Bayesian Probabilistic Modeling for Ecology

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You are a Bayesian if you quantify uncertainty with probability

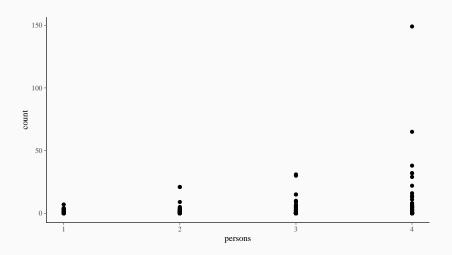
The Bayes Theorem

$$p(\theta \mid y) = \frac{p(y \mid \theta) \, p(\theta)}{p(y)}$$

Rethinking the Bayes Theorem

$$p(\theta \mid y) \propto p(y \mid \theta) p(\theta) = p(y, \theta)$$

Example: Catching Fish



Catching Fish: Modeling

Likelihood:

$$p(y_n \mid \theta) = \mathsf{Poisson}(\exp(\theta_1 + \theta_2 x_n))$$

Example weakly informative priors:

$$p(\theta_1) = \operatorname{normal}(0,3), \quad p(\theta_2) = \operatorname{normal}(0,1)$$

Flat ("uninformative") priors:

$$\theta_1 \propto 1, \quad \theta_2 \propto 1$$

Joint model:

$$p(y,\theta) = \left(\prod_n^N p(y_n \mid \theta)\right) \, p(\theta_1) \, p(\theta_2)$$

The Innocent Marginal Likelihood

$$p(y) = \int p(y \mid \theta) p(\theta) d\theta$$

Expectations to summarize distributions

(Almost) all we care about are expectations

Expectation of some function f over the posterior $p(\theta \mid y)$:

$$\mathbb{E}_{\theta \mid y}(f) = \int f(\theta) \, p(\theta \mid y) \; \mathrm{d}\theta$$

Things that are (behave like) expectations:

- Mean
- Variance / standard deviation
- Median
- Other quantiles

Monte-Carlo Estimator

Samples can be used to approximate expectations

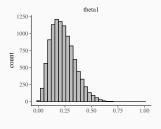
Having obtained S random samples $\{\theta^{(s)}\}$ from $p(\theta \mid y)$:

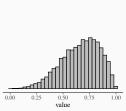
$$\frac{1}{S} \sum_{s=1}^S f(\theta^{(s)}) \sim \text{Normal}\left(\mathbb{E}_{\theta|y}(f), \sqrt{\frac{\text{Var}_{\theta|y}(f)}{\text{S}}}\right)$$

Propagation of Uncertainty

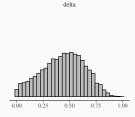
Uncertainty can be propagated easily using samples

Suppose, we are interesting in the posterior of $\delta=|\theta_1-\theta_2|$ Just evaluate $\delta^{(s)}=|\theta_1^{(s)}-\theta_2^{(s)}|$ per sample:





theta2



Markov-Chain Monte-Carlo (MCMC) Sampling

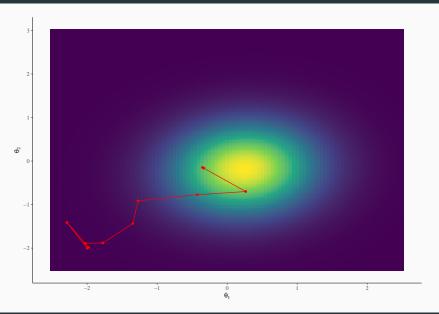
We can't simply draw samples from the posterior

A Markov Chain is a sequence of values where the value at position s is based only on the former value at position s-1:

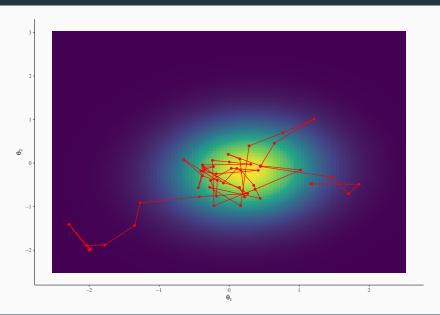
$$\theta^{(1)} \rightarrow \theta^{(2)} \rightarrow \theta^{(3)} \rightarrow \ldots \rightarrow \theta^{(S)}$$

If the transition~distribution is set up correctly, the values $\theta^{(1)},\dots,\theta^{(S)}$ will represent (dependent) samples from the posterior

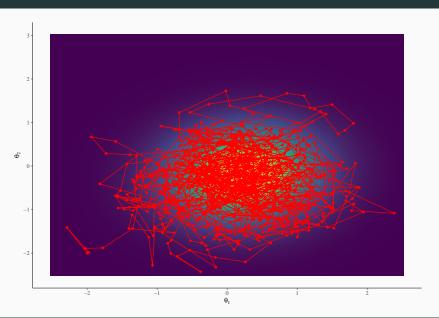
MCMC Sampling: A Single Chain (10 Iterations)



MCMC Sampling: A Single Chain (50 Iterations)



MCMC Sampling: A Single Chain (1000 Iterations)



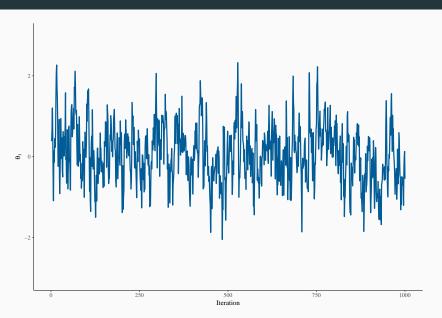
Markov-Chain Monto-Carlo Estimator

MCMC samples can be used to approximate expectations

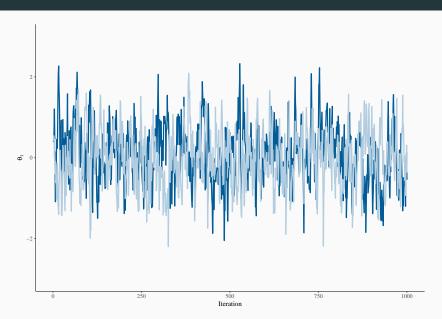
Assume well-behaved MCMC samples $\{\theta^{(s)}\}$ over $p(\theta \mid y)$:

$$\frac{1}{S} \sum_{s=1}^S f(\theta^{(s)}) \sim \text{Normal}\left(\mathbb{E}_{\theta|y}(f), \sqrt{\frac{\text{Var}_{\theta|y}(f)}{\text{ESS}}}\right)$$

Trace Plots: Visualizing a Single Chain



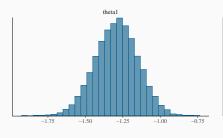
Trace Plots: Visualizing Multiple Chains

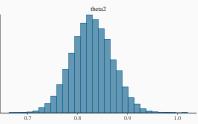


Probabilistic Programming Languages



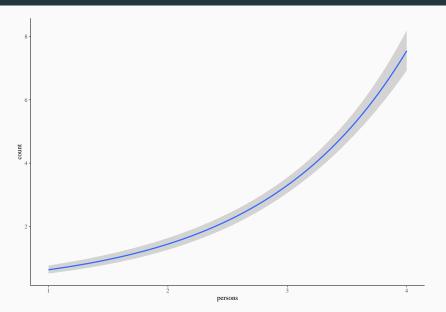
Catching Fish: Results



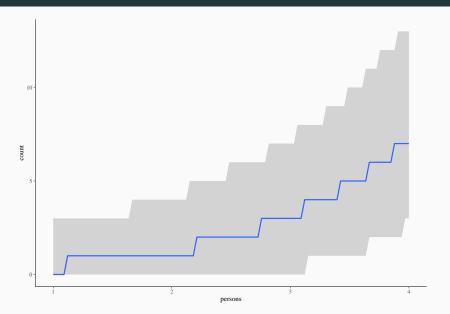


variable	mean	median	sd	q5	q95
theta1	-1.29	-1.29	0.14	-1.53	-1.06
theta2	0.83	0.83	0.04	0.76	0.90

Catching Fish: Expected posterior predictions



Catching Fish: Posterior prediction



What I like and don't like about Bayesian inference

What I like:

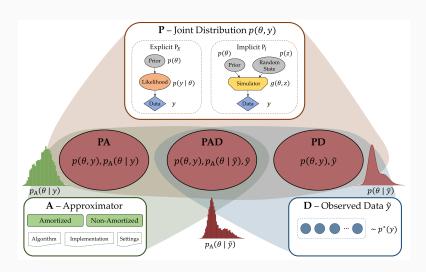
- Intuitive approach to expressing uncertainty
- Ability to incorporate prior information
- A lot of modeling flexibility
- Joint posterior distribution of parameters
- Easy propagation of uncertainty

What I don't like:

Slow Speed of model estimation

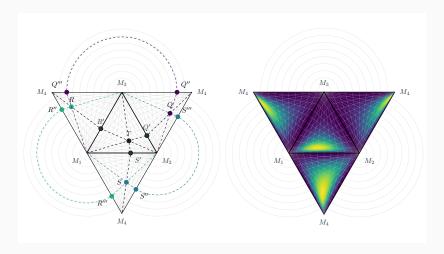
A brief look into my own research

What actually is a Bayesian Model?



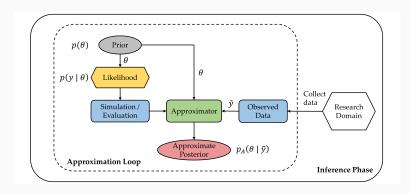
Uncertainty of Uncertainty (Meta-Uncertainty)

How can we combine Bayesian and frequentist uncertainties?



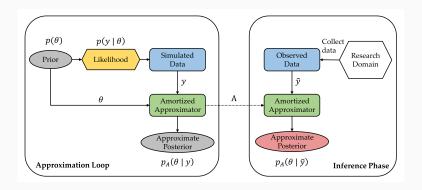
Non-Amortized (Standard) Inference

How can we improved the standard inference setting?



Amortized Inference

How far can we scale amortized inference?



More about me and my research



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