Class 12

2018/06/14, 6/21

- Reaction-Diffusion Models and very basic methods for their numerical solutions
 - 12.1 How to derive reaction-diffusion model (for population dynamics)
 - 12.2 Discretization of diffusion equation
 - 12.3 Stability condition for discretization (skip)
 - 12.4 Initial boundary value problems for diffusion equation
 - 12.5 Initial boundary value problems for reaction-diffusion equation

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Step by step understanding of population dynamics in space

- **Local population dynamics (without space)**

 - Equation of continuity (mass conservation)
 Random movement of particles (diffusion)



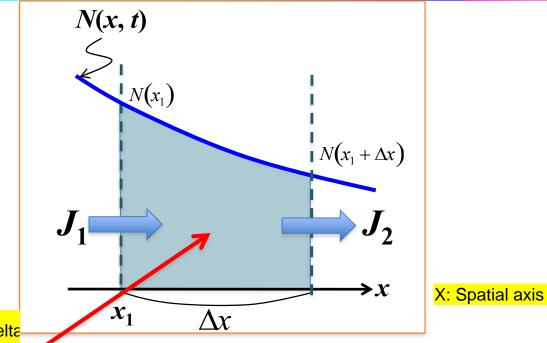
Population dynamics in space: reaction-diffusion model

12.1 Local population dynamics (without space)

$$\frac{dN(t)}{dt} = \frac{\text{(reproduction)} - \text{(mortality)}}{\text{Population}}$$
Population
Growth Rate

Local Dynamics

Equation of continuity (mass conservation) 12.1



Delta x: aprox 0N(x1, t + delta t)* (Delta

$$N(x_1,t_1+\Delta t)\cdot\Delta x-N(x_1,t_1)\cdot\Delta x \approx J_1\cdot\Delta t - J_2\cdot\Delta t$$

Temporal changes in total number of organisms

in grayed region

Outflux

$$= \left[J(x_1) - J(x_1 + \Delta x) \right] \cdot \Delta t$$

$$\therefore \frac{\partial N}{\partial t} = -\frac{\partial J}{\partial x}$$
 (Equation of continuity)

Influx

12.1 Random movement of particles (diffusion)

How to derive the flux due to diffusion

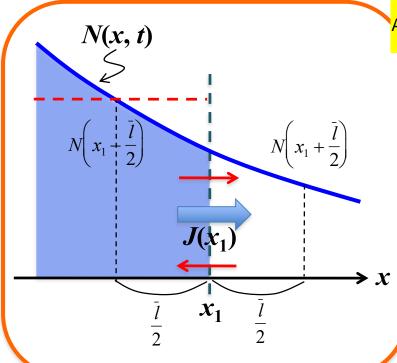
 \overline{l} The 'mean free path' of a organism, defined as the average distance of movement between successive encounter events

 $\overline{\Lambda t}$ The average time between successive encounter events

$$\overline{v} \equiv \overline{l}/\overline{\Delta t}$$

The average rate of movement in short timescale when the effect of encounters can be neglected





Assumptions:Random movement of particles.No interaction between

$$J(x_1) \cdot \overline{\Delta t} \approx \alpha \cdot \overline{v} \cdot \overline{\Delta t} \cdot \left[N\left(x_1 - \frac{\overline{l}}{2}\right) - N\left(x_1 + \frac{\overline{l}}{2}\right) \right]$$

$$= -\alpha \cdot \overline{v} \cdot \overline{\Delta t} \cdot \frac{\partial N}{\partial x} \bigg|_{x=x_1} \cdot \overline{l}$$

alpha: fraction of particles that moves right

$$\therefore J(x_1) = -\alpha \cdot \bar{v} \cdot \bar{l} \cdot \frac{\partial N}{\partial x} \equiv -\underline{D} \frac{\partial N}{\partial x}$$

Diffusion Coefficient

12.1 Population dynamics in space: reaction-diffusion model

Local population dynamics (without space)

$$\frac{dN(t)}{dt} = (reproduction) - (mortality)$$



• Equation of continuity (mass conservation)

$$\frac{\partial N}{\partial t} = -\frac{\partial J}{\partial x}$$

Random movement of particles (diffusion)

$$J = -D \frac{\partial N}{\partial x}$$



$$\frac{\partial N(x,t)}{\partial t} = \text{(reproduction)} - \text{(mortality)} + D\frac{\partial^2 N}{\partial x^2}$$
Population

Growth Rate

Local Dynamics

Diffusion

12.2 Discretization of diffusion equation

$$\frac{\partial N(x,t)}{\partial t} = D \frac{\partial^2 N}{\partial x^2}$$
 (1)

Diffusion Equation

Discretization by approximation with Taylor expansion

$$N(x, t + \Delta t) \approx N(x, t) + \frac{\partial N}{\partial t} \Delta t$$
 (2)

$$N(x + \Delta x, t) \approx N(x, t) + \frac{\partial N}{\partial x} \Delta x + \frac{1}{2} \frac{\partial^2 N}{\partial x^2} (\Delta x)^2$$
 (3)

$$N(x - \Delta x, t) \approx N(x, t) + \frac{\partial N}{\partial x} \left(-\Delta x \right) + \frac{1}{2} \frac{\partial^2 N}{\partial x^2} \left(\Delta x \right)^2$$
 (4)

12.2 Discretization of diffusion equation

From (2),

$$\frac{\partial N}{\partial t} \approx \frac{N(x, t + \Delta t) - N(x, t)}{\Delta t} \tag{5}$$

From (3)+(4),

$$\frac{\partial^2 N}{\partial x^2} \approx \frac{N(x + \Delta x, t) - 2N(x, t) + N(x - \Delta x, t)}{\left(\Delta x\right)^2} \tag{6}$$

Substituting (5) & (6) into (1) gives

$$N(x,t+\Delta t) \approx N(x,t) + D\Delta t \frac{N(x+\Delta x,t) - 2N(x,t) + N(x-\Delta x,t)}{\left(\Delta x\right)^2}$$
(7)

12.2 Discretization of diffusion equation

From (2),

$$N(x,t+\Delta t) \approx N(x,t) + D\Delta t \frac{N(x+\Delta x,t) - 2N(x,t) + N(x-\Delta x,t)}{\left(\Delta x\right)^2}$$
(7)

With the following notation,

$$x_{i+1} - x_i = \Delta x$$
$$t_{j+1} - t_j = \Delta t$$

Let n(i,j) as the approximation of $N(x_i,t_i)$, and then we have

$$n(i,j+1) = n(i,j) + D\Delta t \frac{n(i+1,j) - 2n(i,j) + n(i-1,j)}{(\Delta x)^2}$$
 (8)

Before thinking about stability condition, we need to think about analytical solution for the following initial boundary value problem for diffusion equation.

$$\frac{\partial N(x,t)}{\partial t} = D \frac{\partial^2 N}{\partial x^2}$$
 (1)
$$0 < x < L, \quad t > 0$$

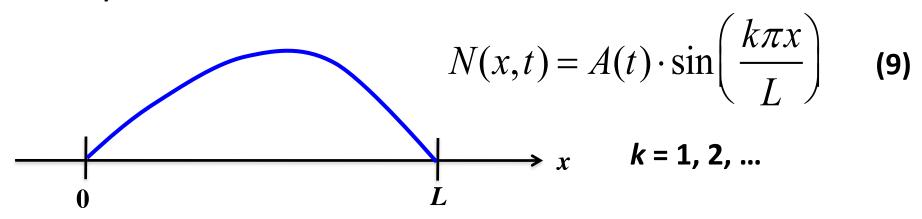
$$0 < x < L$$
, $t > 0$

(Spatial and Temporal Domain)

[Initial Conditions]
$$N(x,0) = N_0(x)$$

[Boundary Conditions]
$$N(0,t) = N(L,t) = 0$$

Solution should have the following form to satisfy the boundary conditions



By substituting (9) into (1), we have

$$\frac{dA(t)}{dt} = -D\left(\frac{k\pi}{L}\right)^2 \cdot A(t)$$

$$\therefore A(t) \rightarrow 0$$
 as $t \rightarrow +\infty$ [All solutions converge to zero.]

This characteristics of analytical solution of (1) should be kept in the numerical solution, constructed by (8).

$$n(i,j+1) = n(i,j) + D\Delta t \frac{n(i+1,j) - 2n(i,j) + n(i-1,j)}{(\Delta x)^2}$$
 (8)

$$n(i,j+1) = n(i,j) + D\Delta t \frac{n(i+1,j) - 2n(i,j) + n(i-1,j)}{(\Delta x)^2}$$
 (8)

We define

$$\Delta x = \frac{L}{M}$$
 $x_i = i\Delta x = \frac{i}{M}L$ $i = 0, 1, 2, ..., M$

Then, we can set the following numerical solution

$$n(i,j) = A_j \sin\left(\frac{i \cdot k\pi}{M}\right) \tag{10}$$

,which satisfies the boundary conditions

$$n(0,j) = n(M,j) = 0$$

Substituting (10) into (8) gives

$$A_{j+1} = A_j \left\{ 1 - \frac{4D\Delta t}{(\Delta x)^2} \sin^2 \frac{k\pi}{2M} \right\}$$

$$\sin \alpha + \sin \beta = 2\sin \frac{\alpha + \beta}{2}\cos \frac{\alpha - \beta}{2}$$
$$\cos 2\alpha = 1 - 2\sin^2 \alpha$$

$$A_{j+1} = A_j \left\{ 1 - \frac{4D\Delta t}{\left(\Delta x\right)^2} \sin^2 \frac{k\pi}{2M} \right\}$$

Then, the numerical solution is

$$n(i,j) = A_j \sin\left(\frac{i \cdot k\pi}{M}\right) = A_0 \left\{ 1 - \frac{4D\Delta t}{\left(\Delta x\right)^2} \sin^2\frac{k\pi}{2M} \right\}^{J} \cdot \sin\left(\frac{i \cdot k\pi}{M}\right)$$

Therefore, to keep the characteristics of convergence,

$$\lim_{j \to +\infty} n(i,j) = 0 \Leftrightarrow \left| 1 - \frac{4D\Delta t}{\left(\Delta x\right)^2} \sin^2 \frac{k\pi}{2M} \right| < 1$$

It gives

$$\frac{D\Delta t}{\left(\Delta x\right)^2} < \frac{1}{2} \Leftrightarrow \Delta t < \frac{1}{2D} \left(\Delta x\right)^2$$
 [stability condition for discretization]

$$\frac{D\Delta t}{\left(\Delta x\right)^2} < \frac{1}{2} \Leftrightarrow \Delta t < \frac{1}{2D} \left(\Delta x\right)^2$$
 [stability condition for discretization]

Time Interval (Δt) should be smaller with higher spatial resolution (smaller Δx) and with higher diffusion coefficient (D).

WHY??

$$N(x,t+\Delta t) \approx N(x,t) + D\Delta t \frac{N(x+\Delta x,t) - 2N(x,t) + N(x-\Delta x,t)}{\left(\Delta x\right)^2}$$
(7)

Let's go back to the original diffusion equation

$$\frac{\partial N(x,t)}{\partial t} = D \frac{\partial^2 N}{\partial x^2}$$
 (1)

$$0 < x < L, \quad t > 0$$

[Initial Conditions]
$$N(x,0) = N_0(x)$$

(No needs to explain)

[Boundary Conditions]

boundary doesn't change with timeParticles 'die' on boundar

[Dirichlet boundary condition] (= fixed boundary condition)

$$N(0,t) = a, N(L,t) = b$$
 (e.g. $N(0,t) = N(L,t) = 0$)

[Neumann boundary condition]

$$\frac{\partial}{\partial x}N(0,t) = a, \frac{\partial}{\partial x}N(L,t) = b$$

Especially when a = b = 0, it is called 'zero flux boundary' or 'reflecting boundary condition'

How to translate boundary conditions to numerical algorism?

[Boundary Conditions]

[Dirichlet boundary condition] (= fixed boundary condition)

$$N(0,t) = N(L,t) = 0$$

$$n(0,j) = n(M,j) = 0$$

$$(= n(-i,j) = n(M+i,j))$$
(12)

[Reflecting boundary condition] (skip)

$$\frac{\partial}{\partial x} N(0,t) = a, \frac{\partial}{\partial x} N(L,t) = b$$

$$\begin{cases}
\frac{n(1,j) - n(-1,j)}{\Delta x} = 2 \cdot a + O\left((\Delta x)^{2}\right) \\
\frac{n(M+1,j) - n(M-1,j)}{\Delta x} = 2 \cdot b + O\left((\Delta x)^{2}\right)
\end{cases}$$
(13)

Remark
$$N(x \pm \Delta x) = N(x,t) \pm N_x(x,t) \cdot \Delta x + (1/2)N_{xx}(x,t) \cdot (\Delta x)^2 + O((\Delta x)^3)$$

Let's numerically solve (1) with <u>fixed boundary</u> condition and following settings

$$\frac{\partial N(x,t)}{\partial t} = D \frac{\partial^2 N}{\partial x^2} \qquad D = 1, \quad 0 < x < L = 10.0, \quad t > 0$$

$$\Delta x = \frac{L}{M} = \frac{10.0}{100} = 0.1$$

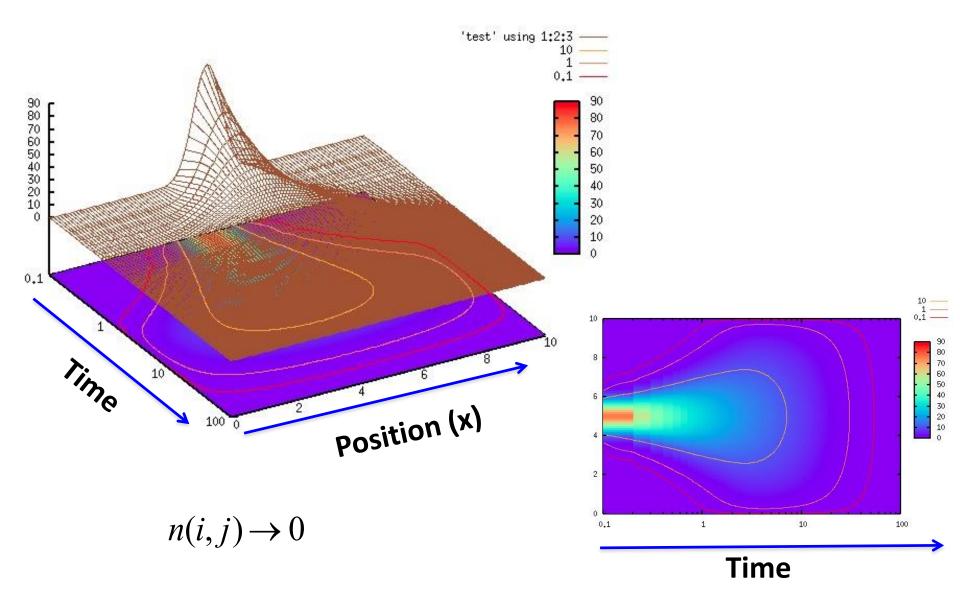
$$\Delta t = \frac{1}{5} \cdot \frac{1}{2D} (\Delta x)^2 = \frac{1}{5} \cdot \frac{1}{2} (0.1)^2 = 1.0e - 3 < \frac{1}{2D} (\Delta x)^2$$

[Initial Conditions]

$$n(i,0) = \begin{cases} 0 & \text{if } i \neq M / 2 = 50 \\ N_0 = 10^3 & \text{if } i = M / 2 = 50 \end{cases}$$

[Boundary Conditions] n(0,j) = n(M,j) = 0.0 = n(-1,j) = n(M+1,j)

Let's numerically solve (1) with fixed boundary condition



Let's numerically solve (1) with <u>reflecting boundary</u> condition and following settings

$$\frac{\partial N(x,t)}{\partial t} = D \frac{\partial^2 N}{\partial x^2} \qquad D = 1, \quad 0 < x < L = 10.0, \quad t > 0$$

$$\Delta x = \frac{L}{M} = \frac{10.0}{100} = 0.1$$

$$\Delta t = \frac{1}{5} \cdot \frac{1}{2D} (\Delta x)^2 = \frac{1}{5} \cdot \frac{1}{2} (0.1)^2 = 1.0e - 3 < \frac{1}{2D} (\Delta x)^2$$

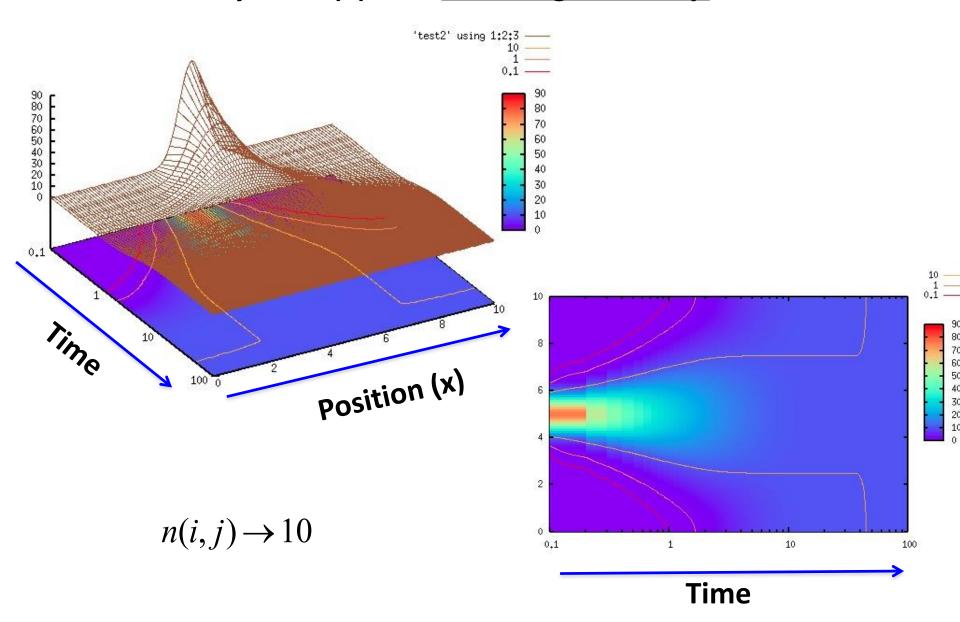
[Initial Conditions]

$$n(i,0) = \begin{cases} 0 & \text{if } i \neq M / 2 = 50 \\ N_0 = 10^3 & \text{if } i = M / 2 = 50 \end{cases}$$

[Boundary Conditions]

$$n(-1, j) = n(1, j), n(M + 1, j) = n(M - 1, j)$$

Let's numerically solve (1) with reflecting boundary condition



12.4 How to save the results into text file for graphics

Tips for R

```
library(scatterplot3d)
test data<-read.csv("test contour.csv")
#A simple 3d scatter plot
scatterplot3d(test data)
#Preparation for contour plot
x<-0:2
y<-0:4
#need to prepare z data as the matrix
test data2<-read.csv("test contour2.csv")
class(test_data2)
test data2<-as.matrix(test data2) #change data.frame to matrix
contour(x, y, test_data2)
image(x,y,test_data2, col=terrain.colors(100))
```

$$\frac{\partial N(x,t)}{\partial t} = f(N) + D \frac{\partial^2 N}{\partial x^2}$$
 (14) $0 < x < L, \quad t > 0$ (Spatial and

Temporal Domain)

[Initial Conditions]
$$N(x,0) = N_0(x)$$

[Boundary Conditions]
$$N(0,t) = N(L,t) = 0$$

We can use the following discretization for space

$$\frac{\partial^2 N}{\partial x^2} \approx \frac{N(x + \Delta x, t) - 2N(x, t) + N(x - \Delta x, t)}{\left(\Delta x\right)^2} \tag{6}$$

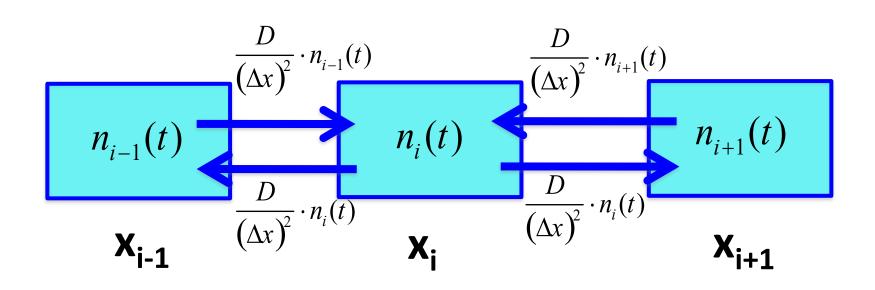
$$x_{i+1} - x_i = \Delta x$$

Let $n_i(t)$ as the approximation of $N(x_i, t)$, and then we have

$$\frac{dn_i(t)}{dt} = f(n_i) + \frac{D}{\left(Dx\right)^2} \left[n_{i+1}(t) - 2n_i(t) + n_{i-1}(t)\right] \quad i = 1, 2, \dots \text{(15)}$$

Now, we convert a reaction-diffusion model (PDE) into a set of ODE.

Then you can use any discretization methods in ODE (e.g. 4th order Runge-Kutta) for numerically solving (15).



Let's numerically solve the following exponential growth model.

$$\frac{\partial N(x,t)}{\partial t} = rN(x,t) + D\frac{\partial^2 N(x,t)}{\partial x^2} \quad D = 1, \quad 0 < x < L = 10.0, \quad t > 0$$

[Fixed Boundary Conditions]

$$N(0,t) = 0, N(L,t) = 0$$

[Initial Conditions]

$$n(i,0) = \begin{cases} 0 & \text{if } i \neq M / 2 = 50 \\ N_0 > 0 & \text{if } i = M / 2 = 50 \end{cases}$$

Confirm that population goes extinct if r is so small that the following condition is satisfied:

$$L < \pi \sqrt{\frac{D}{r}}$$

12.5 Final Homework

Let's numerically solve the following exponential growth model.

$$\frac{\partial N(x,t)}{\partial t} = rN(x,t) + D\frac{\partial^2 N(x,t)}{\partial x^2} \quad D = 1, \quad 0 < x < L = 10.0, \quad t > 0$$

[Fixed Boundary Conditions]

$$N(0,t) = 0, N(L,t) = 0$$
 [Initial Conditions]
$$n(i,0) = \begin{cases} 0 & \text{if } i \neq M \ / \ 2 = 50 \\ N_0 > 0 & \text{if } i = M \ / \ 2 = 50 \end{cases}$$
 [Spatial discritization]
$$\Delta x = \frac{L}{M} = \frac{10.0}{100} = 0.1$$

Plot the spatio-temporal dynamics of the population in two cases: (1) leading to extinction with small r and (2) leading to exponential growth with large r.