The first step in this process is reversible: two molecules of  $G$  can combine, but in the absence of any molecules of  $R$ , the combined form will dissociate back into  $G$ . So, in the absence of  $R$ , the quantity of  $G$  will not change much. In the presence of  $R$ , the sequence of reactions will proceed, producing one molecule of  $G$  for each molecule of  $R$  that is consumed. Due to the first reaction  $2G \xrightarrow{k_{slow}} I_G$ , the transfer will occur at a rate that is super-linear in the quantity of  $G$ ; this speeds up the transfer and so provides positive feedback. Suppose that the initial quantity of  $R$  is set to some non-zero amount and the initial quantity of the other types is set to zero. We will get an oscillation among the quantities of  $R$ ,  $G$ ,  $B$ , and  $V$ .

One requirement for a clock in synchronous computation is that different clock phases should not overlap. A two-phase clock is used for synchronous structures: concentrations of molecular types representing clock phase 0 and clock phase 1 should not be present at the same time. To this end, two non-adjacent phases, say  $R$  and  $B$  in a four-phase  $RGBV$  oscillator, are chosen as the clock phases. The scheme for chemical oscillation works well. Figure 4 shows the concentrations of  $R$  and  $B$  as a function of time, obtained through differential equation simulations of the Reactions (18), (19), (20), and (21). It may be noted that the two phases  $R$  and  $B$  are essentially non-overlapping.

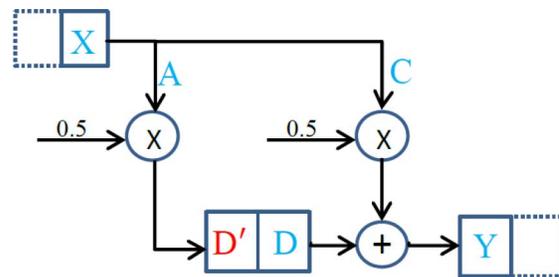


Fig. 5. Block diagram for synchronous implementation of the moving-average filter [26].

| S1                                       | S2                                    |
|--|---------------------------------------|
| $B + X \xrightarrow{k_{slow}} A + C + B$ | $R + D' \xrightarrow{k_{slow}} D + R$ |
| $2A \xrightarrow{k_{fast}} D'$           |                                       |
| $2C \xrightarrow{k_{fast}} Y$            |                                       |
| $B + D \xrightarrow{k_{slow}} Y + B$     |                                       |

Fig. 6. Set of molecular reactions for the synchronous implementation of the moving-average filter [26].

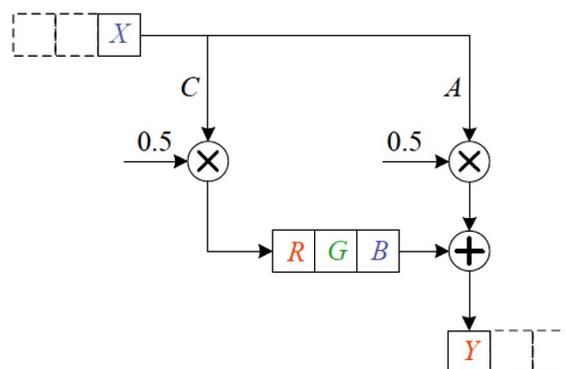


Fig. 7. Block diagram for the asynchronous implementation of the moving-average filter [26].

The delay and computation elements for the moving average filter in Figure 5 are implemented by the reactions in Figure 6. As Figure 5 shows each delay element,  $D$ , is modeled by two molecular types,  $D$  and  $D'$ . In the presence of  $B$ , the input signal  $X$  is transferred to molecular types  $A$  and  $C$ ; these are both reduced to half and transferred to  $D'$  and  $Y$ , respectively. In the presence of  $R$ ,  $D'$  is transferred to  $D$ . Therefore, in the following phase  $B$ , half of the new sample adds with the half of the previous sample stored in  $D$ .

### B. Globally-Synchronous Locally-Asynchronous Framework

The globally-synchronous locally-asynchronous framework is illustrated in Figure 7. It contains no clock signal; rather it is “self-timed” in the sense that a new phase of the computation begins when an external sink removes the entire quantity

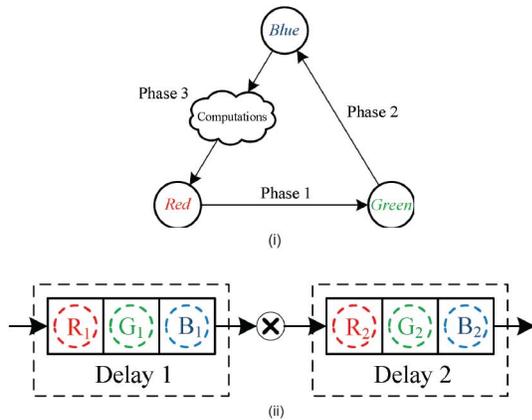


Fig. 8. (i) Implementing delay elements using the 3-phase asynchronous scheme. (ii) Cascaded delay elements implemented using asynchronous scheme [26].

of molecules  $Y$ , i.e., the previous output value, and supplies a new quantity of molecules  $X$ , i.e., the current input value. Each delay element in this framework is modeled by three molecular types, namely  $RGB$ . Figure 8 shows how the computations in asynchronous framework are performed in three phases and how delay elements are implemented using three molecular types  $R_i, G_i, B_i$ . In this framework, the moving-average filter is implemented by the reactions in Figure 9. The molecular types corresponding to signals are  $X, A, C, R, G, B$ , and  $Y$ . To illustrate the design, we use colors to categorize some of these types into three categories:  $Y$  and  $R$  in red;  $G$  in green; and  $X$  and  $B$  in blue. The group of the first three reactions shown in the  $S1$  column of Figure 9 transfers the concentration of  $X$  to  $A$  and to  $C$ , a *fanout* operation. The concentrations of  $A$  and  $C$  are both reduced to half, scalar multiplication operations. The concentration of  $A$  is transferred to the output  $Y$ , and the concentration of  $C$  is transferred to  $R$ . The transfer to  $R$  is the first phase of a delay operation. Once the signal has moved through the delay operation, the concentration of  $B$  is transferred to the output  $Y$ . Since this concentration is combined with the concentration of  $Y$  produced from  $A$ , this is an addition operation. The final group of three reactions shown in the  $S1$  column of Figure 9 implements the delay operation. The concentration of  $R$  is transferred to  $G$  and then to  $B$ . Transfers between two color categories are enabled by the absence of the third category: red goes to green in the absence of blue; green goes to blue in the absence of red; and blue goes to red in the absence of green. The reactions are enabled by molecular types  $r, g$ , and  $b$  that we call absence indicators. The absence indicators ensure that the delay element takes a new value only when it has finished processing the previous value. In the group of reactions shown in the  $S2$  column of Figure 9 molecules of types  $R', G'$ , and  $B'$  are generated from the signal types that we color-code red, green, and blue, respectively. The concentrations of the signal types remain unchanged. This generation/consumption process ensures that equilibria of the concentrations of  $R', G',$  and  $B'$  reflect the total concentrations of red, green, and blue color-coded types, respectively. Accordingly, we call  $R', G',$  and  $B'$  color concentration indicators. They serve to speed up signal transfers between color categories, and provide global synchronization.

In the group of reactions shown in the  $S3$  column of Figure 9, molecules of the absence indicator types  $r, g$ , and  $b$  are generated from external sources  $S_r, S_g$ , and  $S_b$ . At the same time, they are consumed when  $R', G'$ , and  $B'$  are present, respectively. Therefore, the absence indicators only persist in the absence of the corresponding signals:  $r$  in the absence of red types;  $g$  in the absence of green types; and  $b$  in the absence of blue types. They only persist in the absence of these types because otherwise “fast” reactions consume them quickly.

Finally, the reactions shown in the  $S4$  column of Figure 9 provide positive feedback kinetics. These reactions effectively speed up transfers between color categories as molecules in one category are “pulled” to the next by color concentration indicators. Note that the concentration of the input  $X$  is sampled in the green-to-blue phase. The output  $Y$  is produced in the blue-to-red phase.

Although the  $RGB$  scheme doesn’t have an independent global clock signal it provides a global synchronization by categorizing signals into three phases, so called  $RGB$  phases. Many local  $RGB$  blocks enable locally-asynchronous computation while global color concentrations,  $R', G', B'$ , provide global synchronization. In fact, they form a nonsymmetric clock dependent on the signal values of local  $RGB$  blocks.

Another fully-asynchronous framework has been proposed in [30]. In a fully-asynchronous system signal transfers and computations start from the input of the system and progress to its output in multiple phases. Each delay element in a fully-asynchronous system is modeled by two molecular types [30] and introduces two phases. Reactions for each phase are fired as soon as the preceding phase is completed. In addition to FIR filter, an IIR filter and an 8-point FFT have been implemented using this framework [26], [30].

#### IV. MOLECULAR SENSING AND DIGITAL COMPUTING SYSTEMS

Although analog computing systems are important due to their efficiency and their application in *in vivo* systems, digital computing systems are more robust [33], [41], [42]. In fact, regardless of the implementation technology, the fundamental reason for the robustness of the digital computation lies in information theory: information is coded across many 1-bit-precise interacting computational channels in the digital approach but on one channel in the analog approach [33].

Although complex molecular digital systems may be impractical today, these will be practical in near future as synthetic biology is seeing remarkable progress for synthesizing more complex systems *in vitro* especially from DNA. As a practical *in vitro* example, implementation of a scalable digital system, so called *seesaw gates*, with DNA strand-displacement reactions have been used to implement simple logical AND/OR gates, and 2-bit-precise square roots in [42].

Roughly speaking, in a digital molecular system, absence or existence of a molecular type defines whether the related signal is logically ‘0’ or ‘1’, respectively. More precisely, if the concentration of a molecular type is close to 0 nM it represents logical ‘0’, while if it is close to a distinguishable nonzero value, it represents logical ‘1’. In this paper, for *in vitro* DNA

| S1                                   | S2                                      | S3                                     | S4                                    |
|--------------------------------------|---|--|---------------------------------------|
| $g + X \xrightarrow{k_{slow}} A + C$ | $2 R \xrightarrow{k_{fast}} 2R + R'$    | $2S_r \xrightarrow{k_{slow}} 2S_r + r$ | $R' + X \xrightarrow{k_{fast}} A + C$ |
| $2 A \xrightarrow{k_{fast}} Y$       | $2 Y \xrightarrow{k_{fast}} 2Y + R'$    | $2S_g \xrightarrow{k_{slow}} 2S_g + g$ | $G' + R \xrightarrow{k_{fast}} G$     |
| $2 C \xrightarrow{k_{fast}} R$       | $2 G \xrightarrow{k_{fast}} 2G + G'$    | $2S_b \xrightarrow{k_{slow}} 2S_b + b$ | $B' + G \xrightarrow{k_{fast}} B$     |
| $b + R \xrightarrow{k_{slow}} G$     | $2 B \xrightarrow{k_{fast}} 2B + B'$    | $R' + r \xrightarrow{k_{fast}} R'$     | $R' + B \xrightarrow{k_{fast}} Y$     |
| $r + G \xrightarrow{k_{slow}} B$     | $2 X \xrightarrow{k_{fast}} 2X + R'$    | $G' + g \xrightarrow{k_{fast}} G'$     |                                       |
| $g + B \xrightarrow{k_{slow}} Y$     | $2 R' \xrightarrow{k_{fast}} \emptyset$ | $B' + b \xrightarrow{k_{fast}} B'$     |                                       |
|                                      | $2 G' \xrightarrow{k_{fast}} \emptyset$ |  |                                       |
|                                      | $2 B' \xrightarrow{k_{fast}} \emptyset$ |  |                                       |

Fig. 9. Set of molecular reactions for the asynchronous implementation of the moving-average filter [26].



Fig. 10. Block diagram of a general system developed in this paper.

implementations, we consider concentrations near 1 nM as the logical value '1' and near 0 nM as logical value '0'.

Molecular digital systems require molecular analog-to-digital conversion (ADC). This paper, presents a new molecular implementations of ADCs and DACs. Figure 10 illustrates a complete digital system.

We present molecular implementations of a  $k$ -bit analog to digital converter and a  $k$ -bit digital to analog converter. We also review the molecular implementation of basic digital logic gates. Using these gates, we demonstrate a 3-bit molecular binary adder including two ADCs required to sample and digitize the two input operands and a DAC to output an analog signal. A DNA implementation of the complete system is also demonstrated in Section V. It can be noted that all of the molecular reactions are rate-independent. In other words, no matter what the speed rates of the reactions are and how they may change during the computation, the steady-state concentrations compute the correct desired outputs.

#### A. Analog to Digital Converter (ADC)

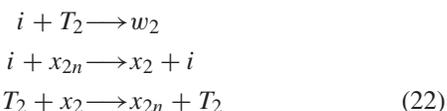
This subsection describes molecular implementation of analog to digital converter. A 3-bit example is considered. Let the input molecular type,  $i$ , have an analog concentration between 0 nM and 8 nM. The output is a 3-bit digital number  $x = x_2x_1x_0$ . Each bit is considered as logical '0' if its concentration is approximately 0 nM and logical '1' if its concentration is approximately 1 nM.

We start with the most significant bit,  $x_2$ . This bit should be set to 1 when  $i$  is larger than 4 nM and to zero when  $i$  is

TABLE I  
STABLE CONCENTRATION OF MOLECULES  $i$ ,  $x_2$ , AND  $w_2$  AFTER COMPLETION OF REACTIONS (22)

|           | $i$       | $w_2$ | $x_2$ |
|-----------|-----------|-------|-------|
| $i_0 < 4$ | 0         | $i_0$ | 0     |
| $i_0 > 4$ | $i_0 - 4$ | 4     | 1     |

less than 4 nM. Reactions (22) implement a one-bit comparator that determines  $x_2$ . The initial concentration of  $T_2$  represents the threshold for the comparator which is set to 4 nM.



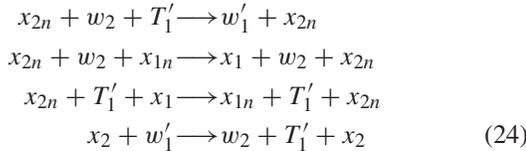
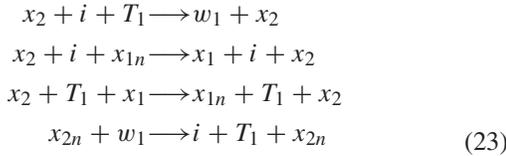
In the first reaction,  $i$  and  $T_2$  molecules combine and the one with larger initial concentration remains and the other one vanishes. The first reaction is independent of the second and third reactions because  $i$  and  $T_2$  remain unaltered in the second and third reactions. However, activation of the second or third reactions depends on the outcome of the first reaction. After completion of the first reaction only one of the second or third reactions is active. If  $i$  is larger than  $T_2$ , the third reaction stops firing while the second reaction transfers all molecules of  $x_{2n}$  to  $x_2$ . Alternately, if  $i$  is less than  $T_2$ , second reaction stops and third reaction transfers  $x_2$  to  $x_{2n}$  completely.  $x_2$  and  $x_{2n}$  are initialized to 0 nM and 1 nM, respectively. Note that in general for a  $k$ -bit ADC, each bit, i.e.,  $x_j$  where  $j = 0, 1, \dots, k - 1$ , is modeled by two molecular types, i.e.,  $x_j$  and  $x_{jn}$ , called the bit and its complement molecular types. All of the  $x_j$  species are initialized to 0 nM and  $x_{jn}$  species are initialized to 1 nM. Furthermore, for each  $j$ , the total concentration of  $x_j$  and  $x_{jn}$ , is constantly 1 nM, i.e., if the concentration of  $x_j$  is  $C$ , then the concentration of  $x_{jn}$  is  $(1 - C)$ , both in nM.

Table I shows the final concentrations for  $i$ ,  $x_2$  and  $w_2$  after completion of Reactions (22).  $i_0$  denotes the initial concentration of  $i$ . If  $i_0 > T_2$  then  $i$  can be used to compute the second bit of  $x$ , i.e.,  $x_1$ . If  $i_0 < T_2$  then  $w_2$  can be used to determine  $x_1$ .

TABLE II  
STABLE MOLECULAR CONCENTRATIONS AFTER COMPLETION  
OF REACTIONS (23) AND (24)

|               | $i$       | $w_2$     | $w_1$     | $w'_1$ | $x_2$ | $x_1$ |
|---------------|-----------|-----------|-----------|--------|-------|-------|
| $i_0 < 2$     | 0         | 0         | 0         | $i_0$  | 0     | 0     |
| $2 < i_0 < 4$ | 0         | $i_0 - 2$ | 0         | 2      | 0     | 1     |
| $4 < i_0 < 6$ | 0         | 4         | $i_0 - 4$ | 0      | 1     | 0     |
| $6 < i_0$     | $i_0 - 6$ | 4         | 2         | 0      | 1     | 1     |

Reactions (23) and (24) determine  $x_1$  for the above two cases. The initial concentrations for both threshold molecules,  $T_1$  and  $T'_1$ , are equal to 2 nM. Similar to Reactions (22), the first three reactions of (23) implement a one-bit comparator. However, here, the molecular concentration of  $i$  and  $T_1$  are compared to determine  $x_1$  when  $x_2$  is nonzero. This is equivalent to comparing initial  $i_0$  to 6 nM. Similarly the first three reactions of (24) compare  $w_2$  and  $T'_1$  to determine  $x_1$  when  $x_2$  is zero. This is equivalent to comparing initial  $i_0$  to 2 nM.

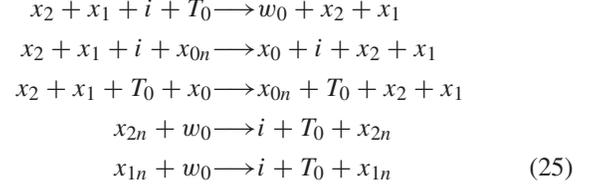


Before the concentration of  $x_2$  reaches its stable value, both  $x_2$  and  $x_{2n}$  may have nonzero concentrations and both sets of Reactions (23) and (24) can be fired. The fourth reactions of (23) and (24) are added to undo undesired reactions fired during the transient time. For example, when the final concentration of  $x_2$  is zero the fourth reaction of (23) transfers  $w_1$  back to  $i$  and  $T_1$  in order to undo the first reaction. The initial concentrations for  $x_1$  and  $x_{1n}$  are 0 nM and 1 nM, respectively. After  $x_1$  and  $x_{1n}$  are stabilized to their final concentrations, depending on the initial value of  $i$ , one of them has the concentration of 1 nM and the other 0 nM.

Except  $i$ , none of the molecular types participating in Reactions (22) is altered by Reactions (23) and (24). However, Reactions (23) and (24) need the final concentrations of  $x_2$  and  $x_{2n}$  from Reactions (22). Thus, the concentrations of molecules of Reactions (23) and (24) reach stable values after reactions in (22) are completed. For different values of  $i_0$ , Table II shows the final concentrations after Reactions (23) and (24) are completed.

Finally, in order to determine the least significant bit (LSB) of  $x$ , i.e.,  $x_0$ , depending on  $i_0$ 's value, the molecular types underlined in Table II are used. For each range of  $i_0$ , the concentration of its related molecular type is compared to 1 nM to determine  $x_0$ . For example when  $i_0 > 6$ , Reactions (25) are used to determine  $x_0$ . The initial concentration of threshold molecules  $T_0$  is 1 nM. Because both  $x_2$  and  $x_1$  are nonzero for  $i_0 > 6$ , the first three reactions compare  $i$  with 1 nM. It is equivalent to comparing  $i_0$  with 7 nM. That is to say, for  $i_0 > 6$ ,  $x_0=1$  nM if  $i_0 > 7$  nM and  $x_0=0$  nM if  $i_0 < 7$  nM. The last two reactions of

(25) are used to undo the undesirable combination of  $i$  and  $T_0$  during the transient time when any of  $x_2$  or  $x_1$  is zero.



Similarly for each range of  $i_0$  five reactions are used to determine  $x_0$ . Due to space limit, these three sets of reactions, each containing five reactions, are not listed here.

The number of bits or the resolution of ADC can be increased by adding the required comparisons and their related undo reactions. In general for  $k$ -bit ADC  $2^{k+1} + 2(k-1)$  molecular types are required while the number of required reactions is  $\sum_{j=1}^k (j+2)2^{j-1} = (k+1)2^k - 1$ . The precision (sensitivity) of ADC depends on its acceptable input range and the number of its output bits.

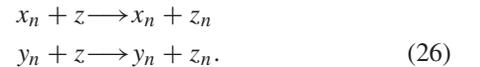
Figure 11 shows results for the mass-action kinetic model simulation of the proposed ADC for different values of  $i_0$ .

## B. Molecular Digital Logic Circuits

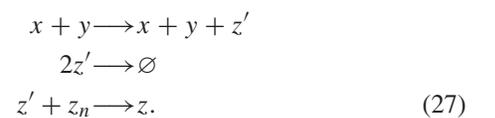
In this section we demonstrate how digital designs can be implemented by molecular reactions. We describe molecular implementations of simple logic AND/OR/XOR gates, a binary adder, and a square-root unit. The method we use here for implementing logical gates is similar to the method presented in [43]. However, in [43] three regulation bit operation reactions are needed for each bit, Whereas these reactions are not required in our complete system implementation due to the self-regulated bits output by the proposed ADC. Here, self-regulated means for each bit only the related molecular type,  $x_j$ , or its complement,  $x_{jn}$ , but not both, has stable non-zero concentration.

1) *Logic Gates*: We only consider two-input gates AND, OR, and XOR. Gates with more than two inputs can be easily implemented by cascading two-input gates. Let  $X$  and  $Y$  denote the inputs of a gate and  $Z$  the output.

*AND Gate*: We start with an AND gate. The output of a logical AND gate is '1' only if both inputs are '1'. It means that if either  $X=0$  or  $Y=0$  then the output  $Z$  should be zero. In other words, when concentration of  $x_n$  or  $y_n$ , i.e., complement molecular types of inputs, is nonzero molecules of  $z$  should be transferred to  $z_n$  in order to set  $Z=0$ . This can be implemented by Reactions (26).



When both  $x$  and  $y$  have stable nonzero concentrations, all molecules of  $z_n$  should be transferred to  $z$  in order to set  $Z=1$ . This can be implemented by Reactions (27).



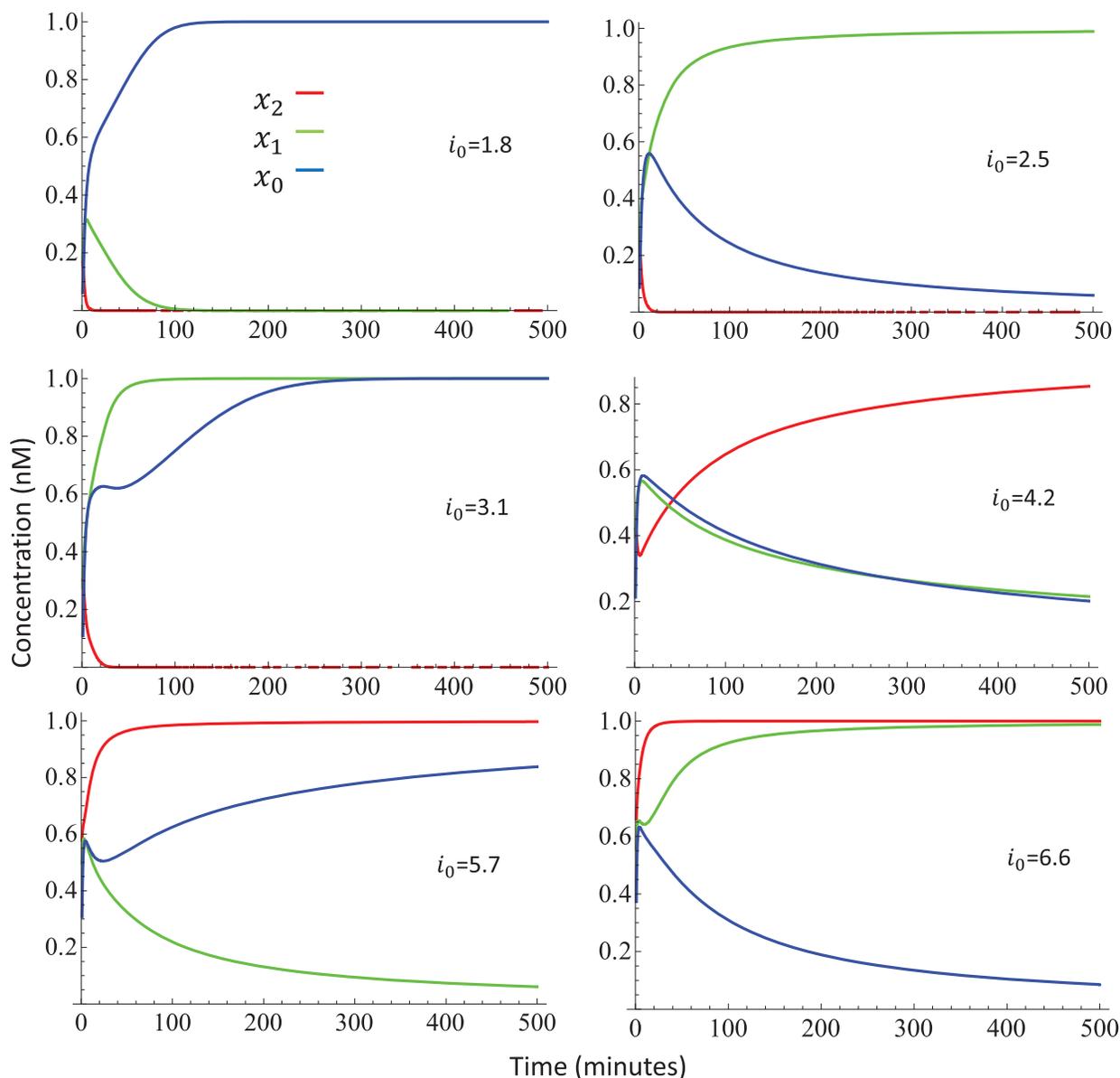


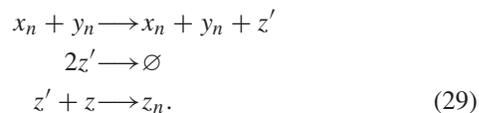
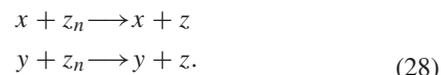
Fig. 11. Simulation results of 3-bit molecular ADC for different input concentrations.

In the first reaction of (27),  $x$  combines with  $y$  to generate  $z'$ , an indicator that  $Z$  should be set to '1'.  $z'$  is transferred to an external sink, denoted by  $\emptyset$ , in the second reaction. (This could be a waste type whose concentration we do not track.) When molecules of both  $x$  and  $y$  are present, these reactions maintain the concentration of  $z'$  at an equilibrium level. When either  $x$  or  $y$  is not present,  $z'$  gets cleared out. In the last reaction,  $z'$  transfers  $z_n$  to  $z$ .

One should note that the input concentrations don't change in logic computations. This enables the outputs of the ADC to be input to other logic gates if needed.

**OR Gate:** The output of an OR gate is '1' if any of its inputs is '1'. For molecular implementation it means that if either  $x$  or  $y$  has nonzero concentration then all molecules of  $z_n$  should be transferred to  $z$ . It is implemented by Reactions (28). In the other case, i.e., when both inputs have zero concentrations,

molecules of  $z$  should be transferred to  $z_n$  as implemented by Reactions (29).



**XOR Gate:** The output of a two-input XOR gate is '1' when inputs are complements of each other. In molecular implementation it means that when either  $x$  and  $y_n$  or  $x_n$  and  $y$  have nonzero concentrations, molecules of  $z_n$  should be transferred to  $z$  as implemented by Reactions (30). For the inputs with the same logical level the output should set to zero and

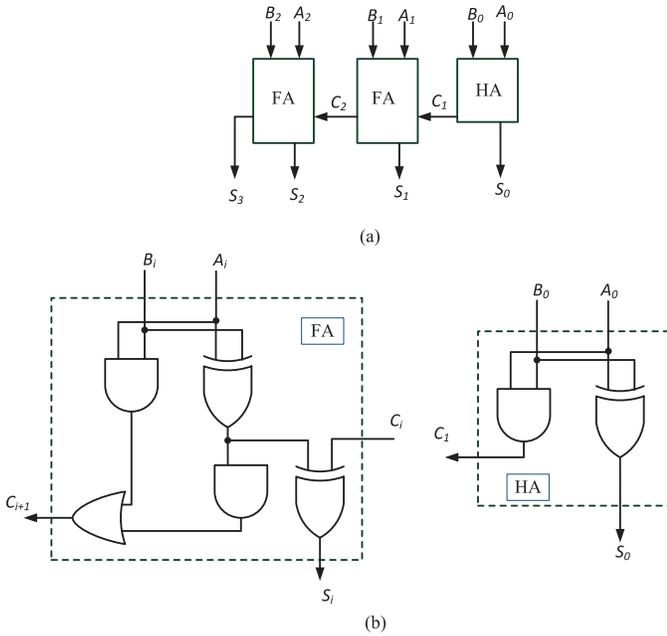
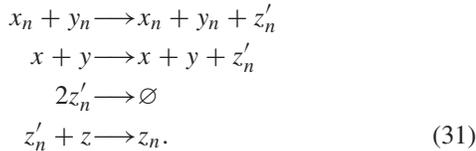
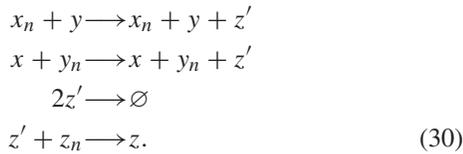


Fig. 12. Schematic of the 3-bit adder; (a) Block diagram, (b) Internal circuits for HA and FA blocks.

molecules of  $z$  should be transferred to  $z_n$ . This is implemented by Reactions (31).



NAND, NOR, and XNOR gates can be implemented by exchanging  $z$  and its complement in the transfer reactions,  $z_n$  in the opposite directions of those of the AND, OR, and XOR gates, respectively.

2) *Binary Adder*: By cascading AND, OR, and XOR gates we implement more complex digital systems such as a 3-bit adder. The adder consists of one half adder (HA) for the LSB and two full adders (FA) as shown in Figure 12a. Internal schematics of HA and FA are shown in Figure 12b. A general  $n$ -bit adder can be easily implemented by extending 3-bit adder using additional FAs for new bits. Cascaded gates for the adder are implemented by molecular reactions presented in Section IV.B. However, other molecular logic gates such as seesaw gates [42] can also be used. In order to verify the functionality of the 3-bit adder we implement the structure shown in Figure 13. Two analog concentrations,  $x$  and  $y$ , are converted to two 3-bit digital data using the proposed ADC. These two digital numbers are added using the 3-bit adder. The output,  $s = s_3s_2s_1s_0$ , is a 4-bit digital number representing the digital sum of  $x$  and  $y$ . Figure 14 shows the simulation results for different concentrations of inputs,  $x$  and  $y$ .

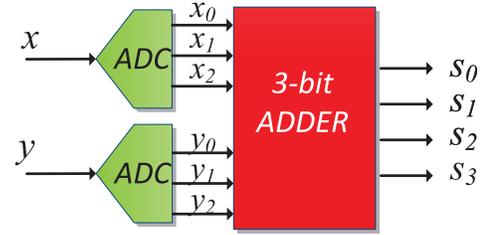
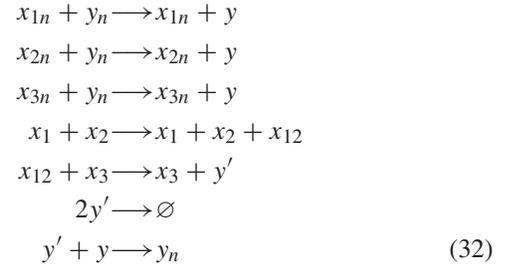


Fig. 13. Block diagram of the system for verifying molecular 3-bit adder.

3) *Square-Root Unit*: As another example of digital computing, we implement square-root of a 4-bit number. Figure 15 shows the schematic of its circuit. In Figure 15, the three-input NAND gate can be implemented by cascading a two-input AND gate with a two-input NAND gate. However, it is more efficient to implement three-input NAND by reactions (32). In these reaction  $x_1, x_2$ , and  $x_3$  are inputs and  $y$  is the output.

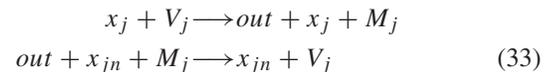


The strategy used for the direct implementation of three-input NAND in (32) is similar to that of two-input NAND.

Figure 16 shows the simulation results for the square root circuit implemented by molecular reactions.

### C. Digital to Analog Converter (DAC)

After performing computations in digital form, in order to convert the computed signal to its analog form, a DAC is required. Using recombinase-based logic and memory, a DAC has been implemented in [44]. For this DAC various digital combinations of the input inducers result in multiple levels of analog gene expression outputs on the basis of the varying strengths of the promoters used and the sum of their respective outputs. This section presents molecular implementations of a  $k$ -bit DAC with controlling the impact of each bit on the analog output concentration. Reactions (33) show a 1-bit template for implementing DAC.



where  $x_j$  and  $x_{jn}$ , respectively, represent the input bit and its complement molecular type.  $out$  is the analog output of DAC with initial concentration of zero. Molecular type  $V_j$  denotes the value of the input bit. In other words, it defines the amount of concentration that is added to the output if input bit,  $x_j$ , is nonzero. If  $x_j$  is the LSB then  $V_j$  is initialized to 1 nM and if it is the bit next to the LSB then  $V_j$  is initialized to 2 nM and so on.

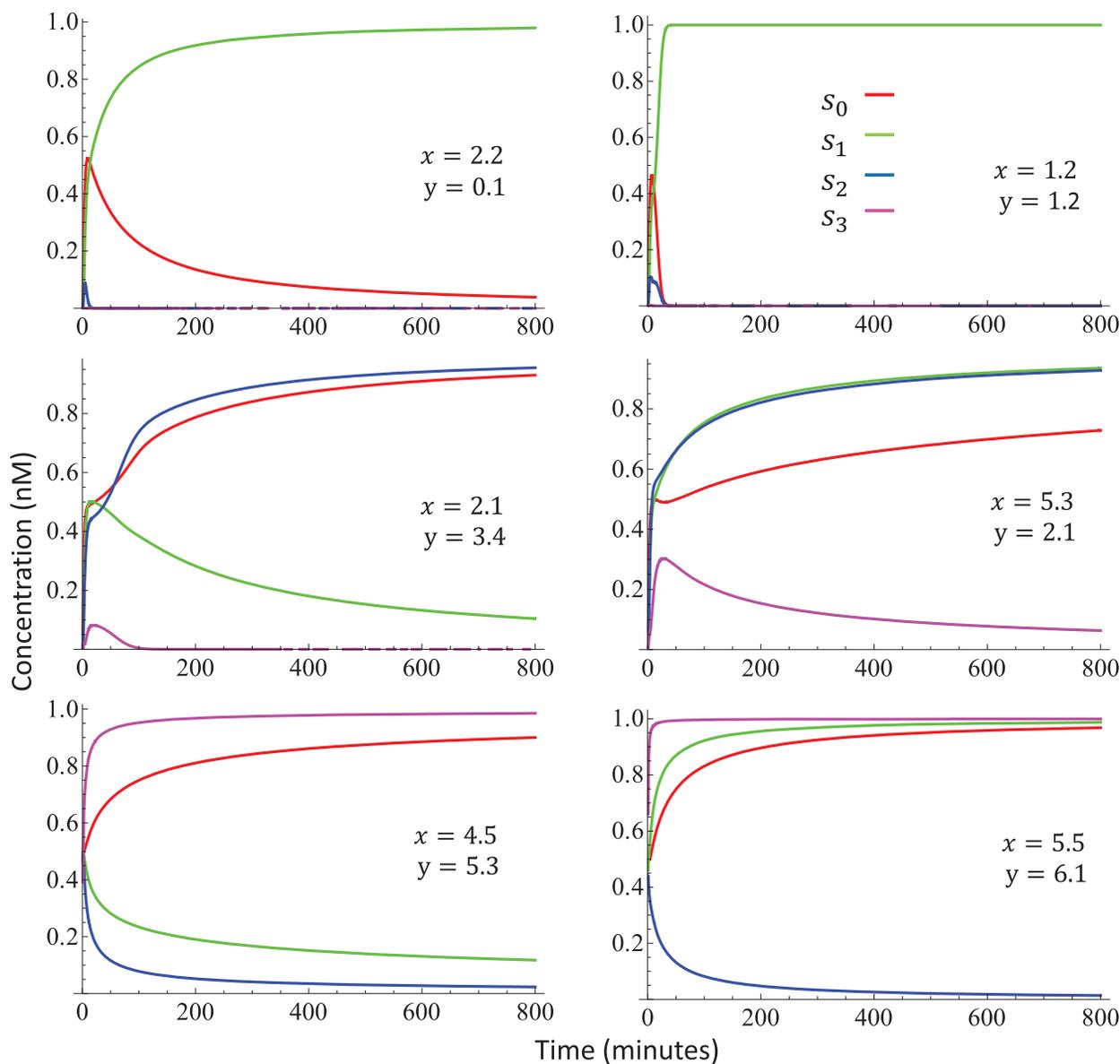


Fig. 14. Simulation results of the molecular implementation of the system shown in Figure 13.

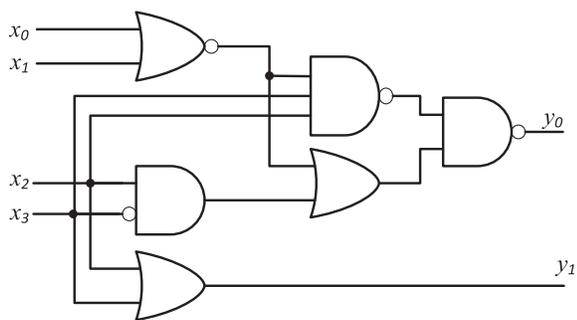


Fig. 15. Schematic for 4-bit Square-root unit.

Even when the stable value of  $x_j$  is zero, during the transient state  $x_j$  may have nonzero concentration. The second reaction of (33) prevents undesired output increase due to the nonzero

concentration of  $x_j$  in transient state.  $M_j$  controls the amount of deducted concentration from the output such that this amount is the same as the amount added to output undesirably during the transient state. In other words, without  $M_j$ , the second reaction continues transferring *out* molecules to  $V_j$  during the steady-state. However, this degrades the effects of other bits on the DAC's output, since the molecular type *out* is common for all bits. The initial concentration for  $M_j$  is zero.

The 1-bit template presented here can be easily extended to a  $k$ -bit DAC; for each additional bit, one instance of Reactions (33) is added. Therefore, to construct a  $k$ -bit DAC, a chemical reaction network including  $k$  copies of the 1-bit template are used with proper initial values of  $V_j$ . As an example, Reactions (34) illustrate a 4-bit DAC using the proposed template. The initial concentrations of  $V_0$ ,  $V_1$ ,  $V_2$ , and  $V_3$  are 1, 2, 4, and 8 nM, respectively.

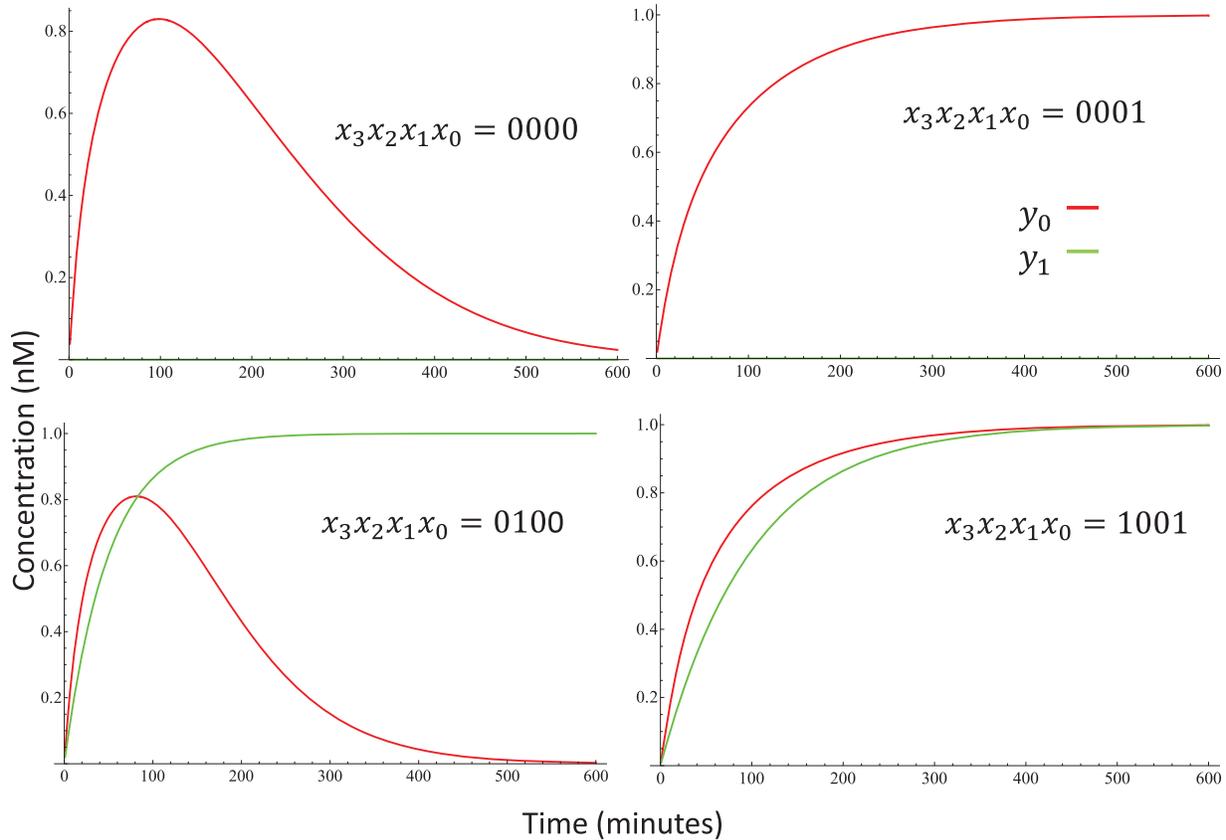
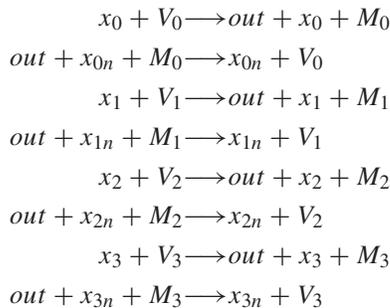


Fig. 16. Kinetics simulations that compute the Square-root of 0, 1, 4, and 9 using the molecular implementation of unit shown in Figure 15.



#### D. A Complete Molecular Digital System

We now illustrate molecular implementation of a digital adder where concentrations of two analog molecules  $x$  and  $y$  are converted to 3-bit digital, then added using a binary adder, and the 4-bit output is converted to an analog value  $s$ . Two molecular ADCs, a molecular digital adder, and a molecular DAC are used to construct a complete system as shown in Figure 17. The functionality of the complete molecular system is verified. Figure 18 shows the simulation results for the complete system illustrated in Figure 17 for different input concentrations.

#### V. DNA IMPLEMENTATION

This section describes mapping of the molecular reactions to DNA. We illustrate mapping the complete digital adder of Section IV.D including ADC, adder and DAC to DNA strand displacement reactions.

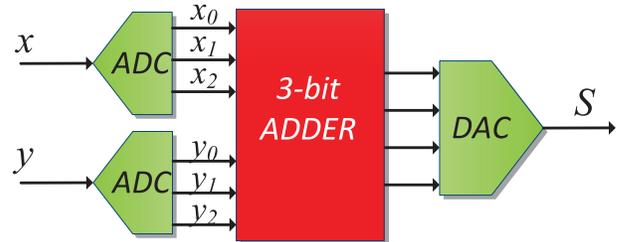


Fig. 17. Block diagram of a simple prototype developed and verified in this paper.

Considering each strand (single or double) of DNA as a molecule, it is possible to implement CRNs with DNA strand-displacement mechanism. For example Figure 19 shows DNA strand-displacement primitive for implementing  $A + B \xrightarrow{f} C + D$ .

Toehold 1 of strand A starts binding to its complement toehold 1\* of B. Then branch migration happens and domain 2 of A displaces domain 2 of strand 2–3. Finally, toehold 3 and 3\* are separated and two new strands (molecules),  $C$  and  $D$ , are produced.

A general method of mapping CRNs to DNA strand-displacement reactions has been presented in [23] by Soloveichik, et. al. In their method based on the number of reactants a chemical reaction is converted to a series of DNA strand-displacement reactions similar to Figure 19. Similarly, for our design we generate the corresponding DNA reactions

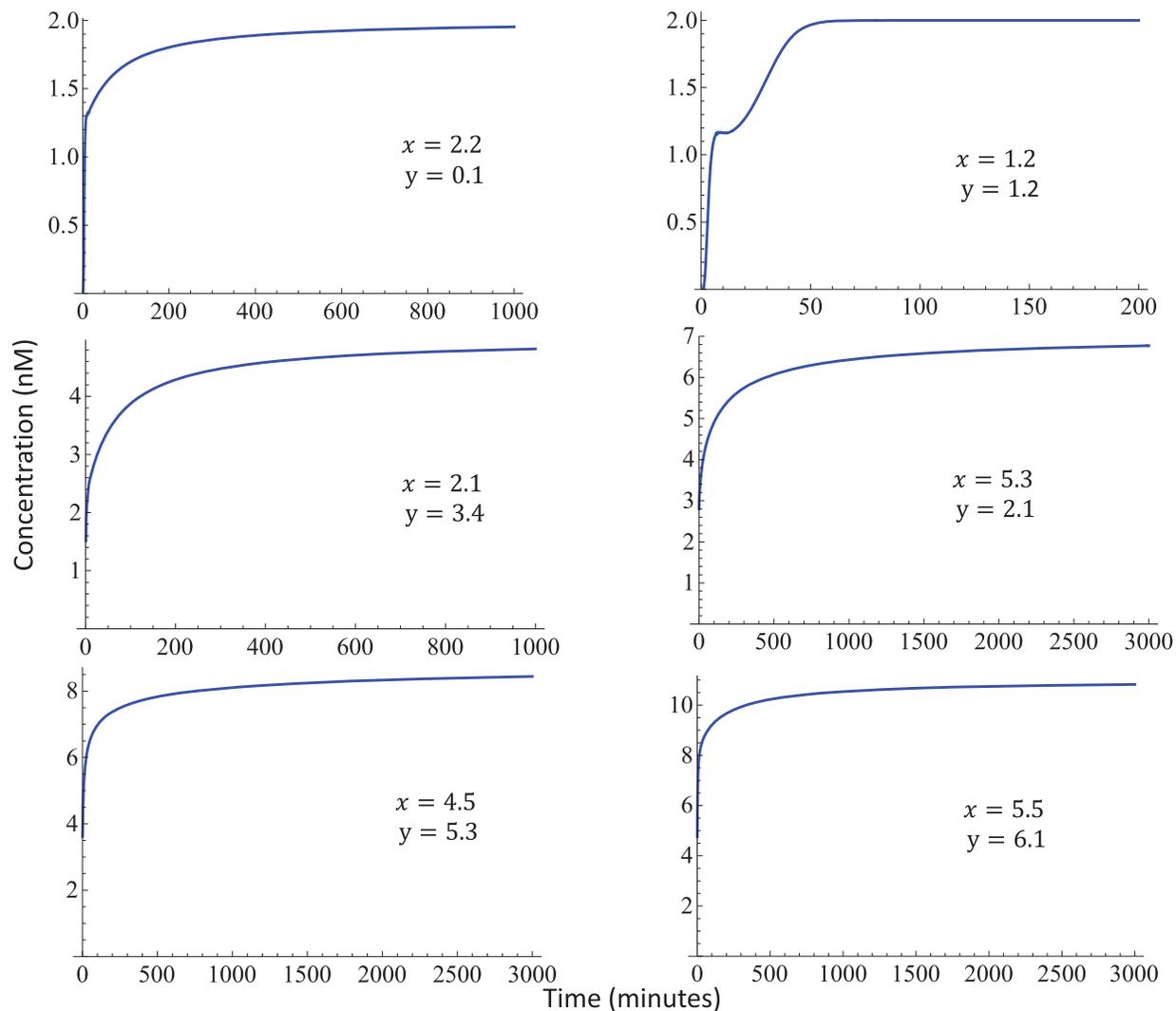


Fig. 18. Simulation results for the system shown in Figure 17.

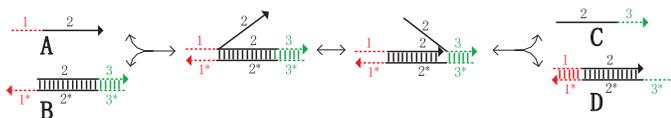


Fig. 19. Implementation of  $A + B \frac{f}{r} C + D$  using DNA strand-displacement mechanism.

and simulate the system using the kinetic differential equations to characterize the behavior of the system.

The initial concentrations of auxiliary complexes is set to  $C_{max} = 10^{-5}M$ , and the maximum strand displacement rate constant is  $q_{max} = 10^6 M^{-1} s^{-1}$ . For all of the reactions the rate constant is considered as  $10^5 M^{-1} S^{-1}$ . Figure 20 shows the ODE simulation results for the DNA implementation of the complete system illustrated in Figure 17 for different inputs.

## VI. COMPARISON BETWEEN MOLECULAR AND ELECTRONIC CIRCUITS

Just as electronic systems implement computation in terms of voltage (*energy per unit charge*), molecular systems compute

in terms of chemical concentrations (*molecules per unit volume*). One of the great successes of electronic circuit design has been in abstracting and scaling the design problem. The physical behavior of transistors is understood in terms of differential equations – say, with models found in tools such as SPICE [45]. However, the design of circuits occurs at more abstract levels – in terms of switches, gates, and modules. Research in molecular computation could benefit from this hierarchical approach.

We point out several fundamental differences in characteristics of molecular and electronic circuits. These are summarized in Table III. Fanout operations in electronic circuits are free while these are expensive in molecular implementations. Addition operations are free in molecular systems, but are expensive in electronic circuits. The critical path of an electronic circuit is typically bounded by computation time; the delay elements enable reduction of critical path and faster computation. However, molecular implementations of delay elements require inherently slow transfer reactions; the speed of molecular systems is bounded by the communication bound as opposed to the computation bound. The computations in molecular systems are inherently highly parallel unlike in electronic

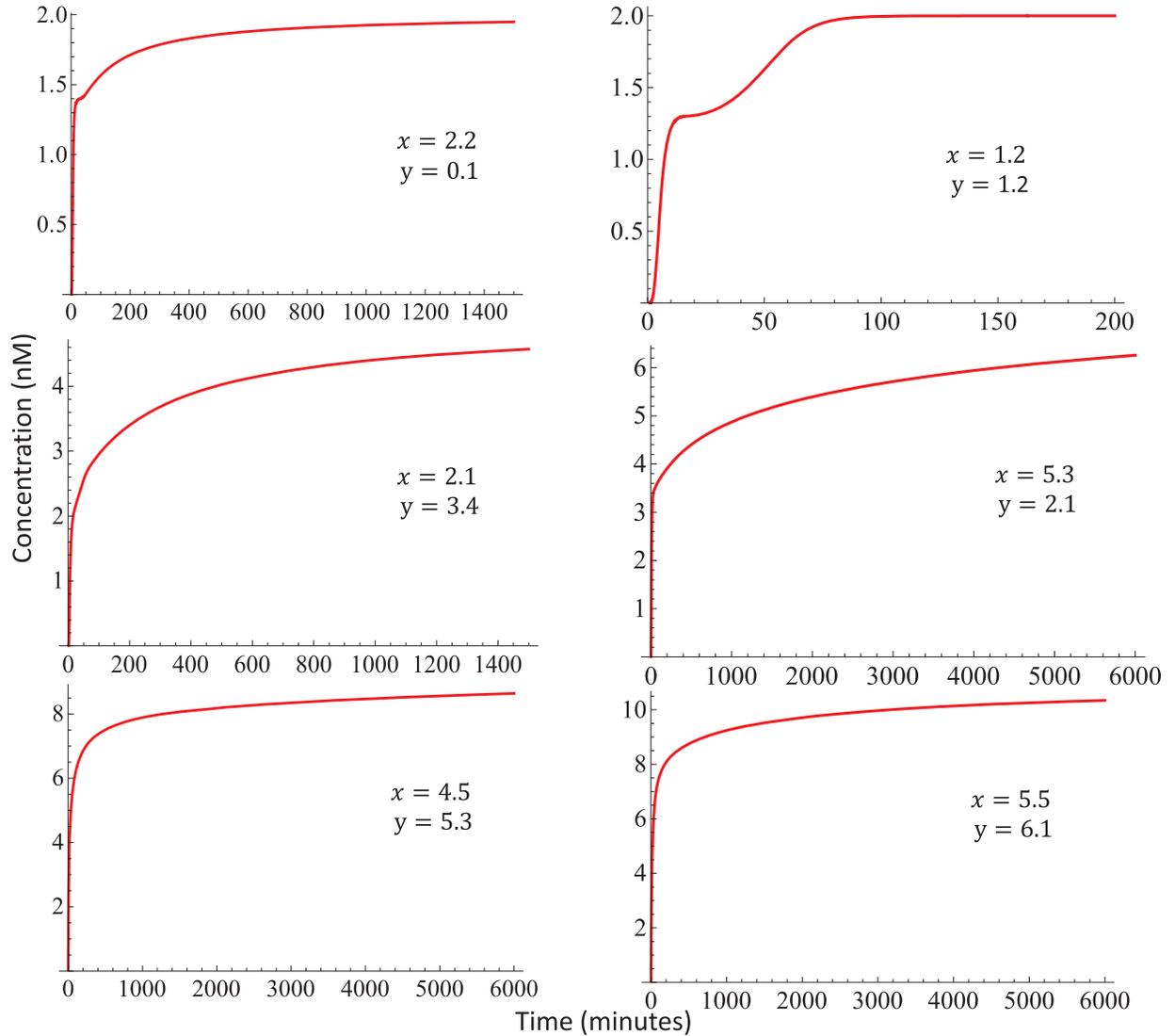


Fig. 20. Simulation results for the DNA implementation of the system shown in Figure 17.

TABLE III  
COMPARISON BETWEEN MOLECULAR (DNA) AND ELECTRONICS (SILICON) COMPUTING SYSTEMS

| Aspect                | DNA-based               |                |           | Silicon-based       |
|-----------------------|-------------------------|----------------|-----------|---------------------|
|                       | analog                  | Discrete-time  | digital   |                     |
| Addition              | free                    | free           | expensive | expensive           |
| Multiplication        | less expensive          | less expensive | expensive | expensive           |
| Fanout                | expensive               | expensive      | free      | free                |
| Delay/Signal Transfer | very expensive          |                |           | almost free         |
| Bound on Performance  | communication bounded   |                |           | computation bounded |
| Speed                 | ultra-slow              |                |           | very fast           |
| Area                  | ~ nm                    |                |           | ~ nm                |
| Parallelism           | highly-parallel         |                |           | less-parallel       |
| Level of Integration  | no integration          |                |           | highly-integrated   |
| Application Area      | <i>in vivo/in vitro</i> |                |           | industrial/consumer |

systems where parallelism requires significant increase in hardware resources. Finally the electronic circuits are highly integrated while the molecular systems are not suitable for highly integrated implementations. DNA and electronic systems also differ fundamentally with respect to storage properties. DNA systems can hold their concentrations indefinitely while the charge or stored value in an electronic system can leak and needs to be refreshed periodically.

## VII. DISCUSSION AND CONCLUDING REMARKS

This paper presented methodologies for implementing continuous-time, discrete-time and digital processing with molecular reactions. Several examples are presented to illustrate the approaches presented in the paper.

Although pertaining to biology, the contributions of this paper are neither experimental nor empirical; rather they are

constructive and conceptual. We design robust digital logic with molecular reactions. For the molecular digital systems, our designs do not depend on specific reaction rates; the computation is accurate for a wide range of rates. This is crucial for mapping the design to DNA substrates.

Intense efforts by the synthetic biology community have been devoted to the implementation of computation and logical functions with genetic regulatory elements [46]–[50]. For example design of robust logical circuits using chemically wired cells have been presented in [41] for single logic gates. Also genetic circuits consisting of multi-layer logical gates have been implemented in single cell in [51]. Yet, progress seems to have stalled at the complexity level of circuits with perhaps 7–15 components. In fact, *in vivo* engineering of such circuits is full of experimental difficulties. In contrast, *in vitro* molecular computation with DNA strand displacement is following a Moore's Law-like trajectory in the scaling of its complexity. Thus, due to their complexity, systems presented in this paper are more likely to be physically realizable *in vitro* than *in vivo*.

The impetus of the field is not computation *per se*; chemical systems will never be useful for number crunching. Rather the field aims for the design of custom, embedded biological "sensors" and "controllers" – viruses and bacteria that are engineered to perform useful tasks *in situ*, such as cancer detection and drug therapy. Exciting work in this vein includes [52]–[55].

One should notice that there is quantization error in the ADC component. This is similar to the quantization error for other types of ADC usually used in digital signal processing systems [56]. The error can decrease the accuracy of system. The quantization error can be reduced by increasing the ADC resolution and, consequently, increasing the number of bits of ADC and DAC components.

In future work, we will perform more detailed studies of the characteristics of biomolecular continuous-time, discrete-time and digital processing systems including noise analysis. For instance, we will study how the resolution correlates with changing molecular concentrations and how robust the designs are to parametric variations. Also, we will develop faster implementations. The main bottleneck in current implementations has been speed. Unlike in electronic systems, where the speed is limited by changes in electric charge, the speed in molecular systems is limited by changes in molecular concentrations, which are inherently slow.

We will investigate new scheduling approaches where multiple computations are mapped into different phases of transfer. Reducing currently achievable sample periods from hundreds of hours to a few hours, or even a few minutes, will enable experimental demonstration of some example signal processing functions using DNA.

## REFERENCES

- [1] R. H. Carlson, *Biology is Technology: The Promise, Peril, and New Business of Engineering Life*. Cambridge, MA, USA: Harvard Univ. Press, 2010.
- [2] R. Carlson, "The pace and proliferation of biological technologies," *Biosecur. Bioterrorism: Biodefense Strategy Pract. Sci.*, vol. 1, pp. 203–214, Sep. 2003.
- [3] M. Conrad, "Molecular computing," *Adv. Comput.*, vol. 31, pp. 235–324, 1990.
- [4] L. Adleman, "Molecular computation of solutions to combinatorial problems," *Science*, vol. 266, no. 11, pp. 1021–1024, 1994.
- [5] J. Chen and D. H. Wood, "Computation with biomolecules," *Proc. Nat. Acad. Sci.*, vol. 97, no. 4, pp. 1328–1330, 2000.
- [6] Y. Benenson, "Biocomputers: From test tubes to live cells," *Mol. Biosyst.*, vol. 5, pp. 675–685, 2009.
- [7] N. Nandagopal and M. B. Elowitz, "Synthetic biology: Integrated gene circuits," *Science*, vol. 333, pp. 1244–1248, 2011.
- [8] A. J. Hockenberry and M. C. Jewett, "Synthetic *in vitro* circuits," *Science*, vol. 16, pp. 253–259, 2012.
- [9] A. J. Genot, T. Fujii, and Y. Rondelez, "Scaling down dna circuits with competitive neural networks," *J. R. Soc. Interface*, vol. 10, no. 85, 20130212, 2013.
- [10] S. A. Salehi, M. D. Riedel, and K. K. Parhi, "Markov chain computations using molecular reactions," in *Proc. IEEE Int. Conf. Digit. Signal Process. (DSP)*, 2015, pp. 689–693.
- [11] L. Qian, D. Soloveichik, and E. Winfree, "Efficient turing-universal computation with DNA polymers," in *Proc. Int. Conf. DNA Comput. Mol. Program.*. Springer Berlin Heidelberg, 2010, pp. 123–140.
- [12] M. N. Stojanovic and D. Stefanovic, "A deoxyribozyme-based molecular automaton," *Nat. Biotechnol.*, vol. 21, pp. 1069–1074, 2003.
- [13] G. Seelig, D. Soloveichik, D. Y. Zhang, and E. Winfree, "Enzyme-free nucleic acid logic circuits," *Science*, vol. 314, pp. 1585–1588, 2006.
- [14] T. Ran, S. Kaplan, and E. Shapiro, "Molecular implementation of simple logic programs," *Nat. Nanotechnol.*, vol. 4, pp. 642–648, 2009.
- [15] Y. Benenson, "Biomolecular computing systems: Principles, progress and potential," *Nat. Rev. Genet.*, vol. 13, pp. 455–468, 2012.
- [16] B. Alberts, A. Johnson, J. Lewis, M. Raff, K. Roberts, and P. Walter, *Molecular Biology of the Cell*, 5th ed. New York, NY, USA: Garland, 2007.
- [17] P. Erdi and J. Toth, *Mathematical Models of Chemical Reactions: Theory and Applications of Deterministic and Stochastic Models*. Manchester Univ. Press, Manchester, U.K., 1989.
- [18] F. Horn and R. Jackson, "General mass action kinetics," in *Archive for Rational Mechanics and Analysis*. Springer, 1972, pp. 81–116, vol. 47.
- [19] W. W. Chen, M. Niepel, and P. K. Sorger, "Classic and contemporary approaches to modeling biochemical reactions," *PMC Genes Dev.*, vol. 24, no. 17, pp. 1861–1875, 2010.
- [20] K. Rinaudo, L. Bleris, R. Maddamsetti, S. Subramanian, R. Weiss, and Y. Benenson, "A universal RNAI-based logic evaluator that operates in mammalian cells," *Nat. Biotechnol.*, vol. 25, no. 7, pp. 795–801, 2007.
- [21] Z. Xie, L. Wroblewska, L. Prochazka, R. Weiss, and Y. Benenson, "Multi-input RNAI-based logic circuit for identification of specific cancer cells," *Science*, vol. 333, no. 6047, pp. 1307–1311, 2011.
- [22] A. Reynolds, D. Leake, Q. Boese, S. Scaringe, W. S. Marshall, and A. Khvorova, "Rational siRNA design for RNA interference," *Nat. Biotechnol.*, vol. 22, no. 3, pp. 326–330, 2004.
- [23] D. Soloveichik, G. Seelig, and E. Winfree, "DNA as a universal substrate for chemical kinetics," *Proc. Nat. Acad. Sci.*, vol. 107, no. 12, pp. 5393–5398, 2010.
- [24] M. Samoilov, A. Arkin, and J. Ross, "Signal processing by simple chemical systems," *J. Phys. Chem. A*, vol. 106, no. 43, pp. 10 205–10 221, 2002.
- [25] K. Oishi and E. Klavins, "Biomolecular implementation of linear *i/o* systems," *IET Syst. Biol.*, vol. 5, pp. 252–260, 2011.
- [26] H. Jiang, S. Salehi, M. Riedel, and K. Parhi, "Discrete-time signal processing with DNA," *ACS Synth. Biol.*, vol. 2, no. 5, pp. 245–254, 2013.
- [27] H. Jiang, M. Riedel, and K. Parhi, "Synchronous sequential computation with molecular reactions," in *Proc. ACM/IEEE Des. Autom. Conf.*, 2011, pp. 836–841.
- [28] H. Jiang, M. Riedel, and K. Parhi, "Digital signal processing with molecular reactions," *IEEE Des. Test Mag. (Special Issue Bio-Des. Autom. Synth. Biol.)*, vol. 29, no. 3, pp. 21–31, 2012.
- [29] H. Jiang, M. Riedel, and K. Parhi, "Asynchronous computations with molecular reactions," in *Proc. Asilomar Conf. Signals. Syst. Comput.*, 2011, pp. 493–497.
- [30] S. A. Salehi, M. D. Riedel, and K. K. Parhi, "Asynchronous discrete-time signal processing with molecular reactions," in *Proc. Asilomar Conf. Signals Syst. Comput.*, 2014, pp. 1767–1772.
- [31] H. M. Sauro and K. Kim, "Synthetic biology: It's an analog world," *Nature*, vol. 497, pp. 572–573, 2013.
- [32] R. Sarpeshkar, "Analog synthetic biology," *Philos. Transact. A. Math. Phys. Eng. Sci.*, vol. 372, no. 20130110, 2014.
- [33] R. Sarpeshkar, "Analog versus digital: Extrapolating from electronics to neurobiology," *Neural Comput.*, vol. 10, pp. 1601–1638, 1998.
- [34] R. Daniel, J. R. Rubens, R. Sarpeshkar, and T. K. Lu, "Synthetic analog computation in living cells," *Nature*, vol. 497, pp. 619–623, 2013.

- [35] S. Hayat and T. Hinze, "Toward integration of in vivo molecular computing devices: Successes and challenges," *HFSP J.*, vol. 5, no. 2, pp. 239–243, 2008.
- [36] T. Kailath, *Linear Systems*. Englewood Cliffs, NJ, USA: Prentice Hall, 1980.
- [37] V. V. Kulkarni, H. Jiang, T. Chanyaswad, and M. Riedel, "A biomolecular implementation of systems described by linear and nonlinear ODE's," in *Proc. Int. Workshop Biodesign Autom.*, 2013, pp. 40–41.
- [38] K. K. Parhi, *VLSI Digital Signal Processing Systems*. Hoboken, NJ, USA: Wiley, 1999.
- [39] I. R. Epstein and J. A. Pojman, *An Introduction to Nonlinear Chemical Dynamics: Oscillations, Waves, Patterns, and Chaos*. London, U.K.: Oxford Univ. Press, 1998.
- [40] P. D. Kepper, I. R. Epstein, and K. Kustin, "A systematically designed homogeneous oscillating reaction: The arsenite-iodate-chlorite system," *J. Amer. Chem. Soc.*, vol. 103, no. 8, pp. 2133–2134, 2008.
- [41] A. Tamsir, J. J. Tabor, and C. A. Voigt, "Robust multicellular computing using genetically encoded nor gates and chemical 'wires'," *Nature*, vol. 469, pp. 212–215, 2011.
- [42] L. Qian and E. Winfree, "Scaling up digital circuit computation with DNA strand displacement cascades," *Science*, vol. 332, pp. 1196–1201, 2011.
- [43] H. Jiang, M. D. Riedel, and K. K. Parhi, "Digital logic with molecular reactions," in *Proc. IEEE Int. Conf. Comput. Aided Des. (ICCAD)*, 2013, pp. 721–727.
- [44] P. Siuti, J. Yazbek, and T. Lu, "Synthetic circuits integrating logic and memory in living cells," *Nat. Biotechnol.*, vol. 31, pp. 448–452, 2013.
- [45] L. Nagel and D. Pederson, "Simulation program with integrated circuit emphasis," in *Proc. Midwest Symp. Circuit Theory*, 1973.
- [46] R. Weiss *et al.*, "Genetic circuit building blocks for cellular computation, communications, and signal processing," *Nat. Comput.*, vol. 2, pp. 47–84, 2003.
- [47] T. S. Gardner, C. R. Cantor, and J. J. Collins, "Construction of a genetic toggle switch in *Escherichia coli*," *Nature*, vol. 403, no. 2000, pp. 339–342, 2000.
- [48] Y. Benenson, B. Gil, U. Ben-Dor, R. Adar, and E. Shapiro, "An autonomous molecular computer for logical control of gene expression," *Nature*, vol. 429, no. 6990, pp. 423–429, 2004.
- [49] D. Endy, "Foundations for engineering biology," *Nature*, vol. 438, pp. 449–453, 2005.
- [50] K. Ramalingam, J. R. Tomshine, J. A. Maynard, and Y. N. Kaznessis, "Forward engineering of synthetic bio-logical and gates," *Biochem. Eng. J.*, vol. 47, nos. 1–3, pp. 38–47, 2009.
- [51] T. Moon, C. Lou, A. Tamsi, B. Stanton, and C. A. Voigt, "Genetic programs constructed from layered logic gates in single cells," *Nature*, vol. 491, pp. 249–253, 2012.
- [52] J. C. Anderson, E. J. Clarke, A. P. Arkin, and C. A. Voigt, "Environmentally controlled invasion of cancer cells by engineered bacteria," *J. Mol. Biol.*, vol. 355, no. 4, pp. 619–627, 2006.
- [53] D. Ro *et al.*, "Production of the antimalarial drug precursor artemisinic acid in engineered yeast," *Nature*, vol. 440, pp. 940–943, 2006.
- [54] S. Venkataramana, R. M. Dirks, C. T. Ueda, and N. A. Pierce, "Selective cell death mediated by small conditional RNAs," *Proc. Nat. Acad. Sci.*, vol. 107, no. 39, pp. 16 777–16 782, 2010.
- [55] D. M. Widmaier *et al.*, "Engineering the *Salmonella* type III secretion system to export spider silk monomers," *Mol. Syst. Biol.*, vol. 5, no. 309, pp. 1–9, 2009.
- [56] A. V. Oppenheim and R. W. Schaffer, *Discrete-Time Signal Processing*, 3rd ed. Englewood Cliffs, NJ, USA: Prentice Hall, 2009.



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