

Fluctuation and thermodynamic uncertainty relations in deterministic chemical reaction networks

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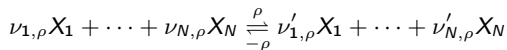
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[KY and Ito, PRL, 2021] see also [KY, Kolchinsky, Dechant, and Ito, arXiv:2205.15227]

Chemical reaction network

- Chemical reaction networks



described by the deterministic rate equation

$$\frac{dc_\alpha}{dt} = \sum_{\rho} (\nu'_{\alpha\rho} - \nu_{\alpha\rho})(\mathcal{K}_{\rho}(c) - \mathcal{K}_{-\rho}(c))$$

$\mathcal{K}_{\pm\rho}(c) \geq 0$: reaction rate, which can be nonlinear

- The theoretical frameworks of deterministic CRNs and stochastic jump processes resemble each other
- Approach based on the analogy has worked, especially to understand the CRN side e.g., Schnakenberg theory [Poletti & Esposito, J. Phys. Chem. 2014], information thermodynamics [Penocchio, *et al.*, arXiv:2204.02815] (see also [KY, Kolchinsky, Dechant, and Ito, arXiv:2205.15227], where we derive several results integrating the two frameworks)

Similarity 1) Equations of motion

- Rate equation $d_t c = \mathbb{S}J(c)$

c : concentration

\mathbb{S} : stoichiometric matrix; $\mathbb{S}_{\alpha\rho} = \nu'_{\alpha\rho} - \nu_{\alpha\rho}$

$J(c)$: reaction current; $J_\rho(c) = \mathcal{K}_\rho(c) - \mathcal{K}_{-\rho}(c)$

- Master equation $d_t p = BJ(p)$

p : probability

B : incidence matrix of the directed graph of states and transitions

given by $B_{i,(jk)} = \delta_{ik} - \delta_{ij}$

J : probability current $J_{ij}(p) = k_{ij}p_i - k_{ji}p_j$

Similarity 2) Thermodynamics

- Affinity/Thermodynamic force

$$\text{CRN: } F_{\rho}(c) = T \ln \frac{\mathcal{K}_{\rho}(c)}{\mathcal{K}_{-\rho}(c)} \quad (R = 1)$$

$$\text{Jump process: } F_{ij}(p) = T \ln \frac{k_{ij}p_i}{k_{ji}p_j} \quad (k_B = 1)$$

- Entropy production rate:

$$\sigma = \frac{1}{T} J^T F = \sum_{\rho} (\mathcal{K}_{\rho} - \mathcal{K}_{-\rho}) \ln \frac{\mathcal{K}_{\rho}}{\mathcal{K}_{-\rho}} \quad \text{or} \quad \sum_{i < j} (k_{ij}p_i - k_{ji}p_j) \ln \frac{k_{ij}p_i}{k_{ji}p_j}$$

Difference: Fluctuates or not

- In stochastic thermodynamics, fluctuations (typically evaluated by the variance) play a central role. e.g., fluctuation theorem, thermodynamic uncertainty relation
- But absent in deterministic CRNs

Q. Aren't there any counterparts that can bring some insights to CRNs?

Results

- $D_{\alpha\beta}(c) := (1/2) \sum_{\rho} \mathbb{S}_{\alpha\rho} \mathbb{S}_{\beta\rho} (\mathcal{K}_{\rho}(c) + \mathcal{K}_{-\rho}(c))$ plays the role of a diffusivity in deterministic CRNs (previously known (e.g., [Ge and Qian, Chem. Phys. 2016]) but has not been utilized effectively)
- TUR [KY and Ito, PRL 2021; KY, *et al.*, arXiv:2205.15227]

$$\sigma \geq \max_{\psi \in \mathbb{R}^N} \frac{1}{\psi^T D(c) \psi} \left(\psi^T \frac{dc}{dt} \right)^2 \quad (\text{arXiv})$$

$$\geq \max_{\psi_{\beta} = \delta_{\alpha\beta}} \max_{\alpha} \frac{(d_t c_{\alpha})^2}{D_{\alpha\alpha}} \quad (\text{PRL})$$

- Speed limit $\tau \geq \frac{|c(0) - c(\tau)|^2}{\bar{D}\Sigma}$ [KY and Ito, PRL 2021]

→ Why is D a diffusion coefficient matrix?

Understanding from analogy

- Traffic $A_{ij}(p) := k_{ij}p_i + k_{ji}p_j$ in a jump process gives a short-time fluctuation as

$$\text{Var}(\Delta R)(t; t + \Delta t) = \sum_{i < j} (R_i - R_j)^2 A_{ij}(p(t)) \Delta t + o(\Delta t)$$

- TUR $\sigma \geq \frac{2(d_t \langle R \rangle)^2}{D_R}$ with $D_R := \sum_{i < j} (R_i - R_j)^2 A_{ij}(p(t))$ [Otsubo, et al., PRE 2020]

- If we define $A_\rho(c) := \mathcal{K}_\rho(c) + \mathcal{K}_{-\rho}(c)$, $D_{\alpha\beta}(c) = \frac{1}{2} \sum_\rho \mathbb{S}_{\alpha\rho} \mathbb{S}_{\beta\rho} A_\rho(c)$
 → analogous to fluctuations

From chemical master/Fokker–Planck equation

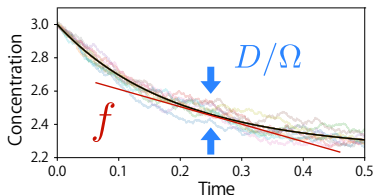
- The system size expansion of the chemical master equation brings the chemical Fokker–Planck equation

$$\frac{\partial}{\partial t} P(c, t) = - \sum_{\alpha} \frac{\partial}{\partial c_{\alpha}} [f_{\alpha}(c) P(c, t)] + \frac{1}{\Omega} \sum_{\alpha, \beta} \frac{\partial^2}{\partial c_{\alpha} \partial c_{\beta}} [D_{\alpha\beta}(c) P(c, t)]$$

with Ω being the volume parameter and

$$f_{\alpha}(c) := \sum_{\rho} S_{\alpha\rho} (\mathcal{K}_{\rho}(c) - \mathcal{K}_{-\rho}(c)), \quad D_{\alpha\beta}(c) = \frac{1}{2} \sum_{\rho} S_{\alpha\rho} S_{\beta\rho} (\mathcal{K}_{\rho}(c) + \mathcal{K}_{-\rho}(c))$$

→ D is the diffusion coefficient of the chemical FP equation times the volume



Proof of TUR

$$\begin{aligned}
\left(\psi^\top \frac{dc}{dt}\right)^2 &= \left(\sum_{\rho} [\mathbb{S}^\top \psi]_{\rho} J_{\rho}(c)\right)^2 = \left(\sum_{\rho} [\mathbb{S}^\top \psi]_{\rho} \frac{J_{\rho}(c)}{F_{\rho}(c)/T} F_{\rho}(c)\right)^2 \\
&\leq \sum_{\rho} \frac{J_{\rho}(c)}{F_{\rho}(c)/T} (c) [\mathbb{S}^\top \psi]_{\rho}^2 \sum_{\rho'} \frac{J_{\rho'}(c)}{F_{\rho'}(c)/T} F_{\rho'}(c)^2 \\
&\leq \frac{1}{2} \sum_{\rho} A_{\rho}(c) [\mathbb{S}^\top \psi]_{\rho}^2 \sigma = (\psi^\top D \psi) \sigma
\end{aligned}$$

Where we used

- Cauchy–Schwarz inequality in the second line
- Inequality between log mean and arithmetic mean in the third line

$$\frac{a-b}{\ln(a/b)} \leq \frac{a+b}{2}$$

with $a = \mathcal{K}_{\rho}(c)$ and $b = \mathcal{K}_{-\rho}(c)$, that is, $\frac{J_{\rho}(c)}{F_{\rho}(c)/T} \leq \frac{A_{\rho}(c)}{2}$

Summary

- We discussed that $D_{\alpha\beta}(c) = \frac{1}{2} \sum_{\rho} \mathbb{S}_{\alpha\rho} \mathbb{S}_{\beta\rho} (\mathcal{K}_{\rho}(c) + \mathcal{K}_{-\rho}(c))$ evaluates the (potential) “fluctuation” of a CRN
- It provides the TUR

$$\sigma \geq \sigma^{\text{ex}} \geq \frac{1}{\psi^{\top} D(c) \psi} \left(\psi^{\top} \frac{dc}{dt} \right)^2$$

and the speed limit

$$\tau \geq \frac{|c(0) - c(\tau)|^2}{\bar{D}\Sigma}$$

- * Please see our paper [KY and S. Ito, PRL 2021] for a bit more details, and [KY, Kolchinsky, Dechant, and Ito, arXiv:2205.15227] for several topics, such as decomposition of the EPR and optimal transport in jump processes & CRNs.

Master equation

- Master equation $d_t p = BJ(p)$

p : probability

B : incidence matrix of the directed graph of states and transitions
 given by $B_{i,(jk)} = \delta_{ik} - \delta_{ij}$

J : probability current $J_{ij}(p) = k_{ij}p_i - k_{ji}p_j$

$$\therefore \sum_{j < k} B_{i,(jk)} J_{jk} = \sum_{j(<i)} J_{ji} - \sum_{k(>i)} J_{ik} = \sum_{j(\neq i)} J_{ji}$$

Open CRN

- α is open $\stackrel{\text{def}}{\iff} d_t c_\alpha = [\mathbb{S}J(c)]_\alpha + \mathcal{J}_\alpha = 0$ by an external flow \mathcal{J}
- The TUR can be stated more generally as

$$\sigma \geq \frac{1}{\psi^\top D(c) \psi} \left(\psi^\top f(c) \right)^2$$

with $f_\alpha(c) = \sum_\rho \mathbb{S}_{\alpha\rho} J_\rho(c)$ regardless of whether α is closed or open

- Specifically, in a steady state c^{ss} , $f_\alpha(c^{\text{ss}}) = 0$ if α is closed, and $f_\alpha(c^{\text{ss}}) = -\mathcal{J}_\alpha$ if open, thus

$$\sigma^{\text{ss}} \geq \frac{1}{\psi^\top D(c^{\text{ss}}) \psi} \left(\psi^\top \mathcal{J} \right)^2$$

Keizer's Fluctuation–Dissipation relation

[Ge and Qian, Chem. Phys. 2016]

- Let c^{ss} be a complex balanced steady state
- Define

$$\Gamma_{\alpha\beta}(c) := \frac{\partial f_{\alpha}}{\partial c_{\beta}}(c), \quad g_{\alpha\beta}(c) := \frac{1}{RT} \frac{\partial^2 G}{\partial c_{\alpha} \partial c_{\beta}} = \frac{\delta_{\alpha\beta}}{c_{\alpha}}$$

- Then, we have

$$-2D(c^{\text{ss}}) = \Gamma(c^{\text{ss}})g(c^{\text{ss}})^{-1} + g(c^{\text{ss}})^{-1}\Gamma(c^{\text{ss}})^{\text{T}}$$

- cf. Lyapunov equation for a stochastic linear system $d_t x = Hx + \Xi$, with
- $$\langle \Xi_i(t) \Xi_j(t') \rangle = 2D_{ij} \delta(t - t')$$

$$\frac{d}{dt} \mathfrak{G}(t) = H\mathfrak{G}(t) + \mathfrak{G}(t)H^{\text{T}} + 2D$$

$$\mathfrak{G}_{ij}(t) = \langle [x_i(t) - \langle x_i(t) \rangle][x_j(t) - \langle x_j(t) \rangle] \rangle$$