Simulation Exercise with Solution

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Solution Q1:

Monte-Carlo approximation for the integral

$$\int_0^\infty \frac{\theta^6 \exp^{-\theta^2 - \theta^3 - 2\theta}}{1 + 3\theta^2} d\theta$$

We aim to write this integral as $\int f(\theta)g(\theta)$ where $g(\theta)$ is a known density that is easy to simulate from. Then, we draw a large number of samples from $g(\theta)$, evaluate $f(\theta)$ for each one of those samples and take the mean, which will be the *Monte-Carlo* approximation for the above mentioned integral. Approach 1: We use Gamma(7,2) as $g(\theta)$.

Now,

$$g(\theta) = \frac{2^7}{\Gamma(7)} \exp^{-2\theta} \theta^{7-1}$$

Thus our $f(\theta)$ will become,

$$f(\theta) = \frac{\Gamma(7)}{2^7} \frac{\exp^{-\theta^2 - \theta^3}}{1 + 3\theta^2} d\theta$$

Approach 2: We use exp(2) as $g(\theta)$.

Now,

$$g(\theta) = 2e^{-2\theta}$$

Thus our $f(\theta)$ will become,

$$f(\theta) = \frac{1}{2} \frac{\theta^6 \exp^{-\theta^2 - \theta^3}}{1 + 3\theta^2} d\theta$$

Output:

Actual result: 0.00359857

Monte-Carlo approximation using Gamma(7, 2) PDF: 0.00362067 Monte-Carlo approximation using Exponential(2) PDF: 0.00361638

Solution Q2:

Monte-Carlo approximation for the integral

$$\int_{1.5}^{6} \frac{\log \theta}{2 + \theta^2} d\theta$$

As per procedure mentioned in *solution-1*, we use unif(1.5, 6) as $g(\theta)$. Now,

$$g(\theta) = \frac{1}{6 - 1.5} = \frac{1}{4.5}$$

Thus our $f(\theta)$ will become,

$$f(\theta) = 4.5 \frac{\log(\theta)}{2 + \theta^2}$$

Output:

Actual result: 0.3734025

Monte-Carlo approximation using Uniform(1.5, 6) PDF: 0.3737958

Solution 3:

Metropolis-Hastings Simulation to draw sample where target density is:

$$\theta \sim Gamma(2,4)$$

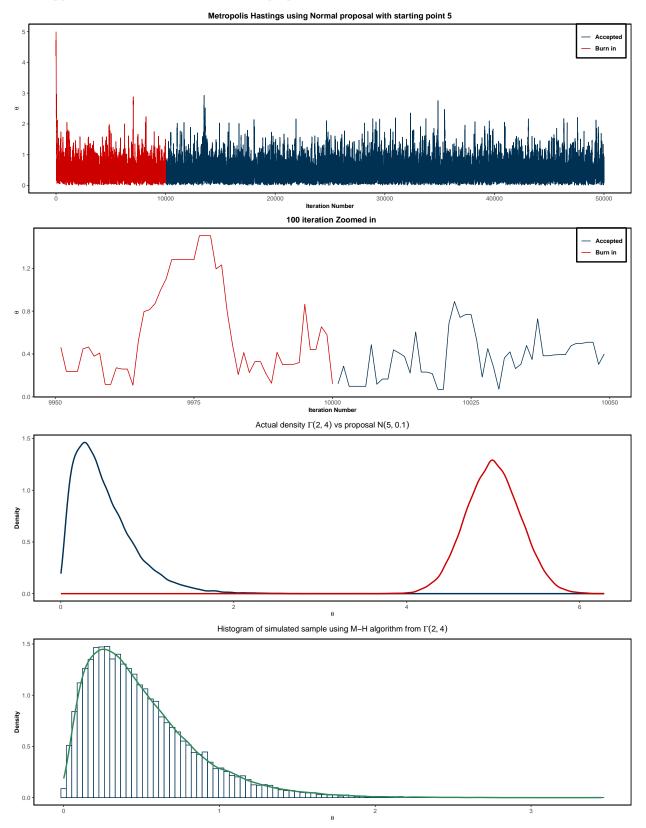
From target distribution True mean is 0.5

• Approach 1: Normal distribution as a proposal with variance = 0.1 and $\theta_0 = 5$

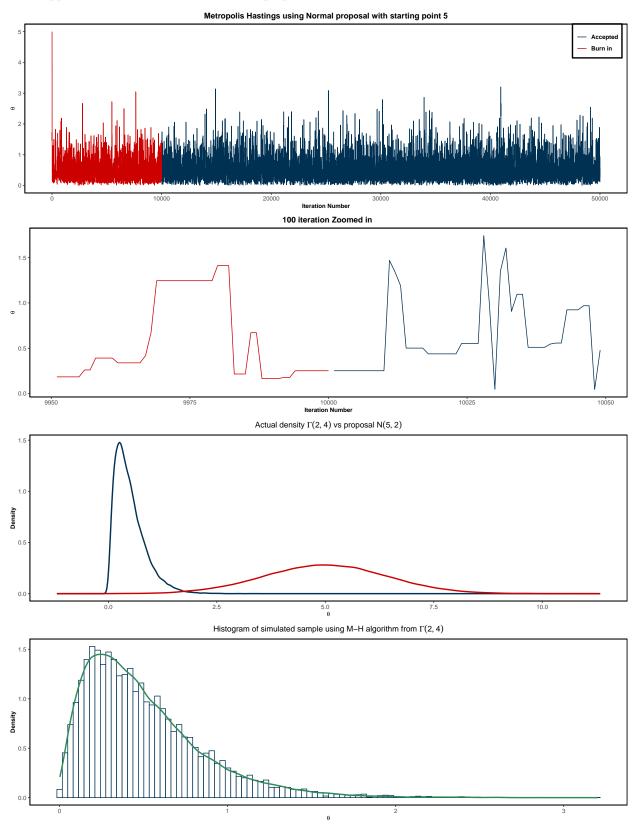
Acceptance Rate: 0.66786 Estimated Mean: 0.4978995

Difference from True Mean: 0.002100459

• Approach 1: Normal distribution as a proposal with variance = 0.1 and $\theta_0 = 5$



• Approach 2: Normal distribution as a proposal with variance = 2 and $\theta_0 = 5$



• Approach 2: Normal distribution as a proposal with variance = 2 and $\theta_0 = 5$

Acceptance Rate: 0.2449 Estimated Mean: 0.5086692

Difference from True Mean: 0.008669169

Solution 5:

Metropolis-Hastings Simulation to draw samples from density:

 $\theta \sim Gamma(2,4)$

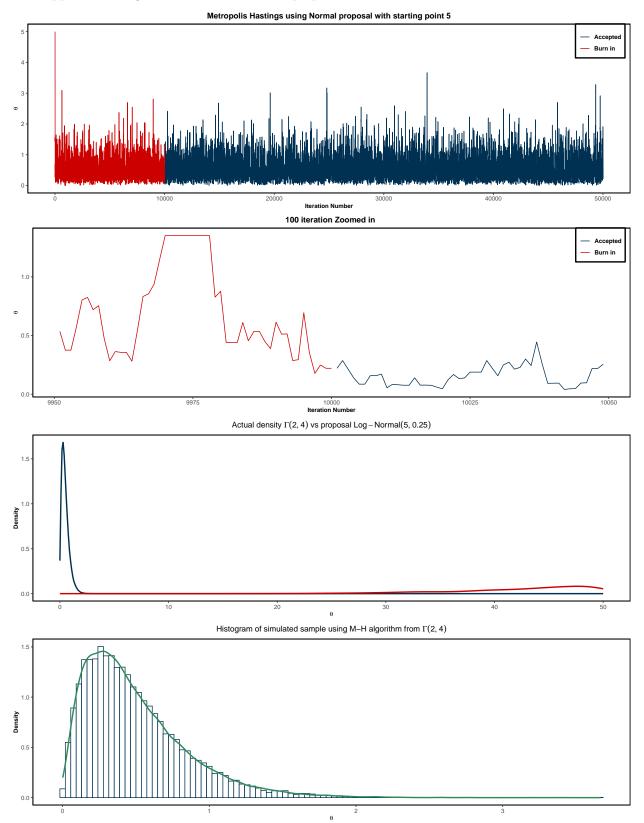
From target distribution True mean is 0.5

• Approach 1: Log-Normal distribution as a proposal with variance = 0.25 and $\theta_0 = 5$

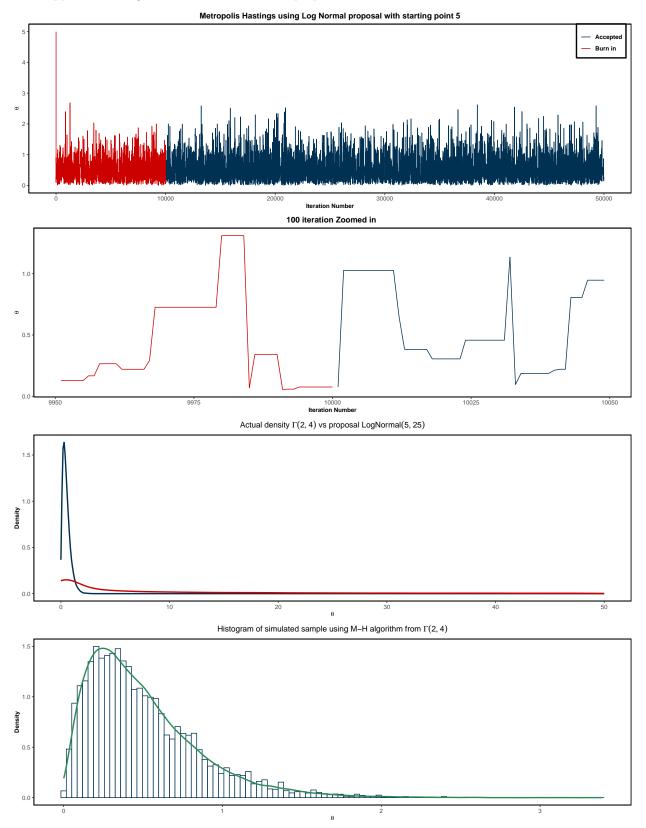
Acceptance Rate: 0.7917 Estimated Mean: 0.4979935

Difference from True Mean: 0.002006515

• Approach 1: Log-Normal distribution as a proposal with variance = 0.25 and $\theta_0=5$



• Approach 2: Log Normal distribution as a proposal with variance = 25 and $\theta_0=5$



• Approach 2: Log Normal distribution as a proposal with variance = 25 and $\theta_0=5$

Acceptance Rate: 0.1868
Estimated Mean: 0.5003575

Difference from True Mean: 0.0003574558

solution 6:

 $f(\theta) = c * \theta^3$

to be a valid PDF, c need to be $4\,$

ii) Simulation using Inverse CDDF method:

Since,

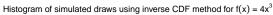
$$F_{\theta}(t) = t^4$$

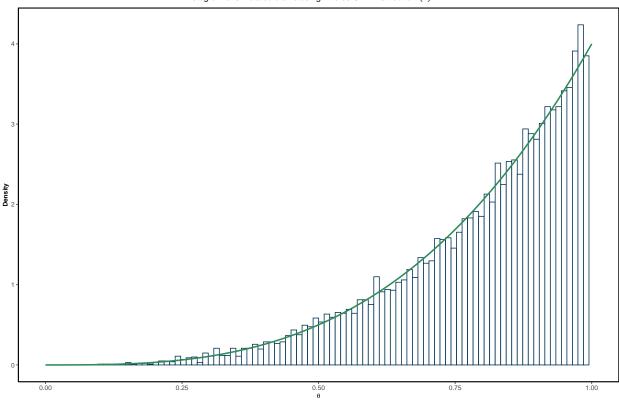
Thus,

$$\theta = F^{-1}(u) = u^{\frac{1}{4}}$$

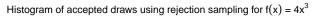
where,

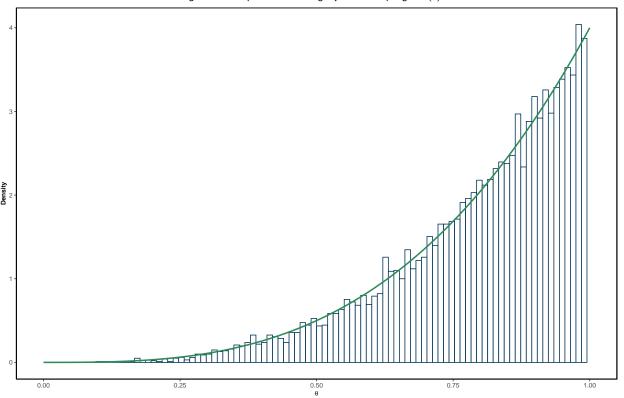
$$U \sim unif(0,1)$$





iii) Simulation usin Rejection sampling:





iv) Summary results from two sets of observations

• Using Inverse CDF:

Empirical Mean: 0.8002587 Variance: 0.02727211 60% Quantile: 0.8820321

• Using Rejection sampling:

Empirical Mean: 0.8020204 Variance: 0.02589842 60% Quantile: 0.8800872