

#1. Write the joint density:

$$f(x, u) \propto M^{-1} \mathbb{1}(u \leq l(x)/M) h(x) \mathbb{1}(u \leq 1)$$

$$\text{where } l(x) = (-\log x)^2 \cdot x^2(1-x)/6$$

$$h(x) = 6x(1-x)$$

Find a bound M for $l(x)$ on $(0, 1)$

Although the analytical supremum of $l(x)$ is not easy to find, we can plot $l(x)$ numerically and pick $M = \frac{1}{60}$, which is not far from the sharp bound.

My algorithm:

1. Obtain X from p.d.f $h(x)$:

the c.d.f of $h(x)$ is $3x^2 - 2x^3 =: H(x)$

$H^{-1}(x)$ on $(0, 1)$ is implemented ^{by} using a

binary search between $(0, 1)$ so that it

can be easily adapted to higher order polynomials

and obtain U from a uniform distribution on $(0, 1)$

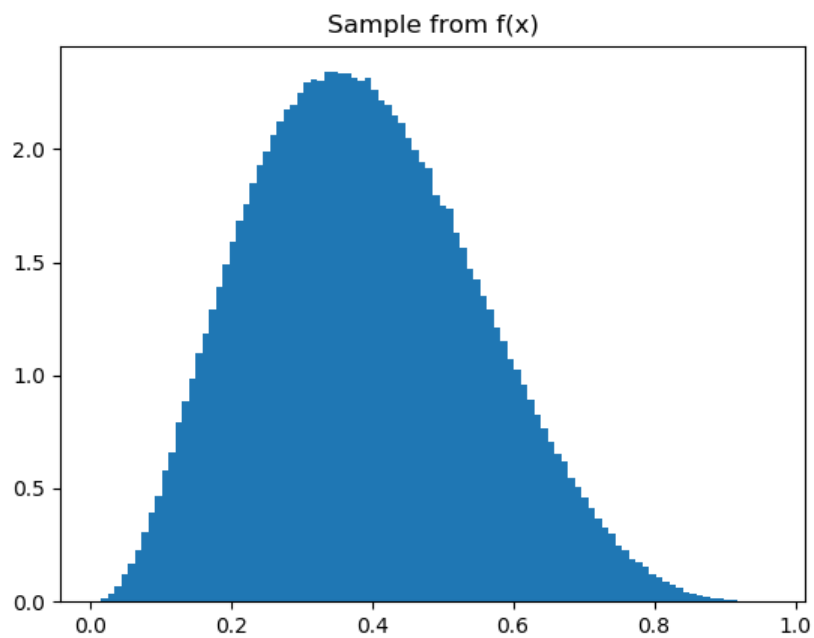
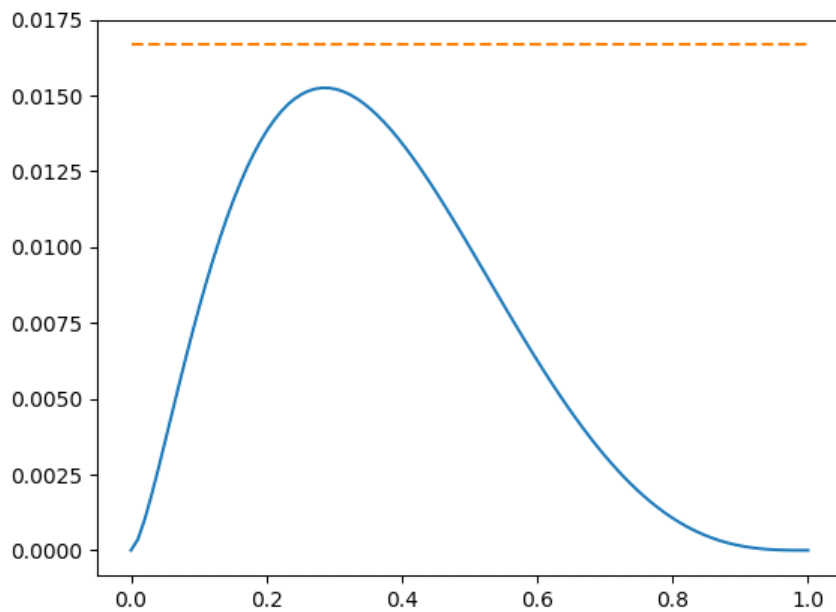
2. check whether $u < \frac{l(x)}{M}$

if this holds, accept X as a sample from $f(x)$

else: reject X and sample again.

SDS 386D HW1

Saturday, February 3, 2018 2:43 PM



Run the rejection sampling algorithm until a million samples are generated. The samples drawn are shown in the above histogram. We can see that it fits well with curve of $f(x)$.

Acceptance probability:

The numerical test shows that the acceptance probability is $\approx 1.03\%$.

To get the actual probability, we need to know the normalizing constant in this problem

$$c = \int_0^1 (-\log x)^2 x^3 (1-x)^2$$

This can be obtained by $\left\{ \begin{array}{l} \text{numerical integration} \\ \text{Monte Carlo integration} \\ \text{Analytically} \end{array} \right.$

$c = \frac{919}{108000}$ from the third method and a MC integration using samples from $f(x)$ confirms this result.

$\Rightarrow \Pr(u < \frac{\mathcal{L}(x)}{n}) = \frac{c}{n} \approx 0.5105$ which is consistent with the test result.

An implementation of adaptive rejection sampling is also tested: this implementation follows the method described in the lecture. The proposed ^{piecewise} probability enveloping the original probability is restricted to have 50 pieces at maximum.

SDS 386D HW1

Saturday, February 3, 2018 2:43 PM

Generate a million accepted samples as in rejection sampling. Only 15089 samples are rejected throughout the whole process and the histogram is similar.

SDS 386D HW1

Saturday, February 3, 2018 2:43 PM

#2. Use the sampling method from #1, a MC integration can be carried out on this problem:

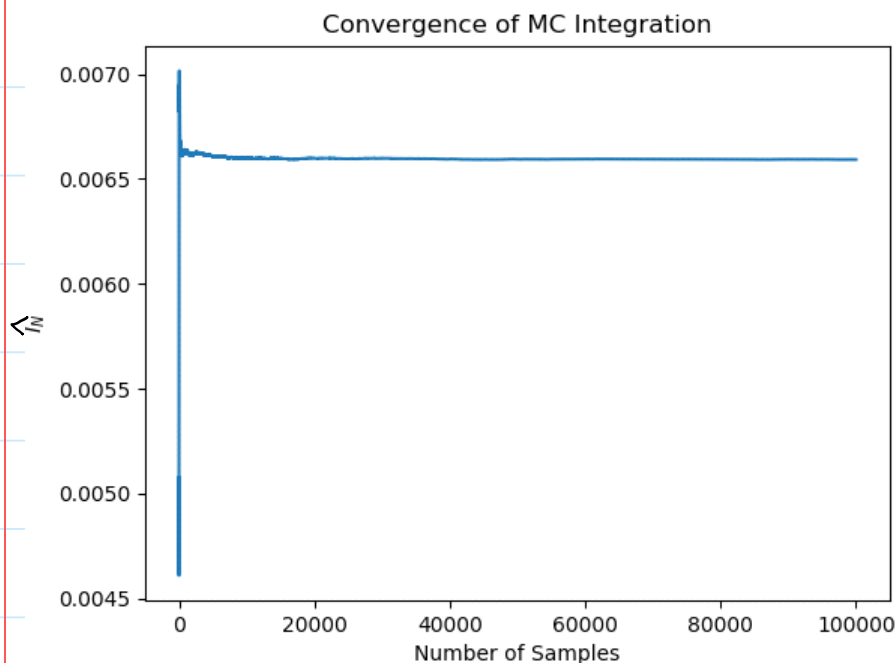
$$I = \int_0^1 c f(x) (1-x)^{1/2} dx.$$

$$\text{where } c = 919/108000$$

Take 10^5 samples from $f(x)$ and use

$$\hat{I}_N = \frac{1}{N} \sum_{i=1}^N c (1-x_i)^{1/2}$$

The numerical ~~est~~ gives 0.00639 which is close to the exact solution.



As shown in the convergence plot, the MC approx converges to the exact integration as $N \uparrow$

Variance:

From lecture, the variance of \hat{I}_N is

$$\text{Var}(\hat{I}_N) = \frac{1}{N} \text{Var}(g(X_1))$$

$$= \frac{1}{N} \cdot [E(g(X)^2) - E(g(X))^2]$$

$$= \frac{1}{N} \left[\int g^2(x) f(x) dx - \left(\int g(x) f(x) dx \right)^2 \right]$$

$$\approx \frac{1}{N} \left[\sum_{i=1}^N g^2(X_i) / N - \hat{I}_N^2 \right] \quad \text{where } X_i \text{ is from p.d.f } f(x).$$

Numerical test shows the $\text{Var}(\hat{I}_N)$ at $N=1000$ is around 8×10^{-7} (Using the same sample $\{X_i\}$ and $\{g(X_i)\}$)

SDS 386D HW1

Saturday, February 3, 2018 2:43 PM

#3. $l(x) = f(x) / h(x) = \pi(1+x^2) \cdot \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$

$$0 = l'(x) = \pi(1+x^2) \frac{1}{\sqrt{2\pi}} \cdot e^{-x^2/2} (-x) + \pi(2x) \frac{1}{\sqrt{2\pi}} \cdot e^{-x^2/2}$$

$$\Rightarrow (1+x^2)x = 2x$$

$x=0$ is a saddle point

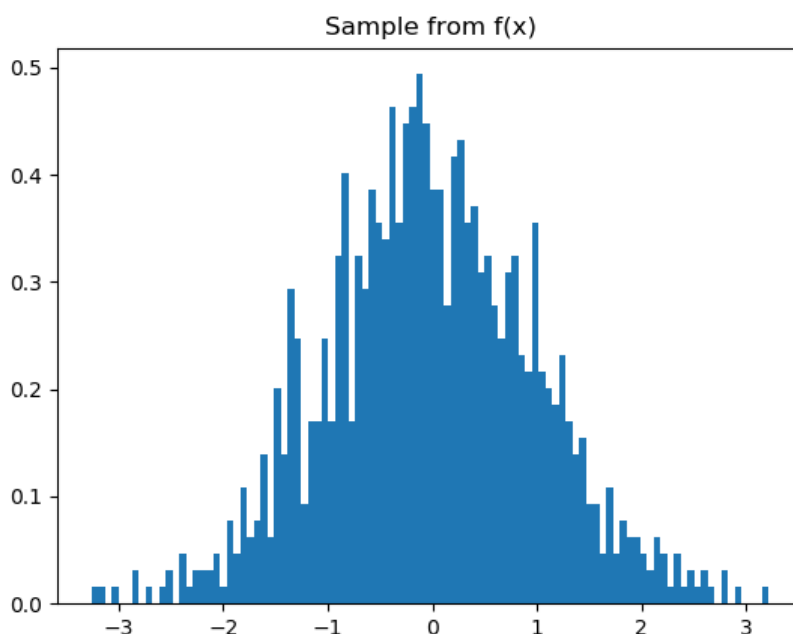
$$x \neq 0 : 1+x^2 = 2 \quad x = \pm 1$$

$$l(\pm 1) = \sqrt{\frac{2\pi}{e}} = \mu$$

$$Pr = \frac{1}{\mu} = \sqrt{\frac{e}{2\pi}} \approx 65.8\%$$

Get a 1000 sample from simulation ; the acceptance probability is above 64% ~ 68% , and the histogram is shown below. As the number of samples is increased,

the acceptance probability will converge to the exact value.



SDS 386D HW1

Saturday, February 3, 2018 2:43 PM