# Stochastic Differential Equations Introduction

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#### Outline

- 1 Background on Shrimpy
- 2 Brownian Motion
- 3 Modeling
- 4 Nondimensionalization

Background on Shrimpy

- Background on Shrimpy
  - The Problem
- - Fundamental Concepts
- Nondimensionalization

#### The Problem

Southern German water routes have had several drastic population changes concerning gammarids[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

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http://www.arkive.org/killer-shrimp/dikerogammarus-villosus/image-G143156.html



(Kinzler, 2008)

Background on Shrimpy

The time to prominence.

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

# **Terminology**

Intraguild Predation

Mutual Predation

Mutual Interference

Intraspecific Predation

Cannibalism

[3]

# 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

#### Basic Goals

# Determine long term population trends!

Does *Dv* totally dominate in the end?

Which species survive?

Is there an equilibrium?

## Outline

- 2 Brownian Motion
  - Fundamental Concepts
  - Integration
  - Itô's Formula
- Nondimensionalization

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

- 1 The number of possible outcomes or states is finite
- 2 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

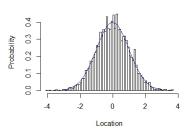
Random Walk Random Walk [1]

### Monte Carlo Simulation - Brownian Motion

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

#### Distribution of Location at time T

Modeling



$$\bar{x} = 0.01931075$$
,  $s^2 = 0.993332$ ,  $p - value = 0.1692423$   
CI:  $\mu \in (-0.0082225, 0.04684401)$ 

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

Modeling

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

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

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

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

# Riemann-Stieltjes

#### (Riemann-Stieltjes Integral):

$$\int_a^b f(g)dg = \lim_{n \to \infty} \sum_{i=1}^N f(g(t_i)) \cdot (g(t_{i+1})) - g(t_i)$$

The main motivation for the Riemann-Stieltjes Integral comes from the concept of Cumulative Distribution Function (CDF) of a random variable.

#### 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

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

Background on Shrimpy

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

Modeling

#### Differential Form

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

#### Integral Form

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

Modeling

## Itô's Formula

Let 
$$F = tR^2$$

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

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

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

$$tB^{2}(t) - aB^{2}(a) = \int_{a}^{t} B^{s} ds + \int_{a}^{t} 2sBdB + \int_{a}^{t} \frac{1}{2}2sds$$
$$= \int_{a}^{b} B^{2} + sds + \int_{a}^{t} 2sBdB$$
OR

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

Modeling

### Outline

- - Fundamental Concepts
- Modeling
- Nondimensionalization

# Shrimp Model

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

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

Modeling

- - Fundamental Concepts
- 4 Nondimensionalization

The initial model is:

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

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

Modeling

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.

$$A = K,$$

$$B = L,$$

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

Modeling

The nondimensionalized system is:

$$\frac{dx}{dt} = rx(1-x) - \alpha xy,$$
  
$$\frac{dy}{dt} = y(1-y) - \beta xy.$$

#### Heun's Method

 Heun's method is a numerical procedure for approximating ordinary differential equations with a given initial value.

Modeling

• First you calculate the intermediate value  $\tilde{y}_{i+1}$  and then the final approximation  $y_{i+1}$  at the next generation point.

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

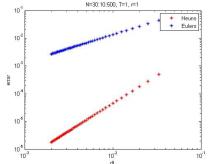
$$y_{i+1} = y_i + \frac{\Delta t}{2} \left[ f(y_i, t_i) + f(\tilde{y}_{i+1}, t_{i+1}) \right]$$

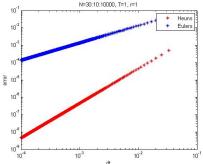
## Heun's Method vs. Euler's Method - Simulation

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

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

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





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