

[https://dicook.github.io/Statistical\\_Thinking/tutorials/lab11/lab11help.pdf](https://dicook.github.io/Statistical_Thinking/tutorials/lab11/lab11help.pdf)

Remember SETU surveys!

## Question One - Monte Carlo Approximations

Your tasks:

1. Sample coordinates from the unit square.
2. Decide if that point is inside the unit circle.
3. Count the number of times the point is inside the circle for  $n \in \{100, 1000, 10000, 100000\}$ .
4. Calculate your approximation for  $\pi$ .

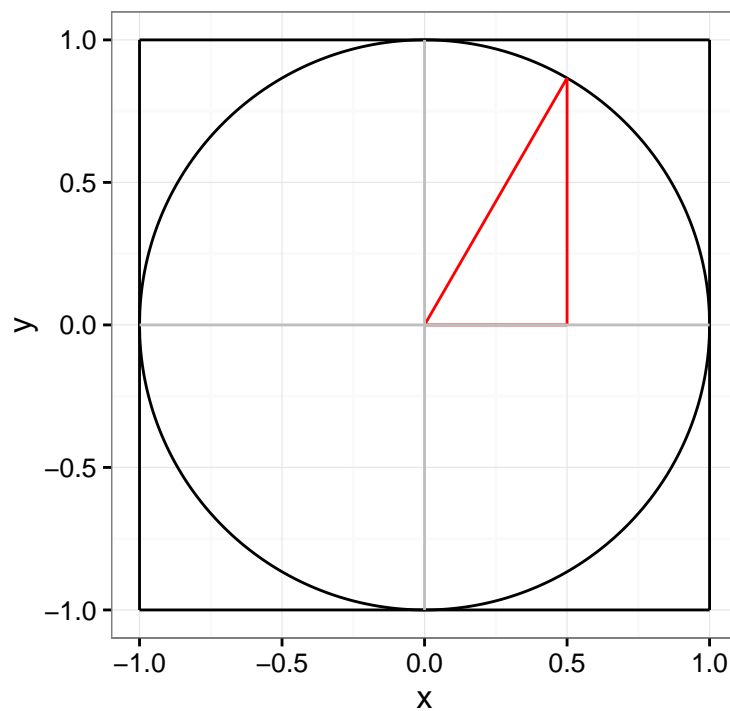


Figure 1: If we have even odds of picking any point in the square, the proportion of points in the circle should be the ratio of the two areas.

## Question Two - Accept/Reject sampling

Part A

1. Sample  $x^{(i)}$  from the proposal distribution  $g(x)$ .
2. Calculate  $\frac{f(x^{(i)})}{cg(x^{(i)})}$ .
3. Sample  $u$  from  $\mathcal{U}(0, 1)$ .
4. Decide to accept or reject  $x^{(i)}$  depending on if  $u < \frac{f(x^{(i)})}{cg(x^{(i)})}$ .

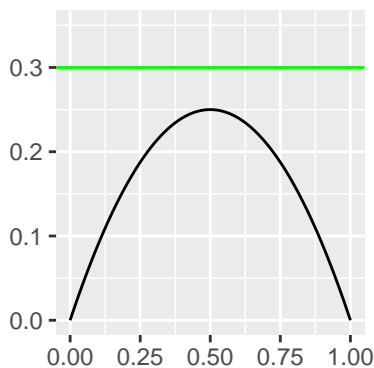


Figure 2: Let  $c = 0.3$ . If we sample a point  $x$  from a uniform distribution, it is accepted as something that could be from  $f(x)$  with probability  $f(x)/c$ , the ratio of the height of the two lines at  $x$ . Points where  $f(x)$  is high are more likely to be accepted, so the accepted sample resembles  $f(x)$ .

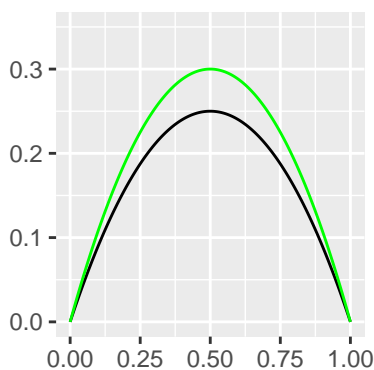


Figure 3:  $g(x)$  is now a  $\text{beta}(2, 2)$  distribution (with  $c = 0.2$ ). The acceptance ratio is higher close to 0 and 1, but now most of our draws will be close to the middle, so the final sample will still resemble  $f(x)$ .

### Part B

1. For each  $c$  run the function 1000 times. Try using `replicate(1000, function)`.
2. Count the number of rejections.
3. Calculate  $f(x)$  for a sequence of  $x$  between 0 and 1.
4. Plot the density of the accepted draws against a line plot of  $(x, f(x))$ .

## Question Three - Metropolis Hastings MCMC

1. Pick an arbitrary starting value that could come from a Cauchy, which is the same as a Students'  $t$  with one degree of freedom. You can get stuck near the starting value for a while but eventually the Markov chain will converge to the posterior so the starting value doesn't really matter.
2. Generate a candidate draw  $y$  from the proposal distribution.
3. Calculate  $r(x, y)$  where  $x$  is the previous draw.
4. Decide to keep the new draw or repeat  $x$ .
5. Calculate the rejection rate.
6. Plot `(1:n, draws)` as a line plot.
7. Plot the draws as a density plot against a Cauchy density.

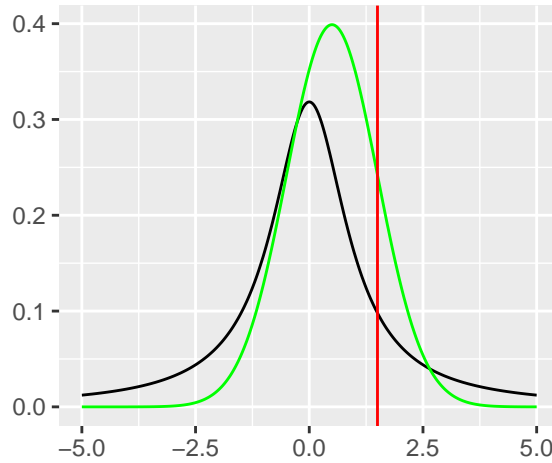


Figure 4: I start with  $x^{(0)} = 0.5$ . Then I sample  $y = 1.5$  from a  $\mathcal{N}(x^{(0)}, 1)$  distribution. Then I calculate  $r(x = 0.5, y = 1.5) = 0.38$  and randomly accept  $x^{(1)} = y$  with probability 0.38 or repeat  $x^{(1)} = x^{(0)}$  with probability  $1 - 0.38$ . Values  $y$  that are more likely to come from a Cauchy distribution than my current draw  $x^{(0)}$  are more likely to be accepted.

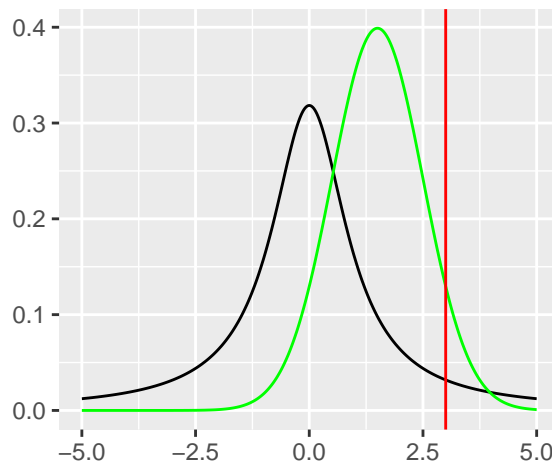


Figure 5: After accepting  $x^{(1)} = 1.5$  I sample  $y = 3$  from a  $\mathcal{N}(x^{(1)}, 1)$  distribution. Now my acceptance probability is 0.325 and it's pretty likely I'll reject  $y$  and set  $x^{(2)} = x^{(1)}$ . I might end up repeating this value until I draw a candidate value  $y$  that is more likely to come from a Cauchy distribution than my current 1.5.