

Stat 499: Expectations and Sampling methods

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In this class we mainly focus on the sampling and its methods as following:

Sampling: the selection of a subset (a statistical sample) of individuals from within a statistical population to estimate characteristics of the whole population.

Sampling method:

1. Importance Sampling

- Approximate $E[f]$ by drawing samples from a "proposal distribution" q , and correcting appropriately by a weighting ratio.
- Suppose dealing with $p(z)$ is harder, i.e., we can't even evaluate $p(z)$ but can only do so up to proportionality constant, and only $\tilde{p}(z)$ can be evaluated. We can still apply importance sampling by applying the importance weight.

2. Rejection Sampling

- Need to set up a proposal function $q(z)$ and M , so that $Mq(z) \geq \tilde{p}(z)$, for all z .
- Simulate $U \sim \text{Unif}(0,1)$ and candidate $X \sim q$ from the candidate density.
- Use $U < \tilde{p}(z)/Mq(z)$ to test if reject candidate X or not.

Bayesian inference:

- A method of statistical inference in which Bayes' theorem is used to update the probability for a hypothesis as more evidence or information becomes available.
- $P(\theta|D) = (P(D|\theta) \times P(\theta)) / P(D)$
- Here, $P(\theta)$ is the prior, $P(D|\theta)$ is the likelihood of observing our result given our distribution for θ . $P(D)$ is the evidence. $P(\theta|D)$ is the posterior belief of our parameters after observing the evidence i.e the number of heads .
- Use $P(\theta|D)$ to estimate the probability of θ given the data.

Maximum likelihood estimation:

- A method of estimating the parameters of a probability distribution by maximizing the likelihood function, so that under the assumed statistical model the observed data is most probable.
- In practice, it is often convenient to work with log likelihood.