

Lecture 8: The Monte Carlo method for estimating expectations

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The uncertainty propagation problem

The uncertainty propagation problem

- characterize the statistics of Y , the model output

- You are given a function $g(x)$ representing a physical model.

x : input, $g(x)$: output

- The inputs of the model are uncertain.

- You represent this uncertainty with a random variable:

$$X \sim p(x)$$

X has $p(x)$ as its pdf

- You would like to quantify your uncertainty about the model output:

$$Y = g(X)$$

may be epistemic

The uncertainty propagation problem

- We would like to estimate the expected value of the output:

$$\mathbb{E}[Y] = \mathbb{E}[g(X)] = \int g(x) p(x) dx$$

Now, how to compute this integral?

The uncertainty propagation problem

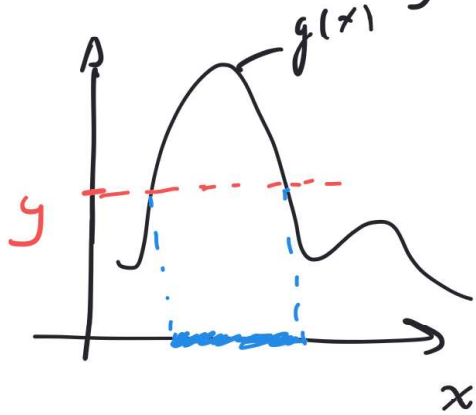
- We would like to estimate the variance of the output:

$$\begin{aligned}\mathbb{V}[Y] &= \int (g(x) - \mathbb{E}[Y])^2 p(x) dx \\ &= \mathbb{E}[\underbrace{[g(x)]^2}_{\text{already have}}] - \left(\mathbb{E}[g(x)]\right)^2 \\ \mathbb{E}[\underbrace{[g(x)]^2}_{\text{already have}}] &= \int g^2(x) p(x) dx\end{aligned}$$

The uncertainty propagation problem

- Or maybe the probability that the output exceeds a threshold:

$$p(Y \geq y) = \int \underbrace{1_{[y, \infty)}(g(x))}_{\epsilon?} p(x) dx = \mathbb{E}[1_{[y, \infty)}(g(X))] = \mathbb{P}[1_{[y, \infty)}(Y)]$$



$$1_{[y, \infty)}(g(x)) = \begin{cases} 1, & \text{if } g(x) \geq y \\ 0, & \text{otherwise} \end{cases}$$

↘ The indicator function of the set $[y, \infty)$

The uncertainty propagation problem

- Notice that all these statistics are essentially expectations of functions of X .
- We must learn how to do such integrals!