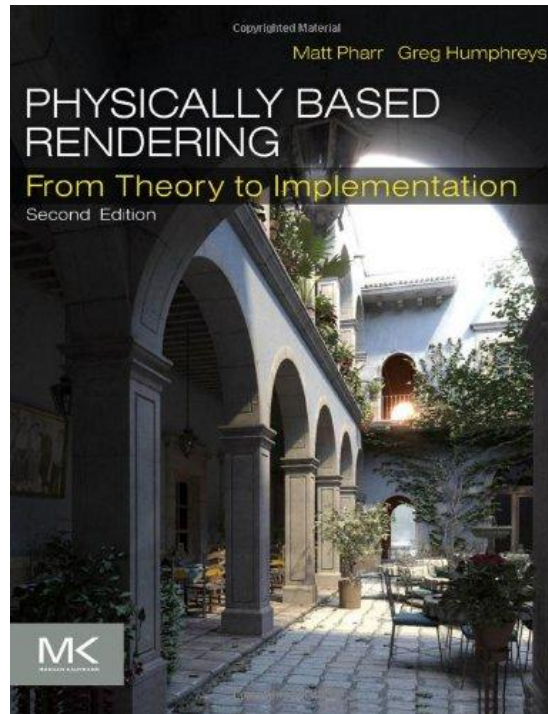


Path Tracing: Monte Carlo integration

Physically based rendering

- Oscar winner



- <https://www.youtube.com/watch?v=7d9juPsv1QU>

Randomize algorithm

- Monte Carlo
 - Monte Carlo integration
- Las Vegas
 - Quick sort

Probability Review

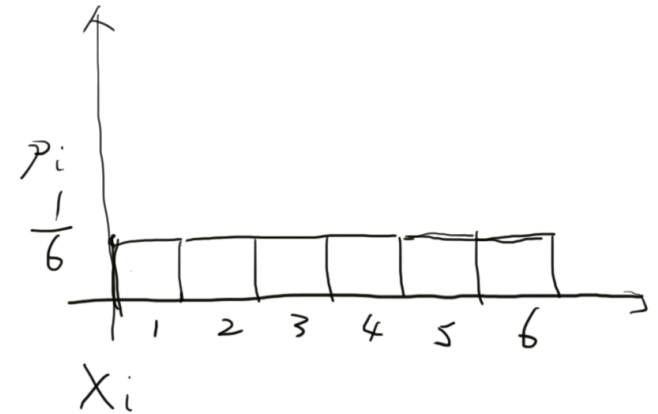
- Random Variable:
 - X
 - A value chosen by some random process
- Discrete
 - Roll a Dice and see its outcome
- Continuous
 - A random person's weight

Dice

$$X_i = \{1, 2, 3, 4, 5, 6\}$$

$$P_i = \frac{1}{6}$$

$$\sum_{i=1}^6 P_i = 1$$



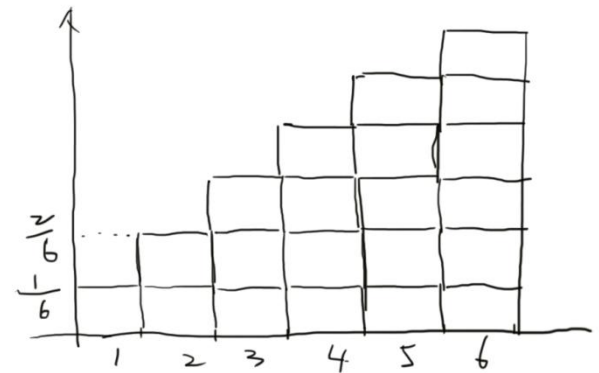
The *cumulative distribution function* (CDF)

$$P(x) = \Pr\{X \leq x\}$$

= outcome of the dice is less or equal to x

example

$$P(2) = \frac{1}{3}$$



Continuous random variables

X_i = weight of a randomly chosen person

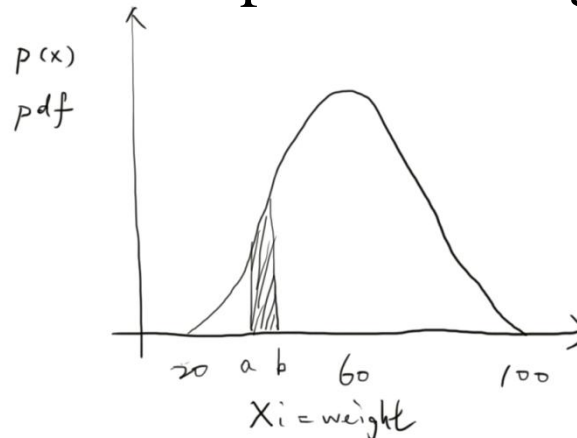
$p(x)$
pdf



$p(x)$ = probability density function

$$P(x \in [a, b]) = \int_a^b p(x) dx$$

= probability of a random person's weight is between a , and b



Continuous random variables

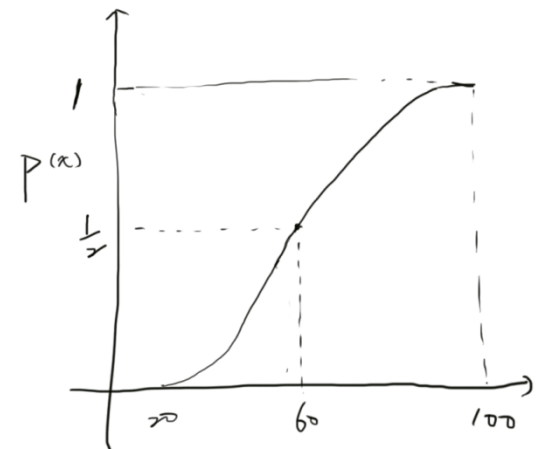
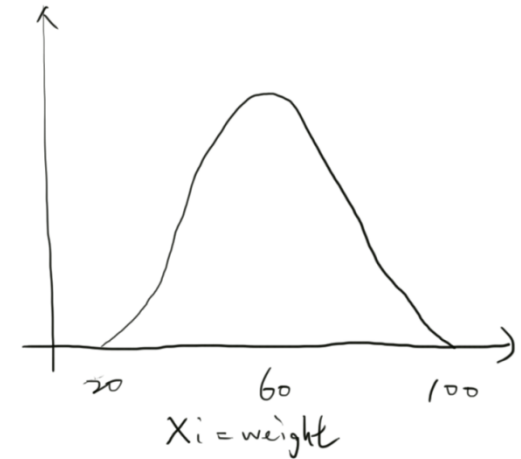
X_i = weight of a randomly chosen person $p(x)$
pdf

The *cumulative distribution function* (CDF)

$$P(x) = \Pr\{X \leq x\}$$

= A randomly chosen person's weight is less than x

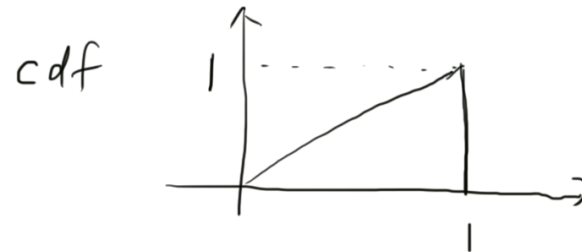
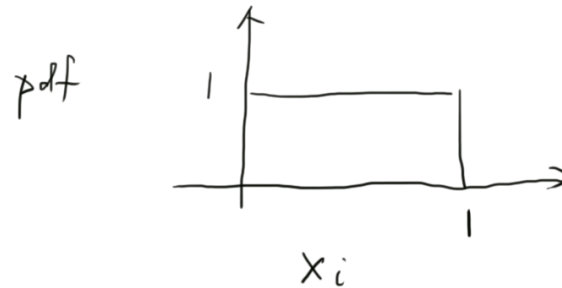
$$= \int_{-\infty}^x p(x) dx$$



Uniform random variable

canonical uniform random variable ξ
(pronounced as Xi)

$$p(x) = \begin{cases} 1 & x \in [0,1) \\ 0 & \text{otherwise} \end{cases}$$



Expected values

- Expected value of a function f
 - Average value of the function over some distribution p

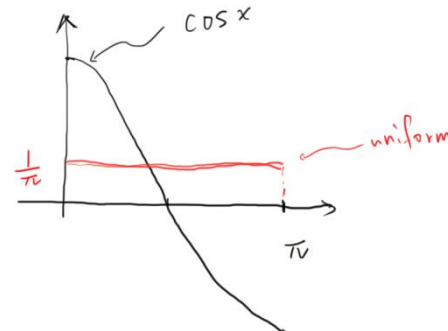
$$E_p[f(x)] = \int_D f(x)p(x)dx$$

- If p is uniform, we drop the subscript p from E_p

- Example

- Expected value of $\cos(x)$, x between 0 and π

$$E[\cos(x)] = \int_0^\pi \frac{\cos(x)}{\pi} dx = \frac{1}{\pi} (-\sin \pi + \sin 0) = 0$$



Properties of E

$$E[af(x)] = aE[f(x)]$$

$$E[\sum_i f(X_i)] = \sum_i E[f(X_i)]$$

The Monte Carlo estimator

- Suppose we want to evaluate the integral

$$\int_a^b f(x)dx$$

- Given uniform random variables $X_i \in [a, b]$
- We claim

$$F_N = \frac{b-a}{N} \sum_{i=1}^N f(X_i) \text{ is the estimator of } \int_a^b f(x)dx$$

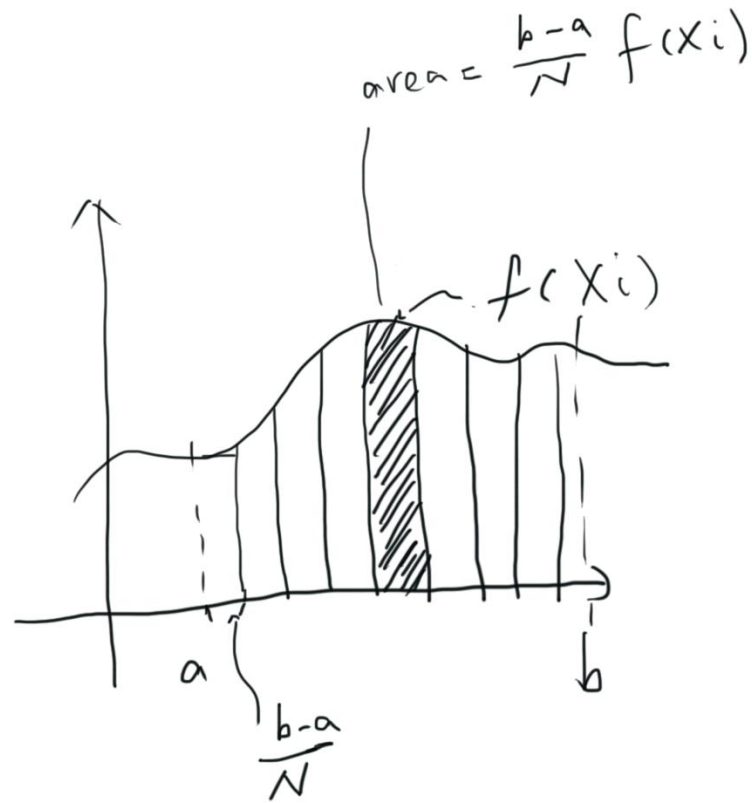
$$(\text{when } N \rightarrow \infty, F_N = \int_a^b f(x)dx)$$

$$(E[F_N] = \int_a^b f(x)dx)$$

Proof

$$\begin{aligned} E[F_N] &= E\left[\frac{b-a}{N} \sum_{i=1}^N f(X_i)\right] \\ &= \frac{b-a}{N} \sum_{i=1}^N E[f(X_i)] \\ &= \frac{b-a}{N} \sum_{i=1}^N \int_a^b f(x) p(x) dx \\ &= \frac{1}{N} \sum_{i=1}^N \int_a^b f(x) dx \\ &= \int_a^b f(x) dx \end{aligned}$$

Geometric interpretation



More general form

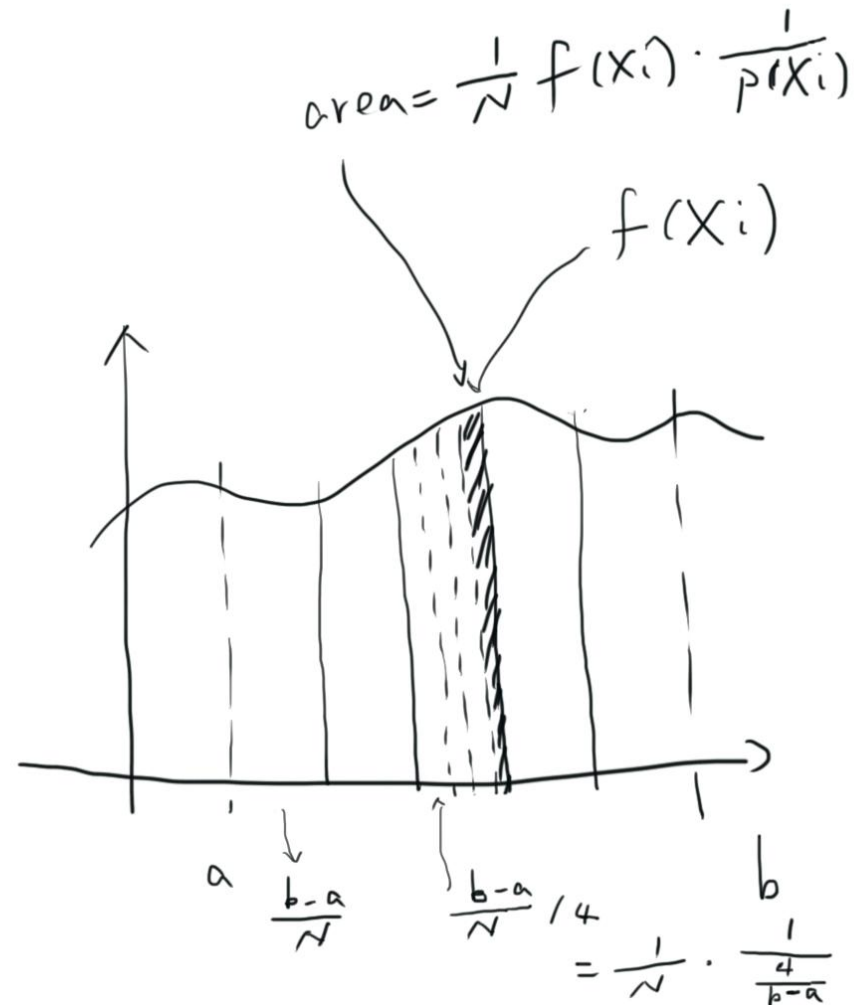
- Given a probability density function $p(x)$

$$F_N = \frac{1}{N} \sum_{i=1}^N \frac{f(X_i)}{p(X_i)} \text{ is the estimator of } \int_a^b f(x) dx$$

proof

$$\begin{aligned} E[F_N] &= E\left[\frac{1}{N} \sum_{i=1}^N \frac{f(X_i)}{p(X_i)}\right] \\ &= \frac{1}{N} \sum_{i=1}^N \int_a^b \frac{f(x)}{p(x)} p(x) dx \\ &= \frac{1}{N} \sum_{i=1}^N \int_a^b \frac{f(x)}{p(x)} dx = \int_a^b f(x) dx \end{aligned}$$

Geometric interpretation



Why we need this form

- Suppose we have a pdf $p(x) = c \cdot f(x)$

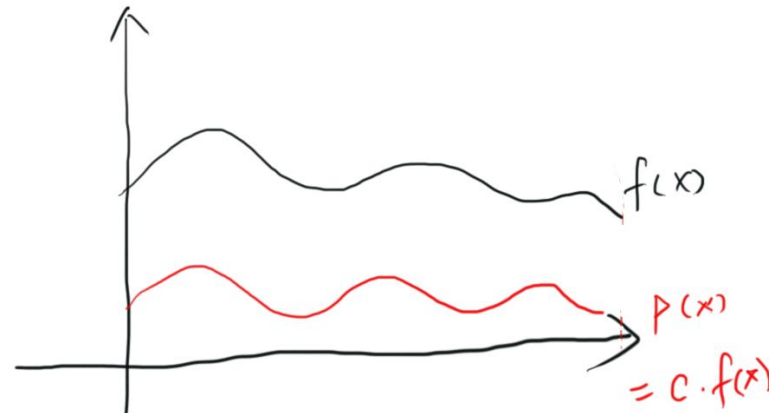
$$\Rightarrow \int p(x)dx = \int c \cdot f(x)dx = 1$$

$$\Rightarrow c = \frac{1}{\int f(x)dx}$$

- Then the Monte Carlo estimator

$$\frac{1}{N} \sum_{i=1}^N \frac{f(X_i)}{p(X_i)} = \frac{1}{N} \sum_{i=1}^N \frac{1}{c} = \int f(x)dx,$$

Perfect estimate!!



Extending to multiple dimensions

- Extending to multiple dimensions is straightforward
- Consider the three-dimensional integral

$$\int_{x_0}^{x_1} \int_{y_0}^{y_1} \int_{z_0}^{z_1} f(x, y, z) dx dy dz$$

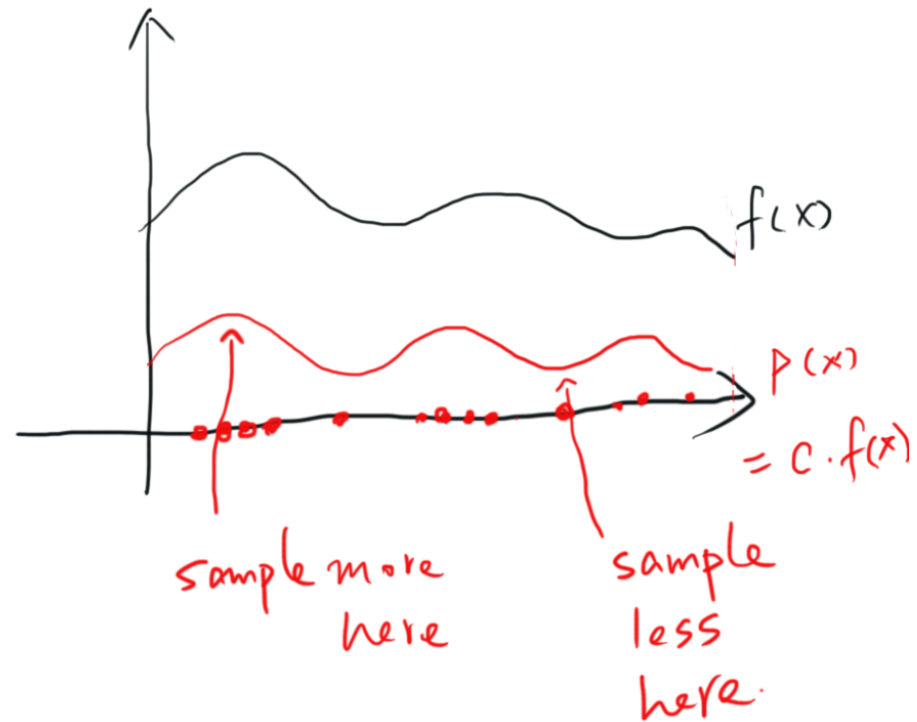
- If we use uniform sampling $p(x, y, z) = \frac{1}{(x_1 - x_0)} \frac{1}{(y_1 - y_0)} \frac{1}{(z_1 - z_0)}$

then the estimator is $\frac{(x_1 - x_0)(y_1 - y_0)(z_1 - z_0)}{N} \sum_i f(X_i)$

Multiple dimensions

- The number of samples N is independent of the dimensionality of the integral
- Error in Monte Carlo estimator decreases at a rate of $O(\sqrt{N})$

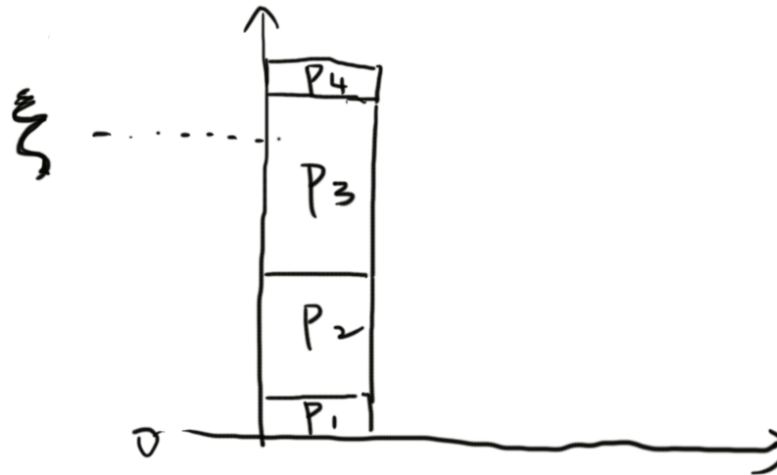
Sampling of random variables



How to sample from a PDF



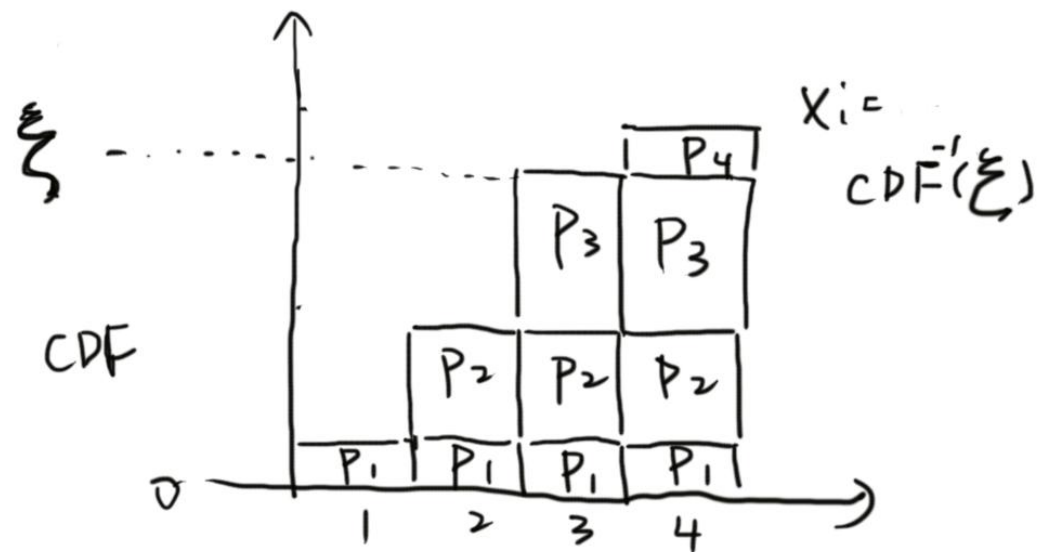
how to sample?



Inversion method



how to sample?



Inversion method

1. Compute the CDF $P(x) = \int_0^x p(x')dx'$
2. Compute the inverse $P^{-1}(x)$
3. Obtain a uniformly distributed random number ξ
4. Compute $X_i = P^{-1}(\xi)$

