

Monte Carlo Methods

MATH 565

Fred Hickernell, Fall 2025

Updated 2025 August 20




Introduction

Assignment 1 Due Sep 5


MATH 565 Monte Carlo Methods, Fred Hickernell, Fall 2025

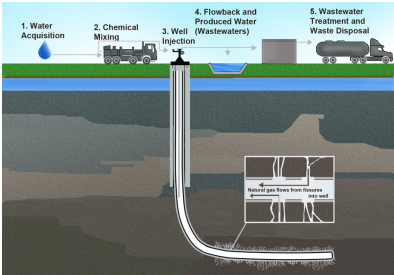




Monte Carlo Helps With Uncertainty

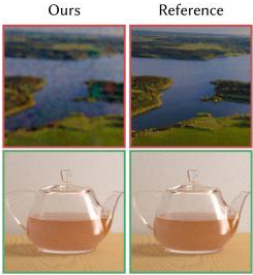
[WSJ 2016]





Ours

Reference



Viral infection

COVID-19
Influenza (H7N9, H5N1, ...)
Haemorrhagic fever
Ebola
Dengue
Rabies

Bacterial infection

Salmonellosis
Tuberculosis
Brucellosis
Anthrax

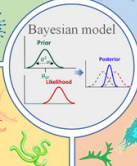
Parasitic infection

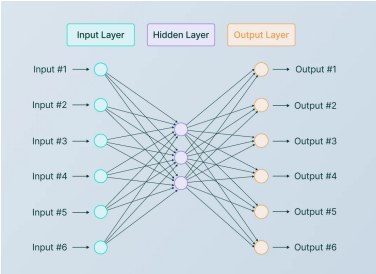
Malaria
Schistosoma
Toxoplasmosis


Other infection

Fungi
Rickettsia
Leptospira


Bayesian model







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Why is there uncertainty?

- Finance — [market forces](#), often modeled by stochastic processes driven by Brownian motion
- Engineering — [variability of system parameters](#), sometimes modeled by Gaussian processes
- Image rendering — [can only trace some rays](#), must rely on a finite sample
- Bayesian inference — the [posterior probability distribution](#) is a combination of a prior and what is learned from data
- Neural networks — many parameters need to be tuned, but one cannot search in [all possible directions](#)
- Queues — [arrival times and service times](#) of customers



How is this expressed quantitatively?

Y = random variable denoting **quantity of interest** = $\left\{ \begin{array}{l} \text{option payoff} \\ \text{fluid pressure} \\ \text{pixel intensity} \\ \text{statistical model parameter} \\ \text{neural network parameter} \\ \text{service time} \end{array} \right.$

= $f(\mathbf{X})$, where

\mathbf{X} = multivariate random variable with a **simpler distribution**

We want to estimate the **mean**, **variance**, **quantile**, or **probability distribution** of Y
using **sample versions**



Are we there yet?

You are visiting your friend and it will require

- A 5 minute walk to the 'L' station
- Waiting for the train, which arrives every 20 minutes
- Traveling 35 minutes by 'L'
- Catching a taxi at the 'L' destination
 - There is a 20% chance that the car is waiting for you
 - Otherwise the average wait time is 10 minutes
- A 12 minute taxi ride

How long should you plan for the trip to take?

Let's look at this Jupyter Notebook [AreWeThereYet](#) on the [class website](#)



Class website and repository

Website

Git repository



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Why should you attend synchronously?

- You will be better keep [pace](#)?
- You will get [real time](#) answers to questions?
- You can [influence](#) the pace and direction of the course?
- You can help your peers learn and benefit from them—partake in a [learning community](#)
- To help me know you, in case you want a [reference](#), or to add me to your [LinkedIn](#) network



Review of probability

Let Y be a **random variable** with a sample space $\mathcal{Y} \subseteq \mathbb{R}$

$\mathbb{P}(Y \in \Omega)$ means the probability of the **event** Ω

F is the **cumulative probability distribution function** $F(y) := \mathbb{P}(Y \leq y)$, $y \in \mathcal{Y}$

Q is the **quantile function** $Q(p) := \inf\{y \in \mathcal{Y} : F(y) \geq p\}$, $0 < p < 1$

$\mathbb{E}[g(Y)] := \int_{\mathcal{Y}} g(y) \, dF(y)$ is the **expectation** of $g(Y)$

discrete

continuous

ρ is the **probability mass function** ρ is the **probability density function**

$$\rho(y) := \mathbb{P}(Y = y)$$

$$\rho(y) := F'(y)$$

$$\mathbb{E}[g(Y)] = \sum_{y \in \mathcal{Y}} g(y)$$

$$\mathbb{E}[g(Y)] = \int_{y \in \mathcal{Y}} g(y) \, dy$$



Moments

$\mu := \mathbb{E}(Y)$ is the **mean** of Y

$\sigma^2 = \text{var}(Y) := \mathbb{E}[(Y - \mu)^2]$ is the **variance** of Y

σ is the **standard deviation**

$$\text{cov}(\mathbf{Y}) := \left(\mathbb{E}[(Y_j - \mu_j)(Y_k - \mu_k)] \right)_{j,k=1}^d$$

is the **covariance matrix** of the random **vector** \mathbf{Y}



Binomial Distribution

This is a coin-flipping distribution with the probability p of heads

$$Y \sim \text{Binomial}(n, p) \quad \text{scipy.stats.binom}(n, p)$$

$$\varrho(y) := \mathbb{P}(Y = y) = \binom{n}{y} p^y (1 - p)^{n-y}, \quad y = 0, 1, \dots, n$$

$$F(y) = \mathbb{P}(Y \leq y) = \sum_{j=0}^{\lfloor y \rfloor} \binom{n}{j} p^j (1 - p)^{n-j}$$

$$Q(p) = \inf\{y \in \{0, 1, \dots, n\} : F(y) \geq p\}$$

$$\mu = \mathbb{E}(Y) = np \quad \sigma^2 = \text{var}(Y) = np(1 - p)$$



Multivariate Normal/Gaussian Distribution

Used in finance and uncertainty quantification

$Y \sim \mathcal{N}(\mu, \Sigma)$ is a d -dimensional random vector

`scipy.stats.multivariate_normal(mean, cov)`

$\mu = \text{mean}$

$\Sigma = \text{covariance matrix}$

$$\varrho(\mathbf{y}) := \frac{1}{(2\pi)^{d/2} |\Sigma|^{1/2}} \exp\left(-\frac{1}{2}(\mathbf{y} - \boldsymbol{\mu})^\top \Sigma^{-1}(\mathbf{y} - \boldsymbol{\mu})\right)$$

$$F(\mathbf{y}) = \int_{(-\infty, \mathbf{y})} \varrho(\mathbf{z}) \, d\mathbf{z}$$



Sampling

Let Y be a random variable, often $Y = f(\mathbf{X})$

Y_1, Y_2, \dots be a **sample**

not necessarily random or independent and identically distributed (IID)

$\hat{\mu}_n := \frac{1}{n} \sum_{i=1}^n Y_i$ is the **sample mean**

$\hat{\sigma}_n^2 := \frac{1}{n-1} \sum_{i=1}^n (Y_i - \hat{\mu}_n)^2$ is the **sample variance**



Under **random** sampling

Let Y be a random variable, often $Y = f(\mathbf{X})$

$$Y_1, Y_2, \dots \sim Y$$

$$\mathbb{E}(\hat{\mu}_n) = \frac{1}{n} \sum_{i=1}^n \mathbb{E}(Y_i) = \mu \text{ so } \hat{\mu}_n \text{ is unbiased}$$

If in addition, Y_1, Y_2, \dots are **uncorrelated**, then

$$\mathbb{E}(\hat{\sigma}_n^2) = \sigma^2 \text{ so } \hat{\sigma}_n^2 \text{ is unbiased}$$

$$\text{mse}(\hat{\mu}_n) := \mathbb{E}[(\mu - \hat{\mu}_n)^2] = \text{var}(\hat{\mu}_n) = \frac{\sigma^2}{n}, \quad \text{rmse}(\hat{\mu}_n) = \frac{\sigma}{\sqrt{n}}$$



Central Limit Theorem

Let Y be a random variable, often $Y = f(\mathbf{X})$

$$Y_1, Y_2, \dots \stackrel{\text{iid}}{\sim} Y$$

The distribution of

$$\frac{\hat{\mu}_n - \mu}{\sigma/\sqrt{n}} \text{ approaches } \mathcal{N}(0, 1) \text{ as } n \rightarrow \infty$$

which means that

$$\hat{\mu}_n \sim \mathcal{N}(\mu, \sigma^2/n) \text{ for large } n$$



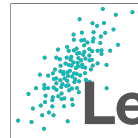
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