## Chapter 2 The Monte Carlo Method

The Monte Carlo Method stands for a broad class of computational algorithms that rely on random samplings. It is often used in physical and mathematical problems and is most useful when it is difficult or impossible to use other algorithms. There are mainly three distinct classes of applications: numerical integration, optimization, and generating draws from a probability distribution. https://cse.sc.edu/~terejanu/files/tutorialMC.pdf )

- Section 2.1 Computing Integrals
- Simple Monte Carlo for finding min and max of $y=f(x)$
* randomly generate points $x_{1}, x_{2}, \ldots$
* Min $=$ Max $=f\left(x_{1}\right)$,
* if $f\left(x_{2}\right)<f\left(x_{1}\right)$, then replace $\operatorname{Min}=f\left(x_{2}\right), \operatorname{Max}=f\left(x_{1}\right)$, continues
- Consider the area below the curve $y=f(x)$ inside the rectangle of area $R$. We randomly generate a point $\left(x_{1}, y_{1}\right)$ inside the rectangle. If $y_{1} \leq f\left(x_{1}\right)$, then this point is below the curve. If in $N$ trials, $M$ points are below the curve, then:

$$
\text { area of under the curve } \approx \frac{M}{N} R
$$



- In theory, as $N \rightarrow \infty$, the limit is exactly the area.
- Monte Carlo Method for integrals:
* Recall that $I=\int_{a}^{b} f(x) d x$ is the signed area between the curve $y=f(x)$ and $x$-axis. If $f(x) \geq 0$ it is the area.
* We may use the above method to approximate the integral as follows:
- Set $c=\min \{f(x): a \leq x \leq b\}, d=\max \{f(x): a \leq x \leq b\}$
- update: $c=\min \{c, 0\}, d=\max \{d, 0\}$
- $N$ is total trials, initialize $N=M=I=0$
- Each trial, randomly generate $a \leq x_{1} \leq b, c \leq y_{1} \leq d$
- If $0 \leq y_{1} \leq f\left(x_{1}\right)$, then $M=M+1$
- If $0 \geq y_{1} \geq f\left(x_{1}\right)$, then $M=M-1$
- The above two step can be combined as: If $\left(f\left(x_{1}\right)-y_{1}\right) y_{1} \geq 0$, then $M=M+\operatorname{sign}\left(y_{1}\right)$
- $I=(M / N)(b-a)(d-c)$
* For double integrals, we use volumes instead.
- Read Example in page 22.
- Exercise 1: Use Monte Carlo to find the following integral within an error of $10^{-5}$. What is $N$ ?

$$
\int_{0}^{2} e^{x^{2}} \cos x d x
$$

- Experiment: try using normal distribution insteat of uniform distribution, compare results.
- Monte Carlo is not very efficient. It is used only for very complicated situations.
- Section 2.2 Mean Time between Failures
- Consider a system has several components. Each has a failure time that is normally distributed with mean $\mu_{i}$ and variance $\sigma_{i}^{2}$. Then the system failure time is

$$
T=\min \left\{T_{1}, T_{2}, \ldots, T_{n}\right\}
$$

- Analytic solution for $E[T]$ : Let $F_{i}$ be the CDF of $T_{i}$. Note that $T(\omega) \leq T_{i}(\omega)$, and

$$
\begin{gathered}
T(\omega)>t \Longleftrightarrow T_{i}(\omega)>t \text { for } i=1,2, \ldots, n \\
\{\omega: T(\omega)>t\}=\cap_{i=1}^{n}\left\{\omega: T_{i}(\omega)>t\right\}
\end{gathered}
$$

Then CDF $F$ of $T$ is (assuming indenpendence)

$$
\begin{aligned}
F(t) & =P(T \leq t)=1-P(T>t) \\
& =1-P\left(\cap_{i=1}^{n}\left\{\omega: T_{i}(\omega)>t\right\}\right) \\
& =1-\prod_{i=1}^{n} P\left(\omega: T_{i}(\omega)>t\right) \\
& =1-\prod_{i=1}^{n}\left(1-F_{i}(t)\right)
\end{aligned}
$$

and

$$
\begin{gathered}
d F(t)=-d\left(\prod_{i=1}^{n}\left(1-F_{i}(t)\right)\right)=\sum_{k=1}^{n}\left(\prod_{i=\neq k}^{n}\left(1-F_{i}(t)\right)\right) d F_{i} \\
E[T]=\sum_{k=1}^{n} \int_{-\infty}^{\infty} t\left(\prod_{i=\neq k}^{n}\left(1-F_{i}(t)\right)\right) d F_{i}(t)
\end{gathered}
$$

- This integral is so difficult.
- Monte Carlo for mean and variance:

$$
\begin{aligned}
\mu & =E[X]=\lim _{N \rightarrow \infty} \frac{X\left(\omega_{1}\right)+X\left(\omega_{2}\right)+\ldots+X\left(\omega_{N}\right)}{N}=\lim _{N \rightarrow \infty}\left(\frac{X\left(\omega_{1}\right)}{N}+\frac{X\left(\omega_{2}\right)}{N}+\ldots+\frac{X\left(\omega_{N}\right)}{N}\right) \\
\sigma^{2} & =E\left[(X-\mu)^{2}\right]=E\left[X^{2}-2 \mu X-\mu^{2}\right]=E\left[X^{2}\right]-2 \mu E[X]+\mu^{2}=E\left[X^{2}\right]-\mu^{2}
\end{aligned}
$$

- Exercise \#2: MC for $E[T]$ (See Routine in page 23)
- Section 2.3 Servicing Requests
- Monte Carlo for random variables, means/variances
- Lemma Suppose that CDF $F_{X}(x)$ of a random variable $X$ is continuous and strictly increasing. Then $F_{X}(X)=Y^{\sim} U(0,1)$, i.e., Y is uniform distribution in $[0,1]$.
* Proof: Using graph of $y=F_{X}(x)$. For any $0<y_{0}<1$, there is a unique $x_{0}$ such that $y_{0}=$ $F_{X}\left(x_{0}\right)$.Since $F_{X}$ is strictly increasing,

$$
\left\{x: F_{X}(x) \leq y_{0}\right\}=\left(-\infty, x_{0}\right]
$$

By definition, $y_{0}=P\left(X \leq x_{0}\right)$.From the above, we see that if

$$
\omega \in\left\{\omega: F_{X}(X(\omega)) \leq y_{0}\right\}
$$

then

$$
\begin{aligned}
& X(\omega) \in\left\{x: F_{X}(x) \leq y_{0}\right\}=\left(-\infty, x_{0}\right] \\
& X(\omega) \leq x_{0}
\end{aligned}
$$

So

$$
\omega \in\left\{\omega: X(\omega) \leq x_{0}\right\} .
$$

Consequently

$$
\left\{\omega: F_{X}(X(\omega)) \leq y_{0}\right\} \subset\left\{\omega: X(\omega) \leq x_{0}\right\} .
$$

Since $F_{X}$ is increasing, we actually have

$$
\left\{\omega: F_{X}(X(\omega)) \leq y_{0}\right\}=\left\{\omega: X(\omega) \leq x_{0}\right\} .
$$

So

$$
F_{Y}\left(y_{0}\right)=P\left(F_{X}(X) \leq y_{0}\right)=P\left(X \leq x_{0}\right)=y_{0}
$$

- Now, for any given $F(x)$, we first randomly generate a sample $y^{\sim} U(0,1)$, then calculate $x=$ $F^{-1}(y)$ as a sample for $F$.
- Estimate $E[g(X)]$ for given CDF $F_{X}(x)$ or $\operatorname{PDF} \rho_{X}$ :
* For any $n$, generate a sequence of random variables $x_{i}$ as above with CDF $F_{X}$
* compute

$$
\frac{1}{n} \sum_{i=1}^{n} g\left(x_{i}\right) \longrightarrow E[g(X)]=\int g(x) d F_{X}=\int_{-\infty}^{\infty} g(x) \rho_{X}(x) d x
$$

* In particular,

$$
\begin{gathered}
\frac{1}{n} \sum_{i=1}^{n} x_{i} \\
\longrightarrow E[X] \\
\frac{1}{n} \sum_{i=1}^{n} x_{i}^{2}-\left(\frac{1}{n} \sum_{i=1}^{n} x_{i}\right)^{2} \\
\longrightarrow E\left[X^{2}\right]-E[X]^{2}=\sigma^{2}
\end{gathered}
$$

* Central Limit Theory: The average of a large number independent random variable converges to a normal distribution. More precisely, let $X_{i}$ be a random variable with mean $\mu$ and variance $\sigma^{2}$, then

$$
\frac{\sum_{i=1}^{n}\left(X_{i}-\mu\right)}{\sigma \sqrt{n}} \longrightarrow N(0,1), \text { as } n \longrightarrow \infty
$$

* Error is $O(1 / \sqrt{n})$

$$
\begin{aligned}
\frac{1}{n} \sum_{i=1}^{n} g\left(X_{i}\right)-E[g(X)] & =\frac{\sum_{i=1}^{n}\left(g\left(X_{i}\right)-E[g(X)]\right)}{n} \\
& =\frac{\sum_{i=1}^{n}\left(g\left(X_{i}\right)-E[g(X)]\right)}{\sigma \sqrt{n}} \frac{\sigma}{\sqrt{n}} \\
& \longrightarrow N(0,1) \frac{\sigma}{\sqrt{n}}
\end{aligned}
$$

- Consider checkout time in a library or grocery story. Assume the time $T$ between requests for services has the exponential distribution

$$
\rho(t)=\frac{e^{-t / a}}{a}, \text { for } t>0
$$

- Assume there are $m$ checkout lines. If the first one is busy, then the requested is handed out to the second, and so on. If all lines are busy, the request is rejected.
- Assume that each line processes each request in time $S$ that is normally distributed with mean $\mu$ and variance $\sigma^{2}$.
- The $C D F$ for $T$ is

$$
y=F(t)=\int_{0}^{t} \frac{e^{-x / a}}{a} d x=1-e^{-t / a}
$$

- The inverse function $F^{-1}$ is

$$
t=-a \ln (1-y)
$$

- So there exists a uniform distribution in $[0,1], Y^{\sim} U[0,1]$

$$
T=-a \ln (1-Y)
$$

- See routine in page 25
- Homework:
- Exercise 1: Using MC to find the integral (see Example in page 22), within an error of $10^{-5}$.What is the smallest $N$ ?
- Exercise 2: Run MC Routine in page 23. Then write your own routine for the following cases:
(a) There are five devices $T_{1}, \ldots, T_{5}$, which have normal distributions with mean $11,12,13,14,15$, and standard deviation $1,2,3,4,5$, respectively.
(b) Repeat (a) with uniform distribution instead (with the same neans and standard deviations). (Hint: $U[\mu-\sigma \sqrt{3}, \mu+\sigma \sqrt{3}]$ has mean $\mu$ and variance $\sigma^{2}$ )
- Exercise 3: Write a MC routine to compute $E[X]$ and $\sigma^{2}[X]$ with given CDF of $X$. Then test your MC routine using exponential distribution with $\lambda=2$ and $\lambda=5$.
- Exercise 4: Write a Matlab routine to generate $N(0,1)$ (e.g., find mean and variance) using using "rand" command only (you cannot use "randn" command) and Lemmas of page 25. Test your results in 2(a).
- mOptionalmDevelop a Matlab routine for MC integration of general integral $\iint_{D} f(x, y) d x d y$ for any function $f(x, y)$ in and domain $D$. Note that $D$ is not necessarily a rectanglar domain. For instance, $D$ could be the unit disk. Then, (accurate up to $10^{-3}$ )
(i) use your program to compute

$$
\int_{0}^{2} d y \int_{-1}^{2} e^{-\left(x^{2} y\right)} \sin x \cos y d x
$$

(ii) Let $B_{1}$ be the unit disk, i.e., $B_{1}: x^{2}+y^{2} \leq 1$, use MC to find

$$
\iint_{B_{1}} e^{-\left(x^{2} y\right)} \sin x \cos y d x d y
$$

