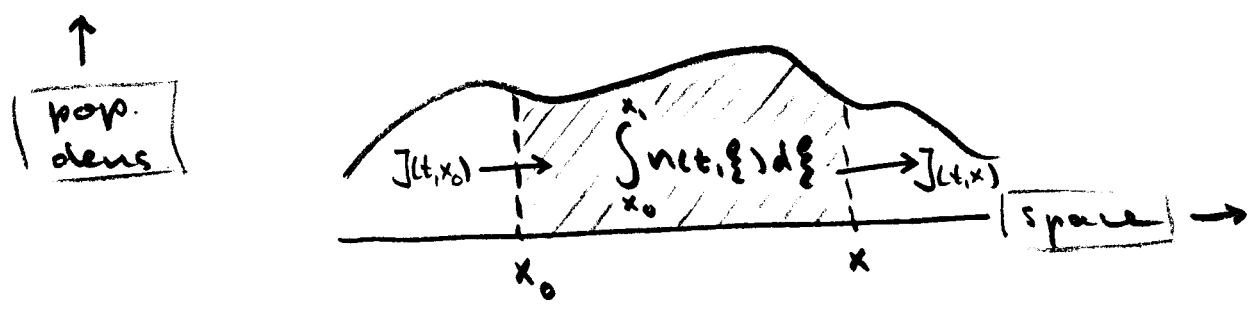


# Movement in space



$n(t, \xi)$  pop. dens. at time  $t$  and location  $\xi$ .

$\int_{x_0}^x n(t, \xi) d\xi$  number of individuals at time  $t$  and between locations  $x_0$  and  $x$

$J(t, \xi)$  net flux of individuals at time  $t$  and location  $\xi$ .

$$\Rightarrow \partial_t \int_{x_0}^x n(t, \xi) d\xi = J(t, x_0) - J(t, x)$$

Differentiation w.r.t.  $x$  gives

$$\textcircled{1} \quad \left| \partial_t n(t, x) = -\partial_x J(t, x) \right|$$

Question is what  $J(t, x)$  looks like.

2

The flux  $J(t, x)$  is a pop-level thing, namely, the number of individuals per unit of time

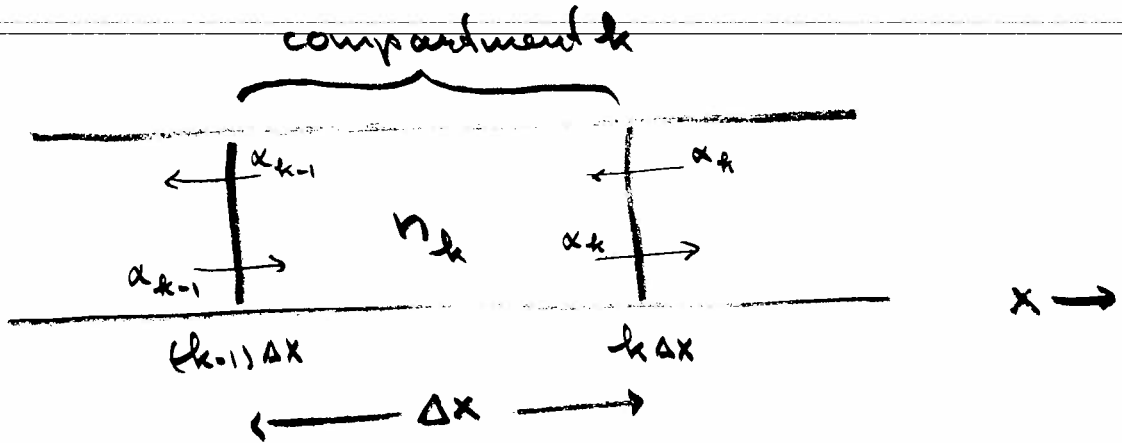
moving through  $x$  from the left minus the number of individuals moving through  $x$  from the right.

How do we connect  $J(t, x)$  to ind-level phenomena.

First we make this connection for randomly moving individuals. Later we extend this to other modes of movement such as advection and taxis.

# Random movement

## Compartment model



i-states |  $N_k$  an individual in comp.  $k \in \mathbb{Z}$ .

i-process |  $N_k \xrightleftharpoons[\alpha_k]{\alpha_{k+1}} N_{k+1}$  (movement between comp.  $k$  and  $k+1$ )

p-state |  $n_k$  number of individuals in compartment  $k \in \mathbb{Z}$ .

Define the net flux between compartments  $k$  and  $k+1$ :

(2)

$$J_k(t) := \alpha_{k+1}(t) n_{k+1}(t) - \alpha_k(t) n_k(t)$$

4

The idea now is to let comp. size  $\Delta x$  go to zero to get a continuous description of space.

As  $\Delta x$  goes to zero, the  $\alpha_x$  should go to infinity, because the prob. per unit of time of hitting the boundary of a compartment goes to infinity as the compartments become smaller.

As  $\Delta x$  goes to zero, we also want to describe pop. size as a pop density rather than in terms of #individuals per compartment, because in the limit  $\Delta x \rightarrow 0$ , the notion of "compartment" becomes meaningless.

To this end we define the following continuously differentiable interpolating functions:

$$\textcircled{*} \begin{cases} n(t, k\Delta x) := \frac{n_k(t)}{\Delta x} & \forall k \in \mathbb{Z} \\ \alpha(t, k\Delta x) := (\Delta x)^2 \alpha_k(t) & \forall k \in \mathbb{Z} \end{cases}$$

and

$$J(t, k\Delta x) := J_k(t) \quad \forall k \in \mathbb{Z}$$

Then, from  $\textcircled{2}$  on page 3, we get

$$\textcircled{3} \quad J(t, k\Delta x) = -\alpha(t, k\Delta x) \frac{n(t, k\Delta x + \Delta x) - n(t, k\Delta x)}{\Delta x}$$

Letting  $\Delta x \rightarrow 0$  and  $k \rightarrow \infty$  such that  $k\Delta x =: x$  stays constant, we get

$$\textcircled{4} \quad J(t, x) = -\alpha(t, x) \partial_x n(t, x)$$

- This is called "Fick's Law" for randomly moving particles.
- $\alpha(t, x)$  is called the diffusion coefficient.

Substitution of (4) (page 5) into (1) (page 1) gives the diffusion equation:

$$(5) \quad \partial_t n(t, x) = \partial_x (\alpha(t, x) \partial_x n(t, x))$$

which is also called the heat equation (because it also describes the transport of heat in a given material).

If  $\alpha(t, x) = D$  is a constant, we get the (maybe) more familiar form

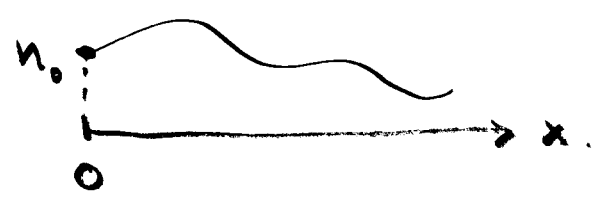
$$(6) \quad \partial_t n(t, x) = D \partial_x^2 n(t, x)$$

From definition  $\otimes$  on page 5 it follows that  $2\alpha(t, x)$  is the expected distance-squared travelled per unit of time per individual.

# Boundary conditions

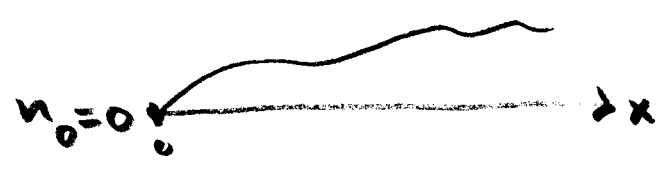
Frequently encountered boundary conditions:

- Constant concentration boundary



$$n(t, 0) = n_0 \quad \forall t$$

Special case: Absorbing boundary



$$n(t, 0) = 0 \quad \forall t$$

- Constant flux boundary

(Case of diffusion with a constant diffusion coefficient)



$$\alpha(t, 0) \partial_x n(t, 0) = D \partial_x n(t, 0)$$

is constant for all t.

Special case: Reflecting boundary  
(again for constant diff coeff.)



$$\begin{aligned} u(t, 0) \partial_x u(t, 0) &= \\ &= D \partial_x u(t, 0) = 0 \quad \forall t. \end{aligned}$$

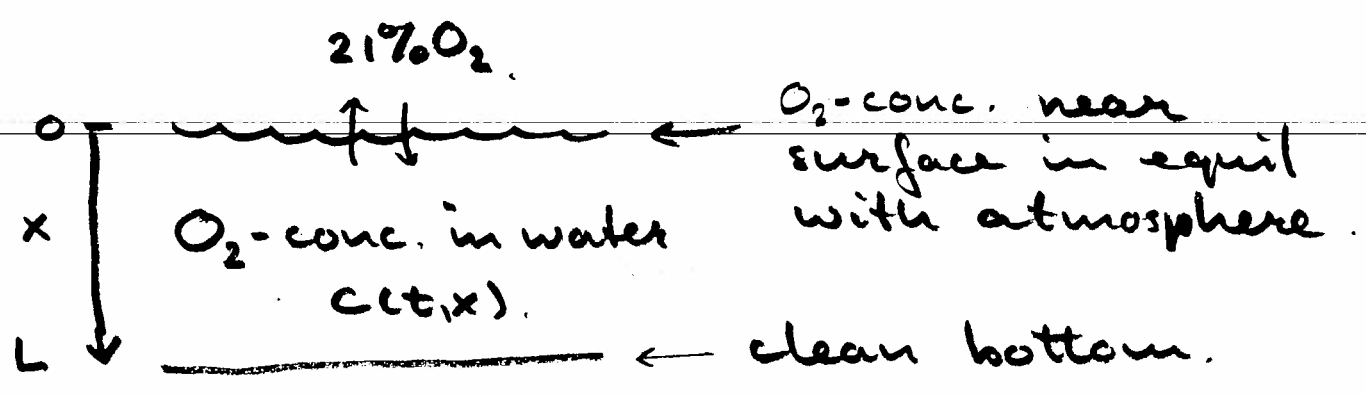


Formulation of the boundary conditions is as much a part of the model formulation as formulation of the flux:

The bnd. cond. do not follow from the mathematics; they follow from the biology.



# Example I



$$\frac{\partial c}{\partial t} = D \frac{\partial^2 c}{\partial x^2} \quad (\text{diffusion})$$

$$c(t,0) = c_0 \quad (\text{const. conc. bnd. at } x=0)$$

$$\frac{\partial c}{\partial x}(t,L) = 0 \quad (\text{refl. bnd. at } x=L)$$

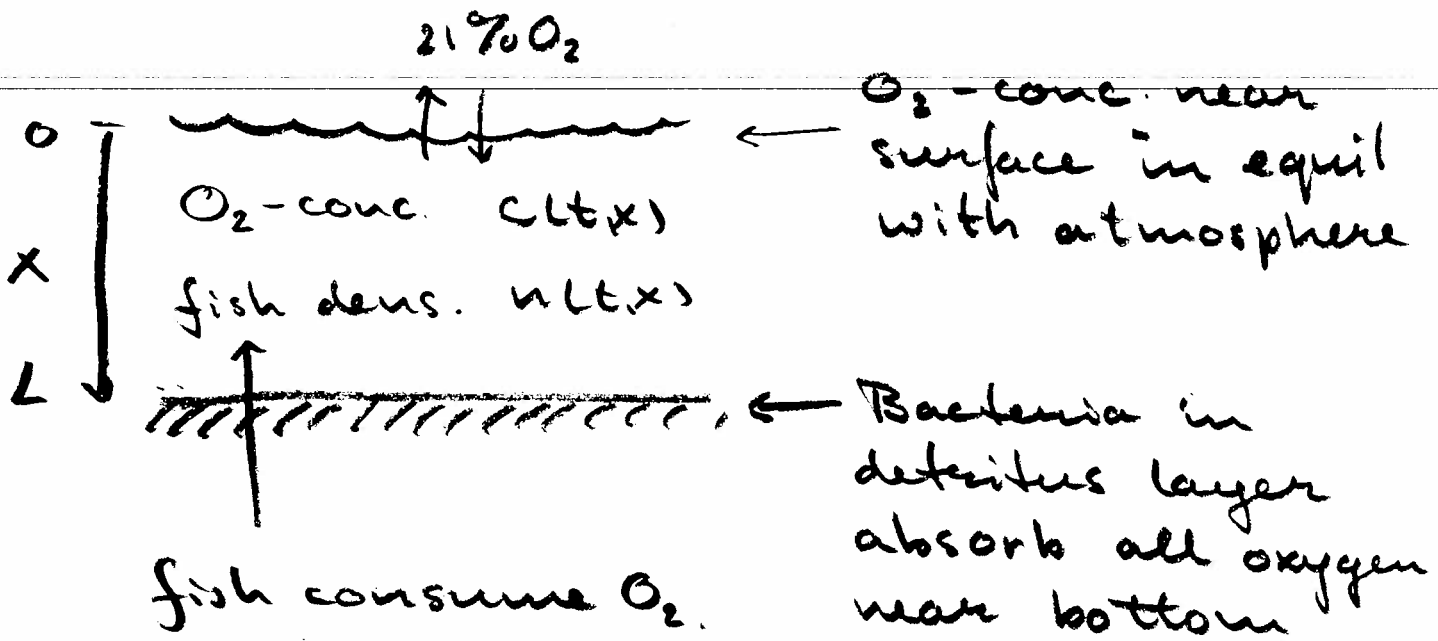
## Reaction-diffusion equations

$$\frac{\partial n}{\partial t} = \frac{\partial}{\partial x} (\alpha \frac{\partial n}{\partial x}) + \text{local reactions}$$

↑

This part is just the kind of mass-action modeling we did in the first half of the course.

# Example II



$$\partial_t c = D_1 \partial_x^2 c - \beta c n$$

$$\partial_t n = D_2 \partial_x^2 n$$

O<sub>2</sub> consumption by fish

$N + O_2 \rightarrow N$  reaction.

$c(t,0) = c_0 > 0$  (constant concentration boundary at  $x=0$ )

$c(t,L) = 0$  (absorbing boundary at  $x=L$ )

$\partial_x n(t,0) = 0$  (reflecting boundary at  $x=0$ )

$\partial_x n(t,L) = 0$  (reflecting boundary at  $x=L$ )

# Equilibrium solution.

We calculate the equilibrium solutions for example II (p. 10).

Setting  $\partial_t c = 0$  and  $\partial_t n = 0$  we get:

⊛ 
$$\begin{cases} 0 = D_1 \partial_x^2 c - \beta c n & \text{(oxygen)} \\ 0 = D_2 \partial_x^2 n & \text{(fish)} \end{cases}$$

From the second equation we get

$$n(x) = a + b x$$

The boundary conditions imply that  $b = 0$ , and so

$$n(x) = a \text{ (constant)}$$

Let  $N_{tot} := \int_0^L n(x) dx$  be the total number of fish, then

$$n(x) = \frac{N_{tot}}{L} \quad \left( \text{i.e., we write } a = \frac{N_{tot}}{L}, \text{ the average density} \right)$$

From the first equation in  $(*)$   
we then get

$$0 = D_1 \partial_x^2 c - \beta \frac{N_{\text{tot}}}{L} c$$

Note that this is a linear  
2<sup>nd</sup>-order ODE for  $c$ .

We can re-write this as  
a system of 1<sup>st</sup>-order ODEs:

$$\begin{cases} \partial_x c = v \\ \partial_x v = \theta^2 c \end{cases} \quad \text{where} \quad \theta := \frac{\beta}{D_1} \frac{N_{\text{tot}}}{L}$$

In matrix notation:

$$\partial_x \begin{pmatrix} c \\ v \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ \theta^2 & 0 \end{pmatrix}}_{\text{matrix } A} \begin{pmatrix} c \\ v \end{pmatrix}$$

Eigen value problem:

$$AB = B\Lambda$$

where  $\Lambda$  is a diagonal matrix with the eigenvalues of  $A$  as entries, and where the columns of  $B$  are the corresponding eigen vectors (See Appendix A).

$$A = B\Lambda B^{-1}$$

$$\Rightarrow \begin{pmatrix} c(x) \\ v(x) \end{pmatrix} = B e^{\Lambda x} B^{-1} \begin{pmatrix} c(0) \\ v(0) \end{pmatrix}$$

$$B = \begin{pmatrix} 1 & 1 \\ -\theta & \theta \end{pmatrix}, \quad B^{-1} = \frac{1}{2} \begin{pmatrix} 1 & \frac{1}{\theta} \\ 1 & -\frac{1}{\theta} \end{pmatrix}$$

$$\Lambda = \begin{pmatrix} -\theta & 0 \\ 0 & \theta \end{pmatrix}$$

$$\Rightarrow c(x) = \frac{v(0)}{\theta} \sinh(\theta x) + c(0) \cosh(\theta x)$$

Applying the boundary conditions  $c(0) = c_0$  and  $c(L) = 0$  given explicit expressions for  $c(x)$  and  $v(x)$  :

$$c(0) = c_0$$

$$v(0) = -c_0 \theta \operatorname{Coth}(L\theta)$$

And so :

$$c(x) = c_0 \frac{\sinh((L-x)\theta)}{\sinh(L\theta)}$$

which is the equil. solution for the  $O_2$ -concentration.

