## 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

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## Outline

- 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 $r<A(t) k \Delta t$ then $A(t+\Delta t)=A(t)-1$;

$$
\text { else } A(t+\Delta t)=A(t)
$$

## Naive SSA: degradation



## 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 $k$ |
| ---: | ---: | :---: | :--- |
| 0 | $\emptyset \xrightarrow{k} A$ | $k \nu$ | $\mathrm{~m}^{-3} \mathrm{sec}^{-1}$ |
| 1 | $A \xrightarrow{k} \emptyset$ | $A(t) k$ | $\mathrm{sec}^{-1}$ |
| 2 | $A+B \xrightarrow{k} \emptyset$ | $A(t) B(t) k / \nu$ | $\mathrm{m}^{3} \mathrm{sec}^{-1}$ |
| 2 | $2 A \xrightarrow{k} \emptyset$ | $A(t)(A(t)-1) k / \nu$ | $\mathrm{m}^{3} \mathrm{sec}^{-1}$ |
| 3 | $A+B+C \xrightarrow{k} \emptyset$ | $A(t) B(t) C(t) k / \nu^{2}$ | $\mathrm{~m}^{6} \mathrm{sec}^{-1}$ |
| 3 | $2 A+B \xrightarrow{k} \emptyset$ | $A(t)(A(t)-1) B(t) k / \nu^{2}$ | $\mathrm{~m}^{6} \mathrm{sec}^{-1}$ |
| 3 | $3 A \xrightarrow{k} \emptyset$ | $A(t)(A(t)-1)(A(t)-2) k / \nu^{2}$ | $\mathrm{~m}^{6} \mathrm{sec}^{-1}$ |

## System with two species

$$
2 A \xrightarrow{k_{1}} \emptyset, \quad A+B \xrightarrow{k_{2}} \emptyset, \quad \emptyset \xrightarrow{k_{3}} A, \quad \emptyset \xrightarrow{k_{4}} B,
$$

Gillespie SSA:
(a4) Generate two random numbers: $r_{1}, r_{2} \sim U(0,1)$
(b4) Compute propensities:

$$
\begin{aligned}
& \alpha_{1}(t)=k_{1} A(t)(A(t)-1), \alpha_{2}(t)=k_{2} A(t) B(t), \\
& \alpha_{3}=k_{3}, \alpha_{4}=k_{4}, \text { and } \alpha_{0}=\alpha_{1}(t)+\alpha_{2}(t)+\alpha_{3}+\alpha_{4}
\end{aligned}
$$

(c4) Next reaction time $\tau=\frac{1}{\alpha_{0}} \ln \left[\frac{1}{r_{1}}\right]$
(d4) Update the numbers of molecules:

|  | $r_{2} \in I_{1}$ | $r_{2} \in I_{2}$ | $r_{2} \in I_{3}$ | $r_{2} \in I_{4}$ |
| :---: | :---: | :---: | :---: | :---: |
| $A(t+\tau)$ | $A(t)-2$ | $A(t)-1$ | $A(t)+1$ | $A(t)$ |
| $B(t+\tau)$ | $B(t)$ | $B(t)-1$ | $B(t)$ | $B(t)+$ |
|  |  |  |  |  |

Set $t:=t+\tau$ and go to (a4)

## System with two species

Trajectories



$$
\begin{aligned}
& A(0)=B(0)=0, k_{1}=10^{-3}, k_{2}=10^{-2}, k_{3}=1.2, k_{4}=1 \mathrm{sec}^{-1} \\
& A_{s}=9.6, B_{s}=12.2 \\
& a_{s}=10, b_{s}=10
\end{aligned}
$$

## System with two species



## General Gillespie SSA

## Notation

$q \ldots$ number of chemical reactions
$\alpha_{j}(t) \ldots$ propensity function of $j$-th reaction, $j=1,2, \ldots, q$
$\alpha_{j}(t) \mathrm{d} t=$ probability that $j$-th reaction occurs in $[t, t+\mathrm{d} t)$

## 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

$$
r_{2} \geq \frac{1}{\alpha_{0}} \sum_{i=1}^{j-1} \alpha_{i}(t) \quad \text { and } \quad r_{2}<\frac{1}{\alpha_{0}} \sum_{i=1}^{j} \alpha_{i}(t)
$$

(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

$$
3 A \stackrel{k_{1}, k_{2}}{\rightleftharpoons} 2 A \quad A \stackrel{k_{3}, k_{4}}{\rightleftharpoons} \emptyset
$$

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

$$
\frac{d a}{d t}=-k_{1} a^{3}+k_{2} a^{2}-k_{3} a+k_{4}
$$

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

$$
\frac{d \bar{A}}{d t}=-\frac{k_{1}}{\nu^{2}} \bar{A}^{3}+\frac{k_{2}}{\nu} \bar{A}^{2}-k_{3} \bar{A}+k_{4} \nu
$$

## Schlögl system


$\frac{k_{1}}{\nu^{2}}=2.5 \times 10^{-4}, \frac{k_{2}}{\nu}=0.18, k_{3}=37.5, k_{4} \nu=2200 \quad\left[\mathrm{~min}^{-1}\right]$

## Schlögl system



## Schlögl system



$$
\frac{k_{1}}{\nu^{2}}=2.5 \times 10^{-4}, \frac{k_{2}}{\nu}=0.18, k_{3}=37.5, k_{4} \nu=2200 \quad\left[\min ^{-1}\right]
$$

## Schlögl system


$\frac{k_{1}}{\nu^{2}}=2.5 \times 10^{-4}, \frac{k_{2}}{\nu}=0.18, k_{3}=37.5, k_{4} \nu=2200 \quad\left[\mathrm{~min}^{-1}\right]$

## Self－induced stochastic resonance

Schnakenberg system

$$
2 A+B \xrightarrow{k_{1}} 3 A \quad \emptyset \stackrel{k_{2}, k_{3}}{\rightleftharpoons} A \quad \emptyset \xrightarrow{k_{4}} B
$$

Concentration：

$$
\begin{aligned}
& \frac{d a}{d t}=k_{1} a^{2} b+k_{2}-k_{3} a \\
& \frac{d b}{d t}=-k_{1} a^{2} b+k_{4}
\end{aligned}
$$

Average numbers of molecules：

$$
\begin{aligned}
\frac{d \bar{A}}{d t} & =\frac{k_{1}}{\nu^{2}} \bar{A}^{2} \bar{B}+k_{2} \nu-k_{3} \bar{A} \\
\frac{d \bar{B}}{d t} & =-\frac{k_{1}}{\nu^{2}} \bar{A}^{2} \bar{B}+k_{4}
\end{aligned}
$$

## Schnakenberg system




$$
\begin{aligned}
& \frac{k_{1}}{\nu^{2}}=4 \times 10^{-5}, k_{2} \nu=50, k_{3}=10, k_{4} \nu=25 \quad\left[\mathrm{sec}^{-1}\right] \\
& A(0)=10, \quad B(0)=10
\end{aligned}
$$

## Schnakenberg system




$$
\begin{aligned}
& \frac{k_{1}}{\nu^{2}}=4 \times 10^{-5}, k_{2} \nu=50, k_{3}=10, \quad k_{4} \nu=100 \quad\left[\mathrm{sec}^{-1}\right] \\
& A(0)=10, \quad B(0)=10
\end{aligned}
$$

## Schnakenberg system




## Stochastic differential equations (SDE)

$$
X(t+\mathrm{d} t)=X(t)+f(X(t), t) \mathrm{d} t+g(X(t), t) \mathrm{d} W
$$

$\mathrm{d} W \ldots$ white noise, $\mathrm{d} W \approx \sqrt{\Delta t} \xi$, with $\xi \sim N(0,1)$
Simulation algorithm
$X(0)=x_{0}, \Delta t>0$ 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:

$$
X(t+\mathrm{d} t)=X(t)+\mathrm{d} W
$$

$$
\begin{aligned}
& X(t+\mathrm{d} t)=X(t)+\mathrm{d} W_{1} \\
& Y(t+\mathrm{d} t)=Y(t)+\mathrm{d} W_{2}
\end{aligned}
$$




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

Trajectories:
$X(t+\mathrm{d} t)=X(t)+\mathrm{d} t+\mathrm{d} W$


## Example 3: two favourable states

Trajectories:

$$
f(x, t)=-k_{1} x^{3}+k_{2} x^{2}-k_{3} x+k_{4}, \quad g(x, t)=k_{5}
$$

$$
k_{1}=10^{-3}, k_{2}=0.75, k_{3}=165, k_{4}=10^{4}, k_{5}=200,
$$

$$
X(t+\mathrm{d} t)=X(t)+f(X(t), t) \mathrm{d} t+g(X(t), t) \mathrm{d} W
$$



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

Stationary probability distribution: $X(t+\mathrm{d} t)=X(t)+\mathrm{d} W$


## Example 3: two favourable states (revisited)

Stationary probability distribution:

$$
\begin{aligned}
& f(x, t)=-k_{1} x^{3}+k_{2} x^{2}-k_{3} x+k_{4}, \quad g(x, t)=k_{5} \\
& k_{1}=10^{-3}, k_{2}=0.75, k_{3}=165, k_{4}=10^{4}, k_{5}=200, \\
& X(t+d t)=X(t)+f(X(t), t) \mathrm{d} t+g(X(t), t) \mathrm{d} W
\end{aligned}
$$



## Example 3: two favourable states (revisited)

Mean exit time:

$$
\begin{aligned}
& \tau_{\mathrm{sim}}=64.7 \\
& \tau_{x_{s_{1}}}=59.45
\end{aligned}
$$




## Stochastic equations for chemical kinetics

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

Notation:

- Well mixed reactor: $N$ chemical species, $q$ reactions $\left(R_{1}, \ldots, R_{q}\right)$
- $\mathbf{X}=\left[X_{1}, \ldots, X_{N}\right], X_{i}(t)=$ number of molecules, $i=1, \ldots, N$
- $\alpha_{j}(x)$ is propensity function of reaction $R_{j}, j=1, \ldots, q$ $\left(\alpha_{j}(\mathbf{x}) \mathrm{d} t=\right.$ probability that one reaction $R_{j}$ occurs in $[t, t+\mathrm{d} t)$, given $\mathbf{X}(t)=\mathbf{x})$
- $\nu_{j i}=\nu_{j i}^{\mathrm{p}}-\nu_{j i}^{\mathrm{r}}$, change of $X_{i}$ during reaction $R_{j}$,
- $\boldsymbol{\nu}_{j}=\left[\nu_{j 1}, \ldots, \nu_{j N}\right]$
- $p(\mathbf{x}, t)=$ probability that $\mathbf{X}(t)=\mathbf{x}$


## Stochastic equations for chemical kinetics

Chemical master equation (CME) - exact
$\frac{\partial}{\partial t} p(\mathbf{x}, t)=\sum_{j=1}^{q}\left[\alpha_{j}\left(\mathbf{x}-\nu_{j}\right) p\left(\mathbf{x}-\nu_{j}, t\right)-\alpha_{j}(\mathbf{x}) p(\mathbf{x}, t)\right]$
Chemical Langevin equation (CLE) - approximate
$\mathrm{d} X_{i}=f_{i}(\mathbf{X}(t)) \mathrm{d} t+\sum_{j=1}^{q} d_{j i}(\mathbf{X}(t)) \mathrm{d} W_{j}$
where $f_{i}(\mathbf{X}(t))=\sum_{j=1}^{q} \nu_{j i} \alpha_{j}(\mathbf{X}(t)), \quad d_{j i}(\mathbf{X}(t))=\nu_{j i} \sqrt{\alpha_{j}(\mathbf{X}(t))}$
Chemical Fokker-Planck equation (CFP) $\Leftrightarrow$ CLE

$$
\begin{aligned}
& \frac{\partial}{\partial t} p(\mathbf{x}, t)=\frac{1}{2} \sum_{i=1}^{N} \sum_{k=1}^{N} \frac{\partial^{2}}{\partial x_{i} \partial x_{k}}\left[\left(\sum_{j=1}^{q} d_{j i}(\mathbf{x}) d_{j k}(\mathbf{x})\right) p(\mathbf{x}, t)\right] \\
&-\sum_{i=1}^{N} \frac{\partial}{\partial x_{i}}\left[f_{i}(\mathbf{x}) p(\mathbf{x}, t)\right]
\end{aligned}
$$

## Schlögl system (revisited)



## Diffusion - position jump process

$$
\begin{aligned}
& X(t+\mathrm{d} t)=X(t)+\sqrt{2 D} \mathrm{~d} W_{x} \\
& Y(t+\mathrm{d} t)=Y(t)+\sqrt{2 D} \mathrm{~d} W_{y} \\
& Z(t+\mathrm{d} t)=Z(t)+\sqrt{2 D} \mathrm{~d} W_{z}
\end{aligned}
$$



## Reflecting boundary condition

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

## Reflecting boundary condition



$D=10^{-4} \mathrm{~mm}^{2} \mathrm{sec}^{-1}, \quad L=1 \mathrm{~mm}$,
$t=4 \mathrm{~min}, \quad h=25 \mu \mathrm{~m}$ $X(0)=0.4 \mathrm{~mm}, \quad \Delta t=0.1 \mathrm{sec}$

## Compartment based model

$t=4$ min
$K=40, h=1 / K, d=D / h^{2}=0.16 \mathrm{sec}^{-1}$
$N_{\text {mol }}=1000, A_{16}(0)=A_{17}(0)=500$
$a(0)=\delta_{0.4}(x)$


10 realizations 1 molecule

## Compartment based reaction-diffusion

$$
A \xrightarrow{k_{1}} \emptyset \text { in }[0, L], \quad \emptyset \xrightarrow{k_{p}} A \text { in }[0, L / 5],
$$



$t=10 \mathrm{~min}$

$$
t=30 \mathrm{~min}
$$

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

$$
k_{1}=10^{-3} \sec ^{-1}, k_{p}=0.012 \mu \mathrm{~m}^{-1} \sec ^{-1}, k_{2}=k_{p} h
$$

$$
A_{i}(0)=0, \quad a(0)=0
$$

## Compartment based reaction-diffusion

$$
\begin{aligned}
& 2 A \xrightarrow{k_{1}} \emptyset, \quad A+B \xrightarrow{k_{2}} \emptyset \\
& \emptyset \xrightarrow{k_{3}} A \text { in }[0, L], \\
& {[0,9 L / 10], \quad \emptyset \xrightarrow{k_{4}} B \text { in }[2 L / 5, K], }
\end{aligned}
$$



$t=30 \mathrm{~min}$
compRD_nonlin.m
$K=40, h=1 / K, d=D / h^{2}=0.16 \mathrm{sec}^{-1}$ $k_{1}=10^{-3}, k_{2}=10^{-2}, k_{3}=1.2, k_{4}=1 \mathrm{sec}^{-1}$ per one compartment $A(0)=B(0)=0, \quad a(0)=b(0)=0$

## Pattern formation - French flag

$A \xrightarrow{k_{1}} \emptyset$ in $[0, L], \quad \emptyset \xrightarrow{k_{p}} A$ in $[0, L / 5], \quad+$ diffusion

deterministic

stochastic

## Pattern formation - Turing instability

Schnakenberg system

$$
2 A+B \xrightarrow{k_{1}} 3 A \quad \emptyset \stackrel{k_{2}, k_{3}}{\rightleftharpoons} A \quad \emptyset \xrightarrow{k_{4}} B
$$

+ diffusion $D_{A}=10^{-5}, D_{B}=10^{-3}\left[\mathrm{~mm}^{2} \mathrm{sec}^{-1}\right]$



$$
\begin{aligned}
& L=1 \mathrm{~mm}, K=40, h=\frac{L}{K}=25 \mu \mathrm{~m} \\
& A_{i}(0)=a_{s}=200, B_{i}(0)=b_{s}=75
\end{aligned}
$$

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# Thank you for your attention 

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