Stan: Probabilistic Modeling & Bayesian Inference

Development Team

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Stan 2.17 (November 2017)

http://mc-stan.org

The Funnel

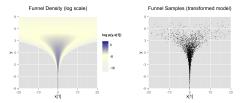
Funnel-Shaped Posteriors (1/2)

- · Arise in hierarchical model with no data (Neal 2003)
 - $\log \sigma \sim \text{Normal}(0, 1.5)$

[hierarchical scale]

[low-level coefficients]

- $\beta_n \sim \text{Normal}(0, \sigma)$ for $n \in 1:9$



 β_1 coefficient (x-axis) vs. $\log \sigma$ (y-axis); *left*) density plot (log scale); *right*) 4000 independent draws

Neal, R. M. 2003. Slice sampling. Annals of Statistics, 31(3):705-767

Funnel-Shaped Posteriors

(2/2)

- · Very challenging for sampling
- Need large step size to explore mouth of funnel
- · Need small step size to explore neck of funnel
- · Even small step sizes lead to divergences
 - numerical failure of Hamiltonian dynamics simulation to conserve the Hamiltonian
- Betancourt and Girolami (2015) analyzed for Hamiltonian Monte Carlo

Betancourt and Girolami. 2015. Hamiltonian Monte Carlo for hierarchical models.

In Current Trends in Bayesian Methodology with Applications. CRC

Non-Centered Parameterization

· The non-centered parameterization of the funnel is

 $\log \sigma \sim \text{Normal}(0, 1.5)$ $\beta_n^{\text{std}} \sim \text{Normal}(0, 1)$ $\beta_n = \sigma \times \beta_n^{\text{std}}$

- Removes dependency of β on σ in prior
- Called it "Matt trick" (after Matt Hoffman) before realizing it was wellknown

Adding Data

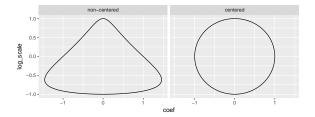
- Use the β in an intercept-only binomial logistic regression,

 $y_j \sim \text{Binomial}(N, \text{logit}^{-1}(\beta_j))$

- i.e., $y_j \in 0$: N is number of successes in K trials
- β_n is log odds of success for group j
- More data lessens dependency between eta and σ
- With informative enough data, centered parameterization is better
 - not size of data, but how much it constrains posterior

Non-Centered + Data = Funnel

- · With more data, centered approaches independent normal
- Non-centering ($\beta^{\text{std}} = \beta / \sigma$) produces a funnel
- · Centered parameterizations dominate with lots of data



Funnel + Data in the Wild

centered vs. non-centered parameterization

hierarchical logistic regression with 10 groups, intercept only, 20000 draws, divergences in orange

