

# Accurate Ratio Computation using Abstract Chemical Reaction Networks

Iuliia Zarubiieva, Joyun Tseng, and Vishwesh Kulkarni

**Abstract**—We present new results on how abstract chemical reactions, viz., catalysis, annihilation and degradation, can be used to implement circuits that accurately compute ratio of two input signals. The input signals can be either constant-valued scalars or time-varying scalars or polynomials. We also characterise the robustness of our circuits to parameter variations, as would be encountered in wet-lab implementations.

**Index Terms**—abstract chemical reaction network, DNA strand displacement, stability, robustness, Circle criterion

## I. INTRODUCTION

An objective of synthetic biology is to design biomolecular circuits for *in situ* monitoring and control. Recently, nucleic acid reactions have been proposed as a potential solution for these purposes [1], [2], [3], [4]. A key advantage of nucleic acid reactions consists in the ease and precision with which these can be implemented, as their design relies essentially on the well-known Watson-Crick base-pairing mechanism (i.e. adenine-thymine and guanine-cytosine pairing), which enables precise programming and timing of molecular interactions simply by the choice of relevant sequences. This approach has allowed the implementation of a number of complex circuits based on DNA strand displacement [5], DNA enzyme [6] and RNA enzyme [7] reactions, and has been used for the modelling and implementation of various nucleic-acids-based circuits such as feedback controllers [8] and predator-prey systems [9]. Recently, it has been shown that any *chemical reaction network* can be closely approximated by a set of suitably designed DNA strand displacement reactions [10]. This logic can be extended to approximate a set of linear ordinary differential equations (ODEs) by a set of idealised *abstract chemical reaction networks* (ACRNs) which can then be approximated by a set of suitably designed DNA strand displacement reactions [4]. In [11], it was shown that ACRNs for catalysis using two reactants can be used to realise a set of nonlinear operators and, in particular, was used to implement a system  $\mathcal{S}_D$  that computes the ratio of two entities - these entities can be two constant-valued scalars or two time-varying signals or two constant-valued polynomials or two time-varying polynomials. However,  $\mathcal{S}_D$  suffers from steady-state errors. In this paper, we present two circuits, viz.,  $\mathcal{S}_{C1}$  (see Fig. 1) and  $\mathcal{S}_{C2}$  (see Fig. 2) to overcome this limitation. We also characterise the input-output stability of  $\mathcal{S}_{C1}$ . Our circuits can be implemented in the wet-lab using DNA strand displacement on the lines

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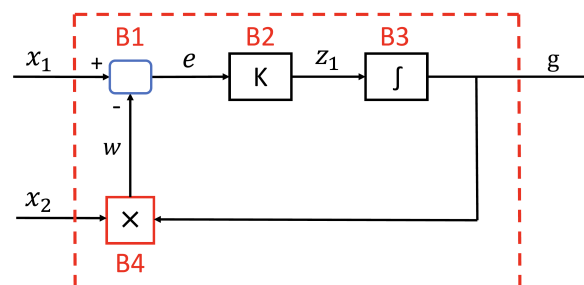
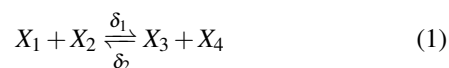


Fig. 1:  $\mathcal{S}_{C1}$ : Block diagram for our first circuit for accurately computing the ratio  $g = x_1/x_2$ . This system takes 2 inputs, viz.,  $x_1$  and  $x_2$ , and produces the output  $g$ . The inputs need not be constant and are allowed to be time-varying. We refer to the blocks B1 and B2 taken together as B1+B2.

described in [4]. In practice, DNA/RNA deactivate rapidly with time, typically with a half-life of 15 minutes, leading to appreciable time delays in the feedback loop which render the feedback system unstable. Hence, we also characterise the tolerance of  $\mathcal{S}_{C1}$  to the time-delays.

## II. NOTATION AND BACKGROUND RESULTS

To ensure consistency, the notation used in [12] and [4] is used throughout in this paper. For example, a bidirectional (i.e., a reversible bimolecular chemical reaction) is represented as



where  $X_i$  are chemical species with  $X_1$  and  $X_2$  being the reactants and  $X_3$  and  $X_4$  being the products. Here,  $\delta_1$  and  $\delta_2$  denote the forward and backward reaction rates, respectively. A unimolecular reaction features only one reactant whereas a multimolecular reaction features two or more reactants. Degradation of a chemical species  $X$  at rate  $K$  (or conversion of  $X$  into an inert form at a rate  $K$ ) is denoted by  $X \xrightarrow{K} \emptyset$ .

### A. Representing signals using differences of concentrations

Whereas signals in systems theory can take both positive and negative values, biomolecular concentrations (with Molar (M) as unit) can only take non-negative values. Thus, following the same approach suggested in [12] and [4], we represent a signal,  $x$  as the difference in concentration of two chemical species,  $x^+$  and  $x^-$ . Here,  $x^+$  and  $x^-$  are respectively the positive and negative components of  $x$  such that  $x = x^+ - x^-$ . The consequence of adopting this scheme is that there is no unique representation for a particular signal. As an example,  $x = 20$  M can be represented by both  $x^+ = 50$  M and  $x^- = 30$  M or equivalently,  $x^+ = 20$  M and  $x^- = 0$  M. In practice,

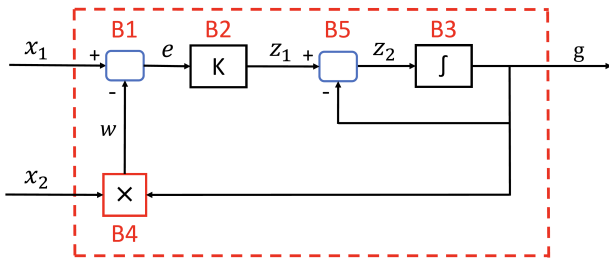


Fig. 2:  $\mathcal{S}_C$ : Block diagram for our second circuit for accurately computing the ratio  $g = x_1/x_2$ . Here, the integrator of  $\mathcal{S}_C$  is replaced by a low-pass filter. This system takes 2 inputs, viz.,  $x_1$  and  $x_2$ , and produces the output  $g$ . The inputs need not be constant and are allowed to be time-varying.

$x^+$  and  $x^-$  can be realised as single strand DNA molecules, as illustrated in [4] where these complementary positive and negative components would annihilate each other at reaction rate  $\eta$  (i.e.  $x^+ + x^- \xrightarrow{\eta} \emptyset$ ). A key advantage of using this scheme is that it allows the realisation of the ‘subtraction’ operation, as discussed further below.

### B. Realising elementary linear system theoretic operators

In [12], results on how to represent linear system theoretic operations such as gain, summation and integration using idealised abstract chemical reactions are presented and it is shown that only three types of elementary chemical reactions, namely, catalysis, annihilation and degradation are needed for such representations. In [4], this set of elementary chemical reactions is further reduced to only two. We here summarise their main results and refer the interested reader to [12] and [4] for details.

Throughout the rest of the paper, equations with superscript  $\pm$  and  $\mp$  are used as shorthand notations that represent the ‘+’ and ‘-’ individual reactions — for example,  $x_i^\pm \xrightarrow{K} x_i^\pm + x_o^\pm$  should be understood as the set of two reactions:  $x_i^+ \xrightarrow{K} x_i^+ + x_o^+$  and  $x_i^- \xrightarrow{K} x_i^- + x_o^-$ . Likewise, the notation  $x_i^\pm \xrightarrow{K} x_i^\pm + x_o^\mp$  is used to represent the set of two reactions:  $x_i^+ \xrightarrow{K} x_i^+ + x_o^-$  and  $x_i^- \xrightarrow{K} x_i^- + x_o^+$ . For brevity and following [12], we will represent such a set of reactions compactly as  $x_i^\pm \xrightarrow{K} x_i^\pm + x_o^\pm$  and  $x_i^\pm \xrightarrow{K} x_i^\pm + x_o^\mp$ .

As noted in [12], one limitation of representing signals as the difference of concentrations is that the requirement of having the same reaction rate,  $K$ , for both positive and negative components may not be easy to implement experimentally. However, as shown in [12], this requirement can be relaxed if the annihilation rate,  $\eta$  in the annihilation reaction,  $x_o^+ + x_o^- \xrightarrow{\eta} \emptyset$  is chosen to be sufficiently large. Hence, we assume this condition of  $\eta \gg K$  throughout the rest of this paper.

#### Lemma 1. [Scalar gain $K$ ]

Let  $x_o = Kx_i$  where  $x_i$  is the input,  $x_o$  is the output and  $K$  is the gain. This operation is implemented using the following set of abstract chemical reactions:

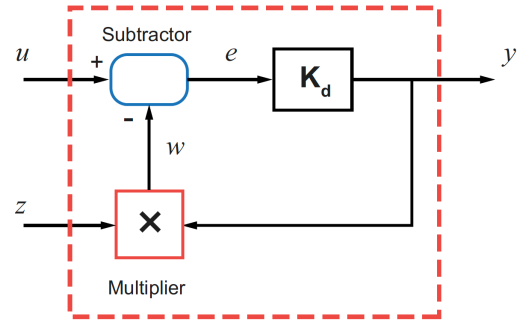
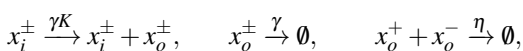


Fig. 3: A block diagram representation of the feedback system  $\mathcal{S}_D$  of [11] that computes the ratio  $y = u/z$  where  $u$  and  $z$  are biomolecular signals.

where  $\gamma K$ ,  $\gamma$  and  $\eta$  are the kinetic rates associated with catalysis, degradation and annihilation respectively.  $\square$

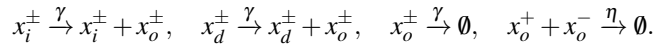
*Proof:* Using mass-action kinetics, it follows that the gain operator realised in this manner is described using the following ODE,

$$\begin{aligned} \frac{dx_o^+}{dt} &= \gamma(Kx_i^+ - x_o^+) - \eta x_o^+ x_o^- \\ \frac{dx_o^-}{dt} &= \gamma(Kx_i^- - x_o^-) - \eta x_o^+ x_o^- \\ \frac{dx_o}{dt} &= \frac{dx_o^+}{dt} - \frac{dx_o^-}{dt} = \gamma(Kx_i - x_o) \end{aligned} \quad (2)$$

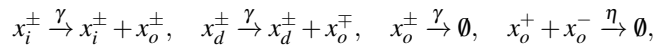
Using the final value theorem, the steady state value of  $x_o$  for constant input  $x_i$  is given by  $\lim_{t \rightarrow \infty} x_o(t) = Kx_i(t)$ .  $\square$

#### Lemma 2. [Summation]

Consider the summation operation  $x_o = x_i + x_d$ , where  $x_i$  and  $x_d$  are the inputs and  $x_o$  is the output. This operation is implemented using the following set of abstract chemical reactions:



Using the following set of abstract chemical reactions:

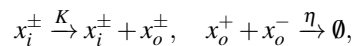


the subtraction  $x_o = x_i - x_d$  is implemented.  $\square$

**Remark 1.** Scaled summation  $x_o = K(x_i + x_d)$ , and scaled subtraction, can be implemented by choosing the catalysis rates in the construct of Lemma 2 to be  $K\gamma$ .  $\square$

#### Lemma 3. [Scaled Integration]

Consider the integrator  $x_o = K \int x_i dt$  where  $x_i$  is the input,  $x_o$  is the output, and  $K$  is the DC gain. Using the following set of abstract chemical reactions:



such an integrator is implemented.  $\square$

*Proof:* Using generalised mass-action kinetics the ODEs for the summation and integrator operations are given by  $\frac{dx_o}{dt} = \gamma(x_i + x_d - x_o)$  and  $\frac{dx_o}{dt} = Kx_i$ , respectively. Then, the proof for Lemmas 2 and 3 can be trivially obtained following the same logic as for the proof of Lemma 1.  $\square$

In [11], circuit to compute the ratio of any given two numbers was derived. The circuit is shown in Fig. 3 and the result is as follows.

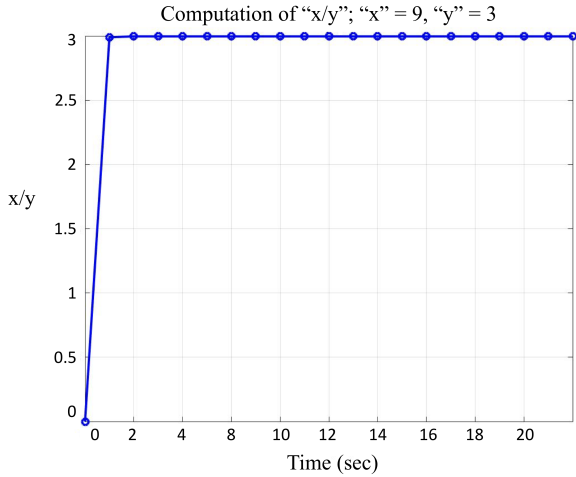


Fig. 4: Computation of the ratio “9/3” using  $\mathcal{S}_{C1}$  for the initial conditions noted down in Fig. 5.

**Lemma 4.** [Divider of [11]]

Consider the system  $\mathcal{S}_D$  shown in Fig. 3. Let the biomolecular signals  $u$  and  $z$  be its inputs. Then for sufficiently large values of  $K_d$ , its output  $y$  approximates the ratio  $u/z$ .  $\square$

*Proof:* From Fig. 3, we have  $e = u - zy$  and  $y = K_d e$ . Substituting the former equation into the latter and rearranging the variables, we get  $y = K_d(u - zy) = \frac{K_d u}{1 + K_d z} = \frac{u}{(1/K_d) + z}$ . If  $K_d$  is chosen large enough,  $y \approx u/z$ .  $\square$

**Remark 2.** The Newton-Raphson method can also be used to implement the ratio  $x/y$  of two numbers as follows. Define  $f(z) = y - 1/z$ . Next, starting with an arbitrarily chosen initial condition  $z_0$ , compute  $1/y$  using the iteration

$$\begin{aligned} z_{n+1} &= z_n - \frac{f(x_n)}{f'(x_n)} \\ &= z_n(2 - y \cdot z_n), \end{aligned}$$

where  $f'(\zeta)$  denotes the derivative of  $f(\zeta)$  with respect to  $\zeta$ . Let  $\bar{z} = \lim_{n \rightarrow \infty} z_n$ . Then, the ratio  $x/y = x \cdot \bar{z}$ . The Newton-Raphson method exhibits a quadratic convergence towards the steady-state value  $\bar{z}$ . This implementation requires one subtraction operation, one integration operation, and three multiplication operations. In contrast,  $\mathcal{S}_D$  requires only one scaled subtractor, one integrator, and one multiplier.  $\square$

III. MAIN RESULT

In [11], the stability and robustness properties of  $\mathcal{S}_D$  were not analysed. Furthermore, the ratio computation using the circuit  $\mathcal{S}_D$  incurs a steady state error since it can compute the ratio precisely if and only if  $K_d$  is arbitrarily large. In practice, reaction rates are finite and hence a wet-lab implementation of  $\mathcal{S}_D$  will always incur a steady-state error. We now present two circuits, viz.,  $\mathcal{S}_{C1}$  and  $\mathcal{S}_{C2}$ , that can overcome this limitation:  $\mathcal{S}_{C1}$  is shown in Fig. 1 and  $\mathcal{S}_{C2}$  is shown in Fig. 2. The ODE representation, the corresponding ACRN representation, and the corresponding DNA implementation for  $\mathcal{S}_{C1}$  is given in Table II (see Appendix). As Fig. 4 shows, this circuit computes the ratio precisely and without exhibiting any overshoot if the circuit parameter set is the initially chosen set of parameter values tabulated in Fig. 5. For poorly chosen parameter values, the

TABLE I: Notation

Symbol	Meaning
$(\mathbb{R}^+)$ $\mathbb{R}$	Set of all (nonnegative) real numbers.
$x \cdot y$	Product of $x$ and $y$ .
$(\cdot)'$ or $(\cdot)^T$	Transpose of a vector or a matrix $(\cdot)$ .
$\langle x, y \rangle$	$= \int_{-\infty}^{\infty} y^T(t)x(t)dt$ .
$\ x\ $	$= \sqrt{\langle x, x \rangle}$ .
$\mathcal{L}_2$	Space of possibly vector valued signals $x$ for which $\ x\  < \infty$ .
$\text{Re } G$ ( $\text{Im } G$ )	Real (imaginary) part of $G$ .
$\dot{x}$	Derivative of $x$ w.r.t. time, i.e., $\frac{dx}{dt}$ .
$\angle G$	Angle of a complex number $G$ .

circuit is unstable and cannot compute the ratio accurately. We now characterize the robustness and stability of  $\mathcal{S}_{C1}$ .

Our notation is summarized in Table I; for the discrete time case, the integration terms are to be appropriately replaced by summation terms, and the discrete time counterpart of  $\mathcal{L}_2$  is referred to as  $\ell_2$ . The set of all  $N$ -dimensional real-valued vectors is denoted  $\mathbb{R}^N$ . We refer to an operator by using a capital letter symbol, such as  $H$ , whereas a signal is referred to by using a small letter symbol, such as  $x$ . Our short-hand notation for  $x(t) = 0 \forall t$  is  $x \equiv 0$ .

**Definition 1.** [finite-gain stability] A system  $\mathcal{S}$  mapping  $u \in \mathcal{L}_2$  into  $y \in \mathcal{L}_2$  is said to be finite gain stable if there exists  $\gamma \geq 0$  such that  $\|y\| \leq \gamma \|u\|$  for all  $u \in \mathcal{L}_2$ . The smallest value of such a  $\gamma$  is said to be the gain of  $\mathcal{S}$ .  $\square$

**Definition 2.** [sector] We say that  $H$  is a sector  $[k_1, k_2]$  operator if it holds that  $\langle Hx - k_1x, Hx - k_2x \rangle \leq 0$  for all  $x \in \mathcal{L}_2$ .  $\square$

Lemma 5 is a straightforward application of the Nyquist stability theorem [13].

**Lemma 5.** Consider  $\mathcal{S}_{C1}$ . Let  $\gamma_1$  be the rate of catalysis and degradation at the scaled summation junction. Let  $\gamma_2$  be the rate of catalysis at the integrator. Let  $x_1$  be an arbitrary time-varying signal in sector(0,  $\alpha$ ) and let  $x_2$  be a constant, where  $\alpha > 0$ . Define

$$G(s) = \frac{K x_2 \gamma_1 \gamma_2}{s(s + \gamma_1)}. \tag{3}$$

Then,  $\mathcal{S}_{C1}$  is stable if and only if the Nyquist plot of  $G(s)$  does not encircle the point  $-1 + j0$  as  $\omega$  varies from  $-\infty$  to  $\infty$ . Furthermore, it can tolerate a time-delay of  $\text{PM}/\omega_g$ , where PM denotes the phase margin of  $G(s)$  and  $\omega_g$  denotes the gain cross-over frequency of  $G(s)$ . Replacing  $G(s)$  given by (3) with

$$\tilde{G}(s) \doteq \frac{K x_2 \gamma_1 \gamma_2}{(s + \gamma_1)(s + \gamma_2)},$$

similar result is obtained for  $\mathcal{S}_{C2}$ .  $\square$

**Remark 3.** As a result of Lemma 5, the finite gain stability of our ratio computation circuit  $\mathcal{S}_{C1}$  is not independent of the denominator  $x_2$  and, in particular, the stability margin decreases with an increase in  $x_2$ .  $\square$

**Remark 4.** Lemma 5 illustrates the main challenge posed by ACRN implementations, viz., an increase in the order of the loop transfer function: in the ODE representation of  $\mathcal{S}_{C1}$ , it

Block	Corresponding catalysis rate	Initially set values	Sets of parameter variations simulated by us: 15 sets in all							
			# 1	# 2	# 3	# 4	# 5	# 6	# 7	# 8
B1	$\gamma K1$	0.01	<b>0.1</b>	<b>1</b>	0.01	0.01	0.01	0.01	0.01	0.01
B2	$\gamma K2$	100	100	100	<b>1000</b>	<b>10000</b>	<b>10</b>	100	100	100
B3	$\gamma K3$	100	100	100	100	100	100	100	<b>150</b>	<b>500</b>
B4	$\gamma K4$	10	10	10	10	10	10	10	10	10
B2	K	250	250	250	250	250	250	<b>25</b>	250	250
Settling time, t.u.		1	-	-	1	1	4	8	1	1
Peak overshoot		-	-	-	-	-	-	-	-	-
Computed steady-state value of the ratio "9/3"		2.996005	<b>unstable</b>	<b>unstable</b>	2.996005	2.996005	2.996005	2.960529	2.996005	2.9960

Block	Corresponding catalysis rate	Initially set values	Sets of parameter variations simulated by us: 15 sets in all						
			#9	# 10	# 11	# 12	# 13	# 14	# 15
B1	$\gamma K1$	0.01	<b>0.1</b>	0.01	<b>0.1</b>	<b>0.1</b>	<b>0.001</b>	0.01	0.01
B2	$\gamma K2$	100	100	100	100	100	100	100	100
B3	$\gamma K3$	100	<b>500</b>	100	<b>500</b>	<b>500</b>	100	<b>10</b>	100
B4	$\gamma K4$	10	10	10	<b>100</b>	<b>100</b>	10	10	<b>100</b>
B2	K	250	250	<b>500</b>	250	<b>500</b>	250	250	250
Settling time, t.u.		1	-	2	1	-	5	6	7
Peak overshoot		-	4.9	-	-	-	-	-	-
Computed steady-state value of the ratio "9/3"		2.996005	3.1510	<b>2.9980</b>	2.996005	2.828692	2.995997	2.995991	2.995990

Fig. 5: Tabulated steady-state response of  $\mathcal{S}_{C1}$  for different sets of parameters values; the time unit (t.u.) is in seconds. All other factors being equal, an increase in a parameter value should be interpreted as a speed-up of the feedback loop of  $\mathcal{S}_{C1}$  and a decrease in the parameter value should be interpreted as a slow-down of the feedback loop of  $\mathcal{S}_{C2}$ .

is  $G(s) = \frac{K x_2}{s}$  but in the ACRN implementation, it is given by  $G(s) = \frac{K x_2}{s} \left( \frac{\gamma_1 \gamma_2}{s + \gamma_1} \right)$ . Consequently, the ACRN representations have diminished stability and robustness properties compared to their ODE counterparts.  $\square$

**Remark 5.** In principle,  $\mathcal{S}_{C1}$  computes the ratio of any two given entities  $x_1$  and  $x_2$  accurately so long as these are either (1) pairs of constant-valued scalars or (2) time-varying scalars or (3) constant-valued polynomials or (4) time-varying polynomials. However, in practice, this can be claimed for only a smaller subset due to the limitations on how fast and how accurately the DNA strand displacement implementation can proceed in the real-world. For example, to compute  $x_1/x_2$  when  $x_2$  is significantly large,  $\gamma_1$  and  $\gamma_2$  must be chosen to be proportionately small while ensuring that  $\gamma_2$  is much larger than  $\gamma_1$ . In practice, since  $\gamma_1$  and  $\gamma_2$  are the catalysis rates, these cannot be chosen to be arbitrarily small or large and consequently,  $\mathcal{S}_{C1}$  will be unstable if  $x_2$  is chosen to be large enough. Our Lemma 5 establishes how large  $x_2$  can be for which  $\mathcal{S}_{C1}$  is finite gain stable.  $\square$

**Theorem 1.** [Stability Analysis of  $\mathcal{S}_{C1}$ ]

Consider  $\mathcal{S}_{C1}$ . Let  $\gamma_1$  be the rate of catalysis and degradation at the scaled summation junction. Let  $\gamma_2$  be the rate of catalysis at the integrator. Let  $\gamma_3$  be the rate of catalysis and degradation at the multiplication junction. Let  $x_1$  be an arbitrary time-varying signal in sector(0,  $\alpha$ ) and let  $x_2$  be an arbitrary time-varying signal in sector(0,  $\beta$ ), where  $\alpha > 0$  and  $\beta > 0$  are constants. Define

$$G(s) = \frac{K \gamma_1 \gamma_2 \gamma_3}{s (s + \gamma_1)(s + \gamma_3)}. \quad (4)$$

Then,  $\mathcal{S}_{C1}$  is stable if and only if

$$\text{Re } G(j\omega) > -\frac{1}{\beta} \quad \forall \omega \in \mathbb{R}. \quad (5)$$

Furthermore, the maximum delay  $\tau^*$  for which  $\mathcal{S}_{C1}$  is stable

is given as

$$\tau^* = \frac{\text{PM} - \cos^{-1}(-1/\beta)}{\omega_c}, \quad (6)$$

where  $\omega_c$  is the gain cross-over frequency of  $G(s)$  and PM is the phase margin of  $G(s)$ . Replacing  $G(s)$  given by (4) with

$$\tilde{G}(s) \doteq \frac{K \gamma_1 \gamma_2 \gamma_3}{(s + \gamma_1)(s + \gamma_2)(s + \gamma_3)},$$

similar result is obtained for  $\mathcal{S}_{C2}$ .  $\square$

*Proof:*  $\mathcal{S}_{C1}$  can be decomposed as a feedback system that has a linear time-invariant (LTI) element  $G(s)$  in the feedforward path and a sector(0,  $\beta$ ) nonlinearity in the feedback path, where  $G(s)$  is given by (4). W.l.o.g., let  $|1/\beta| < 1$ . By the Circle Theorem [14], this feedback system is stable if and only if  $\text{Re } G(j\omega) > -1/\beta \quad \forall \omega$ .

To derive the upper bound (6) on the tolerable time-delay, observe that the effect of the time-delay is to rotate the Nyquist plot of the feedback loop in the clock-wise direction whilst leaving the magnitude unchanged. So, it suffices to focus on the gain cross-over frequency  $\omega_c$ , since at that frequency  $|G(j\omega_c)| = 1$ , and compute the additional phase lag needed to make the system unstable. The phase lag induced by the constant-valued time delay  $\tau$  at the frequency  $\omega$  equals  $\tau\omega$ . Let  $H$  denote the hyperplane in the  $s$ -plane that is defined as  $H \doteq \{x : \text{Re}(x) < -1/\beta\}$ . At the gain cross-over frequency,  $\angle G(j\omega_c) = -180^\circ + \text{PM}$ . Hence, rotating  $G(j\omega_c)$  clock-wise through  $\theta \doteq \text{PM} - \cos^{-1}(-1/\beta)$  makes it intersect with the hyperplane  $H$ , which makes the system  $\mathcal{S}_{C1}$  unstable as a consequence of the Circle Theorem. The time-delay that causes this additional phase lag is  $\theta/\omega_c$ . Hence the proof.  $\square$

**Remark 6.** An illustration of the argument used to prove the bound (6) on time-delay is given in Fig. 7.  $\square$

**Remark 7.** Using the logic explained in [11], it readily follows that  $x_1$  and  $x_2$  can be allowed to be real-valued

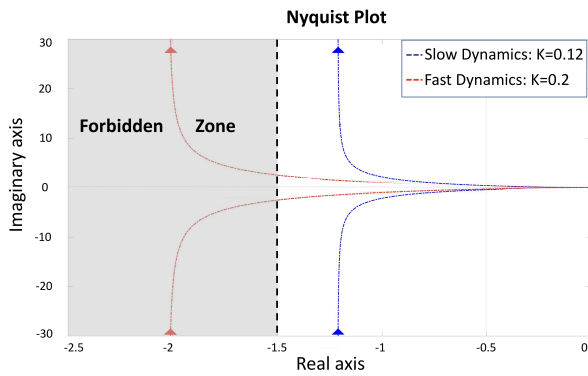


Fig. 6: An illustration of the finite gain stability analysis of  $\mathcal{S}_{C1}$  using our Theorem 1. If the feedback loop is slow enough, the Nyquist plot of the feedback loop is outside the forbidden zone specified by the Circle Criterion and, consequently,  $\mathcal{S}_{C1}$  is stable and computes the ratio  $x_1/x_2$  for the given  $x_1$  and  $x_2$ . However, if the feedback loop is speeded up to ensure that the circuit computes the ratio faster, the Nyquist plot of the feedback loop enters the forbidden zone and, consequently,  $\mathcal{S}_{C1}$  is unstable and cannot compute the ratio  $x_1/x_2$  for the given  $x_1$  and  $x_2$ .

polynomials as well.  $\square$

#### IV. CONCLUSIONS

We have presented two circuits  $\mathcal{S}_{C1}$  and  $\mathcal{S}_{C2}$ , and their abstract chemical reaction network representations, that accurately compute the ratio of two entities; the entities can be either two constant-valued scalars or two time-varying scalars or two constant-valued polynomials or two time-varying polynomials. We have also characterised their finite gain stability and have derived an upper bound on the time-delay that can be safely tolerated by these circuits.

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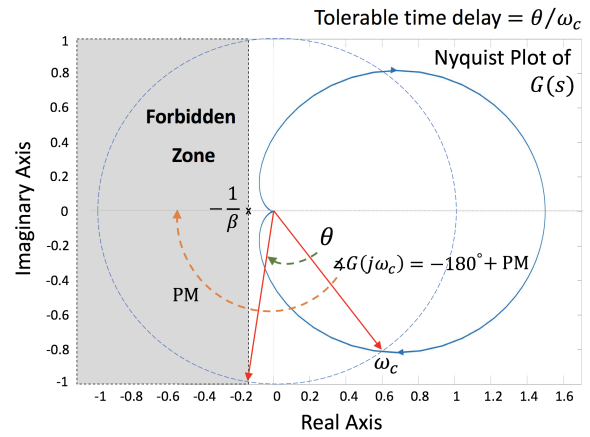


Fig. 7: A graphical illustration of our derivation of the upper bound (6) on the amount of time delay that can be tolerated by  $\mathcal{S}_{C1}$ . At the gain cross-over frequency  $\omega_c$ ,  $|G(j\omega_c)| = 1$  and  $\angle G(j\omega_c) = -180^\circ + PM$ . Here, the maximum delay that can be tolerated is  $\tau^* = \theta/\omega_c$ .

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#### V. CORRECTIONS TO THE PRINTED VERSION

On May 9, 2018, the online version of this paper was revised from the printed version as follows: (1) " $\sin^{-1}(PM)$ " in the caption of Fig. 7 was corrected to "PM", (2) " $\dot{z}_1 = \delta_2 g$ " in Table II was corrected to " $\dot{g} = \delta_2 z_1$ ", (3) more graphical details were added to Fig. 6 and Fig. 7, (4) the sentence "W.l.o.g., let  $|1/\beta| < 1$ ." was added to the proof of Theorem 1, (5) the sentence "We refer to the blocks B1 and B2 taken together as B1+B2: in other words, the block B1+B2 has  $x_1$  and  $w$  as its inputs and produces  $z_1$  as its output." was added to the APPENDIX, and (6) the caption of Fig. 1 was revised slightly to reflect the fifth correction.

#### APPENDIX

Here, in Table II, we note down the ODE's, ACRNs, and DNA implementation details for  $\mathcal{S}_{C1}$ . We refer to the blocks B1 and B2 taken together as B1+B2: in other words, the block B1+B2 has  $x_1$  and  $w$  as its inputs and produces  $z_1$  as its output. The annihilation reaction rate  $\eta$  is to be chosen arbitrarily large.

TABLE II: Ratio Computation Using  $\mathcal{S}_{C1}$ : DNA Implementation, the ACRNs and the ODEs

DNA Implementation	Formal CRNs	ODEs
<b>Block B1+B2</b>		
$\left. \begin{array}{l} x_1^\pm + G_1^\pm \xrightarrow{q_1} \emptyset + O_1^\pm \\ O_1^\pm + T_1^\pm \xrightarrow{q_{max}} x_1^\pm + z_1^\pm \\ w^\mp + G_2^\pm \xrightarrow{q_2} \emptyset + O_2^\pm \\ O_2^\pm + T_2^\pm \xrightarrow{q_{max}} w^\mp + z_1^\pm \\ z_1^\pm + G_3^\pm \xrightarrow{q_3} \emptyset \end{array} \right\} \dots$	$\left. \begin{array}{l} x_1^\pm \xrightarrow{K\delta_1} x_1^\pm + z_1^\pm \\ w^\mp \xrightarrow{K\delta_1} w^\mp + z_1^\pm \\ z_1^\pm \xrightarrow{\delta_1} \emptyset \end{array} \right\}$	$\dot{z}_1 = \delta_1(K(x_1 - w) - z_1)$
$\left. \begin{array}{l} x_1^+ + L_{1x} \xrightleftharpoons[q_{max}]{q_{max}} H_{1x} + B_{1x} \\ x_1^- + LS_{1x} \xrightleftharpoons[q_{max}]{q_{max}} HS_{1x} + BS_{1x} \\ x_1^- + H_{1x} \xrightarrow{q_{max}} \emptyset \end{array} \right\}$	$x_1^+ + x_1^- \xrightarrow{\eta} \emptyset$	
$\left. \begin{array}{l} w^+ + L_{1w} \xrightleftharpoons[q_{max}]{q_{max}} H_{1w} + B_{1w} \\ w^- + LS_{1w} \xrightleftharpoons[q_{max}]{q_{max}} HS_{1w} + BS_{1w} \\ w^- + H_{1w} \xrightarrow{q_{max}} \emptyset \end{array} \right\}$	$w^+ + w^- \xrightarrow{\eta} \emptyset$	
$\left. \begin{array}{l} z_1^+ + L_{1z} \xrightleftharpoons[q_{max}]{q_{max}} H_{1z} + B_{1z} \\ z_1^- + LS_{1z} \xrightleftharpoons[q_{max}]{q_{max}} HS_{1z} + BS_{1z} \\ z_1^- + H_{1z} \xrightarrow{q_{max}} \emptyset \end{array} \right\}$	$z_1^+ + z_1^- \xrightarrow{\eta} \emptyset$	
<b>Block B3</b>		
$\left. \begin{array}{l} z_1^\pm + G_7^\pm \xrightarrow{q_7} \emptyset + O_7^\pm \\ O_7^\pm + T_7^\pm \xrightarrow{q_{max}} z_1^\pm + g^\pm \\ g^+ + L_{3g} \xrightleftharpoons[q_{max}]{q_{max}} H_{3g} + B_{3g} \\ g^- + LS_{3g} \xrightleftharpoons[q_{max}]{q_{max}} HS_{3g} + BS_{3g} \\ g^- + H_{3g} \xrightarrow{q_{max}} \emptyset \end{array} \right\}$	$\left. \begin{array}{l} z_1^\pm \xrightarrow{\delta_2} z_1^\pm + g^\pm \\ g^+ + g^- \xrightarrow{\eta} \emptyset \end{array} \right\}$	$\dot{g} = \delta_2 z_1$
<b>Block B4</b>		
$\left. \begin{array}{l} x_2^\pm + L_{3x2} \xrightleftharpoons[q_{max}]{q_{3'}} H_{3x2} + B_{3x2} \\ g^\pm + H_{3g} \xrightarrow{q_{max}} O_5^\pm \\ O_5^\pm + T_5^\pm \xrightarrow{q_{max}} w^\pm \\ w^\pm + G_6^\pm \xrightarrow{q_6} \emptyset \\ x_2^+ + L_{1x2} \xrightleftharpoons[q_{max}]{q_{max}} H_{1x2} + B_{1x2} \\ x_2^- + LS_{1x2} \xrightleftharpoons[q_{max}]{q_{max}} HS_{1x2} + BS_{1x2} \\ x_2^- + H_{1x2} \xrightarrow{q_{max}} \emptyset \end{array} \right\} \dots$	$\left. \begin{array}{l} x_2^\pm + g^\pm \xrightarrow{\delta_3} x_2^\pm + g^\pm + w^\pm \\ w^\pm \xrightarrow{\delta_3} \emptyset \\ x_2^+ + x_2^- \xrightarrow{\eta} \emptyset \end{array} \right\}$	$\dot{w} = \delta_3(x_2 * g - w)$