

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.

Our Model

Mass

Mass is observed as:

- n Sites and T discrete Times
- Growth parameter α
- Decay parameter β
- 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:

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

Where

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

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

Concentration

Concentration is calculated from Mass where:

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

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, δ

Parameters to be estimated

- Growth Parameter, α
- Decay Parameter, β
- Event site, I_k
- Event Starting time, T_k
- Event time length, W_k
- Event size, S_k
- Precision τ

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:

$$a, b = 1$$

$$r_w = 6$$

$$p_w = 0.75$$

Pseudocode

Find M_{New}

Inputs:

- M
- α
- β
- P
- A
- B

$$growth = \alpha \times (AM)$$

$$transport = \beta \times (P \cdot M)$$

$$M_{New} = B + growth + transport$$

Outputs:

- M_{New}

RunSimulation

Inputs:

- n
- T
- P

$$\alpha = 2$$

$$\beta = 0.8$$

$$N = 1$$

For x in 1: N :

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

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

$$w \sim \text{NegBinom}(6, 0.75) + 1$$

$$s \sim \text{Gamma}(1, 1)$$

$$A_{t:\min(t+w, T), i} = 1$$

$$B_{t, i} = s$$

$$[M_{0,0}, \dots, M_{n,0}] = 0$$

For t in 1: T :

$$M_{\text{New}} = \text{Find}M_{\text{New}}([M_{0,t-1}, \dots, M_{n,t-1}], \alpha, \beta, P_{i:n, i:n}, [A_{t,1}, \dots, A_{t,n}], [B_{t,1}, \dots, B_{t,n}])$$

$$[M_{0,t}, \dots, M_{n,t}] = M_{\text{New}}$$

Outputs:

- $[[M_{0,1}, \dots, M_{n,1}], \dots, [M_{0,T}, \dots, M_{n,T}]]$
- α
- β
- N
- $[I_0, \dots, I_N]$
- $[T_0, \dots, T_N]$
- $[W_0, \dots, W_N]$
- $[S_0, \dots, S_N]$

GetSample

Inputs:

- n
- T
- P
- α
- β
- N
- $[I_1, \dots, I_N]$
- $[t_1, \dots, t_N]$
- $[W_1, \dots, W_N]$
- $[s_1, \dots, s_N]$

For x in $1:N$:

$$A_{t_x: \min(t_x + W_x, t), I_x} = 1$$

$$B_{t,i} = s_x$$

$$[M_{0,0}, \dots, M_{n,0}] = 0$$

For t in $1:T$:

$$M_{New} = FindM_{New}([M_{0,t-1}, \dots, M_{n,t-1}], \alpha, \beta, P_{i:n, i:n}, [A_{t,1}, \dots, A_{t,n}], [B_{t,1}, \dots, B_{t,n}])$$

$$[M_{0,t}, \dots, M_{n,t}] = M_{New}$$

Outputs:

- $[[M_{0,1}, \dots, M_{n,1}], \dots, [M_{0,T}, \dots, M_{n,T}]]$

ConvertConcentration

Inputs:

- n
- T
- M
- V
- δ

For i in $1:n$:

$$C_{i, 1:T} = \frac{M_{i, 1:T}}{V_i}$$

$$\text{Log}C = \log(C + \delta)$$

Outputs:

- $\text{Log}C$

ConvertY

Inputs:

- C
- τ

$$\text{Log}Y \sim N\left(C, \frac{1}{\sqrt{\tau}}\right)$$

Outputs:

- $\text{Log}Y$

Metropolis-Hastings Algorithm

Inputs:

- P

$$n = 10$$

$$T = 100$$

$$\delta = 1^{-10}$$

$$\tau_{True} \sim \text{Gamma}(1, 1)$$

$$[V_1, \dots, V_n] \sim \text{NegBinom}(1, 1) + 1$$

$$[M_{True}, \alpha_{True}, \beta_{True}, N_{True}, I_{True}, T_{True}, W_{True}, S_{True}] = \text{runSimulation}(n, T, P)$$

$$C_{True} = \text{ConvertCalculation}(n, T, M_{True}, V, \delta)$$

$$Y_{True} = \text{ConvertY}(C_{True}, \tau_{True})$$

$$\alpha_0 \sim \text{Gamma}(1, 1)$$

$$\beta_0 \sim \text{Beta}(1, 1)$$

$$N = N_{True}$$

$$[I_{0,1}, \dots, I_{0,N}] \sim \text{DiscreteUniform}(1, n)$$

$$[t_{1,1}, \dots, t_{0,N}] \sim \text{DiscreteUniform}(1, T)$$

$$[w_{0,1}, \dots, w_{0,N}] \sim \text{NegBinom}(6, 0.75)$$

$$[s_{0,1}, \dots, s_{0,N}] \sim \text{Gamma}(1, 1)$$

$$\tau_0 \sim \text{Gamma}(1, 1)$$

$$\Phi = [\alpha_0, \beta_0, N, [I_{0,1}, \dots, I_{0,N}], [t_{0,1}, \dots, t_{0,N}], [w_{0,1}, \dots, w_{0,N}], [s_{0,1}, \dots, s_{0,N}], \tau_0]$$

For each parameter $_0$ in Φ :

$$\text{Array}_{parameter} = \text{parameter}_0$$

$$\text{Accept}_{parameter} = 0$$

For s in 1: 200000:

Alpha

$$\alpha_{Candidate} \sim N(\alpha_{Candidate}, 1)$$

$$M_{Candidate} = \text{GetSample}(n, T, \alpha_{Candidate}, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = \text{GetSample}(n, T, \phi_1, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Current} = \text{ConvertConcentration}(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = \text{Gamma}(1, 1).PDF(\alpha_{Candidate})$$

$$p_{prior_{Current}} = \text{Gamma}(1, 1).PDF(\phi_1)$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Q_{Candidate}} = N(\phi_1, 1).PDF(\alpha_{Candidate})$$

$$p_{Q_{Current}} = N(\alpha_{Candidate}, 1).PDF(\phi_1)$$

$$ratio_{\alpha} = \min \left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1 \right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\Phi_1 = \alpha_{Candidate}$$

$$Accept_\alpha = Accept_\alpha + 1$$

$$Likelihood_{\alpha_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

else:

$$Likelihood_{\alpha_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

Beta

$$\beta_{Candidate} \sim \text{Beta}(1, 1)$$

$$M_{Candidate} = \text{GetSample}(n, T, \Phi_1, \beta_{Candidate}, \Phi_3, \Phi_4, \Phi_5, \Phi_6)$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = \text{GetSample}(n, T, \Phi_1, \Phi_2, \Phi_3, \Phi_4, \Phi_5, \Phi_6)$$

$$C_{Current} = \text{ConvertConcentration}(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = \text{Beta}(1, 1).PDF(\beta_{Candidate})$$

$$p_{prior_{Current}} = \text{Beta}(1, 1).PDF(\Phi_2)$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Q_{Candidate}} = \text{Beta}(1, 1).PDF(\beta_{Candidate})$$

$$p_{Q_{Current}} = \text{Beta}(1, 1).PDF(\phi_2)$$

$$ratio_{\beta} = \min\left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1\right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\phi_2 = \beta_{Candidate}$$

$$Accept_{\beta} = Accept_{\beta} + 1$$

$$Likelihood_{\beta_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log}\left(\delta + N\left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}}\right) \cdot PDF(Y_{True})\right) \right) \right)$$

else:

$$Likelihood_{\beta_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log}\left(\delta + N\left(C_{Current}, \frac{1}{\sqrt{\phi_7}}\right) \cdot PDF(Y_{True})\right) \right) \right)$$

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for k in 1: n :

$$I_{k_{Candidate}} \sim \text{DiscreteUniform}(1, n)$$

$$I_{Candidate} = \phi_3$$

$$I_{Candidate_k} = I_{k_{Candidate}}$$

$$M_{Candidate} = \text{GetSample}(n, T, \phi_1, \phi_2, I_{Candidate}, \phi_4, \phi_5, \phi_6)$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = GetSample(n, T, \phi_1, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Current} = ConvertConcentration(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = DiscreteUniform(1, n).PMF(I_{Candidate_k})$$

$$p_{prior_{Current}} = DiscreteUniform(1, n).PMF(\phi_{3_k})$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}} \right) \right) \cdot PDF(Y_{True}) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\phi_7}} \right) \right) \cdot PDF(Y_{True}) \right) \right)$$

$$p_{Q_{Candidate}} = DiscreteUniform(1, n).PDF(I_{Candidate_k})$$

$$p_{Q_{Current}} = DiscreteUniform(1, n).PDF(\phi_{3_k})$$

$$ratio_{I_k} = \min \left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1 \right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\phi_3 = I_{Candidate}$$

$$Accept_I = Accept_I + \frac{1}{N}$$

$$Likelihood_{I_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}} \right) \right) \cdot PDF(Y_{True}) \right) \right)$$

else:

$$Likelihood_{I_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

T

for k in $1:n$:

$$T_{k_{Candidate}} \sim DiscreteUniform(1, T - 1)$$

$$T_{Candidate} = \Phi_4$$

$$T_{Candidate_k} = T_{k_{Candidate}}$$

$$M_{Candidate} = GetSample(n, T, \Phi_1, \Phi_2, \Phi_3, T_{Candidate}, \Phi_5, \Phi_6)$$

$$C_{Candidate} = ConvertConcentration(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = GetSample(n, T, \Phi_1, \Phi_2, \Phi_3, \Phi_4, \Phi_5, \Phi_6)$$

$$C_{Current} = ConvertConcentration(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = DiscreteUniform(1, T).PMF(T_{Candidate_k})$$

$$p_{prior_{Current}} = DiscreteUniform(1, T).PMF(\Phi_{4_k})$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Q_{Candidate}} = \text{DiscreteUniform}(1, T - 1).PDF(T_{Candidate_k})$$

$$p_{Q_{Current}} = \text{DiscreteUniform}(1, T - 1).PDF(\phi_{4_k})$$

$$ratio_{T_k} = \min\left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1\right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\phi_4 = T_{Candidate}$$

$$Accept_T = Accept_T + \frac{1}{N}$$

$$Likelihood_{T_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log}\left(\delta + N\left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}}\right).PDF(Y_{True})\right) \right) \right)$$

else:

$$Likelihood_{T_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log}\left(\delta + N\left(C_{Current}, \frac{1}{\sqrt{\phi_7}}\right).PDF(Y_{True})\right) \right) \right)$$

W

for k in $1:n$:

$$W_{k_{Candidate}} \sim \text{Binom}\left(T - \phi_{4_k}, \frac{1 + \phi_{5_k}}{2 + T}\right)$$

$$W_{Candidate} = \phi_5$$

$$W_{Candidate_k} = W_{k_{Candidate}}$$

$$M_{Candidate} = \text{GetSample}(n, T, \phi_1, \phi_2, \phi_3, \phi_4, W_{Candidate}, \phi_6)$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = \text{GetSample}(n, T, \phi_1, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Current} = \text{ConvertConcentration}(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = \text{DiscreteUniform}(1, T).PMF(W_{Candidate_k})$$

$$p_{prior_{Current}} = \text{DiscreteUniform}(1, T).PMF(\phi_{5_k})$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

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$$p_{Q_{Candidate}} = \text{Binom} \left(T - \phi_{4_k}, \frac{1 + \phi_{5_k}}{2 + T} \right) \cdot PMF(W_{Candidate_k})$$

$$p_{Q_{Current}} = \text{Binom} \left(T - \phi_{4_k}, \frac{1 + W_{Candidate_k}}{2 + T} \right) \cdot PMF(\phi_{5_k})$$

$$ratio_{W_k} = \min \left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1 \right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\phi_5 = W_{Candidate}$$

$$Accept_W = Accept_W + \frac{1}{N}$$

$$Likelihood_{W_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

else:

$$Likelihood_{W_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

S

for k in 1: n :

$$S_{k_{Candidate}} \sim N(\Phi_{6_k}, 1)$$

$$S_{Candidate} = \Phi_6$$

$$S_{Candidate_k} = S_{k_{Candidate}}$$

$$M_{Candidate} = \text{GetSample}(n, T, \Phi_1, \Phi_2, \Phi_3, \Phi_4, \Phi_5, S_{Candidate})$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = \text{GetSample}(n, T, \Phi_1, \Phi_2, \Phi_3, \Phi_4, \Phi_5, \Phi_6)$$

$$C_{Current} = \text{ConvertConcentration}(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = \text{Gamma}(1, 1).PDF(S_{Candidate_k})$$

$$p_{prior_{Current}} = \text{Gamma}(1, 1).PDF(\Phi_{6_k})$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}', \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}', \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Q_{Candidate}} = N \left(\Phi_{6_k}, 1 \right) \cdot PDF \left(S_{Candidate_k} \right)$$

$$p_{Q_{Current}} = N \left(S_{Candidate_k}, 1 \right) \cdot PDF \left(\Phi_{6_k} \right)$$

$$ratio_{S_k} = \min \left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1 \right)$$

$$u \sim U(0, 1)$$

if $ratio > u$:

$$\Phi_{6_k} = S_{Candidate_k}$$

$$Accept_S = Accept_S + \frac{1}{N}$$

$$Likelihood_{S_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}', \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

else:

$$Likelihood_{S_{s,k}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}', \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

Tau

$$\tau_{Candidate} \sim N(\phi_7, 1)$$

$$M_{Candidate} = \text{GetSample}(n, T, \phi_1, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Candidate} = \text{ConvertConcentration}(n, T, M_{Candidate}, V, \delta)$$

$$M_{Current} = \text{GetSample}(n, T, \phi_1, \phi_2, \phi_3, \phi_4, \phi_5, \phi_6)$$

$$C_{Current} = \text{ConvertConcentration}(n, T, M_{Current}, V, \delta)$$

$$p_{prior_{Candidate}} = \text{Gamma}(1, 1).PDF(\tau_{Candidate})$$

$$p_{prior_{Current}} = \text{Gamma}(1, 1).PDF(\phi_7)$$

$$p_{Likelihood_{Candidate}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\tau_{Candidate}}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

$$p_{Q_{Candidate}} = N(\phi_7, 1).PDF(\tau_{Candidate})$$

$$p_{Q_{Current}} = N(\tau_{Candidate}, 1).PDF(\phi_7)$$

$$ratio_{\alpha} = \min \left(\frac{p_{prior_{Candidate}} \times p_{Likelihood_{Candidate}}}{p_{prior_{Current}} \times p_{Likelihood_{Current}}} \times \frac{p_{Q_{Current}}}{p_{Q_{Candidate}}}, 1 \right)$$

$$u \sim U(0, 1)$$

if ratio > u:

$$\Phi_7 = \tau_{Candidate}$$

$$Accept_{\tau} = Accept_{\tau} + 1$$

$$Likelihood_{\tau_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Candidate}, \frac{1}{\sqrt{\tau_{Candidate}}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

else:

$$Likelihood_{\tau_s} = \sum_{y=1}^T \left(\sum_{x=1}^N \left(\text{Log} \left(\delta + N \left(C_{Current}, \frac{1}{\sqrt{\Phi_7}} \right) \cdot PDF(Y_{True}) \right) \right) \right)$$

Proposal Distributions

$$\alpha^* \sim N(\phi_1, 1)$$

$$\beta^* \sim \text{Beta}(1, 1)$$

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

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

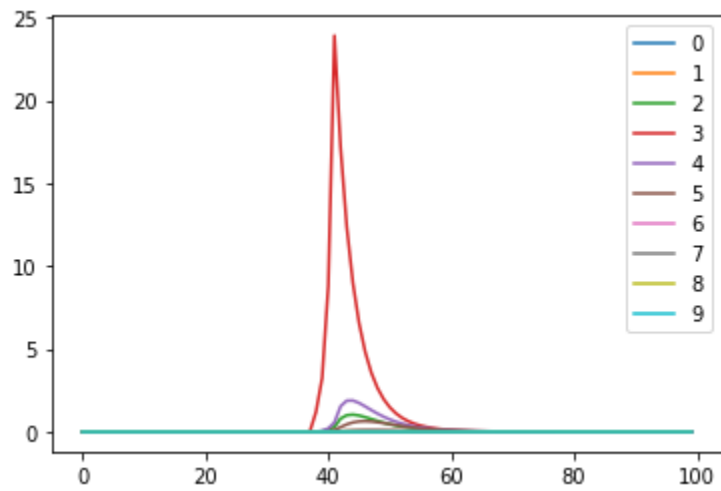
$$W_k^* \sim \text{Binom}\left(T - \phi_{4_k}, \frac{1 + \phi_{5_k}}{2 + T}\right)$$

$$S_k^* \sim N(\phi_{6_k}, 1)$$

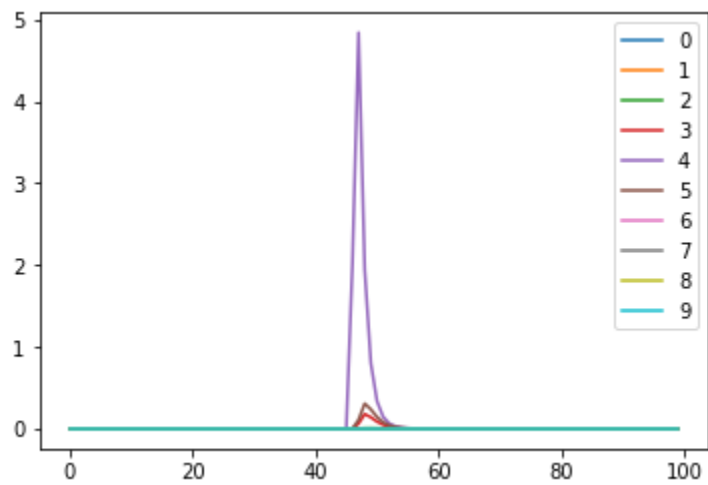
$$\tau^* \sim N(\phi_7, 1)$$

Results

Mass - True values



Mass - Estimated Parameters



Parameter P(Acceptance)		Walk of value	Walk of Likelihood	Scatter Plot of Likelihood	Histogram	Prior PDF
α	0.5216					
β	0.9296					
I_0	0.1004					
T_0	0.9095					
W_0	0.6281					
S_0	0.5242					
τ	0.5314					