

# Calibrated Approximate Bayesian Inference

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- Approximation schemes are key to Bayesian inference.
- Does a  $\alpha$  level approximate credible set have the right coverage?
- Let  $C_y$  and  $\tilde{C}_y$  be the exact and approximate  $\alpha$  level credible set,

$$\alpha = E_{\pi}(\mathbb{1}_{\phi \in C_y}) = \int_{\Omega} \mathbb{1}_{\phi \in C_y} \pi(\phi|y) d\phi.$$

$$\alpha = E_{\tilde{\pi}}(\mathbb{1}_{\theta \in \tilde{C}_y}) = \int_{\Omega} \mathbb{1}_{\theta \in \tilde{C}_y} \tilde{\pi}(\theta|y) d\theta.$$

- We also define

$$b(y) = \Pr(\phi \in \tilde{C}_Y | Y = y) = \int_{\Omega} \mathbb{1}_{\phi \in \tilde{C}_y} \pi(\phi | y) d\phi$$

be the *operational* coverage  $\tilde{C}_Y$  achieves.

- We want to estimate  $b(y_{obs})$ , the true Bayesian coverage of the approximate credible set, as it measures the reliability of approximation at the observed data.

- **Regression approach** Let  $\{\phi_i, y_i\}_{i=1}^M$  be samples from the generative model  $\pi(\phi)p(y|\phi)$ , let  $\tilde{C}_{y_i}$  be an approximate credible set for  $y_i$ , and  $c_i = \mathbb{1}_{\phi_i \in \tilde{C}_{y_i}}$ . Conditional on  $y_i$ ,

$$c_i \sim \text{Bernoulli}(b(y_i)), \quad b(y_i) = \Pr(\phi_i \in \tilde{C}_{Y_i} | Y_i = y_i)$$

- **Weighted-sample approach** Estimate  $b(y)$  by first approximately sampling from the exact posterior using Annealed Importance Sampling algorithm. This leverages our ability to draw samples from the approximate posterior  $\tilde{\pi}(\phi|y_{obs})$  as a good starting point for the AIS iteration.

# Example

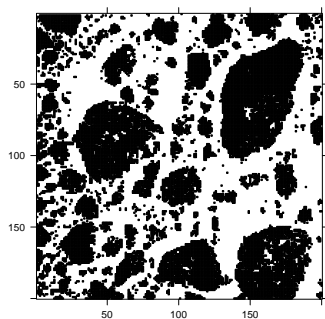


Figure: Icefloe image

We approximate the free boundary condition likelihood by a toroidal boundary condition likelihood.

# Example

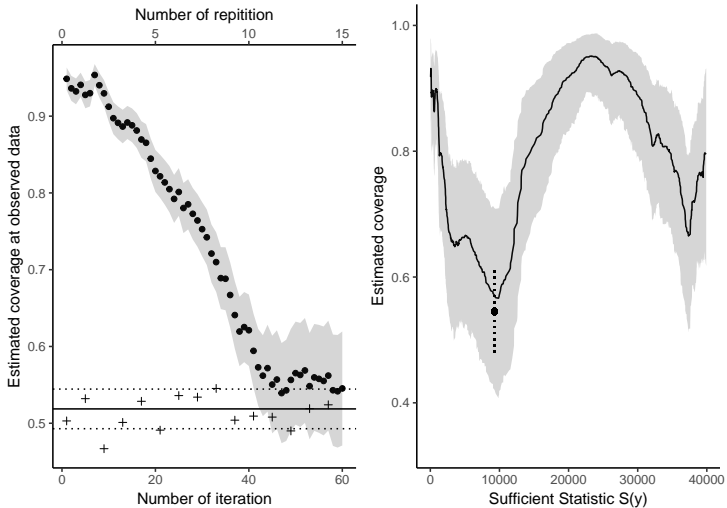


Figure: Left: AIS Right:Regression