Reaction-diffusion systems

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Based on lecture notes of Radek Erban http://www.maths.ox.ac.uk/courses/course/19651/material http://www.maths.ox.ac.uk/cmb/education

> Centre for Mathematical Biology Mathematical Institute



Summer school, Prague, 6–8 August, 2013

Outline

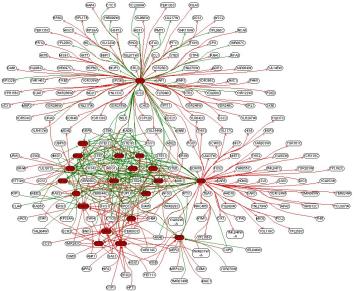


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- Motivation
- Reaction kinetics: deterministic and stochastic models
- Models of diffusion
- Application to circadian rhythms
- Application to pattern formation

Motivation - gene regulatory networks

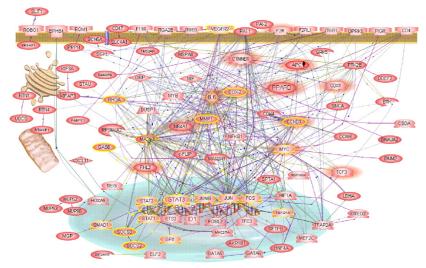




Neighbourhood of mating response genes [Rung, Schlitt, et al, 2002]

Motivation - gene regulatory networks





Angiogenic signaling network. [Abdollahi et al, PNAS 2007]

Stochastic models of reaction kinetics



Degradation

$$A \xrightarrow{k} \emptyset$$

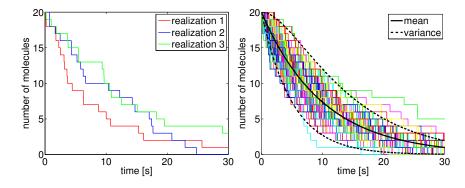
Naive stochastic simulation algorithm (SSA):

Initialization: $\Delta t > 0$ small, for t = 0 set $A(0) = n_0$.

(a1) Generate a random number r uniformly distributed in (0,1)
(b1) If r < A(t)kΔt then A(t + Δt) = A(t) - 1; else A(t + Δt) = A(t)

Naive SSA: degradation





 $k = 0.1, A(0) = 20, \Delta t = 0.005$

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Gillespie SSA for degradation



$$A \xrightarrow{k} \emptyset$$

Initialization: set $A(0) = n_0$.

(a2) Generate a random number r uniformly distributed in (0,1)

- (b2) Compute the next reaction time $\tau = \frac{1}{A(t)k} \ln \left[\frac{1}{r}\right]$
- (c2) Update the number of molecules: $A(t + \tau) = A(t) 1$ Set $t := t + \tau$ and go to (a2)

Chemical reactions of higher-order



order	reaction	propensity	units of <i>k</i>
0	$\emptyset \xrightarrow{k} A$	kν	${\rm m}^{-3}{ m sec}^{-1}$
1	$A \xrightarrow{k} \emptyset$	A(t)k	sec^{-1}
2	$A+B \stackrel{k}{\longrightarrow} \emptyset$	A(t)B(t)k/ u	${\rm m}^3 { m sec}^{-1}$
2	$2A \xrightarrow{k} \emptyset$	A(t)(A(t)-1)k/ u	${ m m}^3{ m sec}^{-1}$
3	$A+B+C \xrightarrow{k} \emptyset$	$A(t)B(t)C(t)k/ u^2$	${\rm m}^6{ m sec}^{-1}$
3	$2A + B \xrightarrow{k} \emptyset$	$A(t)(A(t)-1)B(t)k/ u^2$	${\rm m}^6{ m sec}^{-1}$
3	$3A \xrightarrow{k} \emptyset$	$A(t)(A(t) - 1)(A(t) - 2)k/\nu^2$	$m^6 sec^{-1}$

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System with two species



$$2A \xrightarrow{k_1} \emptyset, \quad A + B \xrightarrow{k_2} \emptyset, \quad \emptyset \xrightarrow{k_3} A, \quad \emptyset \xrightarrow{k_4} B,$$

Gillespie SSA:

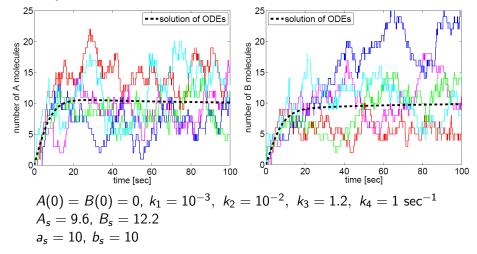
- (a4) Generate two random numbers: $r_1, r_2 \sim U(0, 1)$
- (b4) Compute propensities: $\alpha_1(t) = k_1 A(t)(A(t) - 1), \ \alpha_2(t) = k_2 A(t)B(t), \ \alpha_3 = k_3, \ \alpha_4 = k_4, \ \text{and} \ \alpha_0 = \alpha_1(t) + \alpha_2(t) + \alpha_3 + \alpha_4$ (c4) Next reaction time $\tau = \frac{1}{\alpha_0} \ln \left[\frac{1}{r_1}\right]$

(d4) Update the numbers of molecules:

System with two species



Trajectories



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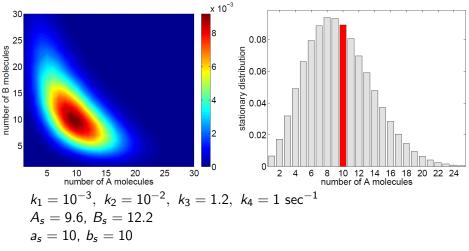
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Stationary distribution





General Gillespie SSA

Notation

 $q \dots$ number of chemical reactions $\alpha_j(t) \dots$ propensity function of *j*-th reaction, $j = 1, 2, \dots, q$ $\alpha_j(t) dt =$ probability that *j*-th reaction occurs in [t, t + dt)

Algorithm

(a5) Generate random numbers r_1 , r_2 uniformly distributed in (0, 1) (b5) Compute propensity $\alpha_j(t)$ of each reaction and $\alpha_0 = \sum_{j=1}^q \alpha_j$ (c5) Next reaction time $\tau = \frac{1}{\alpha_0} \ln \left[\frac{1}{r_1}\right]$

(d5) Compute which reaction occurs at time $t + \tau$. Find j such that

$$r_2 \geq rac{1}{lpha_0}\sum_{i=1}^{j-1}lpha_i(t) \quad ext{and} \quad r_2 < rac{1}{lpha_0}\sum_{i=1}^j lpha_i(t)$$

(e5) The *j*-th reaction takes place. Update numbers of molecules. Set $t := t + \tau$ and go to (a5) System with multiple favourable states



Schlögl system

$$3A \stackrel{k_1,k_2}{\rightleftharpoons} 2A \qquad A \stackrel{k_3,k_4}{\rightleftharpoons} \emptyset$$

oncentration: $a(t) = A(t)/\nu$

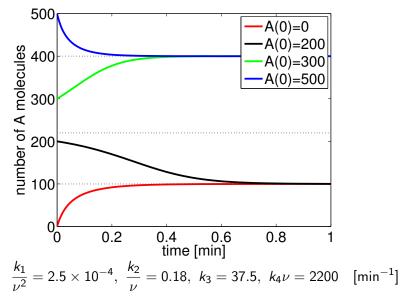
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$$\frac{da}{dt} = -k_1 a^3 + k_2 a^2 - k_3 a + k_4$$

Average number of molecules: $\overline{A}(t) = a(t)\nu$

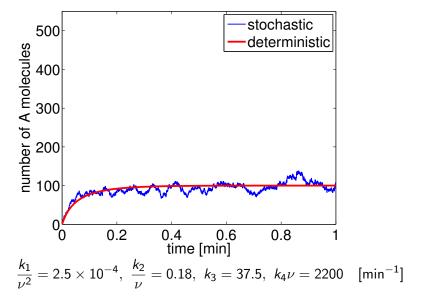
$$\frac{d\overline{A}}{dt} = -\frac{k_1}{\nu^2}\overline{A}^3 + \frac{k_2}{\nu}\overline{A}^2 - k_3\overline{A} + k_4\nu$$





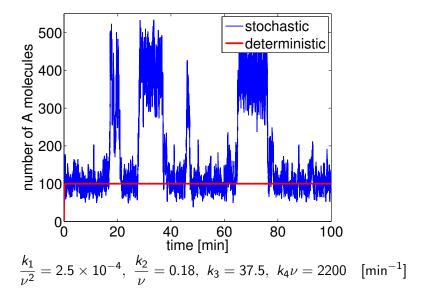
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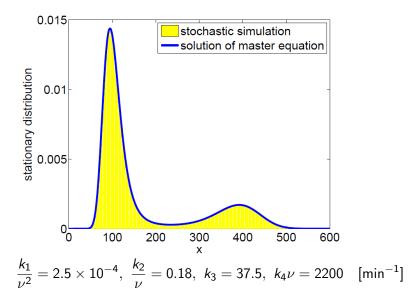
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Self-induced stochastic resonance



Schnakenberg system

$$2A + B \stackrel{k_1}{\rightarrow} 3A \qquad \emptyset \stackrel{k_2,k_3}{\rightleftharpoons} A \qquad \emptyset \stackrel{k_4}{\rightleftharpoons} B$$

Concentration:

$$\frac{da}{dt} = k_1 a^2 b + k_2 - k_3 a$$
$$\frac{db}{dt} = -k_1 a^2 b + k_4$$

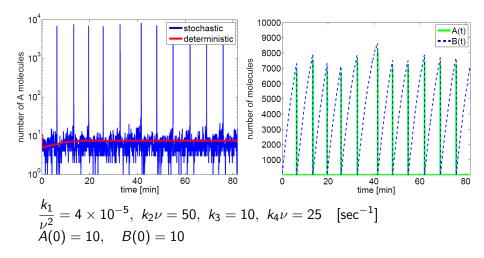
Average numbers of molecules:

$$\frac{d\overline{A}}{dt} = \frac{k_1}{\nu^2} \overline{A}^2 \overline{B} + k_2 \nu - k_3 \overline{A}$$
$$\frac{d\overline{B}}{dt} = -\frac{k_1}{\nu^2} \overline{A}^2 \overline{B} + k_4$$

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Schnakenberg system

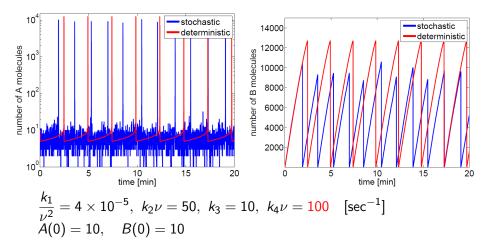




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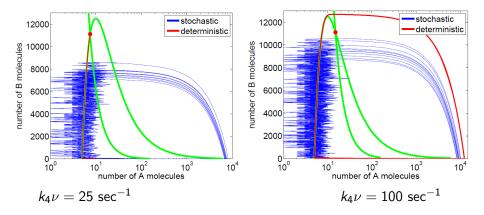
Schnakenberg system





Schnakenberg system





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Stochastic differential equations (SDE)



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$$X(t + dt) = X(t) + f(X(t), t) dt + g(X(t), t) dW$$

 $dW \dots$ white noise, $dW \approx \sqrt{\Delta t} \xi$, with $\xi \sim N(0, 1)$

Simulation algorithm

$$X(0) = x_0, \ \Delta t > 0 \text{ small}$$
(a6) $\xi \sim N(0, 1)$
(b6) $X(t + \Delta t) = X(t) + f(X(t), t)\Delta t + g(X(t), t)\sqrt{\Delta t}\xi$
Set $t := t + \Delta t$ and go to (a6)

Example 1: f(x, t) = 0, g(x, t) = 1



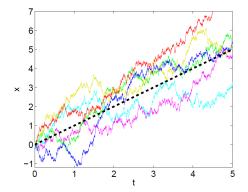
Trajectories: $X(t + \mathrm{d}t) = X(t) + \mathrm{d}W_1$ $X(t + \mathrm{d}t) = X(t) + \mathrm{d}W$ $Y(t + dt) = Y(t) + dW_2$ 0.6 1.5 0.4 0.2 0.5 y [mm] × -0.5 -0.2-1 -0.4 -1.5 -0.6 0 0.2 0.4 0.6 0.8 -0.6 -0.2 0 0.2 0.4 0.6 x [mm] ŧ

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Example 2: f(x, t) = 1, g(x, t) = 1



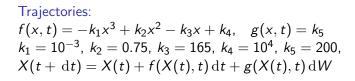
Trajectories: X(t + dt) = X(t) + dt + dW

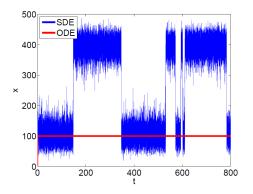


Example 3: two favourable states



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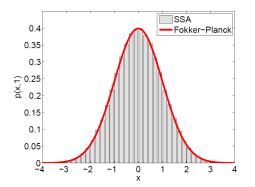




Example 1: f = 0, g = 1 (revisited)



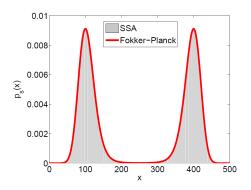
Stationary probability distribution: X(t + dt) = X(t) + dW



Example 3: two favourable states (revisited)



Stationary probability distribution: $f(x, t) = -k_1 x^3 + k_2 x^2 - k_3 x + k_4$, $g(x, t) = k_5$ $k_1 = 10^{-3}$, $k_2 = 0.75$, $k_3 = 165$, $k_4 = 10^4$, $k_5 = 200$, X(t + dt) = X(t) + f(X(t), t) dt + g(X(t), t) dW



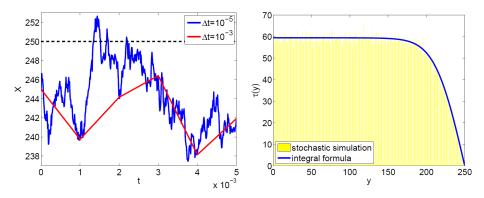
Example 3: two favourable states (revisited)



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Mean exit time: $\tau_{\rm sim} = 64.7$ $\tau_{\rm x_{s_1}} = 59.45$



Stochastic equations for chemical kinetics



$$\sum_{i=1}^{N} \nu_{ji}^{\mathrm{r}} X_{i} \xrightarrow{k_{j}} \sum_{i=1}^{N} \nu_{ji}^{\mathrm{p}} X_{i}, \qquad j = 1, 2, \dots, q$$

Notation:

• Well mixed reactor: N chemical species, q reactions (R_1, \ldots, R_q)

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- ► $\mathbf{X} = [X_1, \dots, X_N]$, $X_i(t) =$ number of molecules, $i = 1, \dots, N$
- α_j(x) is propensity function of reaction R_j, j = 1,..., q
 (α_j(x) dt = probability that one reaction R_j occurs in [t, t + dt), given X(t) = x)
- ν_{ji} = ν^p_{ji} − ν^r_{ji}, change of X_i during reaction R_j,
 ν_i = [ν_{i1},..., ν_{iN}]

•
$$p(\mathbf{x}, t) = \text{probability that } \mathbf{X}(t) =$$

Stochastic equations for chemical kinetics

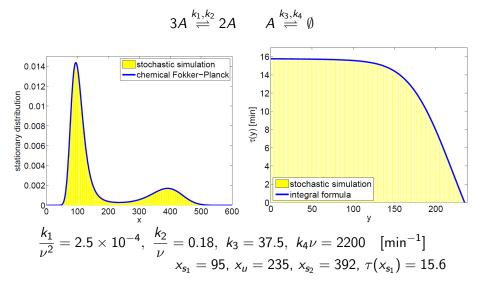


Chemical master equation (CME) – exact $\frac{\partial}{\partial t} p(\mathbf{x}, t) = \sum_{j=1}^{q} \left[\alpha_j (\mathbf{x} - \boldsymbol{\nu}_j) p(\mathbf{x} - \boldsymbol{\nu}_j, t) - \alpha_j(\mathbf{x}) p(\mathbf{x}, t) \right]$ Chemical Langevin equation (CLE) - approximate $\mathrm{d}X_i = f_i(\mathbf{X}(t))\,\mathrm{d}t + \sum_{j=1}^{7} d_{ji}(\mathbf{X}(t))\,\mathrm{d}W_j$ where $f_i(\mathbf{X}(t)) = \sum_{j=1}^{7} \nu_{ji} \alpha_j(\mathbf{X}(t)), \quad d_{ji}(\mathbf{X}(t)) = \nu_{ji} \sqrt{\alpha_j(\mathbf{X}(t))}$ Chemical Fokker-Planck equation (CFP) ⇔ CLE $\frac{\partial}{\partial t}\rho(\mathbf{x},t) = \frac{1}{2}\sum_{i=1}^{N}\sum_{k=1}^{N}\frac{\partial^{2}}{\partial x_{i}\partial x_{k}}\left|\left(\sum_{i=1}^{q}d_{ji}(\mathbf{x})d_{jk}(\mathbf{x})\right)\rho(\mathbf{x},t)\right|$ $-\sum_{i=1}^{N}\frac{\partial}{\partial x_{i}}[f_{i}(\mathbf{x})\rho(\mathbf{x},t)]$

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Schlögl system (revisited)



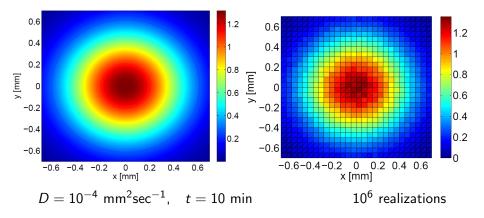


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Diffusion – position jump process



$$X(t + dt) = X(t) + \sqrt{2D} dW_x$$
$$Y(t + dt) = Y(t) + \sqrt{2D} dW_y$$
$$Z(t + dt) = Z(t) + \sqrt{2D} dW_z$$



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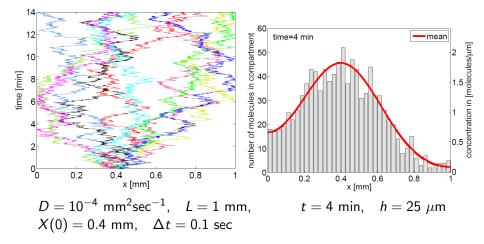
Reflecting boundary condition



Simulation algorithm $X(0) = x_0, \ \Delta t > 0 \text{ small}$ (a7) $\xi \sim N(0, 1)$ (b7) $X(t + \Delta t) = X(t) + \sqrt{2D\Delta t}\xi$ (c7) If $X(t + \Delta t) < 0$ then $X(t + \Delta t) = -X(t) - \sqrt{2D\Delta t}\xi$ If $X(t + \Delta t) > L$ then $X(t + \Delta t) = 2L - X(t) - \sqrt{2D\Delta t}\xi$ Set $t := t + \Delta t$ and go to (a7)

Reflecting boundary condition





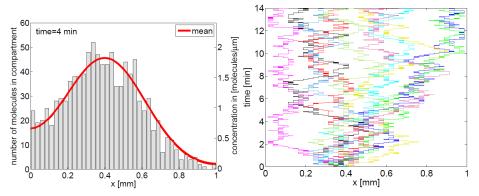
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Compartment based model

 $t = 4 \min$

 $a(0) = \delta_{0.4}(x)$





 $K = 40, h = 1/K, d = D/h^2 = 0.16 \text{ sec}^{-1}$

 $N_{\rm mol} = 1000, A_{16}(0) = A_{17}(0) = 500$

10 realizations 1 molecule

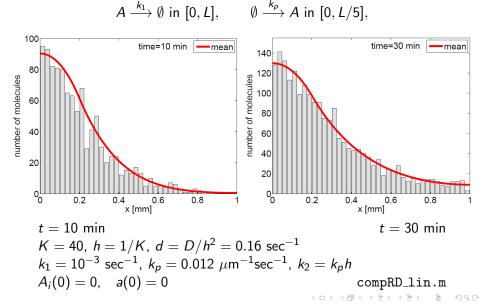
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Compartment based reaction-diffusion

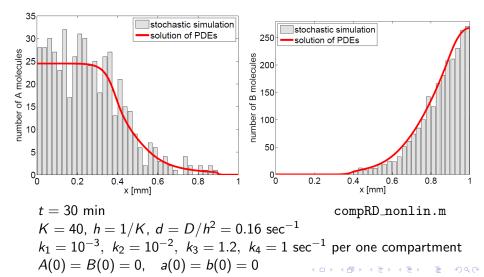




Compartment based reaction-diffusion



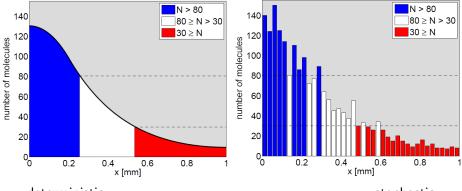
$$2A \xrightarrow{k_1} \emptyset, \quad A + B \xrightarrow{k_2} \emptyset \quad \text{in } [0, L],$$
$$\emptyset \xrightarrow{k_3} A \text{ in } [0, 9L/10], \qquad \emptyset \xrightarrow{k_4} B \text{ in } [2L/5, K],$$



Pattern formation – French flag



 $A \xrightarrow{k_1} \emptyset$ in [0, L], $\emptyset \xrightarrow{k_p} A$ in [0, L/5], +diffusion



deterministic

stochastic

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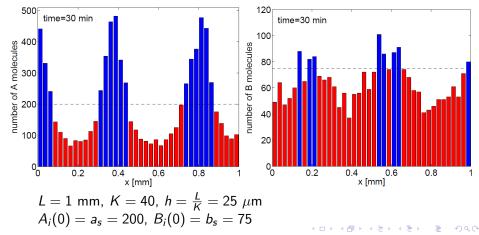


Pattern formation – Turing instability

Schnakenberg system

$$2A + B \xrightarrow{k_1} 3A \qquad \emptyset \stackrel{k_2,k_3}{\rightleftharpoons} A \qquad \emptyset \stackrel{k_4}{\rightleftharpoons} B$$

+ diffusion $D_A = 10^{-5}$, $D_B = 10^{-3} \text{ [mm}^2 \text{sec}^{-1]}$



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Thank you for your attention

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Based on lecture notes of Radek Erban http://www.maths.ox.ac.uk/courses/course/19651/material http://www.maths.ox.ac.uk/cmb/education

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