## IV Southern-Summer School on Mathematical Biology

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Lecture V

São Paulo, January 2015



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Outline









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## Space

- Up to this point, all models that we have studied assume implicitly that all individuals are in certain region of space.
- This region has been supposed not to be very important .
- We think of homogeneous regions.
- Well-mixed populations.
- HOWEVER...
- Individuals move, generating possibly the spatial redistribution of the population.
- And space may be heterogeneous due to several factors :
  - climate
  - soil
  - vegetation
  - composition
  - salinity....



## Density

- Let us consider a population in space.
- Let space be homogeneous. How do populations spread over space?.
- First point: we will not speak of number of individuals.
- Instead we will speak of density of individuals.
- The number of individuals per unit space.
- The usual notation is  $\rho(\vec{x}, t)$  for density. It is a function of time and space.
- In some contexts, we use the term concentration.



## Diffusion

- Our main hypothesis is that individuals move randomly.
- In some sense, they behave as molecules in a gas.
- If we look at such population from a <u>space scale much larger</u> than the typical scale of the movement of the individuals we will see the macroscopic phenomenon called diffusion.
- Particles in a gas obey Fick's law.
- We will assume the same for a population.
- So, what's Fick's law?



## Fick

• The Fickian diffusion law states that:

The flux  $\vec{J}$  of "material"( animals, cells,..) is proportional to to the gradient of the density of the material:

$$\vec{J} = -D\vec{\nabla}
ho \equiv -D(rac{\partial
ho}{\partial x},rac{\partial
ho}{\partial y})$$

- where we took a two-dimensional space.
- But to simplify the calculations let us consider the one-dimensional case:

$$J \sim -\frac{\partial \rho}{\partial x}$$



## Mass/number of individuals conservation

- Let us impose a conservation law:
  - The rate of change in time of the quantity of individuals in a region of space is equal to the flux through the borders.
- that is, (in one dimension,  $(x_0 x_1)$  being the size of the region):

$$\frac{\partial}{\partial t}\int_{x_0}^{x_1}\rho(x,t)dx=J(x_0,t)-J(x_1,t)$$



## The diffusion equation



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# The diffusion equation

# $\tfrac{\partial \rho}{\partial t} = D \tfrac{\partial^2 \rho}{\partial x^2}$

- The above equation is known as the diffusion equation.
- In two dimensions we would have:

$$\frac{\partial \rho}{\partial t} = D\nabla^2 \rho$$

where  $\nabla^2 \rho \equiv \frac{\partial^2 \rho}{\partial x^2} + \frac{\partial^2 \rho}{\partial y^2}$ 

- $\bullet\,$  It is the same equation that describes heat diffusion if  $\rho$  is taken as temperature.
- Let us recall some facts about it.

# Diffusion Equation

- The diffusion equation is a partial differential equation, a PDE.
- It is linear, and the coefficients are constants.
- It can be solved analytically.

## Mathematical comment

- In order to speak of a solution of a differential equation, we need to specify supplementary conditions.
- In the case of the diffusion equation we should give an initial condition  $\rho(x, 0)$  and the values of either  $\rho(x, t)$  or  $\frac{\partial \rho(x, t)}{\partial x}$  at the borders or for  $x \to \pm \infty$ .
- To solve it analytically, means that we can find a formula connecting  $\rho(x, t)$  to  $\rho(x, 0)$ .



## Gauss

- There is a distinctive soolution: a Gaussian function.
- In one dimension we have, for t > 0:

$$\rho(x,t) = \frac{Q}{2(\pi Dt)^{1/2}} e^{-x^2/(4Dt)}$$

where Q os a constant.

- It is a Gaussian that "widens" with time.
- Corresponds to an initial condition concentrated in x = 0.
- Here is a plot.



## Gauss: plots

## Solution to the 1D diffusion equation





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## Gauss: 2D plot

## Solution to the 2D diffusion equation





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# Diffusion, biology

### Let us put some biology in this lecture!

- Let us give a biological sense to all that.
- Suppose that at t = 0 a population of N individuals is released at x = 0.
- After a certain amount of time we want to know the the extension occupied by the population.
- Let's be more specific: we want the extension of the region containing 95% of the population.



# Diffusion, biology

• Knowing the density of a population allows us to calculate the total population in a given area. In the 1D case, we have:

Population between 
$$-L$$
 and  $L = N_L = \int_{-L}^{+L} \rho(x, t) dx$ .

- If we use the Gaussian for  $\rho(x, t)$ , perform the integral, we obtain that 95% of the population is a region of size  $2\sqrt{2Dt}$ .
- Which grows in time proportional to  $t^{1/2}$ .
- Or, at a speed which goes like  $t^{-1/2}$ . Decreasing.



## Diffusion + Growth

• The previous case corresponds to a non-growing population.

• Let us incorporate growth:

$$\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial x^2} + a \rho(x, t)$$

Still linear.

• But, as we already learning, some saturation mechanism should become relevant for large enough populations Say:

$$\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial x^2} + a \rho(x, t) - b \rho^2(x, t)$$



## Fisher-Kolmogorov

# $\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial x^2} + a \rho(x, t) - b \rho^2(x, t)$



Figura : Robert. A. Fisher



Figura : Andrei N. Kolmogorov

- The above equation is called Fisher-Kolmogorov equation.
- It is the simplest equation with diffusion, growth and self-regulation of a species.
- It is nonlinear.
- It is a representative of the class of "reaction-diffusion" equations.
  - This name comes from chemistry.
- The 2D version is obvious:

$$\frac{\partial \rho}{\partial t} = D \nabla^2 \rho + a \rho - b \rho^2$$

Image: Image:

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# Fisher-Kolmogorov

# $\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial x^2} + a \rho(x, t) - b \rho^2(x, t)$

- Let us again look at the problem of a population released at a point (x = 0).
- Suppose it obeys the Fisher-Kolmogorov equation (and not anymore the simple diffusion equation).
- No explicit formula.
- But look at the plot::



## Fisher-Kolmogorov



- This pattern can be made the basis of experimental verification.
- Our observations should concentrate on the front's speed. .

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## Skellam

- The speed does not depend on *b*.
- Therefore, the constant wavefront speed is not related to density dependence. The nonlinear term is there to avoid infinities.
- A equation

$$\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial x^2} + a \rho(x, t)$$

is called the Skellam equation.



## The classic example

## Muskrat

- The muskrat, an species native of North-america, was introduced in Europe.
- In 1905, five individuals were introduced in Prague.
- Today, there are millions in Europe
- In what follow, we see the expansion of the muskrat's range around Prague over 17 years..



## 1905





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## 1909





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## 1913





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## 1917





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# 1921 ES



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## Skellam !

- From these observations we can estimate the speed of invasion as a function of time.
- Here it is:



• A straight line. Constant speed. Skellam dixit! REJOICE!.

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## Micro X macro

- From the theory of the Brownian motion we can see *D* as the mean square displacement per unit of time.
- We could try to track individuals and calculate it .
- Beware!, it is likely that you get a wrong value for D. Too large.
- Why?



## Home range effects

- Many species have home ranges.
- This comes from several factors: the need to find food, the need to find shelter .
- This slows down the diffusion process.
- In general, a mechanistic study of *D* is difficult. In most studies it is taken as a phenomenological constant.



## Example: Hantavirus

- In 2000, a new species of Hantavirus was discovered, being the etiological agent of f a respiratory syndrome. It is fatal in up to 60% of cases
- The host is Oligoryzomys fulvescens. Take a look at him:



- Where you find the rat, you find the Hantavirus
- The disease "follows"the spread of the rat.



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## Hantavirus II

- The diffusion of the hosts is well modeled by the usual models,.
- But D is small.
- Oligoryzomys fulvescens has a limited home-range.
- The population spreads through juvenile migrants.
- A statistically rare event.
- But determinant for the spatial redistribution of the population.
- The diffusion coefficient appearing in the equations is a proxy of all these processes.



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## **Online** Resources

- http://www.ictp-saifr.org/mathbio4
- http://ecologia.ib.usp.br/ssmb/

Thank you for your attention



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