

# Parameter Estimation for Network Processes

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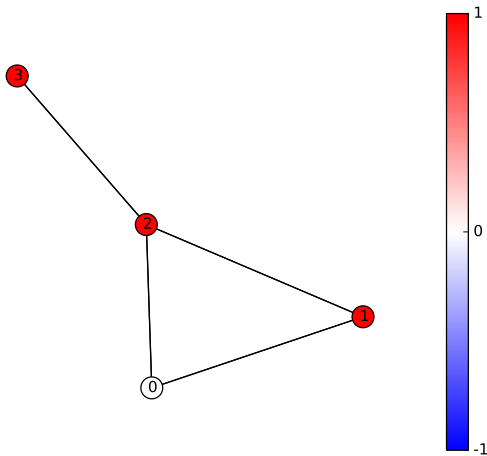
# Overview

- 1 Graph Attributes
- 2 Reaction Networks
- 3 Gibbs Sampling

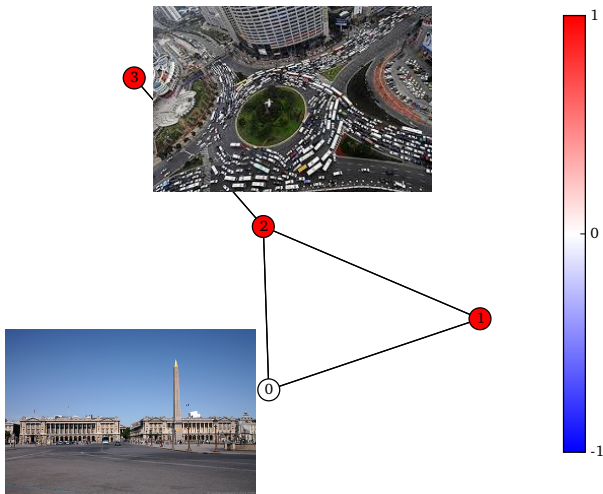
# Section 1

## Graph Attributes

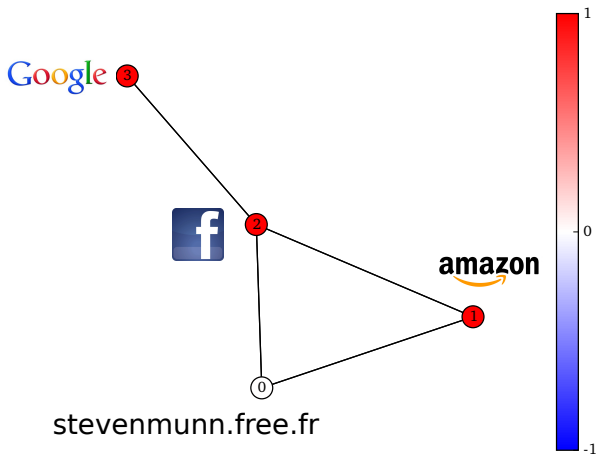
# Attributed Graphs



# Traffic Example



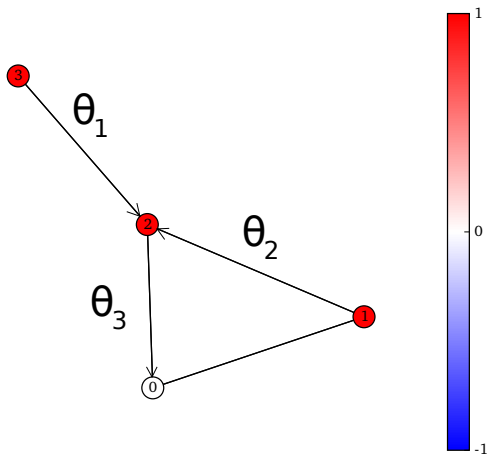
# Website Visits Example



## Section 2

# Reaction Networks

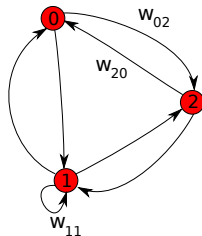
# Reaction Attributes





# Graph Weights

$$\mathbf{W} = \begin{bmatrix} w_{00} & w_{01} & w_{02} \\ w_{10} & w_{11} & w_{12} \\ w_{20} & w_{21} & w_{22} \end{bmatrix}$$



$$dM^T = \begin{bmatrix} 1 & 1 & 1 & \dots \end{bmatrix} \times \begin{bmatrix} w_{00} & w_{01} & w_{02} & \dots \\ w_{10} & w_{11} & w_{12} & \dots \\ w_{20} & w_{21} & w_{22} & \dots \\ \dots & & & \dots \end{bmatrix} \otimes \begin{bmatrix} 1 & M_0 & M_0 & \dots \\ M_1 & 1 & M_1 & \dots \\ M_2 & M_2 & 1 & \dots \\ \dots & & & \dots \end{bmatrix}$$

# Example: Homework One

$$\begin{aligned}dM^T &= \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & \theta_1 \\ 0 & 0 & \theta_1 \\ 0 & \theta_2 & 0 \end{bmatrix} \otimes \begin{bmatrix} 1 & M_0 & M_0 \\ M_1 & 1 & M_1 \\ M_2 & M_2 & 1 \end{bmatrix} \\&= \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & \theta_1 M_0 \\ 0 & 0 & \theta_1 M_1 \\ 0 & \theta_2 M_2 & 0 \end{bmatrix} \\&= \begin{bmatrix} 1 & \theta_2 M_2 & \theta_1 (M_0 + M_1) \end{bmatrix}\end{aligned}$$

## Section 3

# Gibbs Sampling

# Multiple Parameters Estimation

Start with parameters  $\theta^{(t)}$  (e.g. sampled from a uniform distribution)

Update each component one after the other as follows:

- Sample  $\theta_1^{(t+1)}$  from  $p(\theta_1|\text{data}, \theta_2, \theta_3, \dots)$
- Sample  $\theta_2^{(t+1)}$  from  $p(\theta_2|\text{data}, \theta_1, \theta_3, \dots)$
- ...

# Gibbs Sampling for Reaction Networks

## Step 1

Define an upper and lower bound (  $\mathbf{W}_{up}$  and  $\mathbf{W}_{low}$  ) on the graph weights.

## Step 2

Initialize the probability for each parameter as a uniform distribution between the upper and lower bounds,

$$p(w_{ij}) = u\left((\mathbf{W}_{up})_{ij}, (\mathbf{W}_{low})_{ij}\right)$$

# Gibbs Sampling for Reaction Networks

We now have an initial guess for the weights matrix  $\mathbf{W}$ . Next, we will update each component one at a time.

# Gibbs Update for first component

We need to compute,

$$p(w_{00} | \text{data}, w_{01}, w_{02}, w_{10}, w_{11}, \dots)$$

to sample a new value for  $w_{00}$ . Currently, we do this using the same Bayesian inference method from homework one.



# An example with two parameters

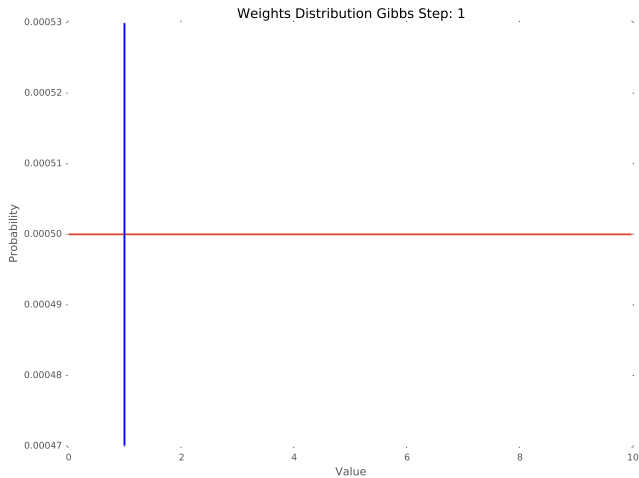
True parameters,

$$\mathbf{W} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & -1.45 & 0 \end{bmatrix}$$

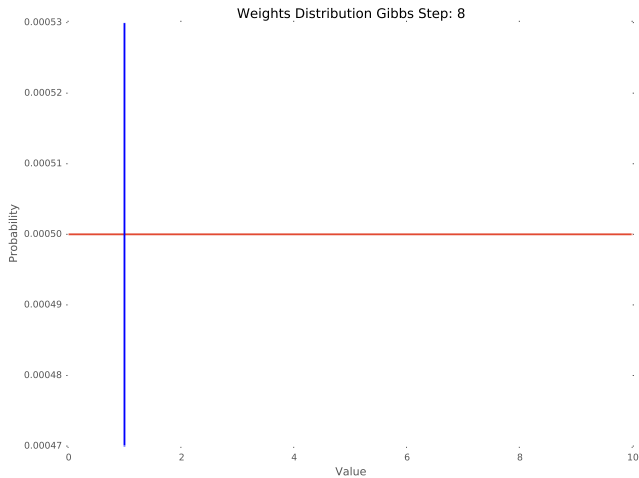
$$\mathbf{W}_{\text{lower}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & -1.45 & 0 \end{bmatrix}$$

$$\mathbf{W}_{\text{upper}} = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 0 & 2 \\ 0 & -1.45 & 0 \end{bmatrix}$$

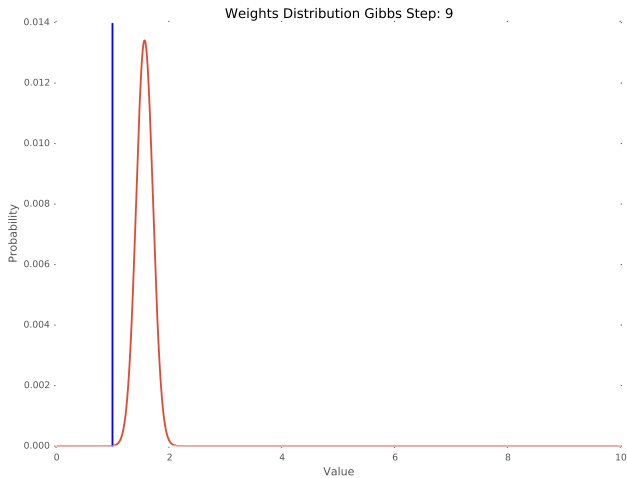
# $w_{02}$ Estimate at step 1



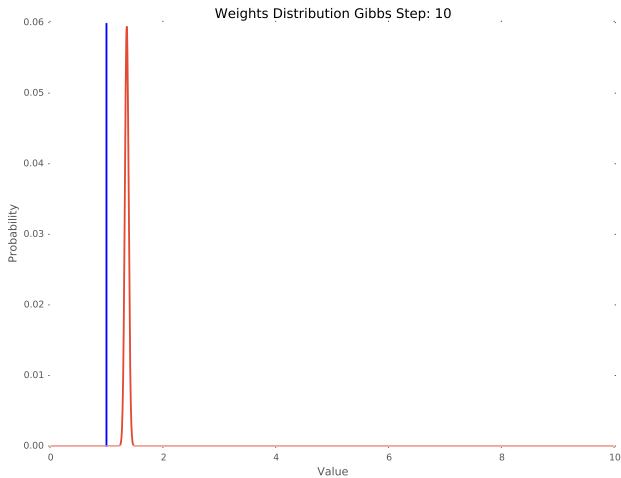
# $w_{02}$ Estimate at step 8



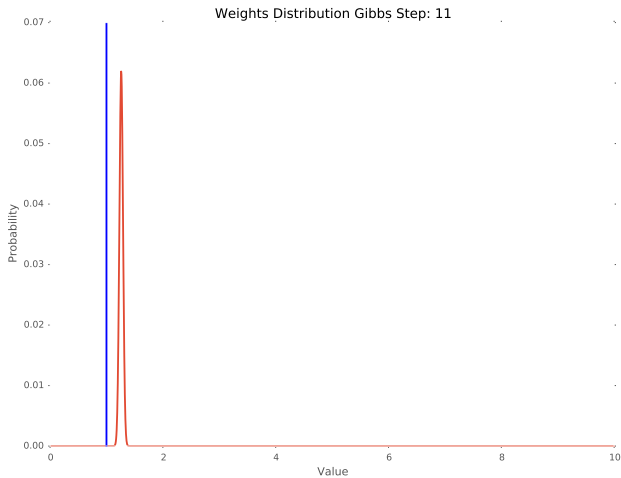
# $w_{02}$ Estimate at step 9



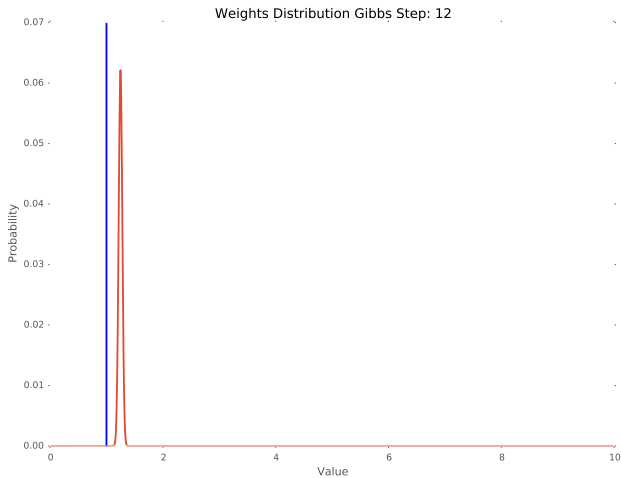
# $w_{02}$ Estimate at step 10



# $w_{02}$ Estimate at step 11



# $w_{02}$ Estimate at step 12



## Probability density in parameter space is too focused

Most of the parameter space has near-zero probability. The space over plausible parameters is very small. This is especially true for runaway reactions (where the rates are all positive).