Lecture 9: Monte Carlo estimates of various statistics

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Estimating the cumulative distribution function



Estimating the variance

- Take a random variable $X \sim p(x)$ and some function g(x).
- Consider the derived random variable Y = g(X).
- We would like to estimate the cumulative distribution function of Y:

$$F(y) = p(Y \le y) = p(g(X) \le y)$$
want to find this using only samples of Y



Estimating the variance

 We would like to estimate the cumulative distribution function of Y:

$$F(y) = p(Y \le y) = p(g(X) \le y)$$

Consider the indicator function of a set A:

$$1_{\underline{A}}(y) = \begin{cases} 1, & \text{y in } A \\ 0, & \text{otherwise} \end{cases}$$

• Using it, we can rewrite F(y) as an expectation:

$$F(y) = p(Y \le y) = p(g(X) \le y) = \text{Figure Laboratory}$$
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Estimating the CDF

$$F(y) = \mathbb{E}\left[1_{(-\infty,y)}(y(x))\right]$$

- Take X_1, X_2, \ldots independent identical copies of X.
- Estimate the CDF using a sample average:

$$\bar{F}_N(y) = \frac{1}{N} \sum_{i=1}^{N} 1_{[-\infty, y]}(g(X_i)) = \frac{\text{number of } g(X_i) \leq y}{N} \leq N$$

This estimate is called the empirical CDF.



Example: 1D CDF



