

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
- 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

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

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

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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

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

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

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

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

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

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

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

Where:

$$a_{\beta}, a_s, a_{\tau}, b = 1$$

$$a_{\alpha} = 2$$

$$r_w = 6$$

$$p_w = 0.75$$

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## Pseudocode

### Find $M_{New}$

Inputs:

- $M$
- $\alpha$
- $\beta$
- $P$
- $A$
- $B$

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

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

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

Outputs:

- $M_{New}$

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## RunSimulation

Inputs:

- $n$
- $T$
- $P$

$$\alpha \sim \text{Gamma}(2, 1)$$

$$\beta \sim \text{Beta}(4, 3)$$

$$N \sim \text{DiscreteUniform}\left(1, \frac{T}{10}\right)$$

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:

- $[[M_{0,1}, \dots, M_{n,1}], \dots, [M_{0,T}, \dots, M_{n,T}]]$
- $\alpha$
- $\beta$
- $N$
- $[I_{0,1}, \dots, I_{n,1}]$
- $[T_{0,1}, \dots, T_{n,1}]$
- $[W_{0,1}, \dots, W_{n,1}]$
- $[S_{0,1}, \dots, S_{n,1}]$

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## GetSample

Inputs:

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

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:

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



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## ConvertConcentration

Inputs:

- $n$
- $T$
- $M$
- $V$
- $\delta$

For  $i$  in  $1:n$ :

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

$$C_{log} = \log(C + \delta)$$

Outputs:

- $C_{log}$

## ConvertY

Inputs:

- $C$
- $\tau$

$$Y_{log} \sim N\left(C, \frac{1}{\sqrt{\tau}}\right)$$

Outputs:

- $Y_{log}$

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## Metropolis-Hastings Algorithm

Inputs:

- $P$

$$n = 10$$

$$T = 100$$

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

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

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

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

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

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

$$\alpha_0 \sim \text{Gamma}(2, 1)$$

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

$$N = N_{True}$$

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

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

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

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

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

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

For each parameter  $\theta_0$  in  $\Phi$ :

$$\text{Array}_{parameter} = \text{parameter}_{\theta_0}$$

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

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For  $s$  in 1: 10000:

### Alpha

$$\alpha_{Candidate} \sim N\left(\alpha_{Candidate}, \frac{1}{2}\right)$$

$$M_{Candidate} = \text{GetSample}(n, T, \alpha_{Candidate}, \phi_{\beta}, \phi_N, \phi_I, \phi_T, \phi_W, \phi_S)$$

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

$$M_{Current} = \text{GetSample}(n, T, \phi_{\alpha}, \phi_{\beta}, \phi_N, \phi_I, \phi_T, \phi_W, \phi_S)$$

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

$$p_{prior_{Candidate}} = \text{Gamma}(2, 1).logPDF(\alpha_{Candidate})$$

$$p_{prior_{Current}} = \text{Gamma}(2, 1).logPDF(\phi_{\alpha})$$

$$p_{Likelihood_{Candidate}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Candidate_{it}}, \frac{1}{\sqrt{\phi_{\tau}}}\right).logPDF(Y_{True_{it}}) \right)$$

$$p_{Likelihood_{Current}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Current_{it}}, \frac{1}{\sqrt{\phi_{\tau}}}\right).logPDF(Y_{True_{it}}) \right)$$

$$p_{Q_{Candidate}} = N\left(\phi_1, \frac{1}{2}\right).logPDF(\alpha_{Candidate})$$

$$p_{Q_{Current}} = N\left(\alpha_{Candidate}, \frac{1}{2}\right).logPDF(\phi_{\alpha})$$

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

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

if  $ratio > \log(u)$ :

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$$\Phi_{\alpha} = \alpha_{Candidate}$$

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

$$Likelihood_{\alpha_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF \left( Y_{True_{it}} \right) \right)$$

else:

$$Likelihood_{\alpha_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Current_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF \left( Y_{True_{it}} \right) \right)$$

## Beta

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

$$M_{Candidate} = GetSample(n, T, \Phi_{\alpha}, \beta_{Candidate}, \Phi_N, \Phi_I, \Phi_T, \Phi_W, \Phi_S)$$

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

$$M_{Current} = GetSample(n, T, \Phi_{\alpha}, \Phi_{\beta}, \Phi_N, \Phi_I, \Phi_T, \Phi_W, \Phi_S)$$

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

$$p_{prior_{Candidate}} = Beta(1, 1) \cdot \log PDF(\beta_{Candidate})$$

$$p_{prior_{Current}} = Beta(1, 1) \cdot \log PDF(\Phi_2)$$

$$p_{Likelihood_{Candidate}} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF \left( Y_{True_{it}} \right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Current_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF \left( Y_{True_{it}} \right) \right)$$

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

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

$$\text{ratio}_\beta = \min \left( p_{\text{prior}_{Candidate}} + p_{\text{Likelihood}_{Candidate}} - p_{\text{prior}_{Current}} - p_{\text{Likelihood}_{Current}} + p_{Q_{Current}} - p_{Q_{Candidate}}, 1 \right)$$

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

if  $\text{ratio} > u$ :

$$\Phi_\beta = \beta_{Candidate}$$

$$\text{Accept}_\beta = \text{Accept}_\beta + 1$$

$$\text{Likelihood}_{\beta_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_\tau}} \right)^2 \cdot \log \text{PDF} \left( Y_{True_{it}} \right) \right)$$

else:

$$\text{Likelihood}_{\beta_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Current_{it}}, \frac{1}{\sqrt{\Phi_\tau}} \right)^2 \cdot \log \text{PDF} \left( Y_{True_{it}} \right) \right)$$

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for  $k$  in  $1:n$ :

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

$$I_{Candidate} = \Phi_3$$

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

$$M_{Candidate} = \text{GetSample}(n, T, \Phi_\alpha, \Phi_\beta, \Phi_N, I_{Candidate}, \Phi_T, \Phi_W, \Phi_S)$$

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

$$M_{Current} = \text{GetSample}(n, T, \Phi_\alpha, \Phi_\beta, \Phi_N, \Phi_I, \Phi_T, \Phi_W, \Phi_S)$$

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

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$$p_{prior\_Candidate} = DiscreteUniform(1, n).logPMF\left(I_{Candidate\_k}\right)$$

$$p_{prior\_Current} = DiscreteUniform(1, n).logPMF\left(\phi_{I\_k}\right)$$

$$p_{Likelihood\_Candidate} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Candidate\_it}, \frac{1}{\sqrt{\Phi_\tau}}\right).logPDF\left(Y_{True\_it}\right) \right)$$

$$p_{Likelihood\_Current} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Current\_it}, \frac{1}{\sqrt{\Phi_\tau}}\right).logPDF\left(Y_{True\_it}\right) \right)$$

$$p_{Q\_Candidate} = DiscreteUniform(1, n).logPMF\left(I_{Candidate\_k}\right)$$

$$p_{Q\_Current} = DiscreteUniform(1, n).logPMF\left(\phi_{I\_k}\right)$$

$$ratio_{I\_k} = \min\left(p_{prior\_Candidate} + p_{Likelihood\_Candidate} - p_{prior\_Current} - p_{Likelihood\_Current} + p_{Q\_Current} - p_{Q\_Candidate}, 1\right)$$

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

if  $ratio > \log(u)$ :

$$\phi_{I\_k} = I_{Candidate}$$

$$Accept_{I\_k} = Accept_{I\_k} + \frac{1}{N}$$

$$Likelihood_{I\_k} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Candidate\_it}, \frac{1}{\sqrt{\Phi_\tau}}\right).logPDF\left(Y_{True\_it}\right) \right)$$

else:

$$Likelihood_{I\_k} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Current\_it}, \frac{1}{\sqrt{\Phi_\tau}}\right).logPDF\left(Y_{True\_it}\right) \right)$$

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**T**

for  $k$  in  $1:n$ :

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

$$T_{Candidate} = \Phi_T$$

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

$$M_{Candidate} = GetSample(n, T, \Phi_{\alpha'}, \Phi_{\beta'}, \Phi_{N'}, \Phi_{I'}, T_{Candidate}, \Phi_{W'}, \Phi_S)$$

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

$$M_{Current} = GetSample(n, T, \Phi_{\alpha'}, \Phi_{\beta'}, \Phi_{N'}, \Phi_{I'}, \Phi_{T'}, \Phi_{W'}, \Phi_S)$$

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

$$p_{prior_{Candidate}} = DiscreteUniform(1, T).logPMF\left(T_{Candidate_k}\right)$$

$$p_{prior_{Current}} = DiscreteUniform(1, T).logPMF\left(\Phi_{T_k}\right)$$

$$p_{Likelihood_{Candidate}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right).logPDF\left(Y_{True_{it}}\right) \right)$$

$$p_{Likelihood_{Current}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Current_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right).logPDF\left(Y_{True_{it}}\right) \right)$$

$$p_{Q_{Candidate}} = DiscreteUniform(1, T).logPMF\left(T_{Candidate_k}\right)$$

$$p_{Q_{Current}} = DiscreteUniform(1, T).logPMF\left(\Phi_{4_k}\right)$$

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$$ratio_{T_k} = \min\left(p_{prior\_Candidate} + p_{Likelihood\_Candidate} - p_{prior\_Current} - p_{Likelihood\_Current} + p_{Q\_Current} - p_{Q\_Candidate}, 1\right)$$

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

if  $ratio > \log(u)$ :

$$\Phi_T = T_{Candidate}$$

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

$$Likelihood_{T_{s,k}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_\tau}}\right) \cdot \log PDF\left(Y_{True_{it}}\right) \right)$$

else:

$$Likelihood_{T_{s,k}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{Current_{it}}, \frac{1}{\sqrt{\Phi_\tau}}\right) \cdot \log PDF\left(Y_{True_{it}}\right) \right)$$

**W**

for  $k$  in  $1:n$ :

$$W_{k\_Candidate} \sim Binom\left(T, \frac{1 + \Phi_{W_k}}{2 + T}\right)$$

$$W_{Candidate} = \Phi_W$$

$$W_{Candidate_k} = W_{k\_Candidate}$$

$$M_{Candidate} = GetSample(n, T, \Phi_\alpha, \Phi_\beta, \Phi_{N'}, \Phi_{I'}, \Phi_{T'}, W_{Candidate}, \Phi_S)$$

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

$$M_{Current} = GetSample(n, T, \Phi_\alpha, \Phi_\beta, \Phi_{N'}, \Phi_{I'}, \Phi_{T'}, \Phi_{W'}, \Phi_S)$$

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



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$$p_{\text{prior}_{\text{Candidate}}} = \text{DiscreteUniform}(1, T). \log \text{PMF}(W_{\text{Candidate}_k})$$

$$p_{\text{prior}_{\text{Current}}} = \text{DiscreteUniform}(1, T). \log \text{PMF}(\phi_{W_k})$$

$$p_{\text{Likelihood}_{\text{Candidate}}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{\text{Candidate}_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right). \log \text{PDF}(Y_{\text{True}_{it}}) \right)$$

$$p_{\text{Likelihood}_{\text{Current}}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{\text{Current}_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right). \log \text{PDF}(Y_{\text{True}_{it}}) \right)$$

$$p_{Q_{\text{Candidate}}} = \text{Binom}\left(T, \frac{1 + \phi_{W_k}}{2 + T}\right). \log \text{PMF}(W_{\text{Candidate}_k})$$

$$p_{Q_{\text{Current}}} = \text{Binom}\left(T, \frac{1 + W_{\text{Candidate}_k}}{2 + T}\right). \log \text{PMF}(\phi_{W_k})$$

$$\text{ratio}_{W_k} = \min\left(p_{\text{prior}_{\text{Candidate}}} + p_{\text{Likelihood}_{\text{Candidate}}} - p_{\text{prior}_{\text{Current}}} - p_{\text{Likelihood}_{\text{Current}}} + p_{Q_{\text{Current}}} - p_{Q_{\text{Candidate}}}, 1\right)$$

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

if  $\text{ratio} > \log(u)$ :

$$\phi_W = W_{\text{Candidate}}$$

$$\text{Accept}_W = \text{Accept}_W + \frac{1}{N}$$

$$\text{Likelihood}_{W_{s,k}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{\text{Candidate}_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right). \log \text{PDF}(Y_{\text{True}_{it}}) \right)$$

else:

$$\text{Likelihood}_{W_{s,k}} = \sum_{t=1}^T \left( \sum_{i=1}^n N\left(C_{\text{Current}_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}}^2\right). \log \text{PDF}(Y_{\text{True}_{it}}) \right)$$

## S

for  $k$  in  $1:n$ :

$$S_{k\_Candidate} \sim N(\phi_{S_k}, 1^2)$$

$$S_{Candidate} = \phi_S$$

$$S_{Candidate_k} = S_{k\_Candidate}$$

$$M_{Candidate} = GetSample(n, T, \phi_{\alpha'} \phi_{\beta'} \phi_{N'} \phi_{I'} \phi_{T'} \phi_{W'} S_{Candidate})$$

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

$$M_{Current} = GetSample(n, T, \phi_{\alpha'} \phi_{\beta'} \phi_{N'} \phi_{I'} \phi_{T'} \phi_{W'} \phi_S)$$

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

$$p_{prior\_Candidate} = Gamma(1, 1).logPDF(S_{Candidate_k})$$

$$p_{prior\_Current} = Gamma(1, 1).logPDF(\phi_{S_k})$$

$$p_{Likelihood\_Candidate} = \sum_{t=1}^T \left( \sum_{i=1}^n N(C_{Candidate_{it}}, \frac{1}{\sqrt{\phi_{\tau}}}^2) \cdot logPDF(Y_{True_{it}}) \right)$$

$$p_{Likelihood\_Current} = \sum_{t=1}^T \left( \sum_{i=1}^n N(C_{Current_{it}}, \frac{1}{\sqrt{\phi_{\tau}}}^2) \cdot logPDF(Y_{True_{it}}) \right)$$

$$p_{Q\_Candidate} = N(\phi_{S_k}, 1^2) \cdot logPDF(S_{Candidate_k})$$

$$p_{Q\_Current} = N(S_{Candidate_k}, 1^2) \cdot logPDF(\phi_{S_k})$$

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$$ratio_{s_k} = \min\left(p_{prior\_Candidate} + p_{Likelihood\_Candidate} - p_{prior\_Current} - p_{Likelihood\_Current} + p_{Q\_Current} - p_{Q\_Candidate}, 1\right)$$

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

if  $ratio > \log(u)$ :

$$\Phi_s = S_{Candidate}$$

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

$$Likelihood_{s,k} = \sum_{y=1}^T \left( \sum_{x=1}^N N\left(C_{Candidate}, \frac{1}{\sqrt{\Phi_\tau}}\right) \cdot \log PDF(Y_{True}) \right)$$

else:

$$Likelihood_{s,k} = \sum_{y=1}^T \left( \sum_{x=1}^N N\left(C_{Current}, \frac{1}{\sqrt{\Phi_\tau}}\right) \cdot \log PDF(Y_{True}) \right)$$

## Tau

$$\tau_{Candidate} \sim N\left(\Phi_\tau, \frac{1}{2}\right)$$

$$M_{Candidate} = GetSample(n, T, \Phi_\alpha, \Phi_\beta, \Phi_{N'}, \Phi_{I'}, \Phi_{T'}, \Phi_{W'}, \Phi_S)$$

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

$$M_{Current} = GetSample(n, T, \Phi_\alpha, \Phi_\beta, \Phi_{N'}, \Phi_{I'}, \Phi_{T'}, \Phi_{W'}, \Phi_S)$$

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

$$p_{prior\_Candidate} = Gamma(1, 1) \cdot \log PDF(\tau_{Candidate})$$

$$p_{prior\_Current} = Gamma(1, 1) \cdot \log PDF(\Phi_\tau)$$

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$$p_{Likelihood_{Candidate}} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Candidate_{it}}, \frac{1}{\sqrt{\tau_{Candidate}}} \right) \cdot \log PDF(Y_{True_{it}}) \right)$$

$$p_{Likelihood_{Current}} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Current_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF(Y_{True_{it}}) \right)$$

$$p_{Q_{Candidate}} = N \left( \Phi_{\tau}, \frac{1}{2} \right) \cdot \log PDF(\tau_{Candidate})$$

$$p_{Q_{Current}} = N \left( \tau_{Candidate}, \frac{1}{2} \right) \cdot \log PDF(\Phi_{\tau})$$

$$ratio_{\tau} = \min \left( p_{prior_{Candidate}} + p_{Likelihood_{Candidate}} - p_{prior_{Current}} - p_{Likelihood_{Current}} + p_{Q_{Current}} - p_{Q_{Candidate}}, 1 \right)$$

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

if  $ratio_{\tau} > \log(u)$ :

$$\Phi_{\tau} = \tau_{Candidate}$$

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

$$Likelihood_{\tau_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Candidate_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF(Y_{True_{it}}) \right)$$

else:

$$Likelihood_{\tau_s} = \sum_{t=1}^T \left( \sum_{i=1}^n N \left( C_{Current_{it}}, \frac{1}{\sqrt{\Phi_{\tau}}} \right) \cdot \log PDF(Y_{True_{it}}) \right)$$

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## Proposal Distributions

$$\alpha^* \sim N\left(\phi_{\alpha}, \frac{1}{2}\right)$$

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

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

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

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

$$S_k^* \sim N\left(\phi_{S_k}, 1\right)$$

$$\tau^* \sim N\left(\phi_{\tau}, \frac{1}{2}\right)$$

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## Acceptance Probabilities

### Alpha

$$p_{\text{prior}_{\text{Candidate}}} = \log(\alpha_{\text{Candidate}}) - \alpha_{\text{Candidate}}$$

$$p_{\text{prior}_{\text{Current}}} = \log(\alpha_{\text{Current}}) - \alpha_{\text{Current}}$$

$$p_{\text{Likelihood}_{\text{Candidate}}} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Candidate}_{it}} \right)^2 \right)$$

$$p_{\text{Likelihood}_{\text{Current}}} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Current}_{it}} \right)^2 \right)$$

$$p_{Q_{\text{Candidate}}} = -2(\alpha_{\text{Candidate}} - \alpha_{\text{Current}})^2$$

$$p_{Q_{\text{Candidate}}} = -2(\alpha_{\text{Current}} - \alpha_{\text{Candidate}})^2$$

$$\text{ratio}_{\log} = \log\left(\frac{\alpha_{\text{Candidate}}}{\alpha_{\text{Current}}}\right) - \alpha_{\text{Candidate}} + \alpha_{\text{Current}} + \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Current}_{it}} \right)^2 \right) - \left( Y_{\text{True}_{it}} - C_{\text{Candidate}_{it}} \right)^2$$

$$P_{\text{Acceptance}}(\alpha) = e^{\text{ratio}_{\log}}$$

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## Beta

$$p_{prior\_Candidate} = 0$$

$$p_{prior\_Current} = 0$$

$$p_{Likelihood\_Candidate} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$p_{Likelihood\_Current} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 \right)$$

$$p_{Q\_Candidate} = 0$$

$$p_{Q\_Candidate} = 0$$

$$ratio_{log} = \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 - \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$P_{Acceptance}(\beta) = e^{ratio_{log}}$$

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$\mathbf{l}_k$

$$p_{prior\_Candidate} = -\log(n)$$

$$p_{prior\_Current} = -\log(n)$$

$$p_{Likelihood\_Candidate} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$p_{Likelihood\_Current} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 \right)$$

$$p_{Q\_Candidate} = -\log(n)$$

$$p_{Q\_Candidate} = -\log(n)$$

$$ratio_{log} = \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 - \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$P_{Acceptance}(I_k) = e^{ratio_{log}}$$



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$\mathbf{T}_k$

$$p_{prior\_Candidate} = -\log(T)$$

$$p_{prior\_Current} = -\log(T)$$

$$p_{Likelihood\_Candidate} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$p_{Likelihood\_Current} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 \right)$$

$$p_{Q\_Candidate} = -\log(T)$$

$$p_{Q\_Candidate} = -\log(T)$$

$$ratio_{log} = \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 - \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$P_{Acceptance}(T_k) = e^{ratio_{log}}$$

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$W_k$

$$p_{\text{prior}_{\text{Candidate}}} = \log\left(\left(W_{\text{Candidate}_k} + 5\right)C\left(W_{\text{Candidate}_k}\right)\right) + 6\log\left(\frac{3}{4}\right) - W_{\text{Candidate}_k} \log(4)$$

$$p_{\text{prior}_{\text{Current}}} = \log\left(\left(W_{\text{Current}_k} + 5\right)C\left(W_{\text{Current}_k}\right)\right) + 6\log\left(\frac{3}{4}\right) - W_{\text{Current}_k} \log(4)$$

$$p_{\text{Likelihood}_{\text{Candidate}}} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Candidate}_{it}} \right)^2 \right)$$

$$p_{\text{Likelihood}_{\text{Current}}} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Current}_{it}} \right)^2 \right)$$

$$p_{Q_{\text{Candidate}}} = \log\left((T)C\left(W_{\text{Candidate}_k}\right)\right) + W_{\text{Candidate}_k} \log\left(W_{\text{Current}_k} + 1\right) + \left(T - W_{\text{Candidate}_k}\right) \log\left(T + 1 - W_{\text{Current}_k}\right) + T \log(T + 2)$$

$$p_{Q_{\text{Current}}} = \log\left((T)C\left(W_{\text{Current}_k}\right)\right) + W_{\text{Current}_k} \log\left(W_{\text{Candidate}_k} + 1\right) + \left(T - W_{\text{Current}_k}\right) \log\left(T + 1 - W_{\text{Candidate}_k}\right) + T \log(T + 2)$$

$$\begin{aligned} \text{ratio}_{\log} &= \log\left(\left(W_{\text{Candidate}_k} + 5\right)C\left(W_{\text{Candidate}_k}\right)\right) + \left(W_{\text{Current}_k} - W_{\text{Candidate}_k}\right) \log(4) - \log\left(\left(W_{\text{Current}_k} + 5\right)C\left(W_{\text{Current}_k}\right)\right) \\ &\quad + \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{\text{True}_{it}} - C_{\text{Current}_{it}} \right)^2 - \left( Y_{\text{True}_{it}} - C_{\text{Candidate}_{it}} \right)^2 \right) + \log\left((T)C\left(W_{\text{Candidate}_k}\right)\right) + W_{\text{Candidate}_k} \log\left(W_{\text{Current}_k} + 1\right) \\ &\quad + \left(T - W_{\text{Candidate}_k}\right) \log\left(T + 1 - W_{\text{Current}_k}\right) - \log\left((T)C\left(W_{\text{Current}_k}\right)\right) - W_{\text{Current}_k} \log\left(W_{\text{Candidate}_k} + 1\right) - \left(T - W_{\text{Current}_k}\right) \log\left(T + 1 - W_{\text{Candidate}_k}\right) \end{aligned}$$

$$P_{\text{Acceptance}}\left(W_k\right) = e^{\text{ratio}_{\log}}$$

---

$S_k$

$$p_{prior\_Candidate} = -S_{Candidate\_k}$$

$$p_{prior\_Current} = -S_{Current\_k}$$

$$p_{Likelihood\_Candidate} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$p_{Likelihood\_Current} = \frac{-\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 \right)$$

$$p_{Q\_Candidate} = -\frac{1}{2} \left( S_{Candidate\_k} - S_{Current\_k} \right)^2$$

$$p_{Q\_Candidate} = -\frac{1}{2} \left( S_{Current\_k} - S_{Candidate\_k} \right)^2$$

$$ratio_{log} = -S_{Candidate\_k} + S_{Current\_k} + \frac{\tau}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 - \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$P_{Acceptance}(S_k) = e^{ratio_{log}}$$

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## Tau

$$p_{prior\_Candidate} = \log(\tau_{Candidate}) - \tau_{Candidate}$$

$$p_{prior\_Current} = \log(\tau_{Current}) - \tau_{Current}$$

$$p_{Likelihood\_Candidate} = \frac{-\tau_{Candidate}}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Candidate\_it} \right)^2 \right)$$

$$p_{Likelihood\_Current} = \frac{-\tau_{Current}}{2} \sum_{t=1}^T \left( \sum_{i=1}^n \left( Y_{True\_it} - C_{Current\_it} \right)^2 \right)$$

$$p_{Q\_Candidate} = -2(\tau_{Candidate} - \tau_{Current})^2$$

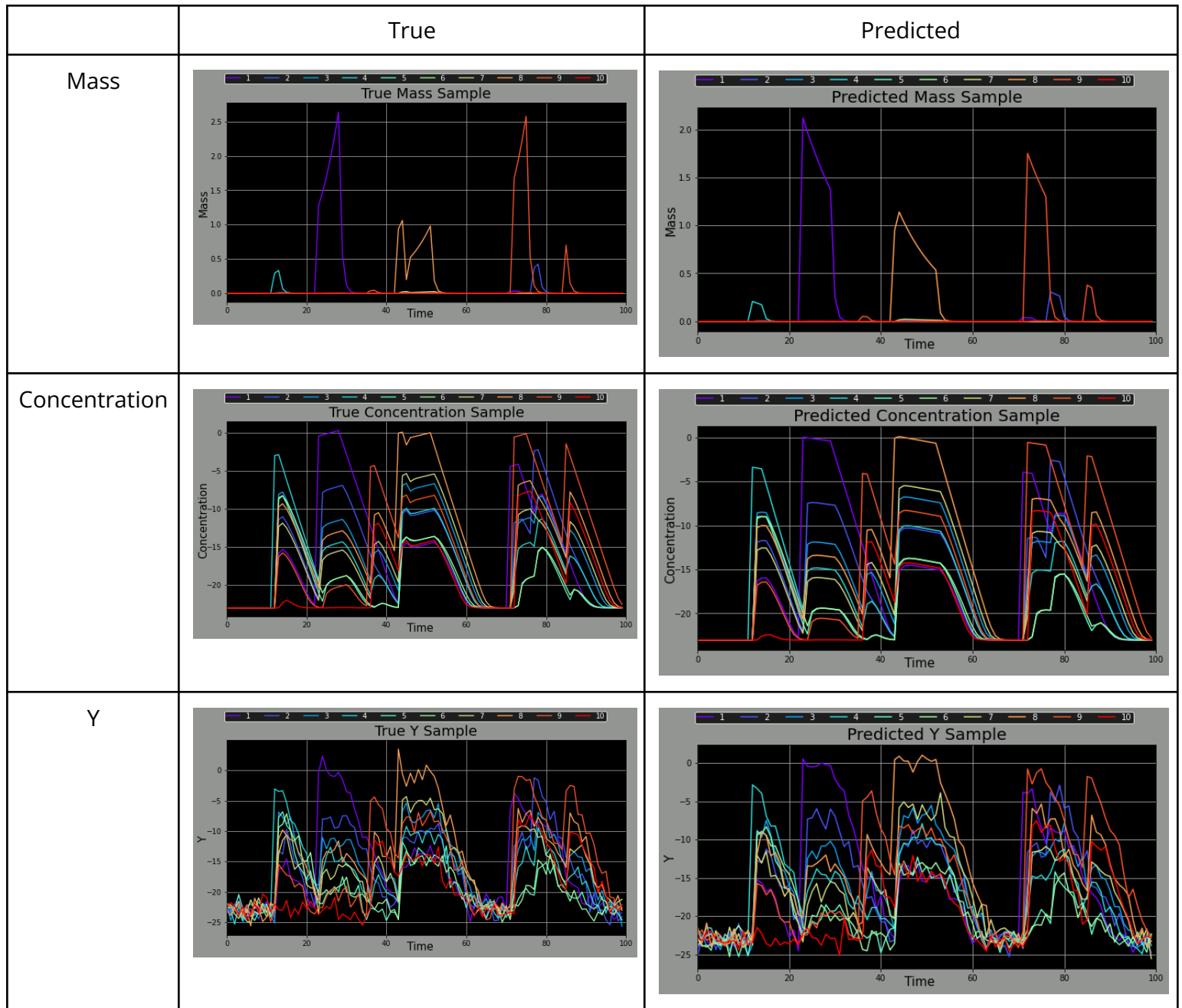
$$p_{Q\_Candidate} = -2(\tau_{Current} - \tau_{Candidate})^2$$

Since for  $\tau$ ,  $C_{Candidate\_it} = C_{Current\_it}$  for all  $i, t$

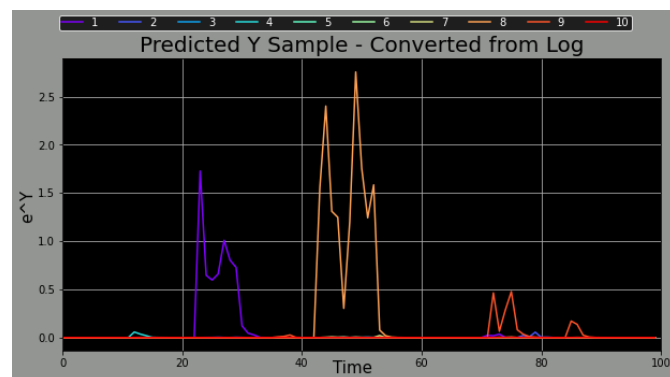
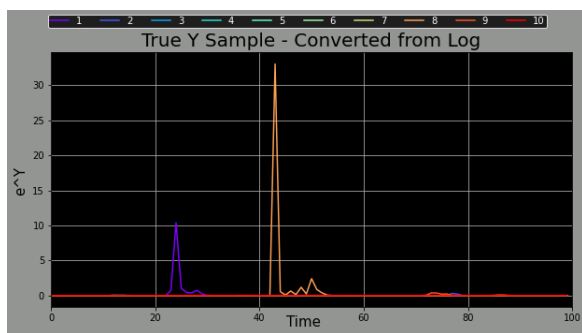
$$ratio_{log} = \log\left(\frac{\tau_{Candidate}}{\tau_{Current}}\right) - \tau_{Candidate} + \tau_{Current}$$

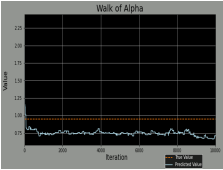
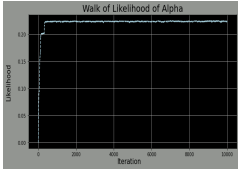
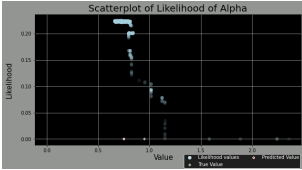
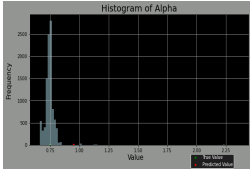
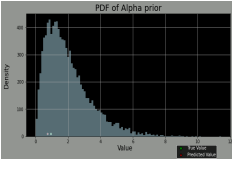
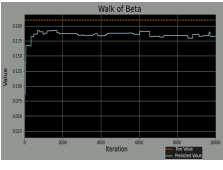
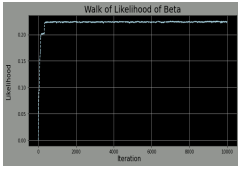
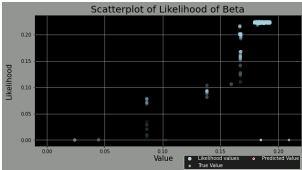
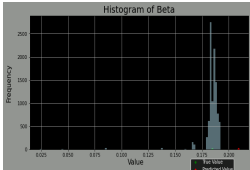
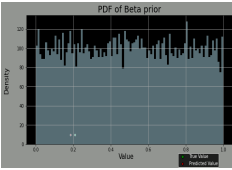
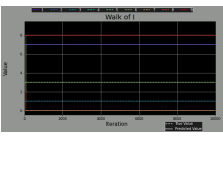
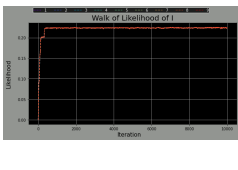
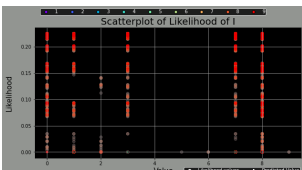
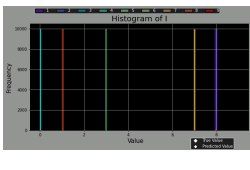


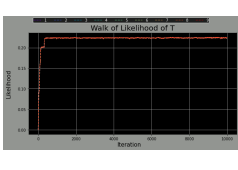
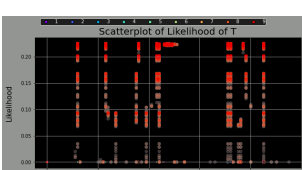
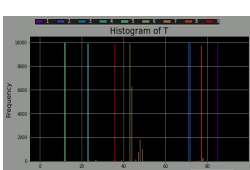
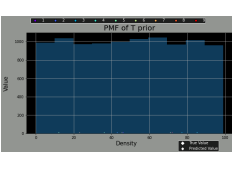
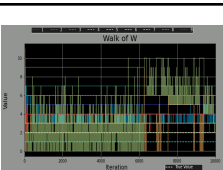

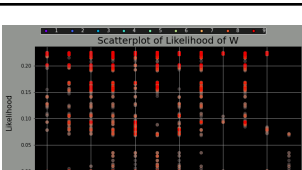
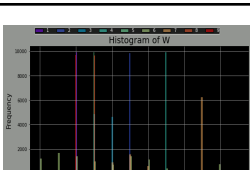
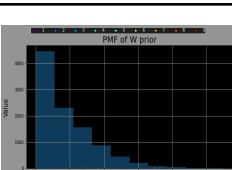
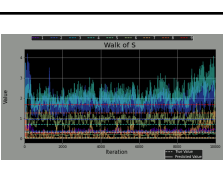
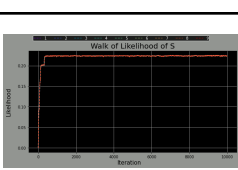

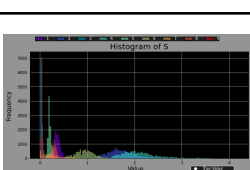
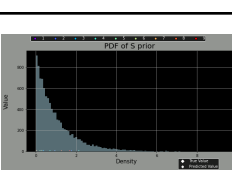
$$P_{Acceptance}(\tau) = e^{ratio_{log}}$$

# Results



$e^Y$



Parameter P(Acceptance)		Walk of value	Walk of Likelihood	Scatter Plot of Likelihood	Histogram	Prior PDF
$\alpha$	0.0257					
$\beta$	0.0056					
$I$	0.0992					
$T$	0.0104					
$W$	0.2327					
$S$	0.1059					
$\tau$	0.0903	