

STT 465

# Monte Carlo Approximations

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# Motivation

⇒ In Bayesian analyses (and in many other applications) we need to compute integrals.

⇒ Examples

- Marginal distribution  $p(y_1) = \int p(y_1, y_2) dy_2 = \int p(y_1 | y_2) p(y_2) dy_2$
- Posterior mean (or other moments)  $E(\theta | y) = \int \theta p(\theta | y) d\theta$
- Posterior probabilities.  $p(\theta \in \Omega | y) = \int_{\theta \in \Omega} p(\theta | y) d\theta$

⇒ In most cases the integrals are difficult or even impossible to compute analytically.

⇒ **MC integration:** we can use sample from a distribution to approximate integrals.

# Idea

⇒ We can use computers to draw samples from the target distribution

$$\{\theta_s\}_{s=1}^S \stackrel{iid}{\sim} p(\theta | y)$$

⇒ The empirical distribution approximates the target distribution

⇒ The quality of the approximation will simply depend on the number of samples (S)

**Example 1:** approximating a normal density

⇒ In most cases we are interested on some integral, e.g.,  $E[g(\theta) | y] = \int g(\theta) \times p(\theta | y) d\theta$

⇒ Law of large numbers

$$\frac{1}{S} \sum_s g(\theta_s) \rightarrow E[g(\theta) | y]$$

# Examples

⇒ Example 2: Computing areas in a unit square

⇒ Example 3: Convergence (Figure 4.2)

- Running Mean
- Evaluating MC-Error by # of samples

⇒ Example 4: Evaluating functions of parameters (e.g., log-odds)