

VI Southern-Summer School on Mathematical Biology

Mathematical Theory of Biological Invasions

Sergei Petrovskii

Department of Mathematics, University of Leicester, UK

<http://www.math.le.ac.uk/PEOPLE/sp237>

IFT/UNESP, Sao Paulo, January 16-27, 2017

Overall goals

- To give you an **overview of the subject**: what it is all about. To consider several real-world **examples** and instructive **case-studies**
- To introduce you to the **array of mathematical approaches** used to study biological invasions
- Starting from simple examples and basic modeling tools, to bring you to the **frontiers in this field**

Interdisciplinary Applied Mathematics 44

Mark A. Lewis
Sergei V. Petrovskii
Jonathan R. Potts

The Mathematics Behind Biological Invasions



 Springer

The Springer logo, which consists of a stylized chess knight piece.

Outline of the course

- Introduction & a glance at field data
- Overview of mathematical tools
- Diffusion-reaction systems
 - ▶ Single-species system: traveling waves, the problem of critical domain, effects of environmental heterogeneity
 - ▶ Predator-prey system and the problem of biological control: traveling waves and pattern formation
 - ▶ Beyond the traveling waves: patchy invasion
- Lattice models
- Kernel-based models (integro-difference equations): fat-tailed kernels, “superspread”, pattern formation
- Extensions, discussion, conclusions

Chapter I

Introduction and examples

General

Biological invasion begins when a new, “alien” or exotic species is brought into a given ecosystem.

Ecology is a science that studies species living in their natural environment. It focuses on the **species interaction** between themselves and with the abiotic factors. Ecology aims to understand the principles and mechanisms of population's **spatiotemporal organization** and to identify factors that may affect species abundance.

Mathematical ecology is application of the tools and techniques of mathematics to ecological problems.

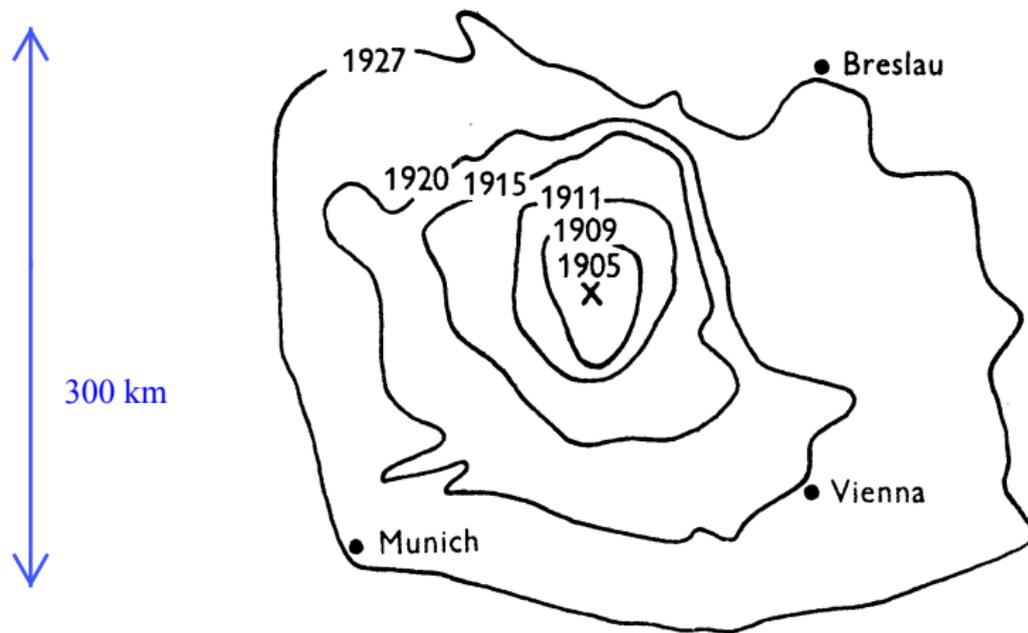
Why biological invasion?

The term **biological invasion** is a common name for a variety of phenomena related to introduction and spatial spread of alien or exotic species, i.e., species that have not been present in a given ecosystem until they are brought in.

Consequences of species invasion:

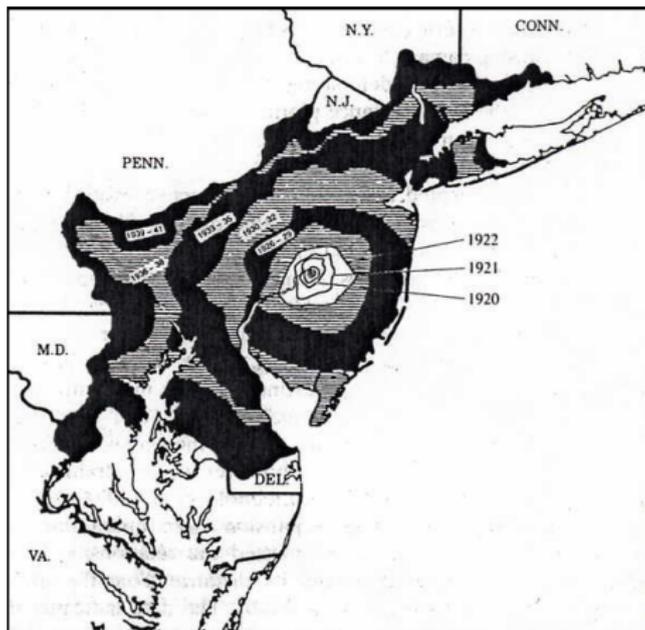
1. A new species often becomes a pest and that can result in **huge economic losses**. For instance, economic loss from the invasion of insect pests in the USA for the period from 1906 to 1991 is estimated to be **\$ 92 billion** (U.S. Office of Technology Assessment, 1993).
2. Severe damage to biodiversity.

A glance at the data I: muskrat in Europe

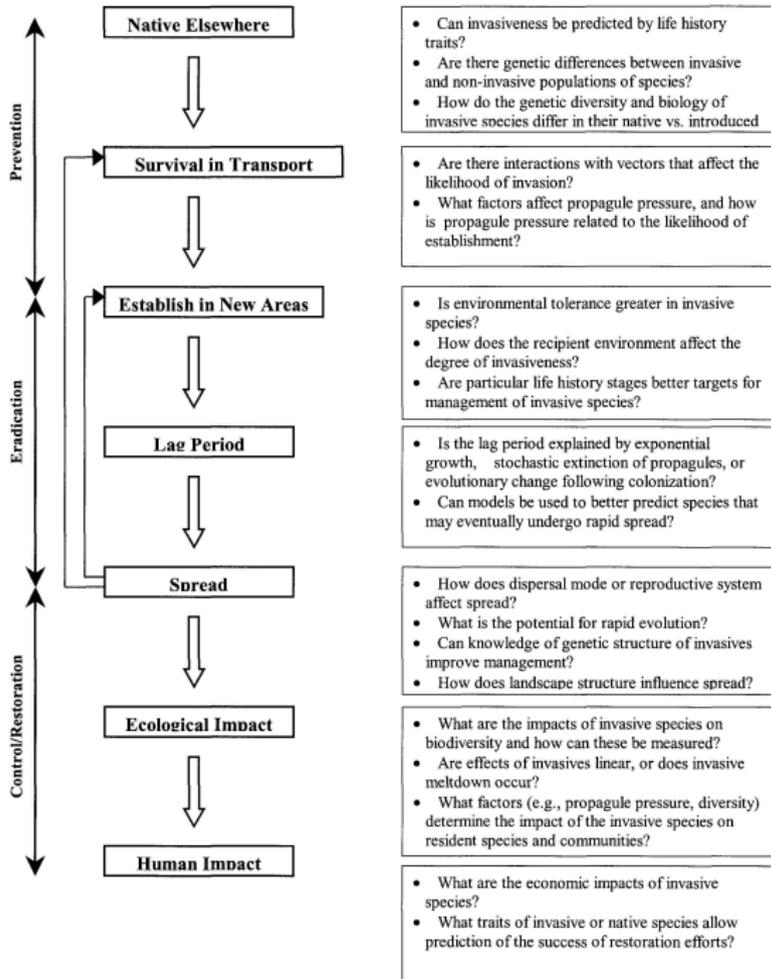


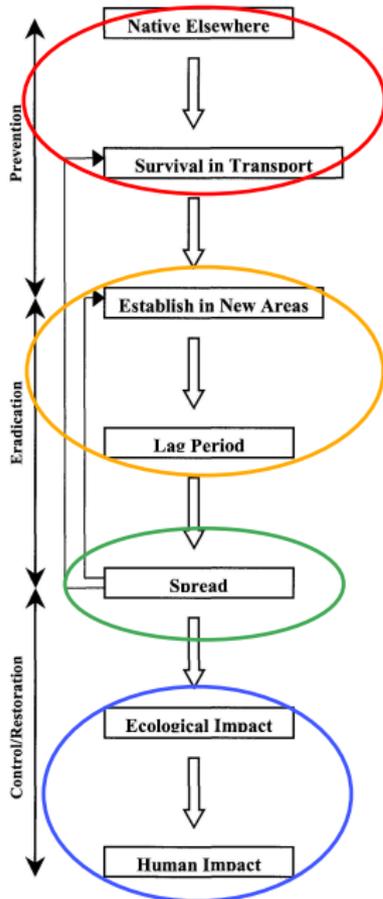
(Skellam, 1951)

A glance at the data II: Japanese beetle in the USA



(United States Bureau of Entomology and Plant Quarantine, 1941)





- Can invasiveness be predicted by life history traits?
- Are there genetic differences between invasive and non-invasive populations of species?
- How do the genetic diversity and biology of invasive species differ in their native vs. introduced

Introduction

- Are there interactions with vectors that affect the likelihood of invasion?
- What factors affect propagule pressure, and how is propagule pressure related to the likelihood of establishment?

- Is environmental tolerance greater in invasive species?
- How does the recipient environment affect the degree of invasiveness?
- Are particular life history stages better targets for management of invasive species?

Establishment

- Is the lag period explained by exponential growth, stochastic extinction of propagules, or evolutionary change following colonization?
- Can models be used to better predict species that may eventually undergo rapid spread?

- How does dispersal mode or reproductive system affect spread?
- What is the potential for rapid evolution?
- Can knowledge of genetic structure of invasives improve management?
- How does landscape structure influence spread?

Spread

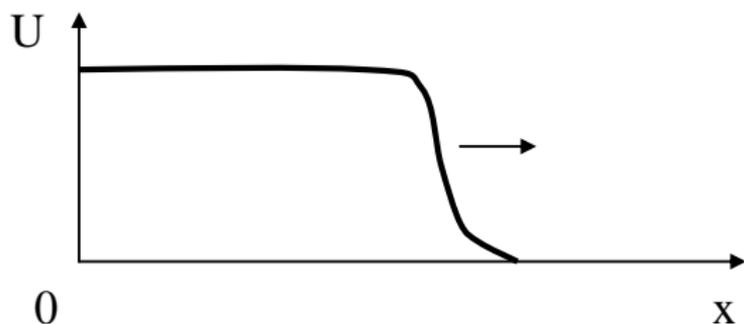
- What are the impacts of invasive species on biodiversity and how can these be measured?
- Are effects of invasives linear, or does invasive meltdown occur?
- What factors (e.g., propagule pressure, diversity) determine the impact of the invasive species on resident species and communities?

Post-factum

- What are the economic impacts of invasive species?
- What traits of invasive or native species allow prediction of the success of restoration efforts?

The stage of geographical spread

Spread of the introduced species over space:



(U is the population density)

Fisher, 1937: existence of the traveling population front

**Kolmogorov, Petrovskii & Piskunov (KPP), 1937:
convergence of the initial condition to the traveling front**

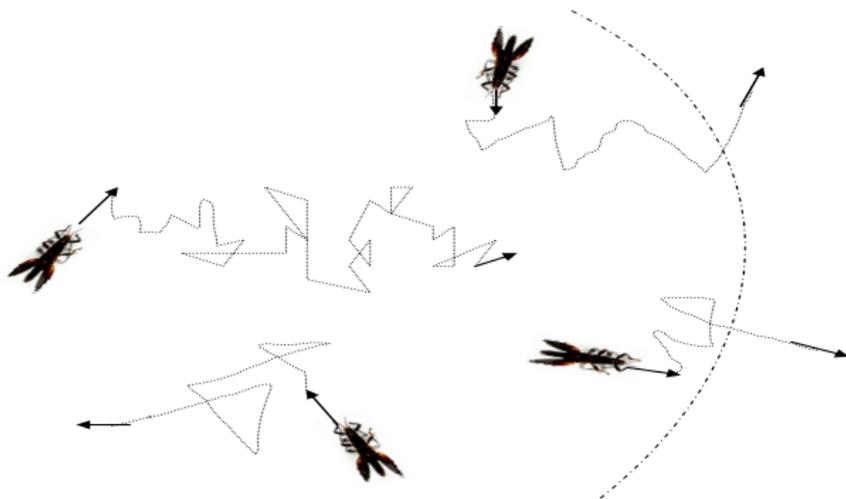
Chapter II

Overview of mathematical tools

Overview of mathematical tools

- ▶ Individual-based or mean-field?

In the **individual-based** modelling (IBM), each individual is described and modelled explicitly (e.g. as given by certain rules):



Overview of mathematical tools

- ▶ Individual-based or mean-field?

In the **mean-field** approach, ecosystem's state is described by an array of its **state variables**:

$$\Omega = \{U_1(\mathbf{r}, t), \dots, U_N(\mathbf{r}, t), \alpha_1(\mathbf{r}, t), \dots, \alpha_M(\mathbf{r}, t)\} .$$

Typically, U_i is the population density of the i th species, and $\alpha_1, \dots, \alpha_M$ are 'parameters' (e.g. the growth rates).

Depending on the **spatial scale**, some of the components may be better described as **discrete**.

Mathematical tools: **nonspatial systems**

How to account for different **reproduction systems**?

Time-continuous systems (overlapping generations):

$$\frac{dU_i(t)}{dt} = f_i(U_1, \dots, U_n), \quad i = 1, \dots, N.$$

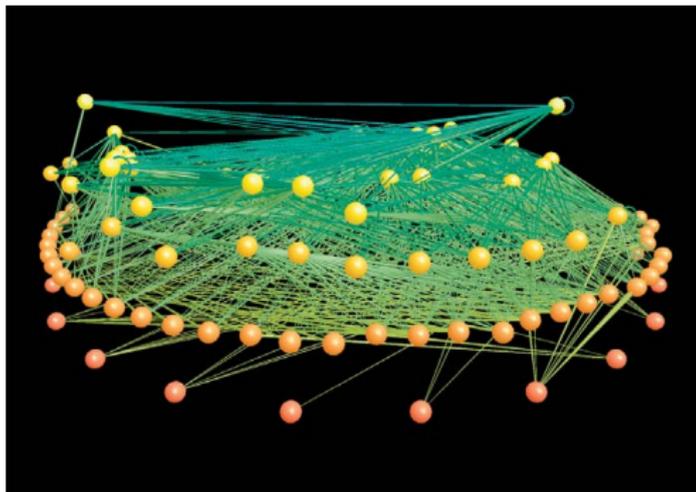
where N is the number of species included into the model.

Time-discrete systems (non-overlapping generations):

$$U_i(t + T) = f_i(U_1(t), \dots, U_N(t)), \quad i = 1, \dots, N,$$

What are the values of N **reasonable** for modeling?

Ecological **communities** usually consist of many species linked together into a complicated **food web**:



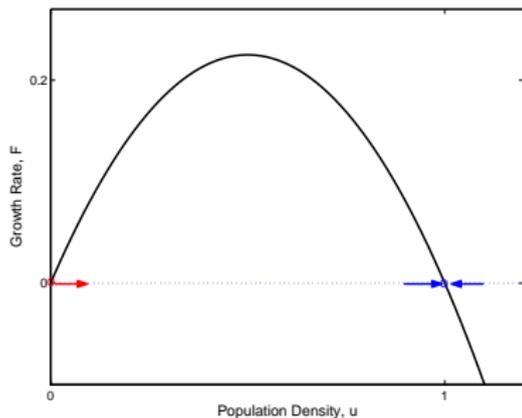
(Lake food web, from Pascual 2005)

Remarkably, however, many important insights can be made based on simple **few-species models**.

Example 1: single-species system

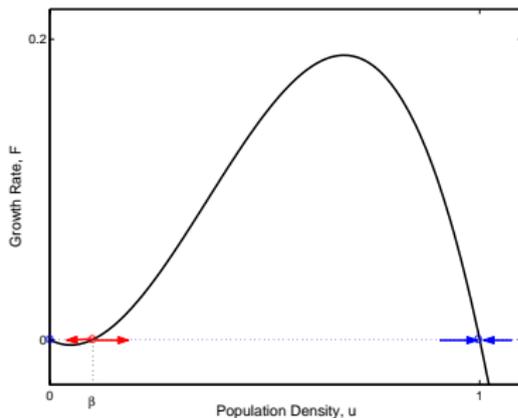
$$\frac{dU(T)}{dT} = f(U)$$

Logistic growth



$$f(U) = \alpha U(K - U)$$

Strong Allee effect



$$f(U) = \gamma U(U - \beta)(K - U)$$

Example 2: predator-prey system

$$\begin{aligned}\frac{dU(T)}{dT} &= P(U) - E(U, V), \\ \frac{dV(T)}{dT} &= \kappa E(U, V) - \mu(V),\end{aligned}$$

where U and V are prey and predator densities.

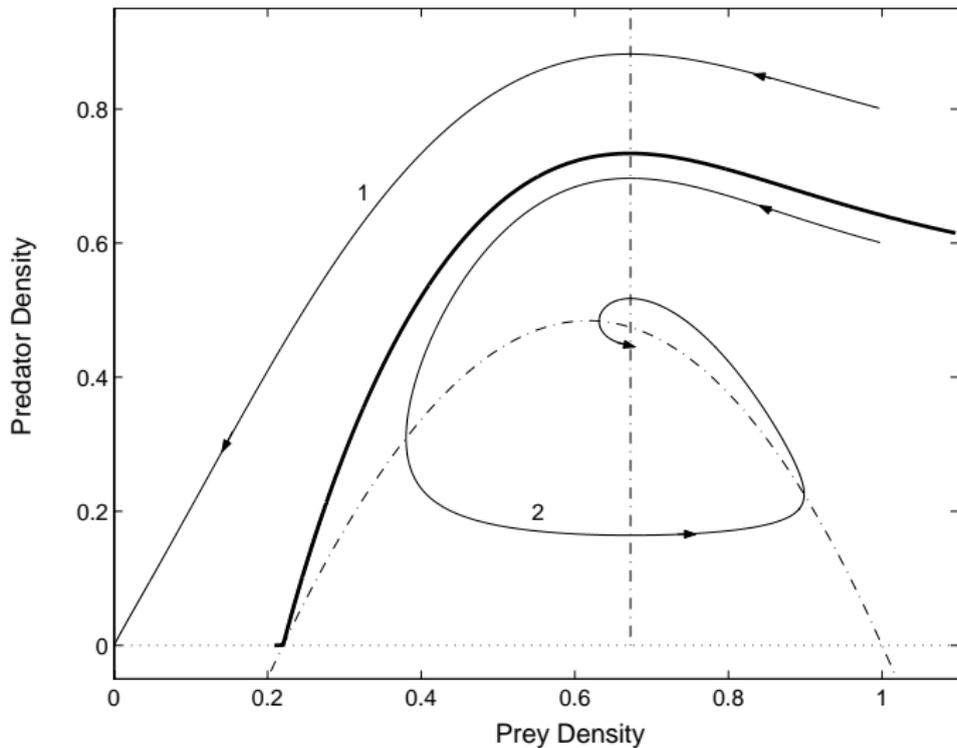
Prey **growth rate** P can be logistic or with the Allee effect.

Predation E can also have different properties, for instance:

$$E(U, V) = A \frac{UV}{U + H} \quad \text{or} \quad E(U, V) = A \frac{U^2 V}{U^2 + H^2},$$

that is, Holling **type II** and Holling **type III**, respectively.

Predator-prey system can have **complicated properties!**



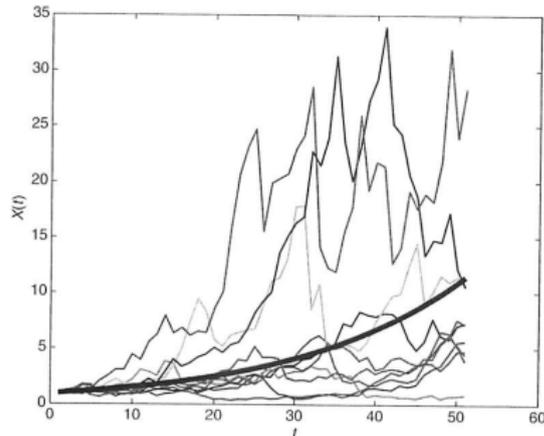
Deterministic or stochastic?

The impact of stochasticity can be described as a **noise** applied to the ‘**deterministic skeleton**,’ for instance:

$$\frac{dU(t)}{dt} = f(U) + \xi(t)U .$$

where $U(t)$ is the population size and ξ is a random variable.

How the system will behave?



A fundamental result (**Central Limit Theorem**):

the relative magnitude of the fluctuations decreases as the number of individuals increases!

“Large scale random phenomena in their collective action create strict, **non random** regularity.”

(Gnedenko & Kolmogorov, 1954)

- Population dynamics is intrinsically stochastic, but that does **not necessarily** mean that the model must be stochastic.
- I will mostly focus on **deterministic** models

Mathematical tools: **spatial systems**

Time-discrete and space-continuous, IDEs:

$$U_i(x, t + T) = \int_{-\infty}^{\infty} K(x - \xi) f_i(U_1(\xi, t), \dots, U_n(\xi, t)) d\xi .$$

Time- and space-continuous, “**diffusion-reaction**” PDEs:

$$\frac{\partial U_i(\mathbf{r}, t)}{\partial t} = D_i \nabla^2 U_i(\mathbf{r}, t) + f_i(U_1, \dots, U_n) .$$

Time- and space-discrete: Coupled Maps Lattices.

Dynamical systems' approach to invasion

Native community before invasion:

$$\frac{dU_i(t)}{dt} = f_i(U_1, \dots, U_n), \quad i = 1, \dots, N. \quad (1)$$

As a result of invasion, a new species U_{n+1} is added:

$$\frac{dU_i(t)}{dt} = f_i(U_1, \dots, U_n, U_{n+1}), \quad i = 1, \dots, N, (N+1). \quad (2)$$

Invasion will be successful if the system (2) allows for the existence of an attractor such as $U_{n+1}(T) > 0$, for instance, a stable steady state:

$$(\bar{U}_1, \bar{U}_2, \dots, \bar{U}_n, \bar{U}_{n+1}) \quad \text{with} \quad \bar{U}_{n+1} > 0.$$

Questions to be asked (and answered)

- How likely the new species will **establish** in the new environment?
- Will it start **spreading** and, if yes, how soon after the introduction? – How large can be the **gap**?
- What are the **rate** and **pattern** of spread?
- What are the **mechanisms** of spread?
- Can we **control** the rate of spread?
- Can we **eradicate** the invading species?

Chapter III

PDE models of biological invasion:

Single-species system

How can we build a model?

What we need is to keep the **balance of mass**:

$$\left(\begin{array}{c} \textit{Change in the} \\ \textit{population density} \end{array} \right) = \left(\begin{array}{c} \textit{Local} \\ \textit{growth} \end{array} \right) + \textit{Dispersal}$$

How can we build a model?

What we need is to keep the **balance of mass**:

$$\left(\begin{array}{c} \text{Change in the} \\ \text{population density} \end{array} \right) = \left(\begin{array}{c} \text{Local} \\ \text{growth} \end{array} \right) + \text{Dispersal}$$

Translate it to the mathematical language:

$$\frac{\partial U(X, T)}{\partial T} = f(U) + D \frac{\partial^2 U}{\partial X^2},$$

How can we build a model?

What we need is to keep the **balance of mass**:

$$\left(\begin{array}{c} \text{Change in the} \\ \text{population density} \end{array} \right) = \left(\begin{array}{c} \text{Local} \\ \text{growth} \end{array} \right) + \text{Dispersal}$$

Translate it to the mathematical language:

$$\frac{\partial U(X, T)}{\partial T} = f(U) + D \frac{\partial^2 U}{\partial X^2},$$

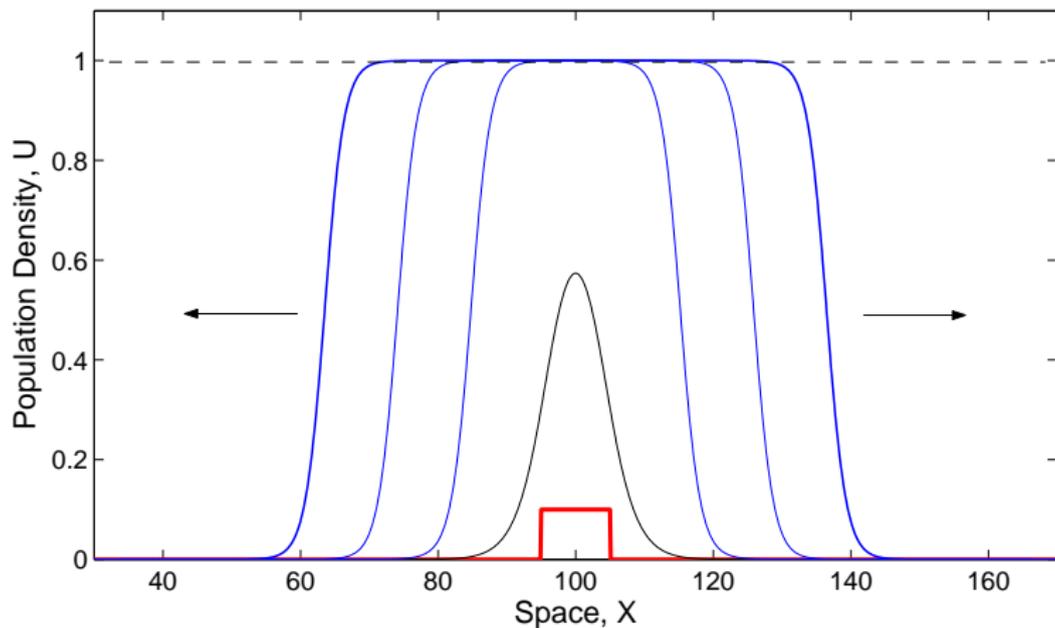
Biological invasion is accounted for by **finite initial conditions**:

$$\begin{aligned} u(x, 0) &= \Phi(x) \geq 0 \quad \text{for } x_1 \leq x \leq x_2, \\ u(x, 0) &= 0 \quad \text{for } x < x_1 \text{ and } x > x_2 \geq x_1. \end{aligned}$$

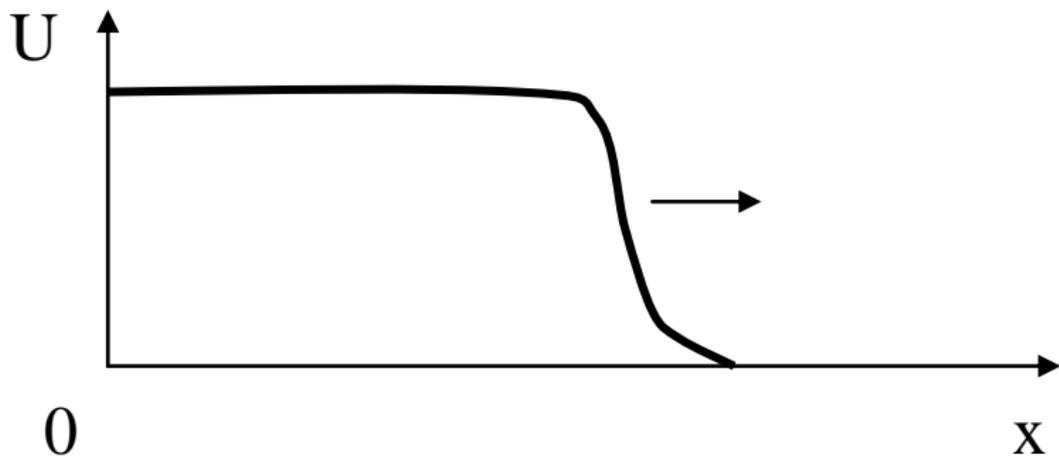
The corresponding **conditions at infinity**: $U(X \rightarrow \pm\infty, t) = 0$.

Consider the **evolution of the initial conditions**.

Easy to do using computer simulations:



The solution eventually converges to a **travelling front**:



Single-species model

$$\frac{\partial U(X, T)}{\partial T} = D \frac{\partial^2 U}{\partial X^2} + f(U), \quad (3)$$

where $f(0) = f(K) = 0$ and $f(U) \geq 0$ for $0 < U < K$,
with corresponding conditions at infinity:

$$U(X \rightarrow -\infty, t) = K, \quad U(X \rightarrow \infty, t) = 0.$$

Single-species model

$$\frac{\partial U(X, T)}{\partial T} = D \frac{\partial^2 U}{\partial X^2} + f(U), \quad (4)$$

where $f(0) = f(K) = 0$ and $f(U) \geq 0$ for $0 < U < K$,
with corresponding conditions at infinity:

$$U(X \rightarrow -\infty, t) = K, \quad U(X \rightarrow \infty, t) = 0.$$

Traveling wave solution – **change of variables**:

$$U(x, t) = \phi(\xi) \quad \text{where} \quad \xi = x - ct$$

where c is the speed of the wave.

The **shape** of the front **does not change** with time!

Using the chain rule,

$$\frac{d}{dx} = \frac{d}{d\xi}, \quad \frac{d^2}{dx^2} = \frac{d^2}{d\xi^2}, \quad \frac{d}{dt} = -c \frac{d}{d\xi},$$

the **partial** differential equation (4) turns into an **ODE**:

$$D \frac{d^2 \phi(\xi)}{d\xi^2} + c \frac{d\phi(\xi)}{d\xi} + f(\phi) = 0, \quad (5)$$

where

$$\phi(\xi \rightarrow -\infty) = K, \quad \phi(\xi \rightarrow \infty) = 0.$$

What is the **direction** of the front propagation:

invasion or retreat?

We multiply Eq. (5) by $\frac{d\phi}{d\xi}$ and integrate it over the whole line:

$$D \int_{-\infty}^{\infty} \frac{d^2\phi}{d\xi^2} \left(\frac{d\phi}{d\xi} \right) d\xi + c \int_{-\infty}^{\infty} \left(\frac{d\phi}{d\xi} \right)^2 d\xi + \int_{-\infty}^{\infty} f(\phi) \left(\frac{d\phi}{d\xi} \right) d\xi = 0.$$

Note that

$$\int_{-\infty}^{\infty} \frac{d^2\phi}{d\xi^2} \left(\frac{d\phi}{d\xi} \right) d\xi = \frac{1}{2} \int_{-\infty}^{\infty} \frac{d}{d\xi} \left(\frac{d\phi}{d\xi} \right)^2 = \frac{1}{2} \left(\frac{d\phi}{d\xi} \right)^2 \Big|_{-\infty}^{\infty} = 0,$$

and

$$\int_{-\infty}^{\infty} f(\phi) \left(\frac{d\phi}{d\xi} \right) d\xi = \int_K^0 f(\phi) d\phi = - \int_0^K f(\phi) d\phi.$$

We therefore obtain:

$$c \int_{-\infty}^{\infty} \left(\frac{d\phi}{d\xi} \right)^2 d\xi = \int_0^K f(\phi) d\phi \equiv M.$$

Note that

$$\int_{-\infty}^{\infty} \left(\frac{d\phi}{d\xi} \right)^2 d\xi > 0.$$

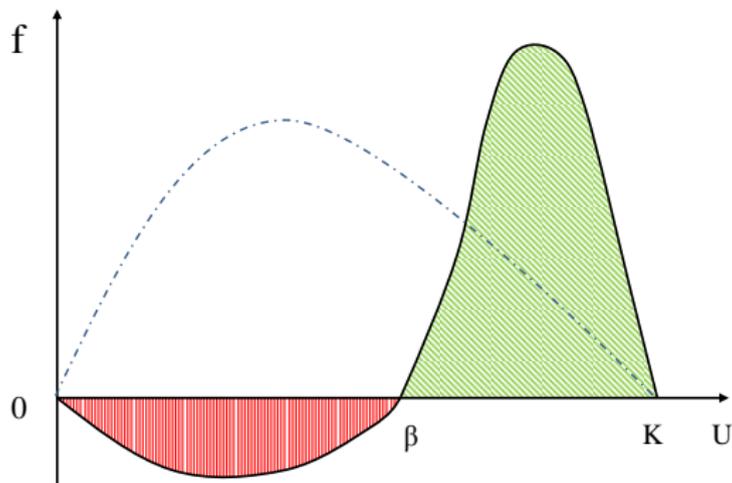
Therefore,

$$c \sim M, \quad \text{sign}(c) = \text{sign}(M).$$

Direction of front propagation is defined by the sign of M .

In case of **logistic growth** invasion is **always successful**: the front propagates from the area where the species is abundant, $U = K$, towards the area where the species is absent, $U = 0$.

However, in case of the strong Allee effect, it is not necessarily so:



$M > 0$: invasion success, species spread

$M < 0$: invasion failure, species retreat

What can we tell about the value of c ?

Consider the solution far in front of the front, i.e. where ϕ is small; then,

$$f(\phi) \approx \alpha\phi$$

and Eqn. (5) becomes **linear**:

$$D \frac{d^2\phi(\xi)}{d\xi^2} + c \frac{d\phi(\xi)}{d\xi} + \alpha\phi = 0, \quad (6)$$

and its **general solution** is known:

$$\phi(\xi) = C_1 e^{\lambda_1 \xi} + C_2 e^{\lambda_2 \xi},$$

where $\lambda_{1,2}$ are the solutions of the characteristic equation:

$$D\lambda^2 + c\lambda + \alpha = 0,$$

so that

$$\lambda_{1,2} = \frac{1}{2D} \left(-c \pm \sqrt{c^2 - 4\alpha D} \right).$$

We are only interested in solutions that are **nonnegative**, which means that the solution cannot oscillate around zero, which means that $\lambda_{1,2}$ **cannot be complex**.

Thus, we obtain $c^2 - 4\alpha D \geq 0$, so that

$$c \geq c_{min} = 2\sqrt{\alpha D}. \quad \text{There are no slow waves!}$$

It only works **if** $\alpha > 0$, i.e. when there is **no** strong Allee effect!

The **actual value** of the speed depends on the initial condition. For ecologically meaningful **compact** initial conditions, $c = c_{min}$.

In case of the **Allee effect**, things are different.

For a general $f(u)$, the wave speed is **not known**.

Consider a specific but meaningful case:

$$f(u) = \gamma U(U - \beta)(K - U),$$

so that Eq. (5) for the traveling wave takes the form:

$$D \frac{d^2 \phi(\xi)}{d\xi^2} + c \frac{d\phi(\xi)}{d\xi} + \gamma \phi(\phi - \beta)(K - \phi) = 0. \quad (7)$$

If we look for a monotone front, then $\frac{d\phi}{d\xi} = \psi(\phi)$,
where ψ is a certain (unknown) function.

Consider the following [ansatz](#):

$$\frac{d\phi}{d\xi} \stackrel{1}{=} \psi(\phi) \stackrel{2}{=} a\phi(\phi - K),$$

and substitute it into Eq. (7), taking into account that

$$\frac{d^2\phi(\xi)}{d\xi^2} = \frac{d}{d\xi} \left(\frac{d\phi}{d\xi} \right) = \frac{d\phi}{d\phi} \cdot \frac{d\psi(\phi)}{d\xi} = \frac{d\phi}{d\xi} \cdot \frac{d\psi(\phi)}{d\phi} = \psi \frac{d\psi(\phi)}{d\phi}.$$

Equation (7) then turns into

$$Da^2\phi(\phi - K)(2\phi - K) + ca\phi(\phi - K) + \gamma\phi(\phi - \beta)(K - \phi) = 0,$$

or, after obvious simplifications,

$$(2Da^2 - \gamma)\phi + (ca - KDa^2 + \gamma\beta) = 0. \quad (8)$$

Equation (8) holds for any ϕ , which is only possible if

$$2Da^2 - \gamma = 0 \quad \text{and} \quad ca - KDa^2 + \gamma\beta = 0,$$

from which we obtain $a = \sqrt{\gamma/(2D)}$ and the **speed**:

$$c = \left(\frac{D\gamma}{2} \right)^{1/2} (K - 2\beta).$$

We therefore observe that $c > 0$ (**invasion**) for $\beta < \frac{1}{2}K$,
but $c < 0$ (**retreat**) for $\beta > \frac{1}{2}K$.

Note that, since $f(U)$ is a cubic polynomial, this is equivalent to $M > 0$ and $M < 0$, respectively.

Note that, considering the ansatz as the ODE for $\phi(\xi)$,

$$\frac{d\phi}{d\xi} = a\phi(\phi - K) ,$$

we can easily obtain the **exact solution** of the problem:

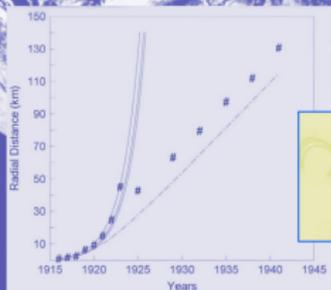
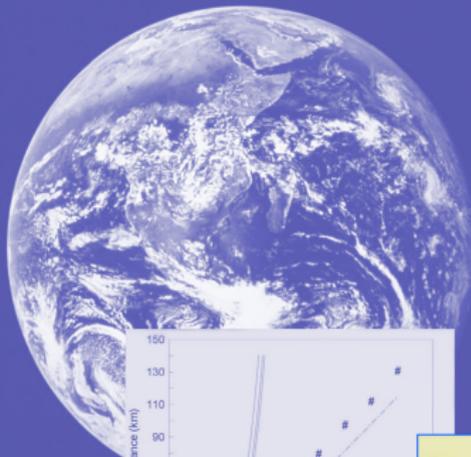
$$U(x - ct) = \phi(\xi) = \frac{K}{1 + A \exp(aK\xi)}$$

where A is an arbitrary constant depending on the initial position of the front.

Thus, ansatz is a **powerful method** to find a special solution of the problem.

Chapman & Hall/CRC Mathematical Biology and Medicine Series

EXACTLY SOLVABLE MODELS OF BIOLOGICAL INVASION

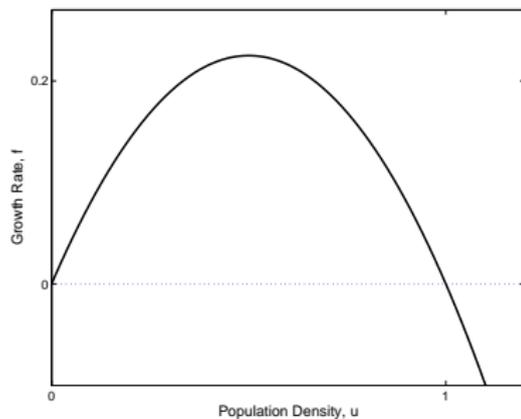


SERGEI V. PETROVSKII AND BAI-LIAN LI

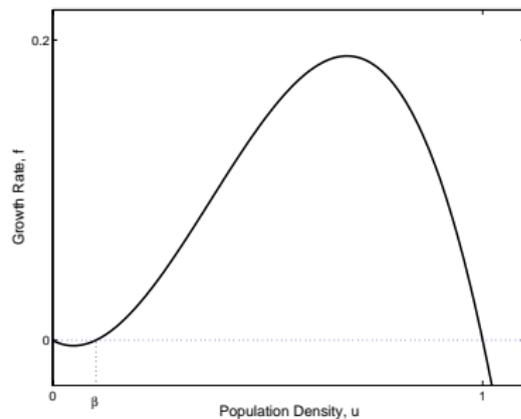
 Chapman & Hall/CRC
Taylor & Francis Group

Single-species invasion – a brief summary

$$\frac{\partial U(X, T)}{\partial T} = D \frac{\partial^2 U}{\partial X^2} + f(U)$$



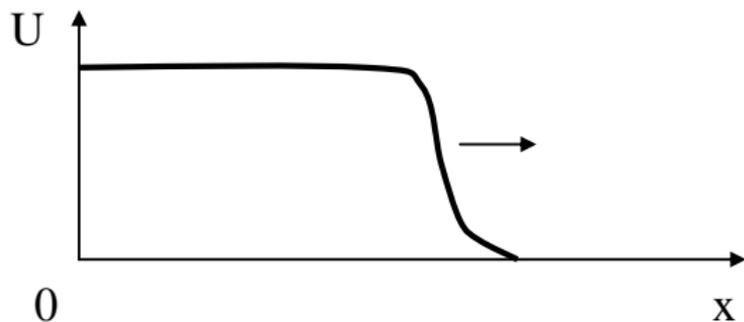
Logistic growth



Strong Allee effect

Single-species invasion – a brief summary

Spread of the introduced species over space:



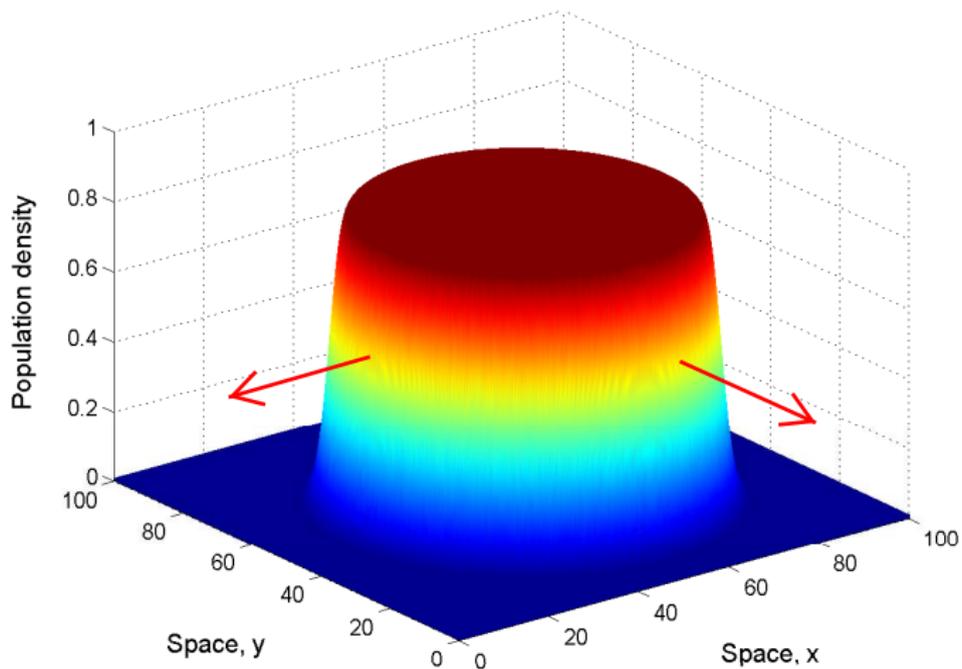
$$c = 2\sqrt{Df'(0)}$$

Logistic growth

$$c \sim \int_0^K f(U) dU$$

Strong Allee effect

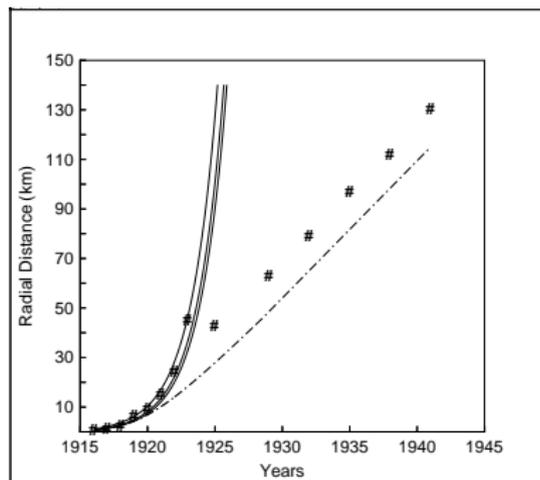
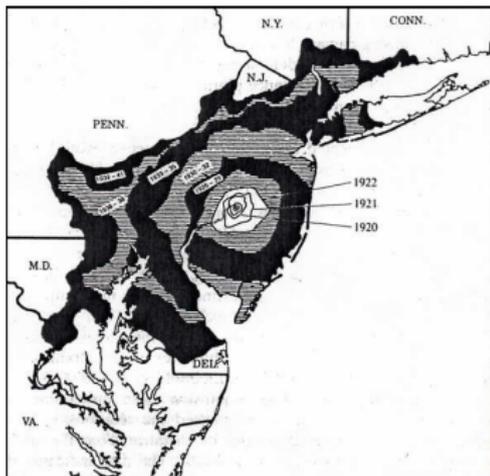
Single-species invasion in a 2D space



Example of historical data

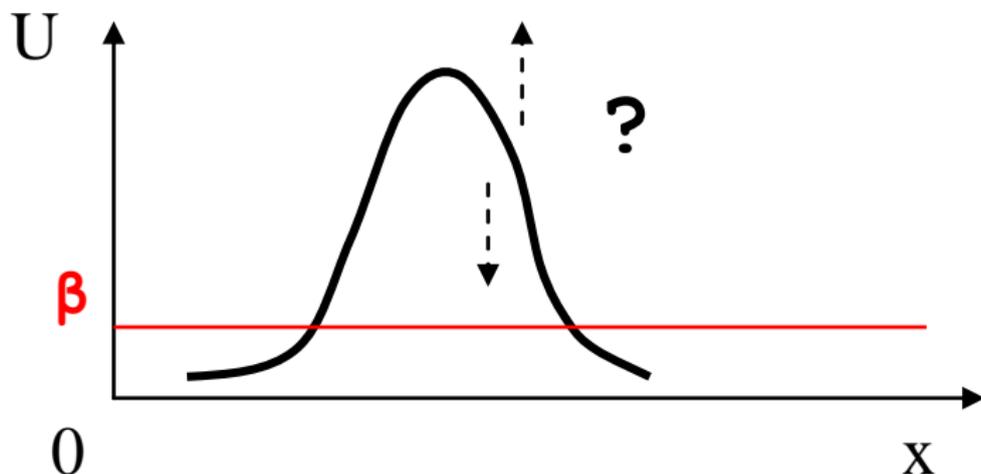
In many cases, this simple model **works very well**.

Invasion of Japanese beetle in the United States:



(Petrovskii & Shigesada, 2001)

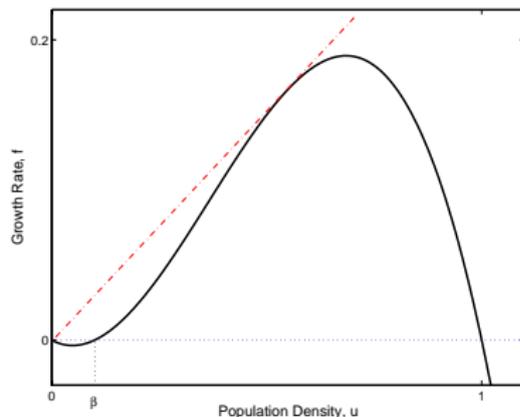
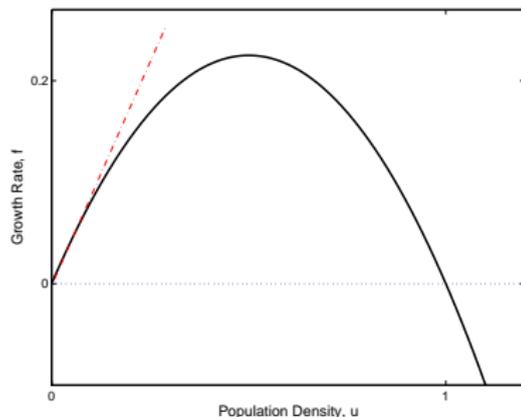
Stage of species establishment



The species will only survive if its maximum density remains above a certain **critical threshold** β .

A Very Simple Model

$$\frac{\partial U(X, T)}{\partial T} = D \frac{\partial^2 U}{\partial X^2} + \alpha U \quad (9)$$
$$(-\infty < X < \infty, T > 0)$$



Solution of Eq. (9) therefore gives an **upper bound** for the solution of the corresponding nonlinear equation.

(by virtue of the comparison principle)

A Very Simple Model

Consider the following initial condition

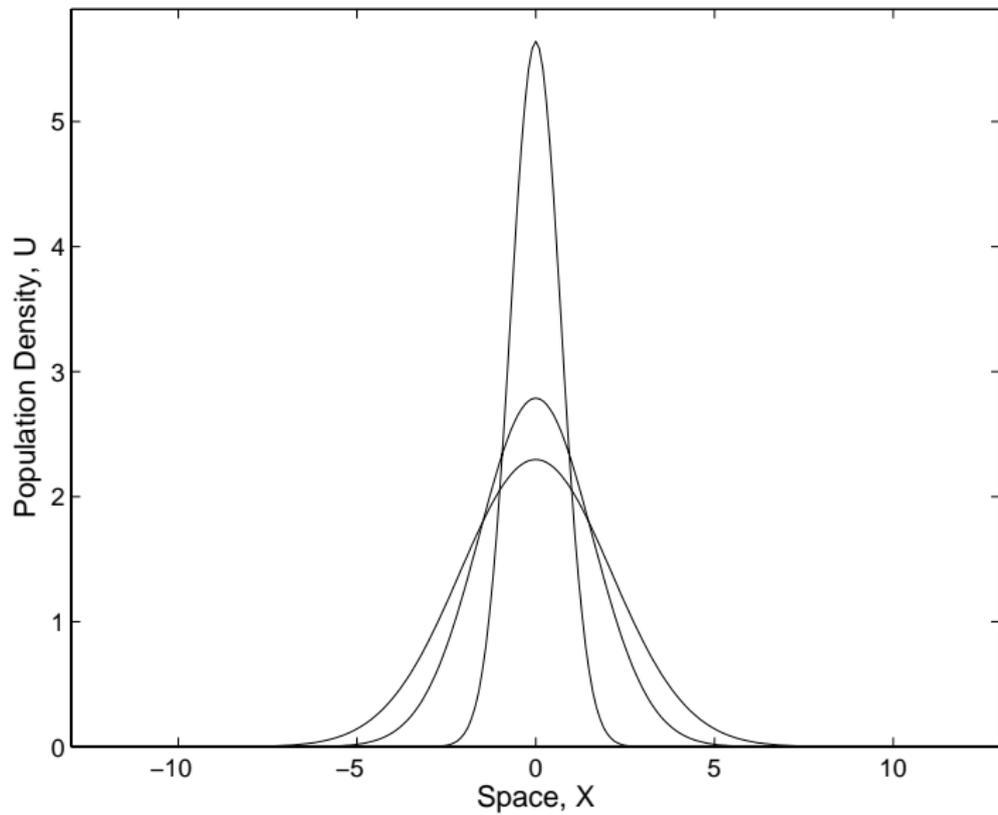
$$U(X, 0) = \frac{G}{\sqrt{4\pi\delta^2}} \exp\left(-\frac{X^2}{4\delta^2}\right),$$

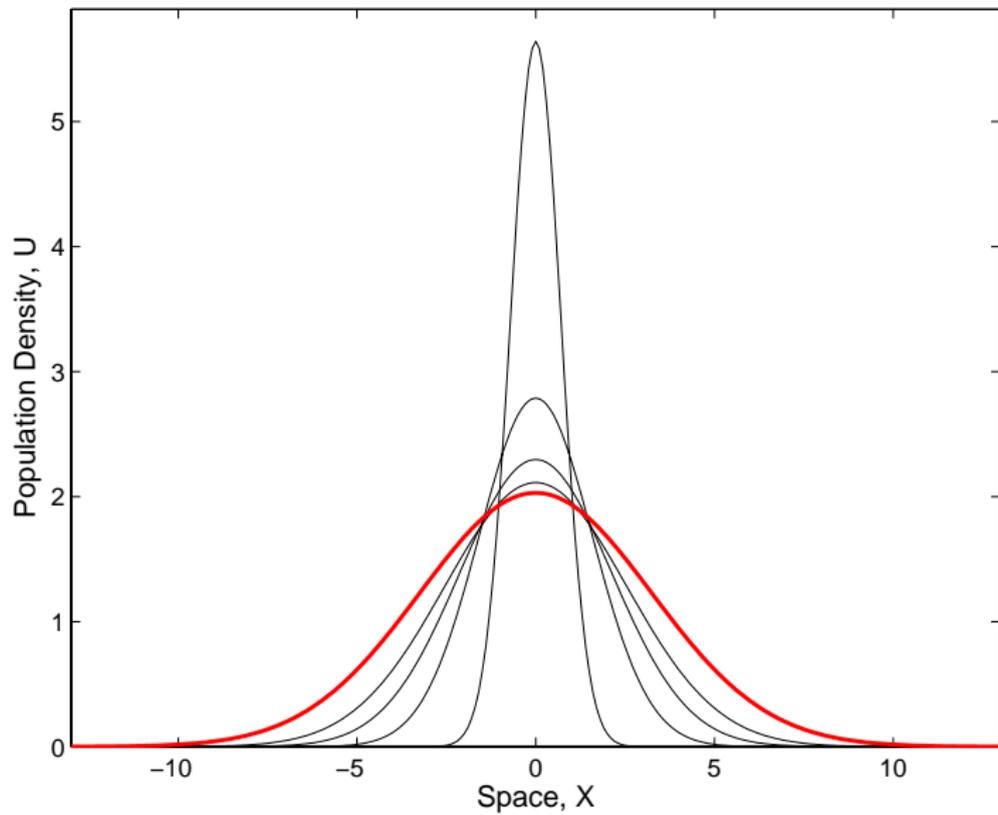
where G is the total initial population size, δ is the **width** and $U_0 = G(4\pi\delta^2)^{-1/2}$ is the **height** of the distribution.

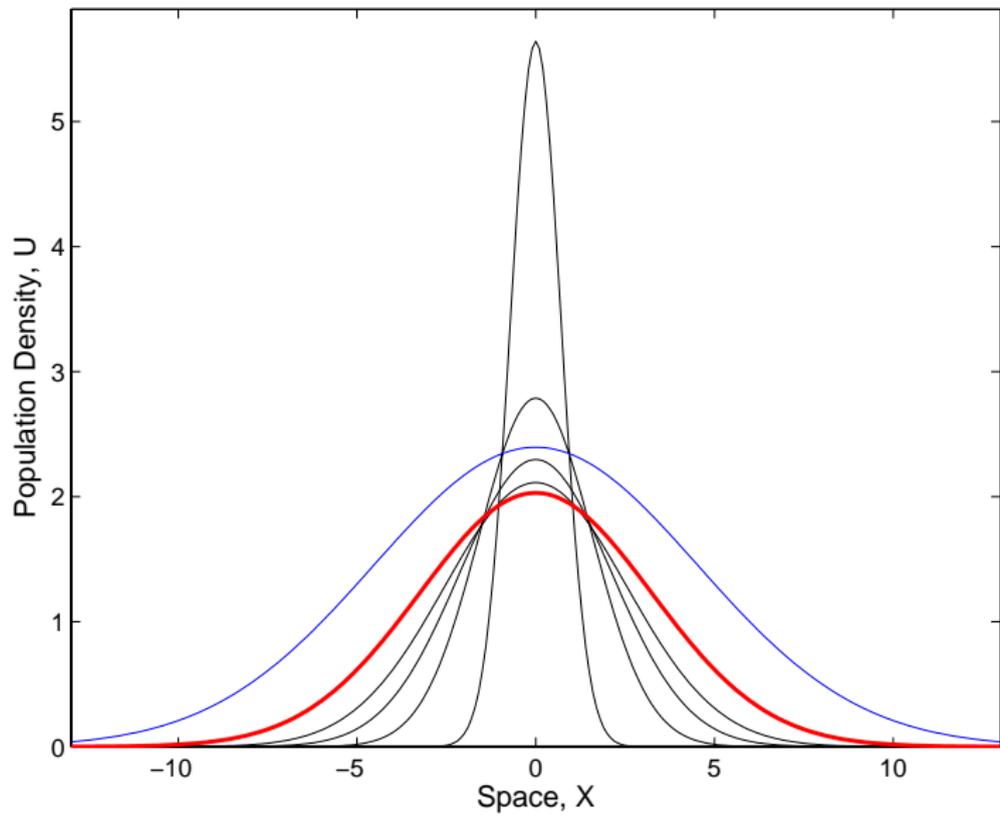
It is readily seen that the corresponding **solution** of the linearized diffusion-reaction equation (9) is

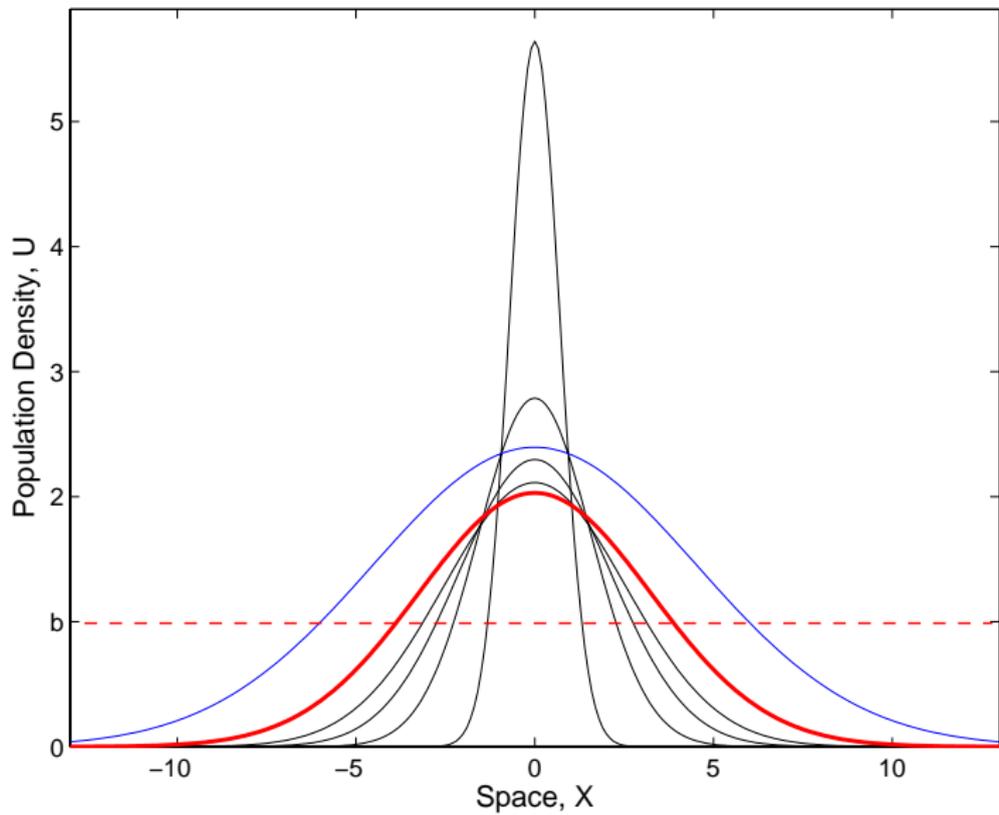
$$U(X, T) = \frac{G}{\sqrt{4\pi(\delta^2 + DT)}} \exp\left(-\frac{X^2}{4(\delta^2 + DT)} + \alpha T\right).$$

What are the **properties** of this solution?









Obviously, the solution has its **maximum** $U_m(T)$ at $X = 0$:

$$U_m(T) = \frac{G \exp(\alpha T)}{\sqrt{4\pi(\delta^2 + DT)}} ,$$

which reaches its minimum at a certain $T_* \geq 0$:

$$U_* = \min_T U_m(T) = U_m(T_*) .$$

Therefore, all we need to do is to **compare** U_* with the extinction threshold β .

It is readily seen that

$$T_* = \frac{1}{2\alpha} - \frac{\delta^2}{D} .$$

Correspondingly, we obtain:

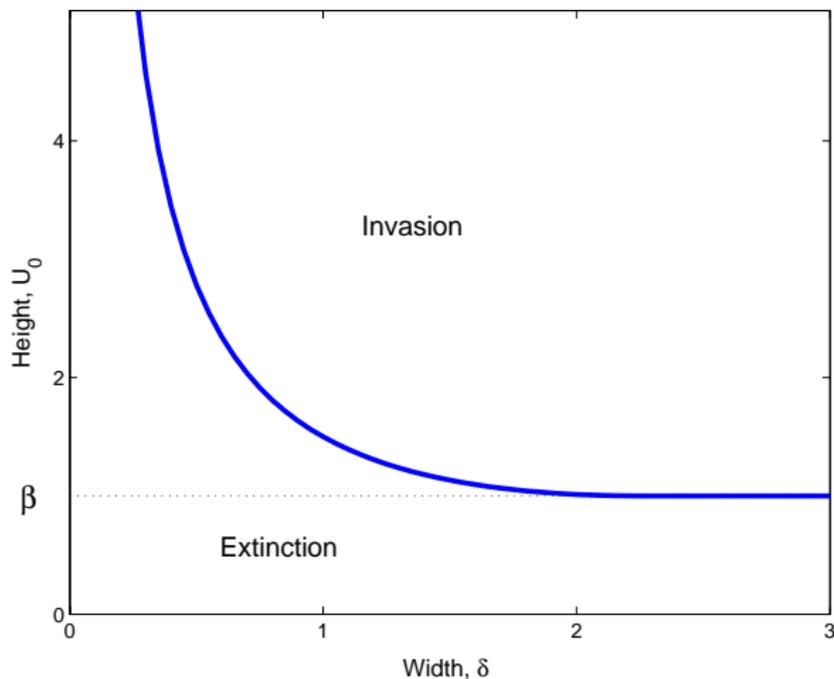
$$U_* = U_m(T_*) \quad \text{for } \delta < \delta_* = \sqrt{\frac{D}{2\alpha}},$$
$$U_* = U_0 \quad \text{for } \delta \geq \delta_*,$$

where

$$U_m(T_*) = G \sqrt{\frac{\alpha}{2\pi D}} \exp\left(\frac{1}{2} - \frac{\alpha\delta^2}{D}\right).$$

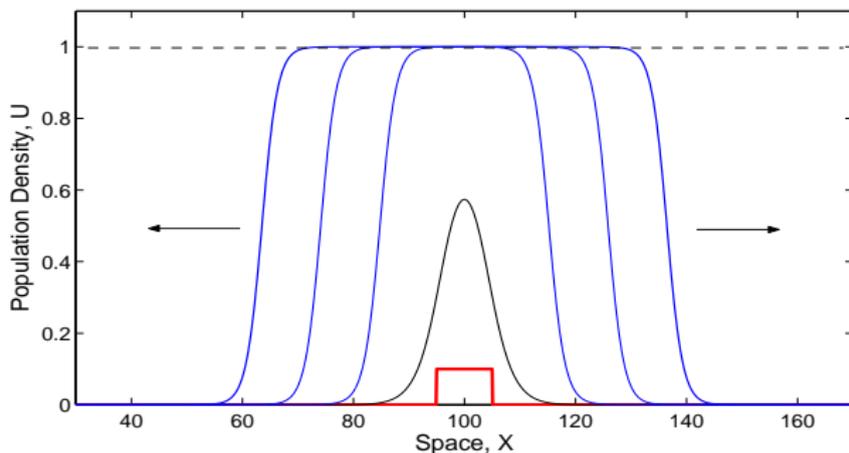
Equating $U_* = \beta$, we obtain the **critical relation** between the height and width of the initial distribution!

Extinction-invasion diagram



More realistic models gives the curve of the same shape.

A **supercritical** initial condition will develop into a traveling wave:



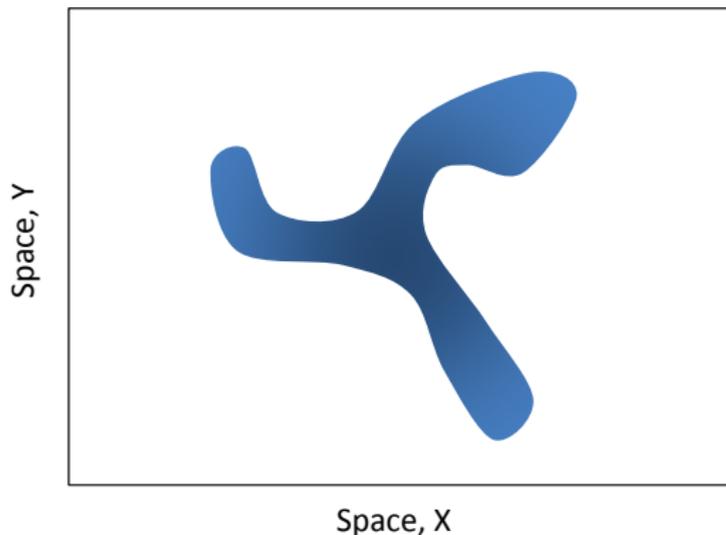
For a **subcritical** initial condition, **strictly speaking**, extinction only can happen in case of the strong Allee effect.

Theorem (Aronson): In case of logistic growth, **any** compact positive initial condition will converge to the traveling wave.

In case of the logistic growth, the threshold is not the inherent property of the model: cf. **“practical stability concept”**

Challenges

The situation is much **less clear** in the 2D case:



The outcome of the species introduction can strongly **depend on the shape** of the originally invaded area!

Heterogeneous environment

$$\frac{\partial U(X, T)}{\partial T} = D(X) \frac{\partial^2 U}{\partial X^2} + f(U, X),$$

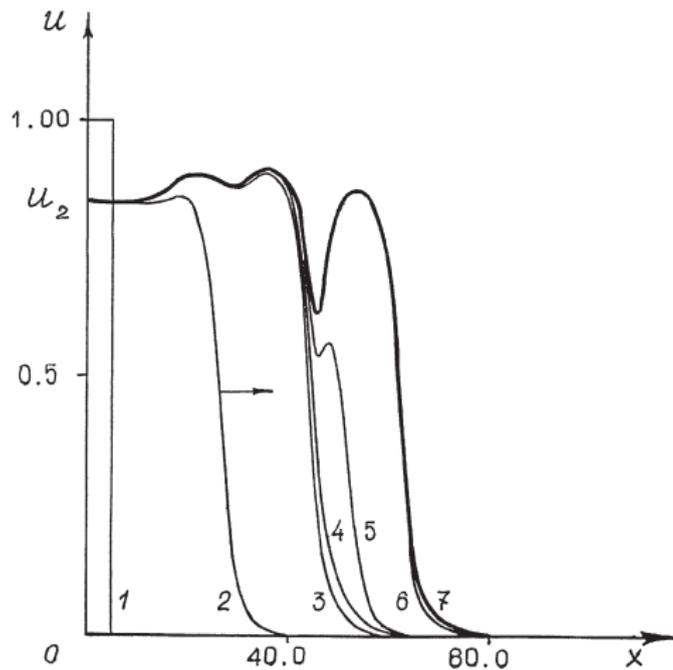
i.e. parameters are now functions of the position in space.

Definition of the traveling wave now becomes less obvious, such as a “non-stationary traveling wave,” $U(X, T) = \phi(\xi, T)$.

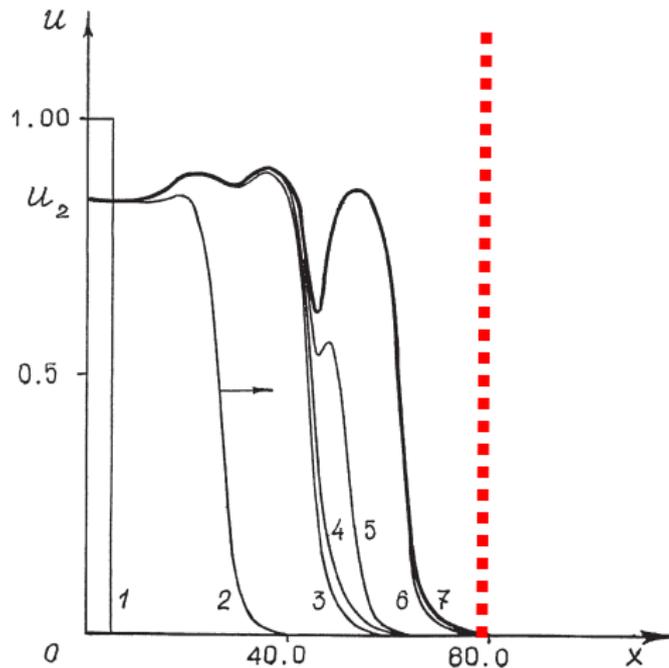
Population distribution behind the front will become heterogeneous, too:

$$U(X, T) \rightarrow \tilde{U}(X) \quad \text{where} \quad D(X) \frac{\partial^2 \tilde{U}}{\partial X^2} + f(\tilde{U}, X) = 0.$$

Heterogeneous environment: a sketch



Heterogeneous environment: a sketch



Is there a possibility of **wave blocking**?

Heterogeneous environment

Example I: **Logistic growth**

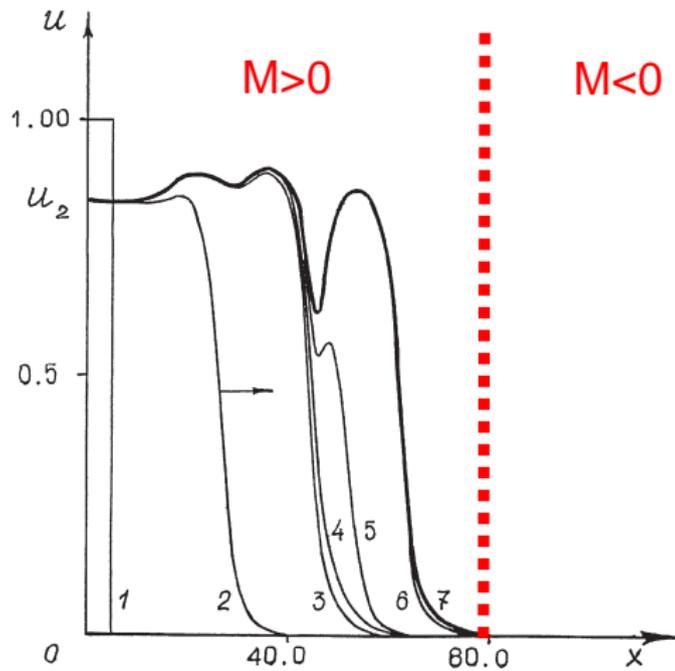
$$f(U, X) = \alpha(X)U \left(1 - \frac{U}{K(X)}\right) .$$

No wave blocking is possible.

Example II: the **strong Allee effect**

$$f(U, X) = \alpha(X)U(\beta - U) \left(1 - \frac{U}{K(X)}\right) .$$

Idea: wave blocking can occur if $M(X) < 0$ in front of the traveling front



A single-hump heterogeneity

Consider $K(X) = \text{const}$ and $\alpha(X) = \alpha_0[1 + \psi(X)]$,

$$f(U, X) = -\alpha_0[1 + \psi(X)]U + \alpha_2 U^2 - \alpha_3 U^3,$$

where $\psi(X)$ is positive only in a compact domain:

$$\psi(X) > 0 \quad \text{for} \quad X_1 < X < X_2,$$

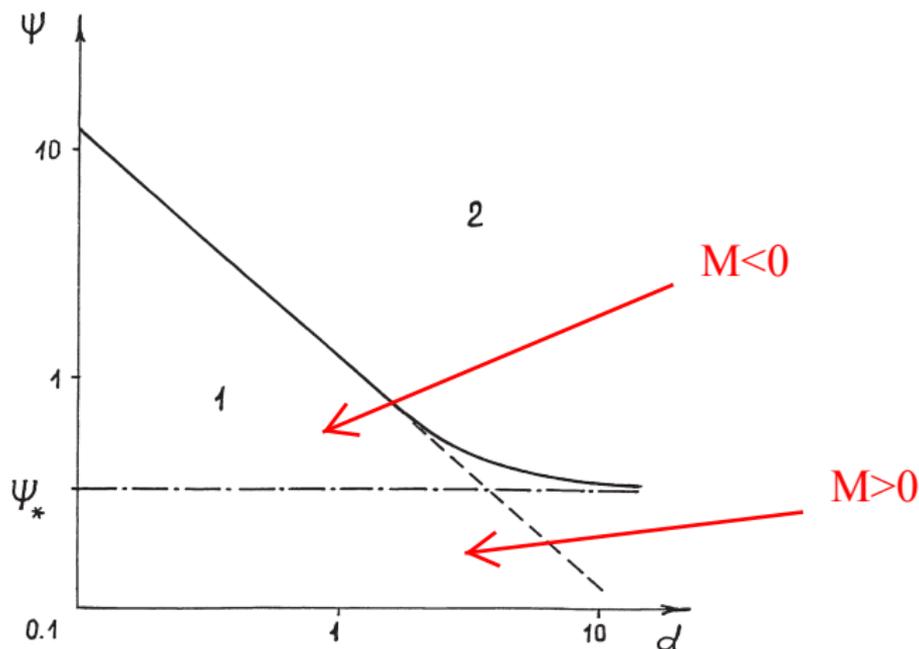
$$\psi(X) \equiv 0 \quad \text{for} \quad X < X_1 \quad \text{or} \quad X > X_2.$$

Heterogeneity leads to a **backward shift** Δ in the front position.

Theorem. Consider $\epsilon = \max \psi(X) \ll 1$. Then $\Delta = \frac{B}{2M} + o(\epsilon)$ where

$$M = \int_0^K f(U) dU \quad \text{and} \quad B = \int_{X_1}^{X_2} \psi(X) dX.$$

The stop-go diagram:



Domain 1 for wave propagation, Domain 2 for **wave blocking**
This property is used in practice (e.g. control of gypsy moth spread)

Chapter IV

PDE models of biological invasion:

Multi-species system

Impact of other species

- This is a very big issue.

Impact of other species

- This is a very **big issue**.
- Most of the **complexity** associated with biological invasions results from the **interspecific interactions**.

(Even in a homogeneous environment: **pattern formation!**)

Impact of other species

- This is a very **big issue**.
- Most of the **complexity** associated with biological invasions results from the **interspecific interactions**.

(Even in a homogeneous environment: **pattern formation!**)
- Being motivated by the problem of **biological control**, we will mostly focus on **two-species** systems such as predator-prey and host-pathogen.

Impact of other species

- This is a very **big issue**.
- Most of the **complexity** associated with biological invasions results from the **interspecific interactions**.

(Even in a homogeneous environment: **pattern formation!**)
- Being motivated by the problem of **biological control**, we will mostly focus on **two-species** systems such as predator-prey and host-pathogen.
- **Biological control** is a strategy in invasive species management when natural factors are used in order **to slow down or to block** the species spread and/or to lead to its eradication

Factors that may affect the pattern of spread

- Heterogeneity of environment
- Human-assisted dispersal (aka “stratified diffusion”)
- Predation?
- Impact of pathogens?
- ... ?

Factors that may affect the pattern of spread

- Heterogeneity of environment
- Human-assisted dispersal (aka “stratified diffusion”)
- Predation?
- Impact of pathogens?
- ... ?

Mathematical model

A predator-prey / host-pathogen system:

$$\frac{\partial U(\mathbf{R}, T)}{\partial T} = D_1 \nabla^2 U(\mathbf{R}, T) + f(U)U - r(U)V ,$$

$$\frac{\partial V(\mathbf{R}, T)}{\partial T} = D_2 \nabla^2 V(\mathbf{R}, T) + \kappa r(U)V - MV$$

Mathematical model

A predator-prey / host-pathogen system:

$$\frac{\partial U(\mathbf{R}, T)}{\partial T} = D_1 \nabla^2 U(\mathbf{R}, T) + f(U)U - r(U)V ,$$

$$\frac{\partial V(\mathbf{R}, T)}{\partial T} = D_2 \nabla^2 V(\mathbf{R}, T) + \kappa r(U)V - MV$$

We consider

- Allee effect for prey/host: $f(U) = \alpha(U - U_0)(K - U)$
- $r(U) = \eta \frac{U}{H+U}$ in case of predation
- $r(U) = \eta U$ in case of pathogens

Mathematical model (contd.)

In dimensionless variables:

$$\frac{\partial u(\mathbf{r}, t)}{\partial t} = \nabla^2 u(\mathbf{r}, t) + \gamma u(u - \beta)(1 - u) - \frac{uv}{1 + \Lambda u},$$

$$\frac{\partial v(\mathbf{r}, t)}{\partial t} = \epsilon \nabla^2 v(\mathbf{r}, t) + \frac{uv}{1 + \Lambda u} - mv.$$

For $\Lambda = 0$, these equations coincide with an SI model of a host-pathogen system.

1D case

Analytical results

The generic pattern of species spread is considered to be the traveling front

Question: can predation **block** or **reverse** the front?

Yes is the answer.

$\epsilon \ll 1$: Owen & Lewis (2001), singular perturbation analysis.

$\epsilon = 1$: Petrovskii et al. (2005), exact solution.

1D case (contd.)

However, the above analytical results were obtained in terms of **traveling waves**. This is a strong assumption.

Can the pattern of spread be **different** from the traveling front?

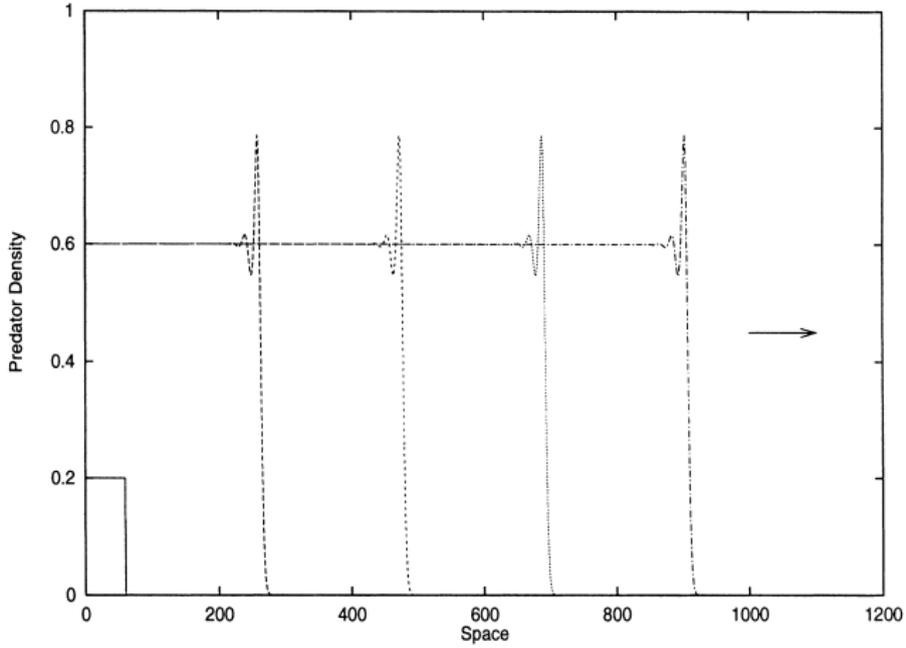
Simulations. Consider the following initial conditions:

$$u(x, 0) = u_0 \quad \text{for } -\Delta_u < x < \Delta_u, \quad \text{otherwise } u(x, 0) = 0,$$

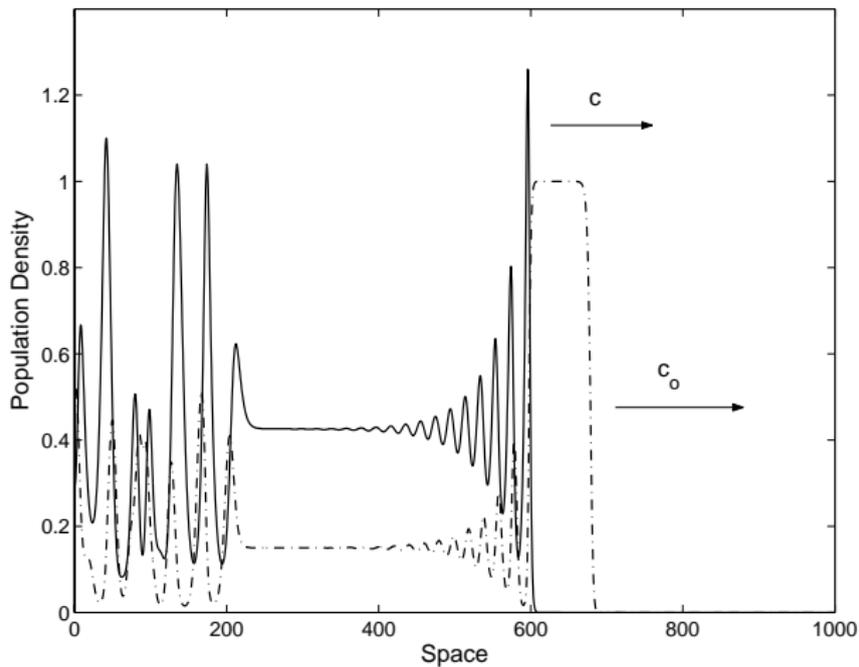
$$v(x, 0) = v_0 \quad \text{for } -\Delta_v < x < \Delta_v, \quad \text{otherwise } v(x, 0) = 0,$$

where u_0 , v_0 are the initial population densities,
 Δ_u and Δ_v give the radius of the initially invaded domain.

1D case (contd.)

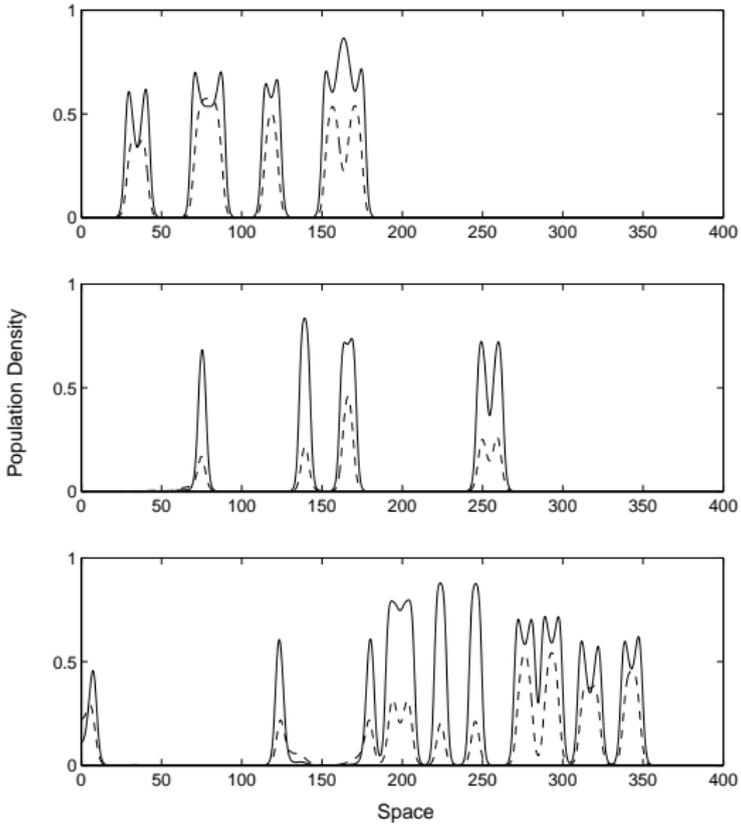


1D case (contd.)

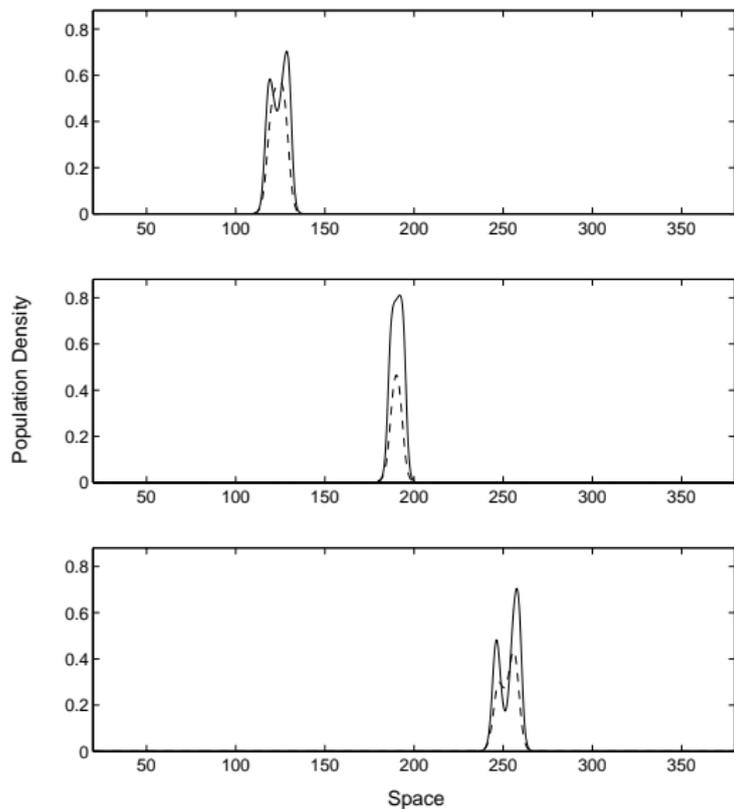


(Sherratt et al., 1995; Petrovskii & Malchow, 2000, 2002; Petrovskii et al., 2001)

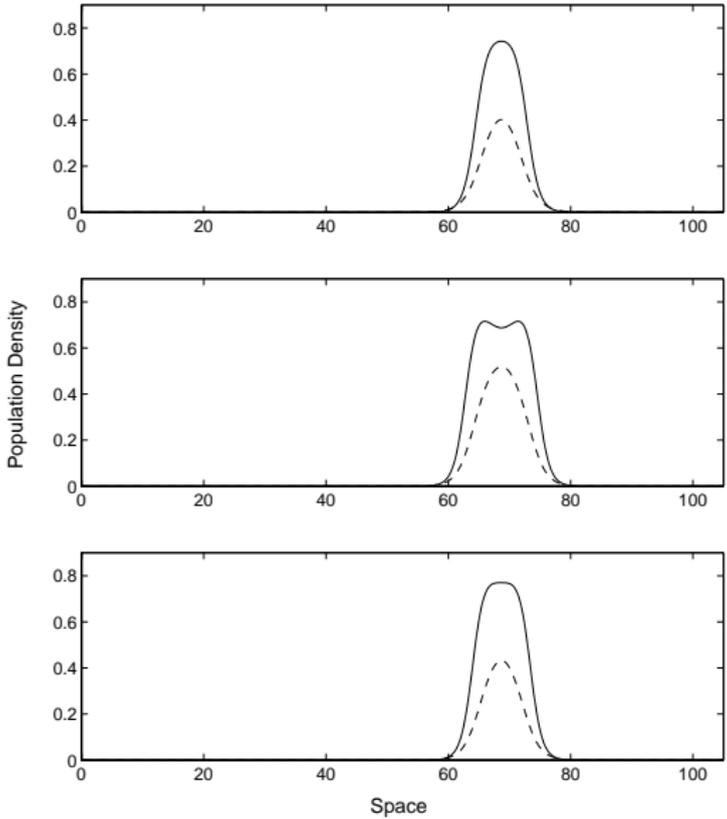
1D case (contd.)



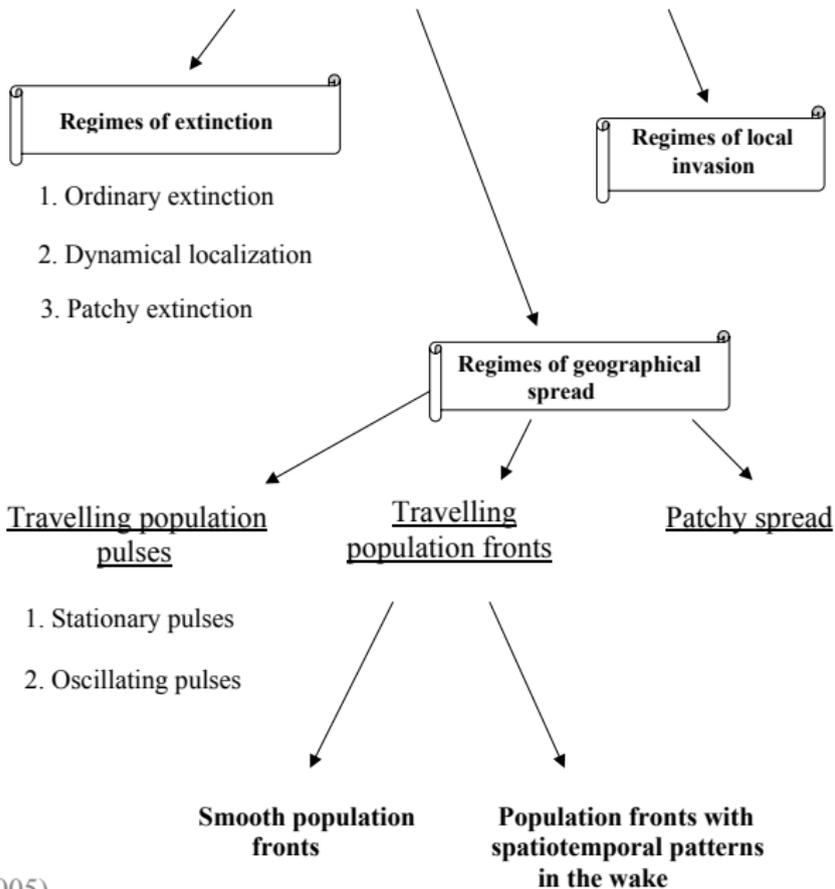
1D case (contd.)



1D case (contd.)



Regimes of biological invasion in a predator-prey system



2D case, simulations

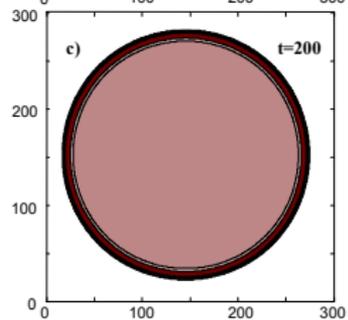
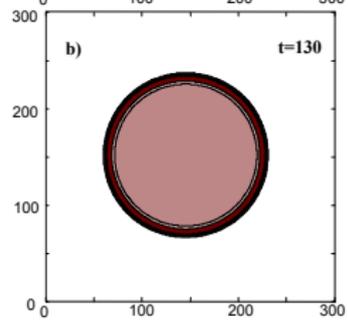
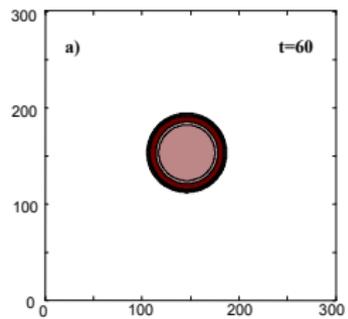
The initial conditions:

$$u(x, y, 0) = u_0 \text{ for } x_{11} < x < x_{12}, y_{11} < y < y_{12},$$

$$u(x, y, 0) = 0 \text{ otherwise,}$$

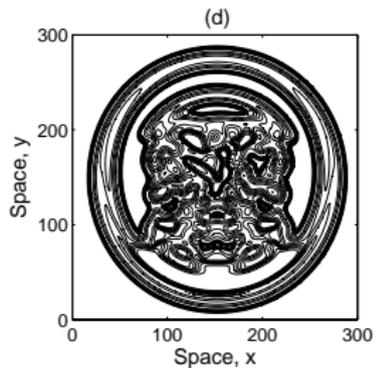
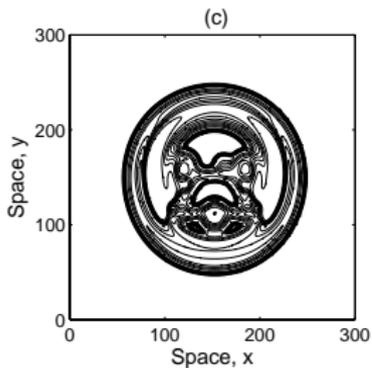
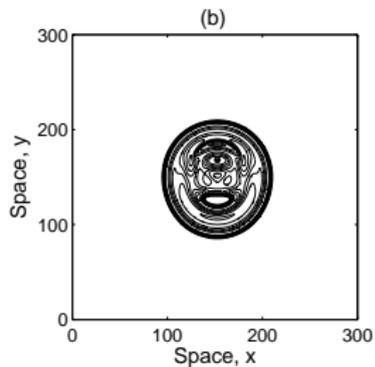
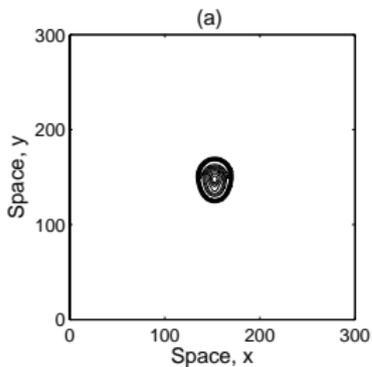
$$v(x, y, 0) = v_0 \text{ for } x_{21} < x < x_{22}, y_{21} < y < y_{22},$$

$$v(x, y, 0) = 0 \text{ otherwise.}$$



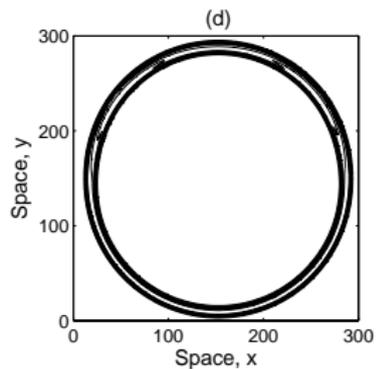
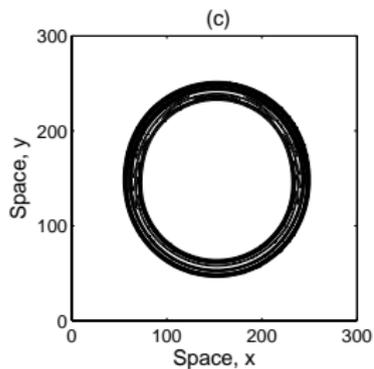
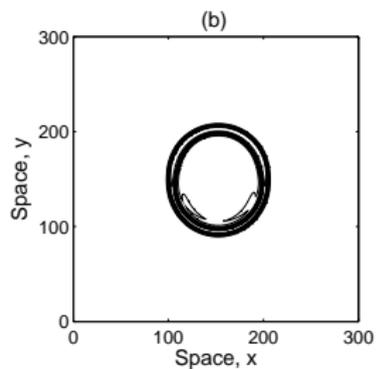
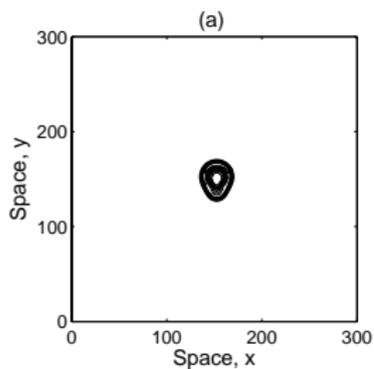
2D case (contd.)

Chaos in the wake

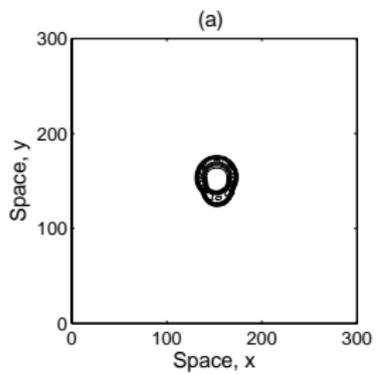


2D case (contd.)

Spread without invasion

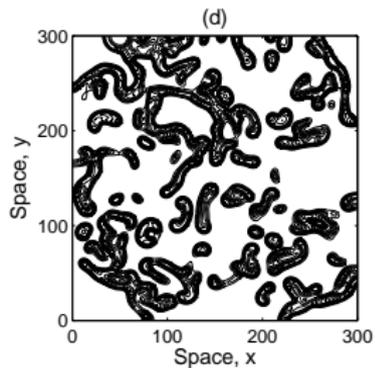
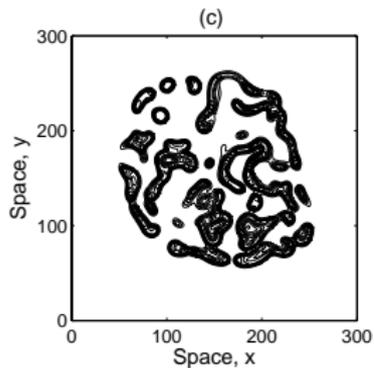
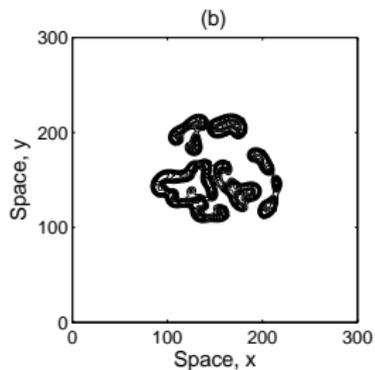
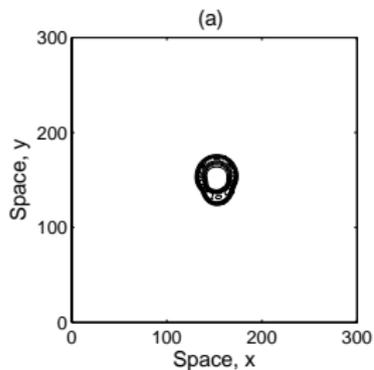


2D case (contd.)



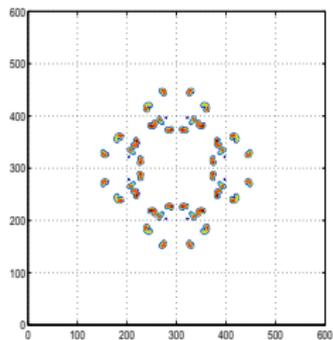
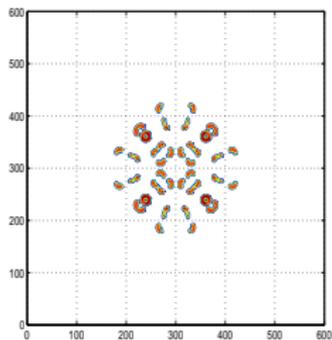
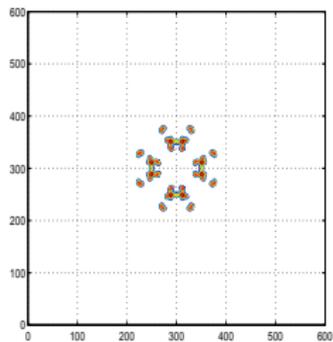
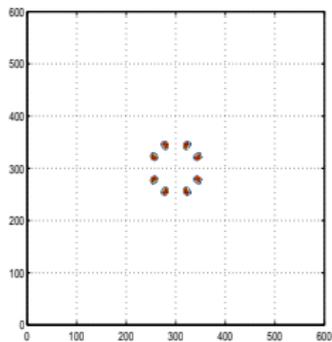
2D case (contd.)

Patchy invasion

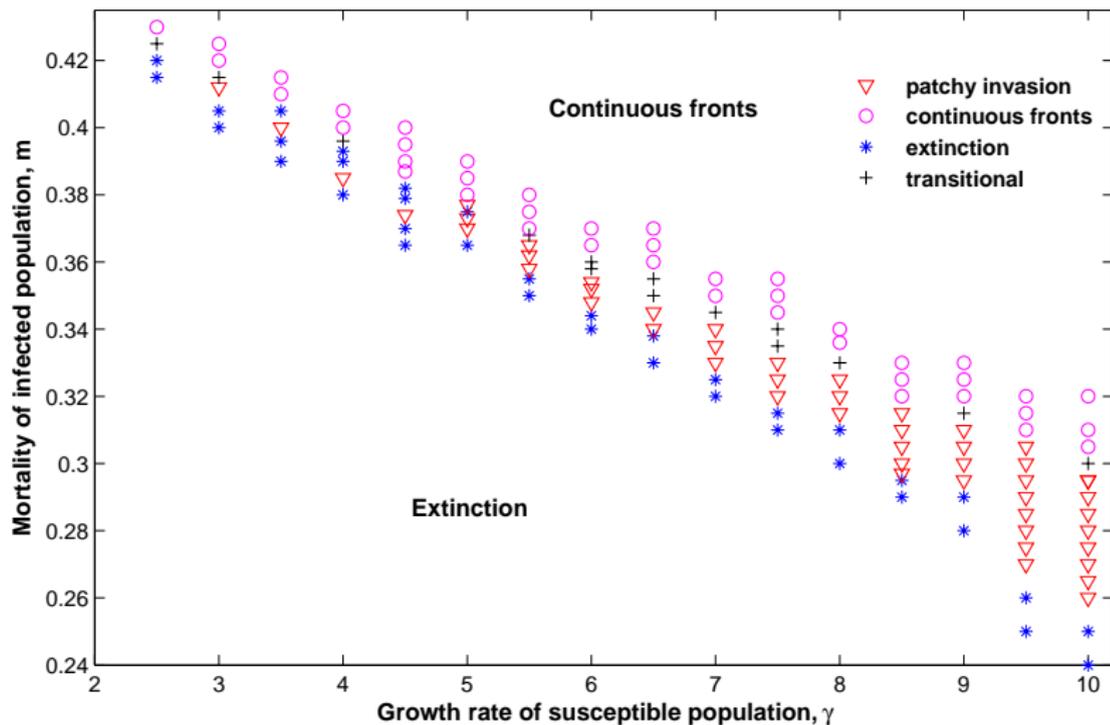


2D case (contd.)

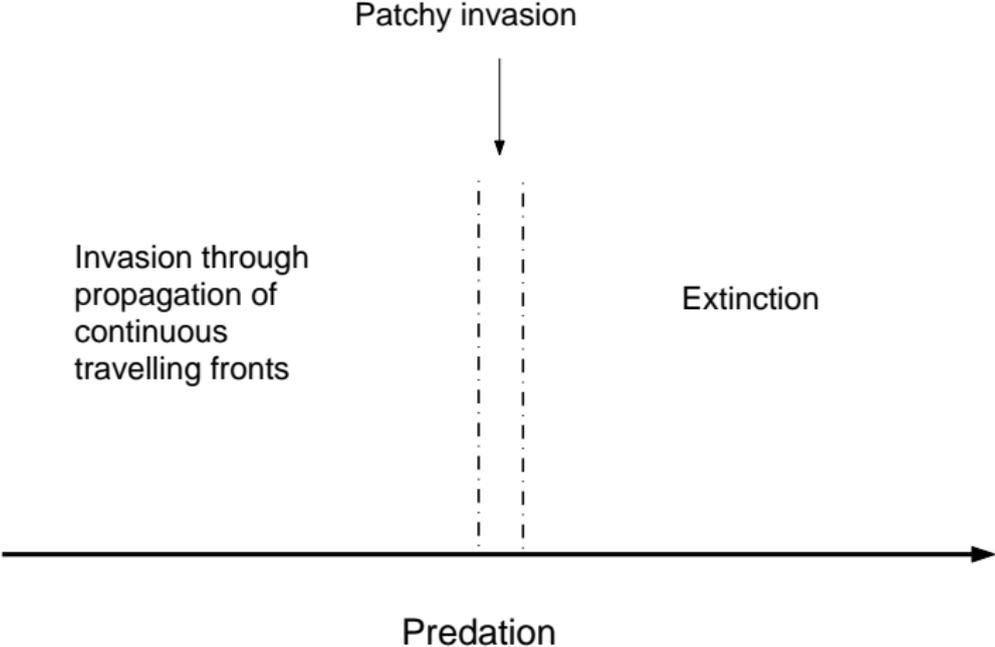
Patchy invasion
for symmetric initial conditions:



Summary of simulation results

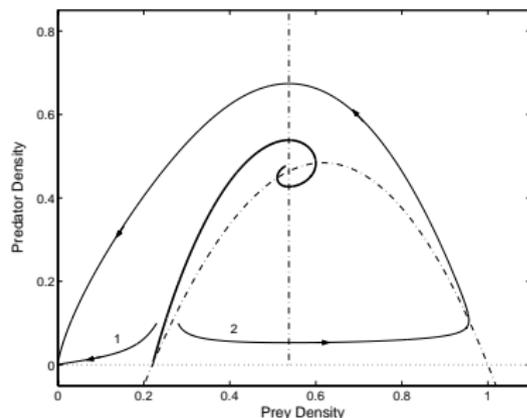
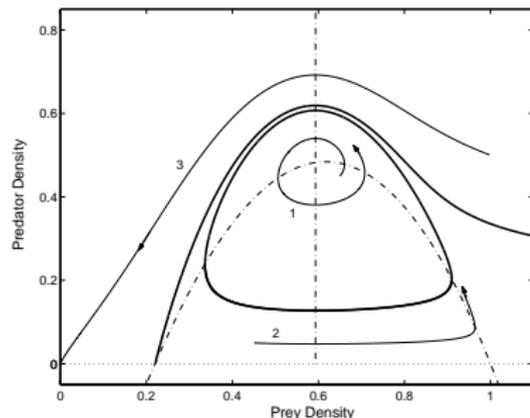


Invasion at the edge of extinction



Without space:

Predation/infection strength increases from left to right:



Limit cycle disappears through a heteroclinic bifurcation

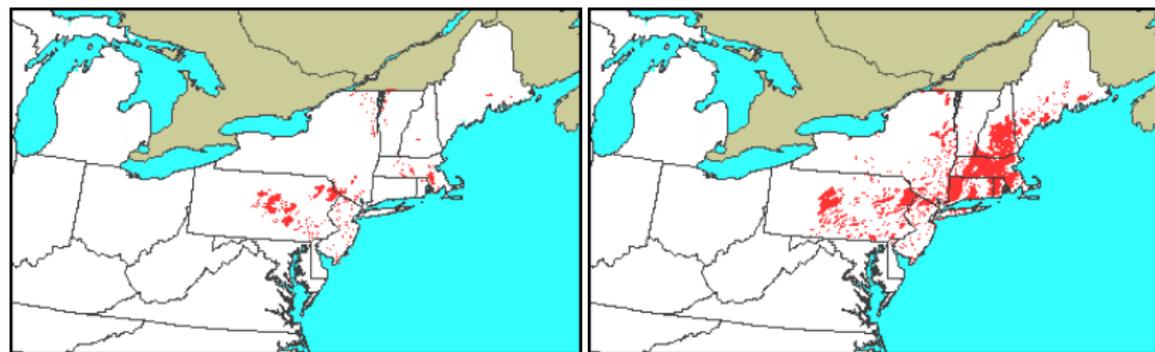
Patchy spread occur when the nonspatial system **is not viable!**

(Malchow, Petrovskii & Venturino, 2008)

Theory against data: gypsy moth in the USA

1977

1981



The red spots show heavily infested areas

(from the US Forest Service, www.fs.fed.us/ne/morgantown)

The **calculated** rate of spread is between 0.36 and 3.3 km/year
against the **observed** rate between 2 and 10 km/year

(Jankovic & Petrovskii, 2013)

The end of Part 1