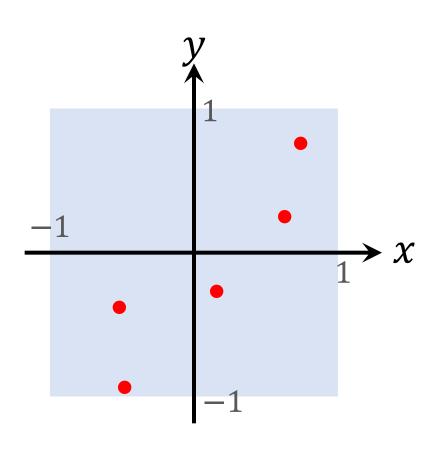
Monte Carlo Algorithms

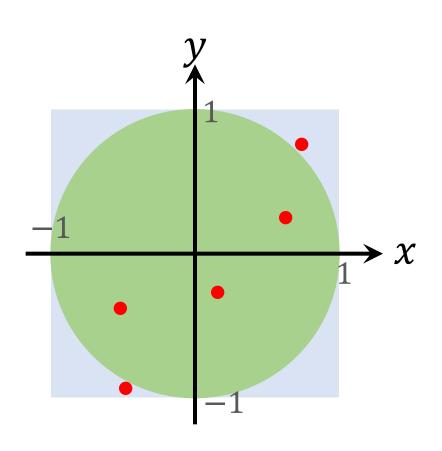
Shusen Wang

Application 1: Calculating Pi

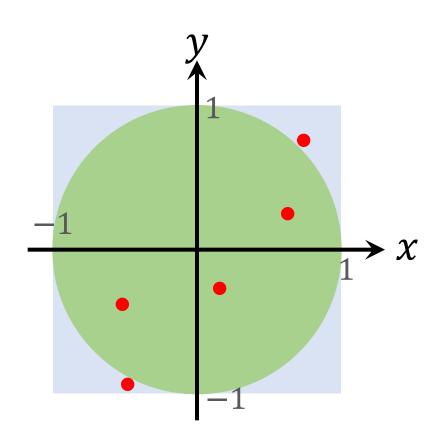
- We already know $\pi \approx 3.141592653589 \dots$
- Pretend we do not know the value of π .
- Can we find it out (approximately) using a random number generator?



• Assume (x, y) is a point sampled from square uniformly at random.

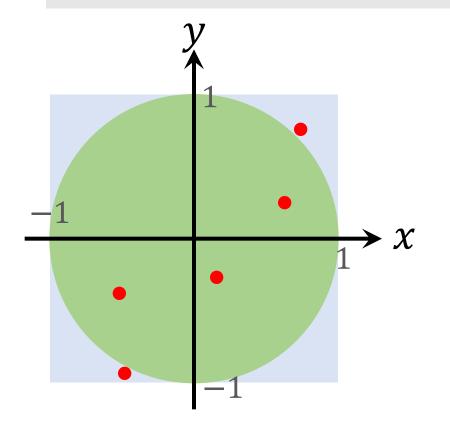


- Assume (x, y) is a point sampled from square uniformly at random.
- What is the probability that (x, y) is in the circle?



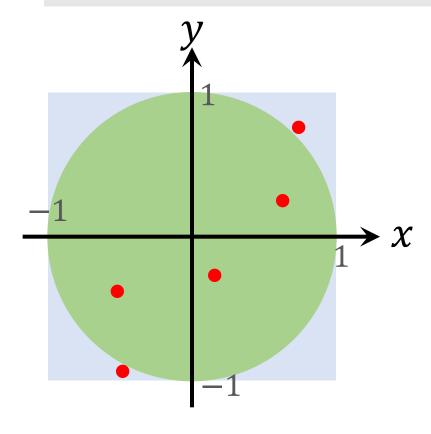
- Assume (x, y) is a point sampled from square uniformly at random.
- What is the probability that (x, y) is in the circle?
- Area of the square is $A_1 = 2^2 = 4$.
- Area of the circle is $A_2=\pi r^2=\pi$.
- Probability: $P = \frac{A_2}{A_1} = \frac{\pi}{4}$.

- Suppose n points are uniformly sampled from $[-1, 1] \times [-1, 1]$.
- Then, in expectation, $P_n = \frac{\pi n}{4}$ points are in the circle.



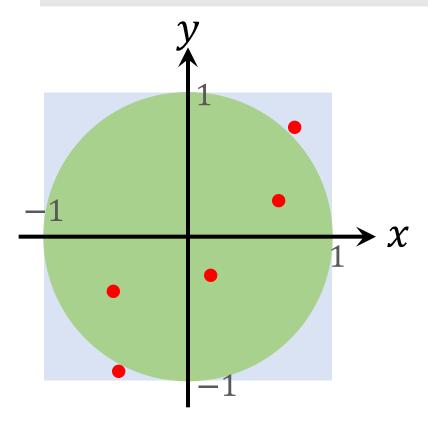
- Assume (x, y) is a point sampled from square uniformly at random.
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- Suppose n points are uniformly sampled from $[-1, 1] \times [-1, 1]$.
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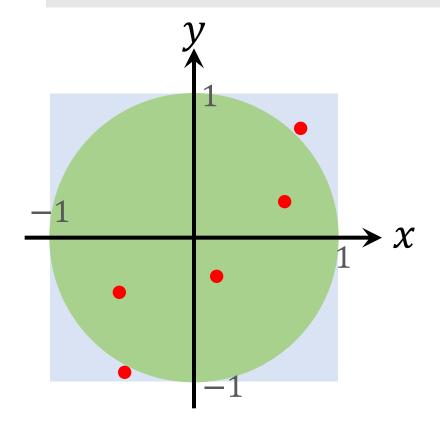
- Given a point (x, y), how do you know whether (x, y) is in the circle?
- If $x^2 + y^2 \le 1$, then it is in the circle.

- Suppose n points are uniformly sampled from $[-1, 1] \times [-1, 1]$.
- Then, in expectation, $Pn = \frac{\pi n}{4}$ points are in the circle.



- We found m points in the circle.
- If n is big, then $m \approx \frac{\pi n}{4}$.
- Thus, $\pi \approx \frac{4m}{n}$.

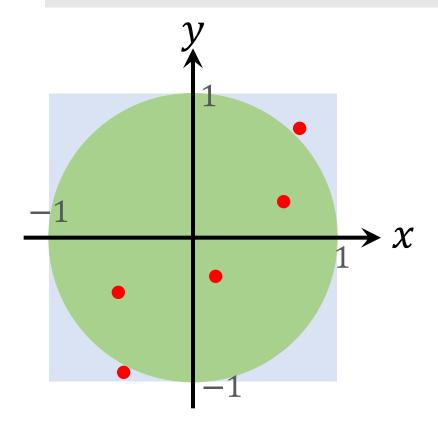
- Suppose n points are uniformly sampled from $[-1, 1] \times [-1, 1]$.
- Then, in expectation, $P_n = \frac{\pi n}{4}$ points are in the circle.



• Law of big number:

$$\frac{4m}{n} \to \pi$$
, as $n \to \infty$.

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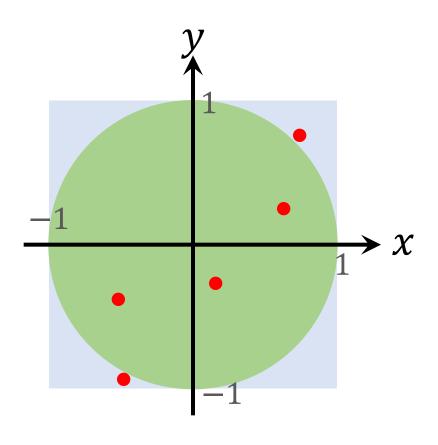
• Law of big number:

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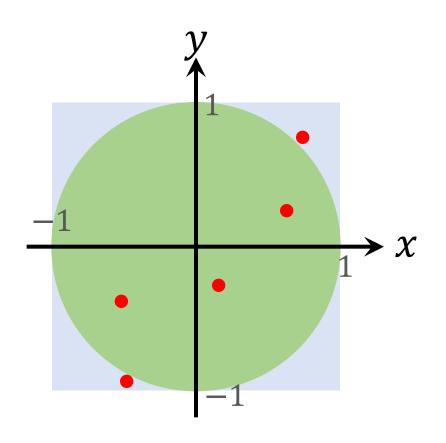
Concentration bound:

$$\left|\frac{4m}{n} - \pi\right| = O\left(\frac{1}{\sqrt{n}}\right).$$

Totally *n* points sampled from the square; *m* are in the circle. Then $\pi \approx \frac{4m}{n}$.



Totally n points sampled from the square; m are in the circle. Then $\pi \approx \frac{4m}{n}$.



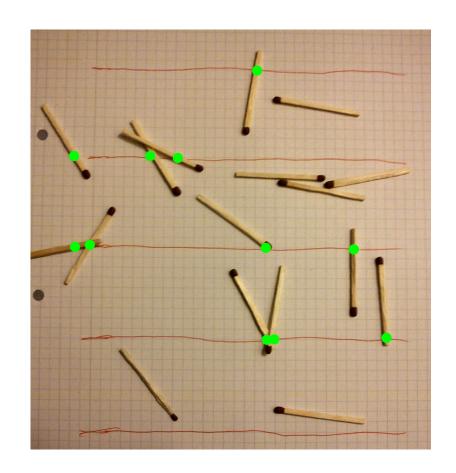
Algorithm

- 1. User specify a big n; set m = 0.
- 2. For i = 1 to n:
 - a) Randomly generate $x \in [-1, 1]$.
 - b) Randomly generate $y \in [-1, 1]$.
 - c) If $x^2 + y^2 \le 1$, then $m \leftarrow m + 1$.
- 3. Return $\pi \approx \frac{4m}{n}$.

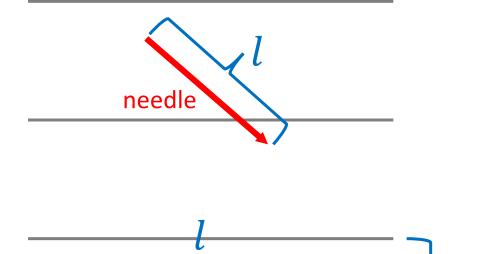
Application 2: Buffon's Needle Problem



Buffon, 1707 – 1788 French scientist

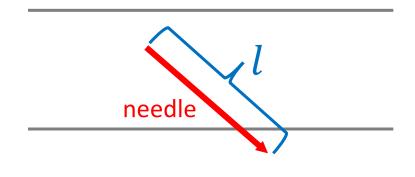


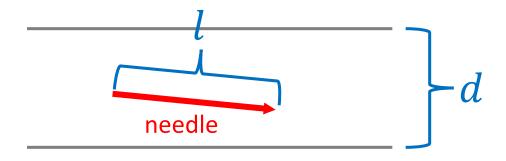
Buffon's Needle Problem



needle

- The parallel lines have distance d.
- Needles have length *l*.
- Randomly throw a needle; the needle may or may not lie across a line.

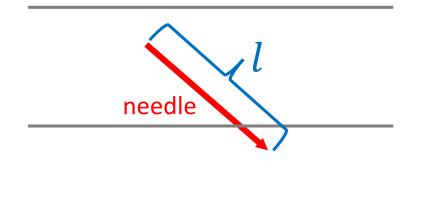


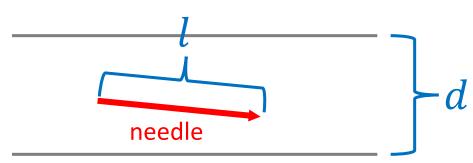


- The parallel lines have distance d.
- Needles have length *l*.
- Randomly throw a needle; the needle may or may not lie across a line.

With probability $P = \frac{2l}{\pi d}$, they are across.

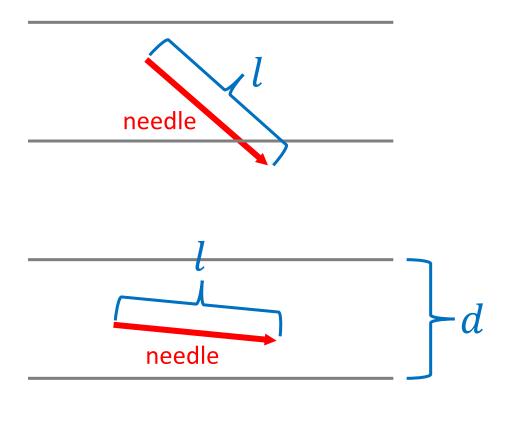
• It can be proved using integral.





With probability $P = \frac{2l}{\pi d}$, they are across.

- Randomly throw a total of n needles.
- In expectation, $P_n = \frac{2ln}{\pi d}$ needles lie across the lines.



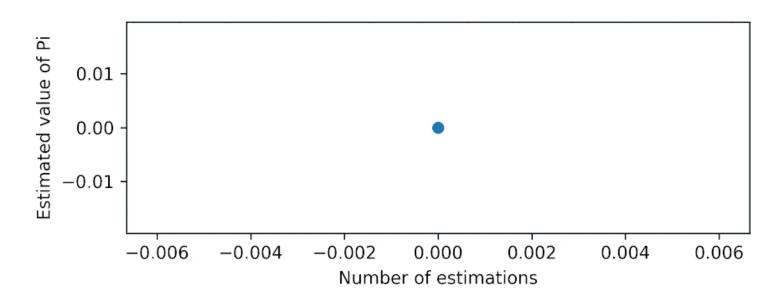
With probability $P = \frac{2l}{\pi d}$, they are across.

- Randomly throw a total of n needles.
- In expectation, $Pn = \frac{2ln}{\pi d}$ needles lie across the lines.
- Actually observe m needles across the lines.
- If n is big, $m \approx \frac{2 l n}{\pi d}$.
- Thus, $\pi \approx \frac{d m}{2 l n}$.

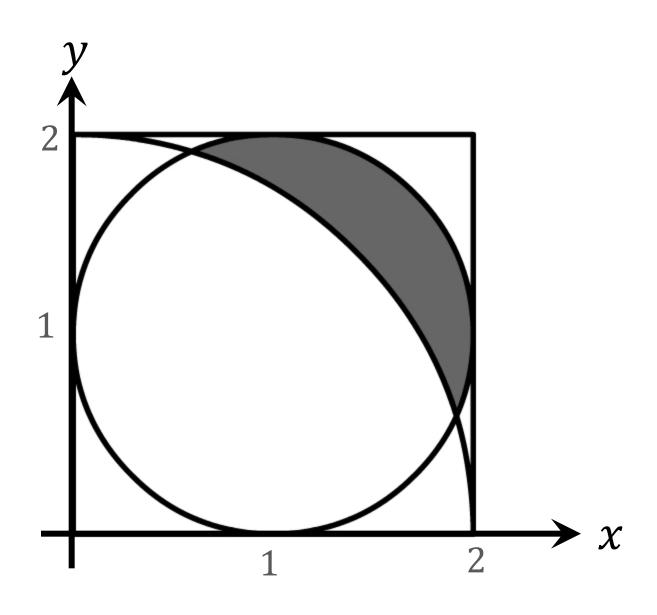
Researcher	Year	n =	m =	Estimate of π
Wolf	1850	5000	2532	3.1596
Smith	1855	3204	1218	3.1554
De Morgan	1860	600	382	3.137
Fox	1884	1030	489	3.1595
Lazzerini	1901	3408	1808	3.1415929
Reina	1925	2520	859	3.1795

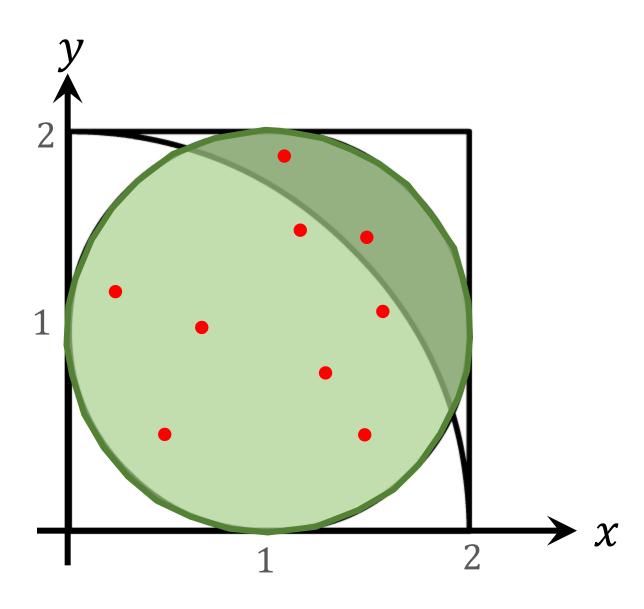






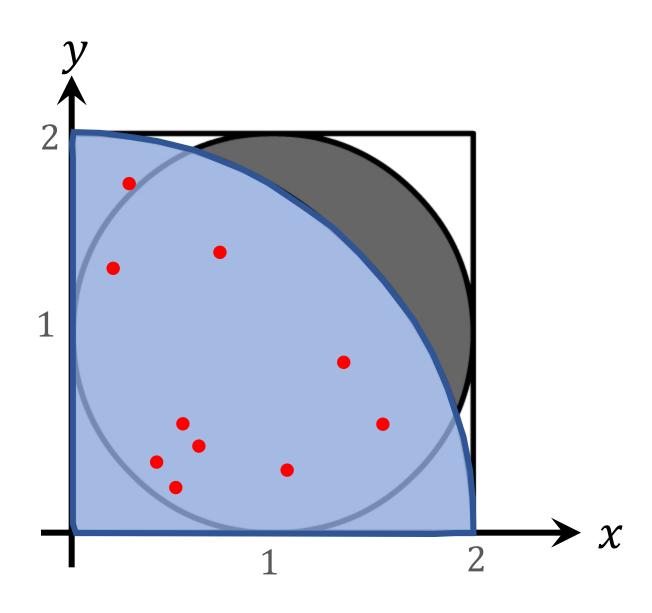
Application 3: Area of A Region





 If a point (x, y) is in the circle, it must satisfy

$$(x-1)^2 + (y-1)^2 \le 1.$$

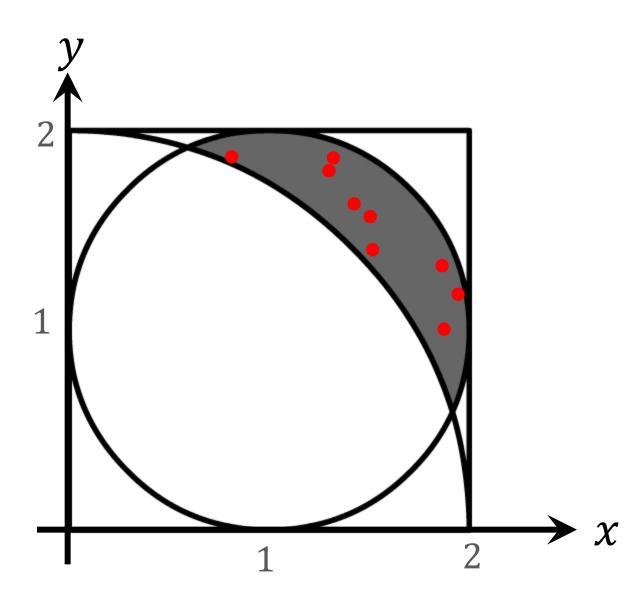


 If a point (x, y) is in the circle, it must satisfy

$$(x-1)^2 + (y-1)^2 \le 1.$$

 If a point (x, y) is in the quarter circle, it must satisfy

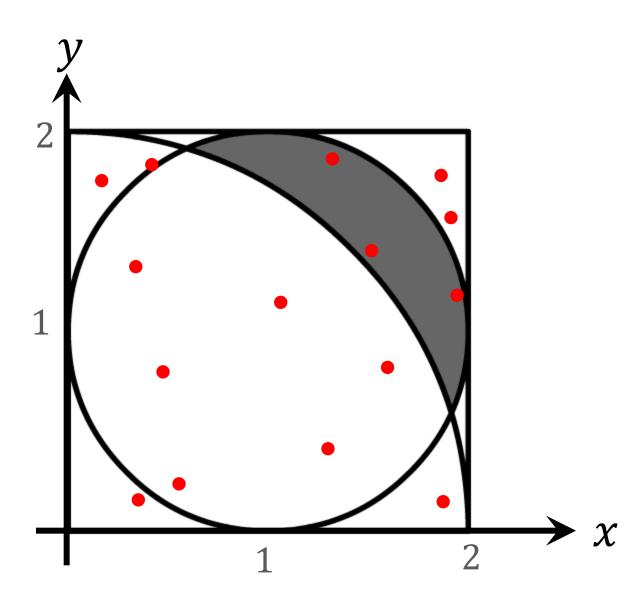
$$x^2 + y^2 \le 2^2.$$



• A point (x, y) in the grey region satisfies:

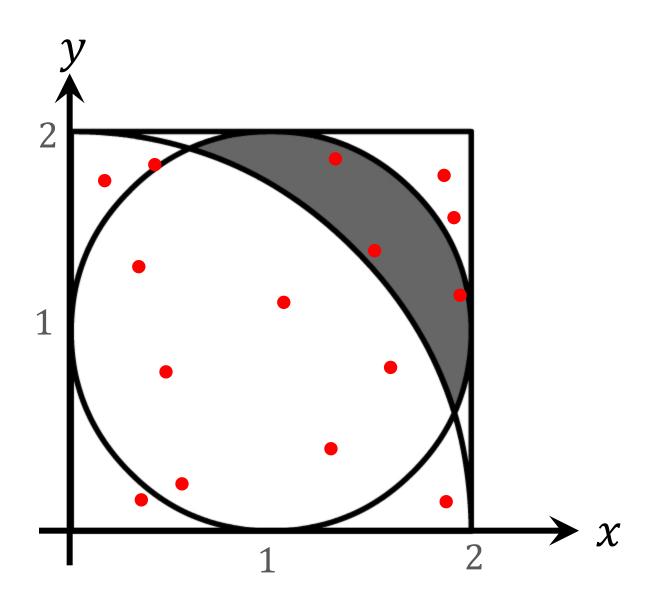
1.
$$(x-1)^2 + (y-1)^2 \le 1$$
.

2.
$$x^2 + y^2 > 2^2$$
.

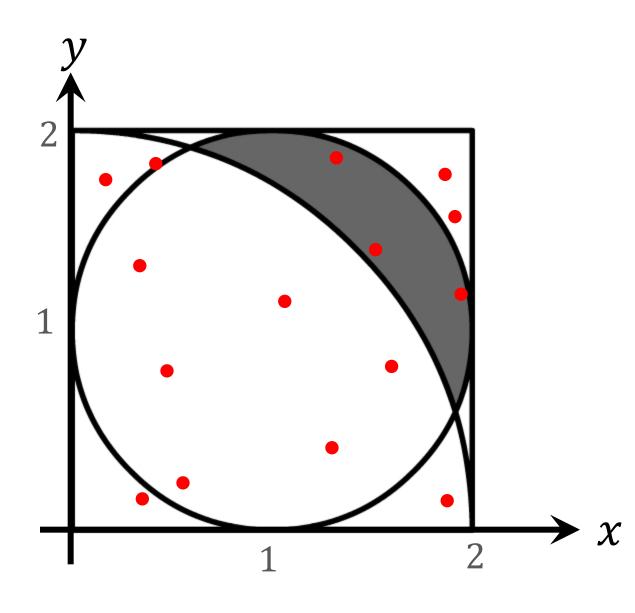


- Area of the square: $A_1 = 2^2 = 4$.
- Area of the grey region: A_2 .
- A point uniformly sampled from the square falls in the grey region w.p.

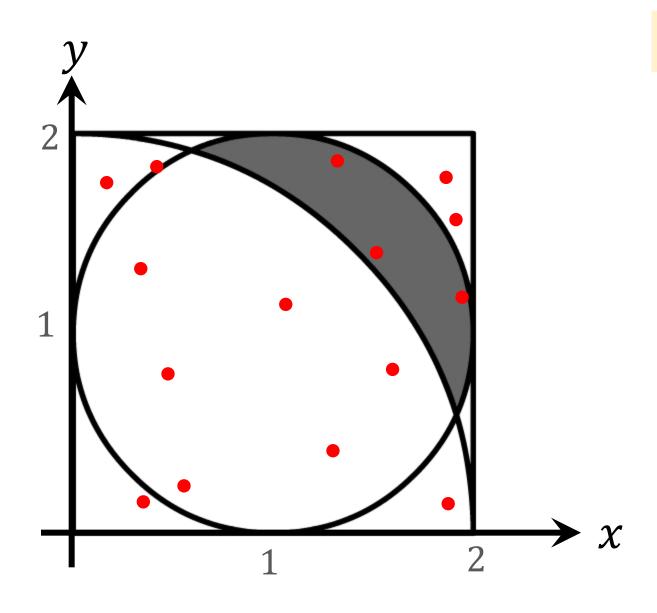
$$P = \frac{A_2}{A_1} = \frac{A_2}{4}$$
.



- Uniformly sample n points from the square $[0, 2] \times [0, 2]$.
- In expectation, $nP = \frac{n A_2}{4}$ points fall in the grey region.



- Uniformly sample n points from the square $[0, 2] \times [0, 2]$.
- In expectation, $nP = \frac{n A_2}{4}$ points fall in the grey region.
- We actually observe m points in the grey region.
- If n is big, then $m \approx \frac{n A_2}{4}$.
- Thus, $A_2 \approx \frac{4m}{n}$.



Algorithm

- 1. User specify a big n.
- 2. For i = 1 to n:
 - a) Randomly generate $x \in [0, 2]$.
 - b) Randomly generate $y \in [0, 2]$.
 - c) If both of the following conditioned are satisfied, then $m \leftarrow m + 1$:

i.
$$(x-1)^2 + (y-1)^2 \le 1$$
.

ii.
$$x^2 + y^2 > 2^2$$
.

3. Return area = $\frac{4m}{n}$.

Application 4: Integration

Integration

- We are given a function, e.g., $f(x) = \frac{1}{1 + \sin(x) \cdot (\log_e x)^2}$.
- Calculate the integral: $I = \int_{0.8}^{3} f(x) dx$.
- If f(x) is very involved, there is no way to analytically calculate the integral.
- For example, many machine learning applications requires such an integral where f(x) is a deep neural network.

Monte Carlo Integration (Univariate)

Task: Given a univariate function f(x), calculate $I = \int_a^b f(x) dx$.

Monte Carlo Integration (Univariate)

Task: Given a univariate function f(x), calculate $I = \int_a^b f(x) dx$.

- 1. Drawn n samples from [a, b] uniformly at random, denote x_1, \dots, x_n .
- 2. Calculate $Q_n = (b-a) \cdot \frac{1}{n} \sum_{i=1}^n f(x_i)$.
- 3. Return Q_n as an approximation to the integral $I = \int_a^b f(x) dx$.

Monte Carlo Integration (Univariate)

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Theory: Law of large numbers guarantees $Q_n \to I$, as $n \to \infty$.

Monte Carlo Integration (Univariate): Example

Task: Given function
$$f(x) = \frac{1}{1+\sin(x)\cdot(\log_e x)^2}$$
, calculate $I = \int_{0.8}^3 f(x) \ dx$.

- 1. Drawn n samples from [0.8, 3] uniformly at random, denote x_1, \dots, x_n .
- 2. Calculate $Q_n = 2.2 \cdot \frac{1}{n} \sum_{i=1}^{n} f(x_i)$.
- 3. Return Q_n as an approximation to the integral $I = \int_{0.8}^3 f(x) \ dx$.

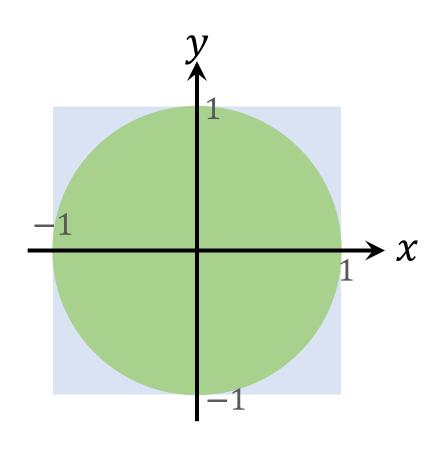
Monte Carlo Integration (Multivariate)

Task: Given a multivariate function $f(\mathbf{x})$, calculate $I = \int_{\Omega} f(\mathbf{x}) \ d\mathbf{x}$. $\mathbf{x} \in \mathbb{R}^d$ is a vector Ω is a subset of \mathbb{R}^d

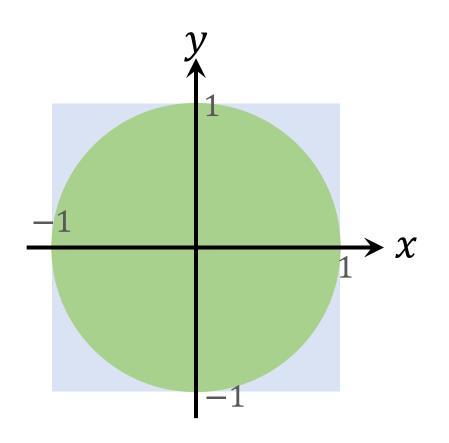
Monte Carlo Integration (Multivariate)

Task: Given a multivariate function $f(\mathbf{x})$, calculate $I = \int_{\Omega} f(\mathbf{x}) d\mathbf{x}$.

- 1. Draw n samples from set Ω uniformly at random, denote $\mathbf{x}_1, \dots, \mathbf{x}_n$.
- 2. Calculate $V = \int_{\Omega} d\mathbf{x}$.
- 3. Calculate $Q_n = V \cdot \frac{1}{n} \sum_{i=1}^n f(\mathbf{x}_i)$.
- 4. Return Q_n as an approximation to the integral $I = \int_{\Omega} f(\mathbf{x}) d\mathbf{x}$.

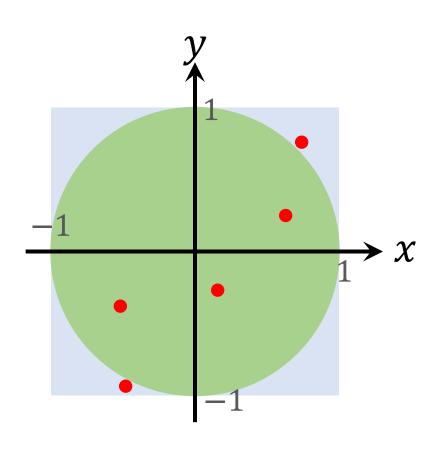


- Let $f(x,y) = \begin{cases} 1, & \text{if } x^2 + y^2 \le 1; \\ 0, & \text{otherwise.} \end{cases}$
- Let $\Omega = [-1, 1] \times [-1, 1]$.
- Question: What is $I = \int_{\Omega} f(x, y) dx dy$?



- Let $f(x,y) = \begin{cases} 1, & \text{if } x^2 + y^2 \le 1; \\ 0, & \text{otherwise.} \end{cases}$
- Let $\Omega = [-1, 1] \times [-1, 1]$.
- Question: What is $I = \int_{0}^{\infty} f(x, y) dx dy$?
 - Obviously, *I* is the area of the circle:

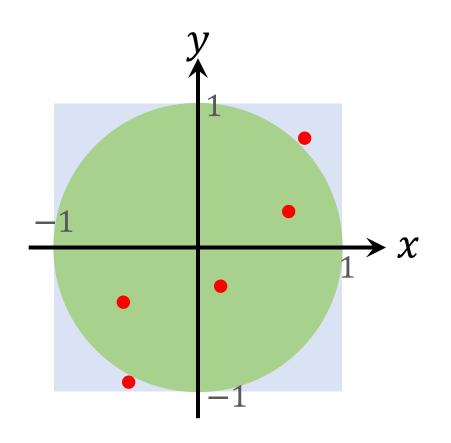
$$I = \pi r^2 = \pi.$$



1. Draw n samples from set Ω uniformly at random, denote

$$(x_1, y_1), \dots, (x_n, y_n).$$

2. Calculate $V = \int_{\Omega} dx \ dy = 4$. (It is the area of set Ω .)



1. Draw n samples from set Ω uniformly at random, denote

$$(x_1, y_1), \dots, (x_n, y_n).$$

- 2. Calculate $V = \int_{\Omega} dx \ dy = 4$. (It is the area of set Ω .)
- 3. Calculate $Q_n = V \cdot \frac{1}{n} \sum_{i=1}^n f(\mathbf{x_i}, \mathbf{y_i})$.
- 4. Return Q_n as an approximation to the integral $\pi = \int_{\Omega} f(x,y) \, dx \, dy$.

Application 5: Estimate of Expectation

Expectation

- Let X be a d-dimensional random vector variable.
- Let p(x) be a probability density function (PDF).
 - Property: $\int_{\mathbb{R}^d} p(\mathbf{x}) d\mathbf{x} = 1$.
 - E.g., PDF of univariate Gaussian is $p(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right]$.

Expectation

- Let X be a d-dimensional random vector variable.
- Let p(x) be a probability density function (PDF).
- Let $f(\mathbf{x})$ be any function of vector variable.
- Expectation: $\mathbb{E}_{X \sim p}[f(X)] = \int_{\mathbb{R}^d} f(\mathbf{x}) \cdot p(\mathbf{x}) d\mathbf{x}$.

Monte Carlo Estimate of Expectation

Task: Estimate the expectation $\mathbb{E}_{X \sim p}[f(X)] = \int_{\mathbb{R}^d} f(\mathbf{x}) \cdot p(\mathbf{x}) d\mathbf{x}$.

Monte Carlo Estimate of Expectation

Task: Estimate the expectation $\mathbb{E}_{X \sim p}[f(X)] = \int_{\mathbb{R}^d} f(\mathbf{x}) \cdot p(\mathbf{x}) d\mathbf{x}$.

- 1. Draw n random samples from probability distribution $p(\mathbf{x})$; denote them by $\mathbf{x}_1, \dots, \mathbf{x}_n$.
- 2. Calculate $Q_n = \frac{1}{n} \sum_{i=1}^n f(\mathbf{x}_i)$.
- 3. Return Q_n as an approximation to the $\mathbb{E}_{X \sim p}[f(X)]$.

Monte Carlo and Beyond

Monte Carlo



Casino de Monte-Carlo, Monaco

 The term "Monte Carlo method" was first introduced in 1947 by Nicholas Metropolis.

Reference

1. Metropolis. The beginning of the Monte Carlo method. Los Alamos Science, 125–130, 1987.

Monte Carlo Algorithms

- Monte Carlo refers to algorithms that rely on repeated random sampling to obtain numerical results.
- The output of Monte Carlo algorithms can be incorrect.
 - In all of our examples, the algorithms' outputs are incorrect.
 - But they are close to the correct solution.

Monte Carlo Algorithms

- Monte Carlo refers to algorithms that rely on repeated random sampling to obtain numerical results.
- The output of Monte Carlo algorithms can be incorrect.
 - In all of our examples, the algorithms' outputs are incorrect.
 - But they are close to the correct solution.
- Las Vegas algorithms are those always produce the correct answer.
 - E.g., random quicksort.
- Atlantic City algorithms are polynomial-time randomized algorithm that answers correctly w.p. greater than 50%.

Thank you!