

Interim Report

# Predicting Algae Bloom spread in the Marlborough Sounds using Bayesian Inference

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## Introduction

This is an interim technical report describing the current state of my Metropolis-Hastings algorithm for Predicting Algae Bloom spread, using a simplified model involving a singular event.

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# Our Model

## Mass

Mass is observed as:

- $n$  Sites and  $T$  discrete Times
- Growth parameter  $\alpha$
- Decay parameter  $\beta$
- Dispersion Matrix  $P$  where  $p_{ij}$  is the proportion of Mass at site  $i$

that travels to site  $j$  (Such that  $\sum_{j=1}^n p_{ij} = 1$  for all  $i \in \{1, \dots, n\}$ )

- $N$  events (Simplified to 1)
- Each of  $k \in \{1, \dots, N\}$  events has:
  - Starting site  $I_k$  with uniform probability  $\frac{1}{n}$
  - Starting time  $T_k \in \{1, \dots, T\}$
  - Length of time  $W_k$
  - Starting size  $S_k$

Where the equation for  $M_{i,t}$  is:

$$M_{i,t} = \left( \alpha A_{i,t} + \beta \sum_{j=1}^n p_{i,j} \right) M_{i,t-1} + B_{i,t}$$

Where

$$A_{i,t} = I \left( \sum_{k=1}^n I(I_k = i) I(T_k \leq t \leq T_k + W_k) > 0 \right)$$

$$B_{i,t} = \sum_{k=1}^n I(I_k = i) I(T_k = t) S_k$$

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## Concentration

Concentration is calculated from Mass where:

- Each site  $i \in \{1, \dots, n\}$  has volume  $V_i$
- $\delta$  is a concentration offset of approximately  $0^+$
- Precision parameter  $\tau$

Where the equation for  $C_{i,t}$  is

$$\log(Y_{i,t}) \sim N\left(\log\left(\delta + \frac{M_{i,t}}{V_i}\right), \frac{1}{\tau^2}\right)$$

## Parameters

### Known Parameters:

- Number of sites,  $n$
- Time period,  $T$
- Dispersion Matrix,  $P$
- Number of events,  $N$
- Volumes,  $V_i$
- Concentration offset,  $\delta$

### Parameters to be estimated

- Growth Parameter,  $\alpha$
- Decay Parameter,  $\beta$
- Event site,  $I_k$
- Event Starting time,  $T_k$
- Event time length,  $W_k$
- Event size,  $S_k$
- Precision  $\tau$

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## Priors

$$\alpha \sim \text{Gamma}(a_{\alpha}, b_{\alpha})$$

$$\beta \sim \text{Beta}(a_{\beta}, b_{\beta})$$

$$N = 1$$

$$I_k \sim \text{DiscreteUniform}(1, n)$$

$$T_k \sim \text{DiscreteUniform}(1, T)$$

$$W_k \sim \text{NegBinomial}(r_w, p_w)$$

$$S_k \sim \text{Gamma}(a_s, b_s)$$

$$\tau \sim \text{Gamma}(a_{\tau}, b_{\tau})$$

Where:

$$a, b = 1$$

$$r_w = 6$$

$$p_w = 0.75$$

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## Metropolis-Hastings Algorithm

*Generate Sample  $Y$  with unknown parameters*

*Generate initial set of parameters  $\theta$  from priors*

*For  $s$  in 1: 10000:*

*For each parameter in  $\{\alpha, \beta, I, T, W, S, \tau\}$ :*

*Generate candidate value from proposal distributions*

*For  $p$  in  $\{p^*, p_{\text{current}}\}$ :*

*Generate sample  $M^*$*

*Calculate  $P(p)$  from prior distribution*

*Calculate  $P(Y | p, \theta/p)$  from Likelihood distribution using  $M^*$*

*Calculate probability =  $P(p) \times P(Y | p, \theta/p)$*

*Calculate ratio =  $\min\left(\frac{\text{probability}_{p^*}}{\text{probability}_{p_{\text{current}}}}, 1\right)$*

*Randomly generate  $u \sim U(0, 1)$*

*If ratio  $> u$ :*

*Set  $p_{\text{current}} = p^*$*

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## Proposal Distributions

$$\alpha^* \sim N(\alpha_{\text{current}}, 1)$$

$$\beta^* \sim N(\beta_{\text{current}}, 0.5)$$

$$I_k^* \sim \text{DiscreteUniform}(1, n)$$

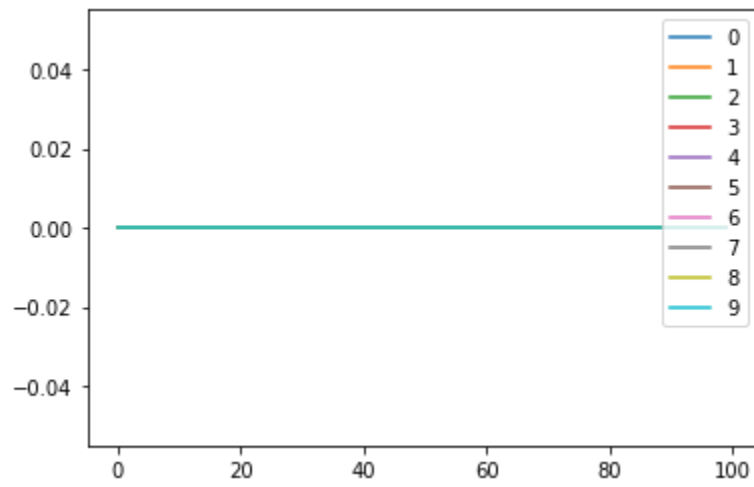
$$T_k^* \sim \text{DiscreteUniform}(1, T)$$

$$W_k^* \sim \text{DiscreteGamma}(1, 1)$$

$$S_k^* \sim \text{Gamma}(S_{\text{current}}, 1)$$

$$\tau^* \sim N(\tau_{\text{current}}, 1)$$

## Mass - True values



## Results

Parameter	Acceptance Probability	Walk	Histogram	Prior PDF
$\alpha$	0.5237			
$\beta$	0.4506			
$I_0$	0.9999			
$T_0$	0.9999			



