Stochastic Differential Equations Invasion of the Killer Shrimp

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Outline

- 1 Background on Killer Shrimp
- 2 Stochastics
- 3 Modeling

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- Background on Killer Shrimp
 - The Problem
- 2 Stochastics
 - Fundamental Concepts
 - Integration
 - Itô's Formula
- 3 Modeling
 - The Mode
 - Nondimensionalization

The Problem

Southern German water routes have had several drastic population changes concerning gammarids (Kinzler, 2008) [2].

Much of this is due to canal construction.

Native Species:

Gammarus pulex (Gp)

Invasive Species:

Dikerogammarus villosus (Dv) Dikerogammarus haemobaphes (Dh) Dikerogammarus bispinosus (Db) Echinogammarus berilloni (Eb)

Killer Shrimp



http://www.arkive.org/killer-shrimp/dikerogammarus-villosus/image-G143154.html

Killer Shrimp

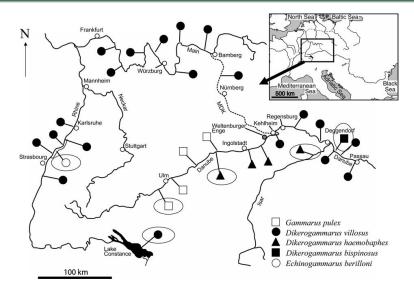


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(Kinzler, 2008)

Time Line

Years	Species
<1976	<i>Gp</i> native
1976-1994	Dh invades
1992-1995	Dv invades, Dh declines
>1995	All but <i>Dv</i> coexist

Kinzler, 2008 Study

Isolated pairs of specimens in a controlled environment

- One freshly moulted (prey)
- One predator
- Total of 279 experiments
- Grouped by age, sex, and species *

Results

- Found Dv to be clear strongest predator
- Found *Dh* to have highest cannibalism rate
- * Intraguild: predation between different species; Intraspecific: predation within species (cannibalism) [3]



Basic Goals

Determine long term population trends!

Does *Dv* totally dominate in the end?

Why are we seeing this?

Why is Dv dominating?

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Markov Process

(Markov Process) is a stochastic process with the following properties:

- The outcome at any stage depends only on the outcome of the previous stage.
- The probabilities are constant over time.

Brownian Motion

Definition

(**Brownian Motion**) is a stochastic process that models random continuous motion. The stochastic process $B = \{B(t), t \ge 0\}$ is standard Brownian Motion if the following holds:

- ① *B* has independent increments.
- ② For $0 \le s < t$,

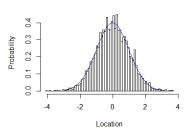
$$B(t) - B(s) \sim N(0, t - s).$$

- 3 The paths of B are continuous with probability 1.
- B(0) = 0

Monte Carlo Simulation - Brownian Motion

Recorded final step of 5000 Brownian Motions over [0,1] We expect the distribution to be N(0,1).

Distribution of Location at time T



$$\bar{x} = 0.019311$$
, $s^2 = 0.9933$, $t - test \ p - value = 0.16924$

CI:
$$\mu \in (-0.00822, 0.046844)$$

Scaling a Brownian Motion

Background on Killer Shrimp

If x(t) is a Brownian Motion then $\frac{x(\lambda t)}{\sqrt{\lambda}}$ is a Brownian motion.

$$P\left(a \le \frac{x(\lambda t)}{\sqrt{\lambda}} - \frac{x(\lambda s)}{\sqrt{\lambda}} \le b\right)$$

$$= P\left(a\sqrt{\lambda} \le x(\lambda t) - x(\lambda s) \le b\sqrt{\lambda}\right),$$

$$= \frac{1}{\sqrt{2\pi(\lambda t - \lambda s)}} \int_{a\sqrt{\lambda}}^{b\sqrt{\lambda}} e^{\frac{-x^2}{2}(\lambda t - \lambda s)} dx,$$

$$= \frac{1}{\sqrt{2\pi(t - s)}} \int_{a}^{b} e^{\frac{-u^2}{t - s}} du.$$

Wiener Process

A standard Wiener process (also called Brownian Motion) is a stochastic process $\{W_t\}_{t\geq 0}$, which has properties mutually consistent with those of Brownian motion.

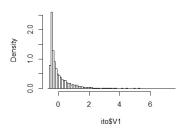
Wiener Integral For a pair $(W_t, f(t))$ of a Wiener Process W_t , a random process f(t), we define the Itô integral

$$I(f) = \int_0^a f(t)dW_t$$

Monte Carlo Simulation - Itô Integral

Recorded approximations of 5000 Itô Integrals for $\int_0^{1.5} WdW$ WHAT DO WE EXPECT???? I DON'T KNOW.

Histogram of ito\$V1



$$\bar{x} = 0.000555, \quad s^2 = 0.4925248$$

Itô's Formula

Itô's Formula is used in Itô Calculus to find the differential of a time-dependent function of a stochastic process.

Differential Form

$$dx_{t} = \left(\frac{\partial x}{\partial t} + \frac{1}{2}\left(\frac{\partial^{2} x}{\partial B^{2}}\right)\right) \cdot dt + \frac{\partial x}{\partial B} \cdot dB$$

Integral Form

$$F(t,B(t)) - F(a,B(a)) = \int_{a}^{t} \frac{\partial F}{\partial s} + \frac{1}{2} \frac{\partial^{2} F}{\partial B^{2}} ds + \int_{a}^{t} \frac{\partial F}{\partial B} dB$$

Itô's Formula

Let
$$F = tB^2$$

$$\frac{dF}{dt} = B^2$$

$$\frac{dF}{dB} = 2t \cdot B$$

$$\frac{d^2F}{dB^2} = 2t$$

$$tB^{2}(t) - aB^{2}(a) = \int_{a}^{t} B^{2} ds + \int_{a}^{t} 2sB dB + \int_{a}^{t} \frac{1}{2}2s ds$$

$$OR$$

$$tB^{2}(t) - aB^{2}(a) = \int_{a}^{t} B^{2} ds + \int_{a}^{t} 2sB dB + \frac{1}{2}t^{2} - \frac{1}{2}a^{2}$$

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Shrimp Model

Our model is

$$\frac{dx}{dt} = rx^{2} \left(1 - \frac{x}{K} \right) - \alpha xy - \frac{x^{2} \gamma_{o}}{x + D},$$

$$\frac{dy}{dt} = \rho y^{2} \left(1 - \frac{y}{L} \right) - \beta xy - \frac{y^{2} \delta_{o}}{y + R},$$

where x is the population of Dv and y is the population of Dh, with parameters $[r, K, \alpha, \gamma_{\circ}, D, \rho, L, \beta, \delta_{\circ}, R]$.

Nondimensionalization

The initial model is

$$\frac{dx}{dt} = rx^{2} \left(1 - \frac{x}{K} \right) - \alpha xy - \frac{x^{2} \gamma_{o}}{x + D},$$

$$\frac{dy}{dt} = \rho y^{2} \left(1 - \frac{y}{L} \right) - \beta xy - \frac{y^{2} \delta_{o}}{y + R}.$$

Let

$$x \to A\hat{x}(s)$$

 $y \to B\hat{y}(s)$
 $t \to \tau \cdot s$

When you substitute and group terms the system becomes nondimensionalized.

After making the following substitutions:

$$A = K,$$

$$B = L,$$

$$\tau = \frac{1}{rK}.$$

The nondimensionalized system is:

$$\frac{dx}{dt} = x^2(1-x) - \alpha xy - \frac{\gamma_0 x^2}{x+D},$$

$$\frac{dy}{dt} = \rho y^2(1-y) - \beta xy - \frac{\delta_0 y^2}{y+R}$$

What happens if we add noise?

$$\frac{dx}{dt} = x^2(1-x) - \alpha xy - \frac{\gamma_o x^2}{x+D} + \text{"noise,"}$$

$$\frac{dy}{dt} = \rho y^2(1-y) - \beta xy - \frac{\delta_o y^2}{y+R} + \text{"noise."}$$

If we add additive noise...

$$\frac{dx}{dt} = x^{2}(1-x) - \alpha xy - \frac{\gamma_{o}x^{2}}{x+D} + v\frac{dW}{dt},$$

$$\frac{dy}{dt} = \rho y^{2}(1-y) - \beta xy - \frac{\delta_{o}y^{2}}{y+R} + \kappa \frac{dB}{dt}.$$

If we add proportional noise...

$$\frac{dx}{dt} = x^2(1-x) - \alpha xy - \frac{\gamma_0 x^2}{x+D} + \upsilon x \frac{dW}{dt},$$

$$\frac{dy}{dt} = \rho y^2(1-y) - \beta xy - \frac{\delta_0 y^2}{y+R} + \kappa y \frac{dB}{dt}.$$

Heun's Method

- Heun's method is a numerical procedure for approximating ordinary differential equations with a given initial value.
- First you calculate the intermediate value \tilde{y}_{i+1} and then the final approximation y_{i+1} at the next generation point.

$$\tilde{y}_{i+1} = y_i + \Delta t \ f(t_i, y_i)$$

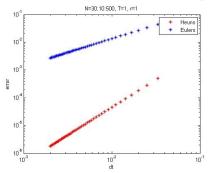
$$y_{i+1} = y_i + \frac{\Delta t}{2} [f(y_i, t_i) + f(\tilde{y}_{i+1}, t_{i+1})]$$

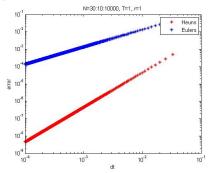
Heun's Method vs. Euler's Method - Simulation

For the DE y' = ry on [0, T],

Heun's:
$$y_{i+1} = y_i + \frac{\Delta t}{2} [f(y_i, t_i) + f(\tilde{y}_{i+1}, t_{i+1})]$$

Euler's: $\tilde{y}_{i+1} = y_i + \Delta t \ f(y_i, t_i)$





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