

# Mathematical Biology

January 20, 2018

<i>CONTENTS</i>	2
-----------------	---

## Contents

<b>0</b> Miscellaneous	<b>3</b>
<b>1</b> Birth-death models	<b>4</b>

## 0 Miscellaneous

Course notes online: Julia Gog([www.damtp.cam.ac.uk/research/dd/teaching](http://www.damtp.cam.ac.uk/research/dd/teaching), 2013-2017), Peter Haynes([www.damtp.cam.ac.uk/user/phh/mathbio.html](http://www.damtp.cam.ac.uk/user/phh/mathbio.html))

Moodle page: Handwritten notes by lecture; Matlab/Python programming examples; solved exercises.

This course involves 3 models: Deterministic temporal models (11 lectures), Stochastic temporal models (5 lectures), Deterministic spatio-temporal models (8 lectures).

The focus of this course is biochemical reactions and population processes.

(some introductory speech)

**Example.** (1, Transient population) If we use  $n(t)$  to denote the size of a population, we may want to model  $\frac{dn}{dt} = f(n)$  by an ODE, or maybe if we have several components  $\mathbf{n}(t)$  then we may want to model  $\frac{d\mathbf{n}}{dt} = \mathbf{f}(\mathbf{n})$  which is a system of ODEs.

Note that although  $n$  should be an integer (discrete), when  $n \gg 1$  we may model it with continuous equations.

**Example.** (2)  $n \rightarrow \partial_t P(n, t) = W \cdot P(n, t)$ , Markov processes. Here  $P(n, t)$  is a probability(?),  $n$  being a state, and  $W$  being the transition matrix.

**Example.** (3)

If we include spatial aspect, we may have  $n(t)$  becoming  $n(x, t)$ . Now there might be 'diffusion':  $\partial_t n(x, t) = f(n(x, t)) + D \nabla^2 n(x, t)$  where  $\nabla^2 = \frac{\partial^2}{\partial x^2}$ ; this is the reaction-diffusion equation.

## 1 Birth-death models

The general idea is that we have a population of size  $n(t)$ ; per capita per unit time, we have births of rate  $b$  and deaths of rate  $d$ . Then we can write

$$n(t + \Delta t) = n(t) + bn\Delta t - dn\Delta t$$

So we have an ODE

$$\frac{dn}{dt} = (b - d)n = rn$$

where  $r = b - d$ . This has an easy solution  $n(t) = n_0 e^{rt}$ , assuming  $r$  is a constant. We see that if  $r$  is positive then the population grows exponentially, and if  $r$  is negative then the population decreases to 0 asymptotically.

Now probably  $b$  and  $d$  are related to  $n$  by  $b(n) = bn$  and  $d(n) = dn^2$  due to competition. Then we have

$$\frac{dn}{dt} = bn - dn^2$$

which we can definitely rewrite as

$$\frac{dn}{dt} = \alpha n(1 - n)$$

by some change of variable on  $n$ . Now

$$\begin{aligned} \frac{dn}{n(1-n)} &= \alpha dt \\ \implies \frac{dn}{n} + \frac{dn}{1-n} &= \alpha dt \\ \implies \ln n - \ln(1-n) &= \alpha t + c \\ \implies n &= \frac{n_0 e^{\alpha t}}{(1-n_0) + n_0 e^{\alpha t}} \end{aligned}$$

where we are given that  $t = 0$ ,  $n = n_0$ . If  $t \gg \frac{1}{\alpha}$ , when  $t \rightarrow \infty$  we have  $n(t) \rightarrow 1$ . Now we can investigate if the population size is stable, and if it has any fixed points.

Let's now define  $\mathbf{n} = (n_1, \dots, n_p)$ , i.e.  $p$  populations, and  $\frac{d\mathbf{n}}{dt} = \mathbf{f}(\mathbf{n})$ . If  $\mathbf{n} = \mathbf{n}^*$  is a fixed point, then  $\frac{d\mathbf{n}}{dt} = 0$ , i.e.  $\mathbf{f}(\mathbf{n}) = 0$ . Now if we apply a small perturbation  $\mathbf{n} = \delta\mathbf{n}^* + \delta\mathbf{n}$ , i.e.

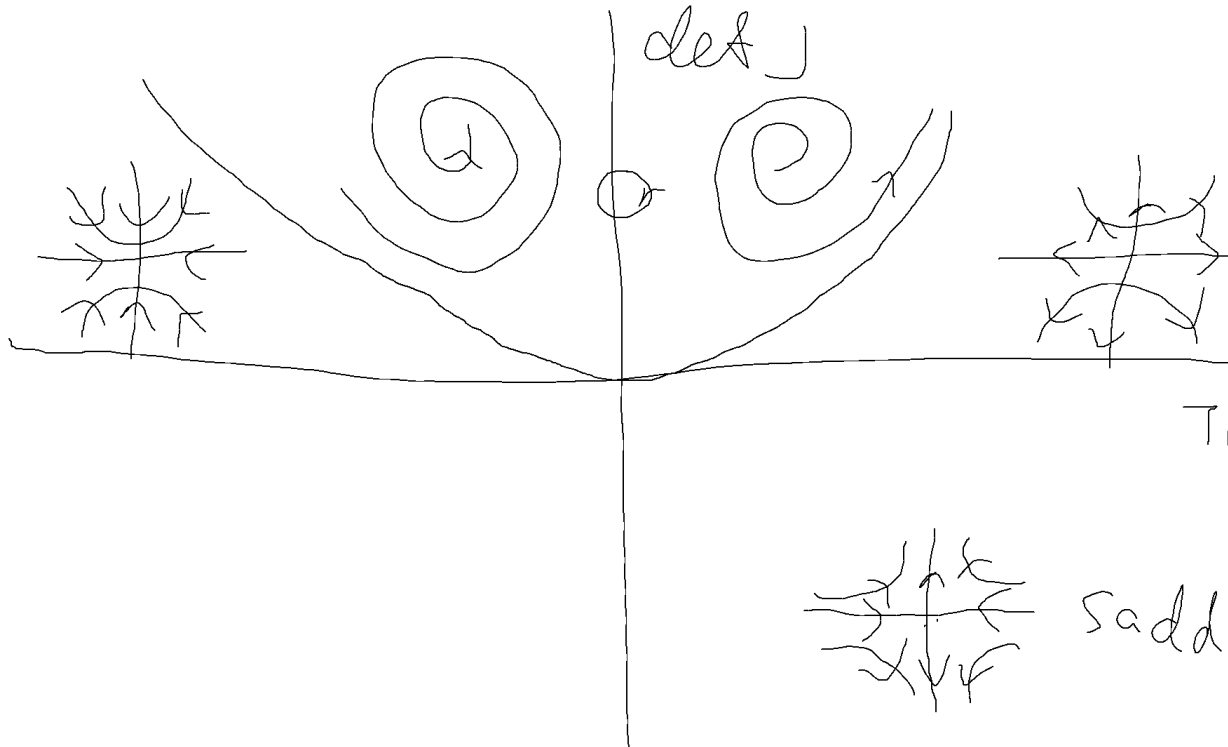
$$\begin{aligned} \frac{d}{dt}\delta\mathbf{n} &= \mathbf{f}(\mathbf{n}^* + \delta\mathbf{n}) \\ a &= \mathbf{f}(\mathbf{n}^*) + \frac{\partial f_i}{\partial n_j} \delta n_j + \frac{1}{2} \frac{\partial^2 f_i}{\partial n_j \partial n_k} \delta n_j \delta n_k \end{aligned}$$

So  $\frac{d}{dt}\delta\mathbf{n} = J \cdot \delta\mathbf{n}$ , so  $\delta n(t) = e^{Jt} \cdot \delta n(0)$ . If  $\lambda_i$ 's are the eigenvalues of  $J$ , we consider the real part of  $\lambda_i$ : if  $\text{Re}(\lambda_i) < 0$ , then if  $p \geq 5$  we only have numerical solutions, if  $3 \leq p \leq 5$  we have analytic solutions, and  $p = 2$  is an easy case (recall  $p$  is the number of populations):

- If  $p = 2$ ,  $\mathbf{n} = (n_1, n_2)$ , then

$$\frac{d}{dt} \begin{pmatrix} \delta_{n_1} \\ \delta_{n_2} \end{pmatrix} = \begin{pmatrix} \frac{\partial f_1}{\partial n_1} & \frac{\partial f_1}{\partial n_2} \\ \frac{\partial f_2}{\partial n_1} & \frac{\partial f_2}{\partial n_2} \end{pmatrix} \cdot \begin{pmatrix} \delta_{n_1} \\ \delta_{n_2} \end{pmatrix}$$

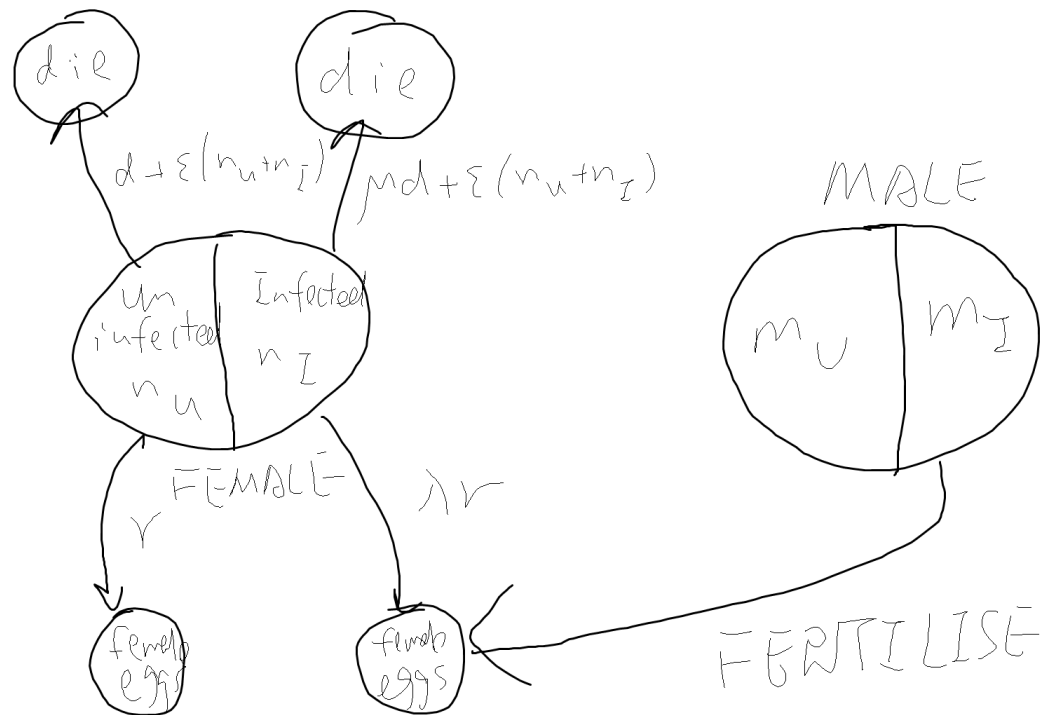
Where the matrix is  $J$ . Now we have  $\lambda_1 \lambda_2 = \det J$  and  $\lambda_1 + \lambda_2 = \text{tr } J$ . Determined by the signs of those two, we have different possible behaviours:



Now let's consider the spread of Dengue. There are several processes going on at the same time:

- (1) Mosquitos carry dengue;
- (2) Wolbachia infect mosquitos;
- (3) Infected mosquitos do not transmit dengue;
- (4) Wolbachia transmission only across generations.

Question: will an initially infected population of mosquitos eventually spread over the entire population as  $t \rightarrow \infty$ ?



F	M	frequency	rate	outcome	
u	u	$n_u \cdot m_u$	$r$	u	$m_u =$
u	i	$n_u \cdot m_i$		X	$\frac{n_u}{n_u + n_i}$
i	u	$n_i \cdot m_u$	$\lambda_r$	i	$m_i =$
i	i	$n_i \cdot m_i$	$\lambda_r$	i	$\frac{n_i}{n_u + n_i}$

We always assume that there are enough males to fertilise the female eggs.

Now consider  $\frac{d}{dt}$  of  $n_U$  and  $n_I$  (uninfected and infected females). From the above tables we should be able to get (hopefully)

$$\begin{aligned}\frac{d}{dt}n_U &= rn_U \frac{n_U}{n_U + n_I} - dn_U - \varepsilon(n_U + n_I)n_U \\ \frac{d}{dt}n_I &= \lambda rn_I \frac{n_U}{n_U + n_I} + \lambda rn_I \frac{n_I}{n_U + n_I} - \mu dn_I - \varepsilon(n_U + n_I)n_I \quad (*)\end{aligned}$$

This is our model when  $p = 2$ .