

Lecture October 22

Monte Carlo Methods

— MC integration.

$$E[f] = \langle f \rangle = \int_{x \in D} f(x) p(x) dx = \mu_f$$

$$\sigma_f^2 = \int_{x \in D} (f(x) - \mu_f)^2 p(x) dx$$

$$\begin{cases} \mu_f = \sum_{i \in D} f(x_i) p(x_i) \\ \sigma_f^2 = \sum_{i \in D} (f(x_i) - \mu_f)^2 p(x_i) \end{cases}$$

Sample mean and other
sample expectation values

$$\mu_f \neq \bar{\mu}_f = \frac{1}{N} \sum_{i=1}^N f(x_i)$$

$$\sigma_f^2 \neq \bar{\sigma}_f^2 = \frac{1}{N} \sum_{i=1}^N (f(x_i) - \bar{\mu}_f)^2$$

standard deviation

$$\sigma_f \sim \frac{1}{\sqrt{N}}$$

Central limit theorem (later)

MC - integration (smarter)

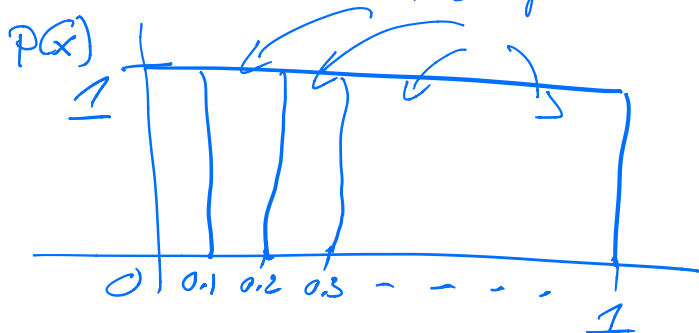
- change of variables
- importance sampling

$$I = \int_a^b f(x) dx \approx \frac{1}{N} \sum_{x_i \in [a,b]} f(x_i)$$

RNG, uniform distribution

RN = Random number

RN $\tilde{x}_i \in [0,1]$ with
uniform PDF $p(x) dx$
1000 points



$$N = 10^4$$

can map $\tilde{x}_i \in [0,1] \rightarrow x_i \in [a,b]$

$$x_i = a + (b-a)\tilde{x}_i$$

$$\int_0^{\infty} \frac{e^{-x^2}}{x^2} dx$$

changes of variables;

$$p(x)dx = \begin{cases} dx & 0 \leq x \leq 1 \\ 0 & \text{else} \end{cases}$$

$$\int_{x \in \mathbb{D}} p(x) dx = 1$$

Probability is conserved

$$p(y)dy = p(x)dx$$

↑ uniform distrib.

$$p(y)dy = dx$$

$$x(y) = \int_0^y p(y) dy =$$

CDF or cumulative distribution function,

$$p(y)dy = \begin{cases} \frac{dy}{b-a} & a \leq y \leq b \\ 0 & \text{else} \end{cases}$$

$$p(y)dy = \frac{dy}{b-a} = dx$$

$$x(y) = \int_0^y \frac{dy}{b-a} \Rightarrow$$

$$y = a + (b-a)x$$

$$y \in [0, 2\pi] \quad \swarrow \quad \frac{x \in [0, 1]}{\text{RNG}}$$

$$y = 2\pi \cdot x$$

Example : $p(y)dy = e^{-y}dy$
 $y \in [0, \infty)$

$$p(y)dy = dx$$

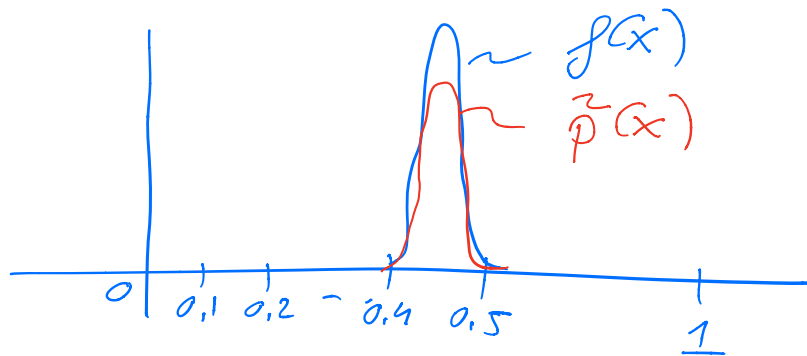
$$x(y) = \int_0^y e^{-y'} dy' = 1 - e^{-y}$$

$$\Rightarrow y(x) = -\ln(1-x)$$

$$x \in [0, 1]$$

Analytically integrable CDF,

$$I = \int_0^1 f(x) dx$$



RNG
 given by
 uniform
 distribution

$$x \in [0, 1]$$

$$N = 10^6$$

$$\begin{matrix} \tilde{p}(x) \\ f(x) \end{matrix} \begin{cases} x \in [0.4, 0.5] \end{cases}$$

$$\frac{f(x)}{\tilde{p}(x)} = \text{constant } T, \quad \text{change of variables}$$

$$I = \int_0^1 f(x) dx = \int_0^1 \tilde{p}(x) \frac{f(x)}{\tilde{p}(x)} dx \quad \text{+ importance sampling}$$

$$= \int_0^1 \tilde{p}(x) F(x) dx$$

$x \in [0, 1]$

$$\sigma_F^2 = \frac{1}{N} \sum_{x \in D} (F(x_i) - \mu_F)^2$$

$$= \frac{1}{N} \sum_{x \in D} (F(x_i))^2 - \mu_F^2$$

$$\mu_F = \text{constant}$$

$$= 0 \quad ?$$

Example

$$I = \int_0^1 \frac{1}{1+x^2} dx = \frac{\pi}{4}$$

$$\text{New } p(x) dx = \frac{1}{3} (4-2x) dx$$

$$\int_0^1 p(x) dx = 1 \quad x \in [0, 1]$$

$$f(x) = \frac{1}{1+x^2}$$

$$\frac{f(1)}{p(1)} = \frac{f(1)}{p(1)} = \frac{3}{4}$$

$$g(x) = \int_0^x p(x') dx' = \frac{1}{3} x(4-x)$$

$$x = 2 - (4-3g)^{3/2}$$

$$g=0 \quad x=0 \quad 1 \quad g=1 \quad x=1$$

$$\int_0^1 f(x) dx = \int_0^1 p(x) \frac{f(x)}{p(x)} dx$$

$$= \int_0^1 \frac{f(x(g))}{p(x(g))} dg$$

$$p(x) dx = \underbrace{p(g) dg}_{\text{uniform dist}} = dg$$

$$\frac{f(x(g))}{p(x(g))} = \text{constant}$$

$$\overline{I} = \text{constant} \Rightarrow$$

$$\sigma_f^2 = 0$$

$$\int_a^b f(x) dx = \int_a^b p(x) \frac{f(x)}{p(x)} dx$$

$$\int_{\tilde{a}}^{\tilde{b}} \frac{f(x(g))}{p(x(g))} dg$$

$$\sigma_F^2 = \frac{1}{N} \sum_{i=1}^N (F(x_i) - \mu_F)^2$$

$$= \frac{1}{N} \sum F(x_i)^2 + \mu_F^2$$

$$- 2 \mu_F \underbrace{\frac{1}{N} \sum F(x_i)}_{\mu_F}$$

$$= \frac{1}{N} \sum F(x_i)^2 - \mu_F^2$$