Here is a Metropolis-Hastings sampler for a (non-normalized) probability distribution function f:

```
theta <- rep(0,1000)  # space to store results
t <- 0  # must pick a starting t with f(t) > 0
for (step in 1:1000) {
  t_proposed <- t + rnorm(1)
  acceptance <- f(t_proposed)/f(t)
  if (runif(1) < acceptance) { t <- t_proposed }
  theta[step] <- t
}
```

1. Let $f(x) = \begin{cases} 1 - x^2 & -1 \le x \le 1\\ 0 & \text{otherwise} \end{cases}$.

- (a) What normalization constant C would make Cf(x) into a PDF?
- (b) Set up and run M-H to sample θ distributed as f.
- (c) Make a density plot of θ and check that it matches the function f(x).
- (d) Make a trace plot of your simulated θ values.
- (e) Estimate the standard deviation of θ .
- 2. The M-H sampler you used for question 1 has a proposal distribution given by adding rnorm(1), a standard normal value with mean 0 and sd 1.
 - (a) Use a proposal SD of 10 (add rnorm(1,0,10)), so the proposed values are taking much larger jumps. Repeat 1b,c,d. Explain what happened.
 - (b) Use a proposal SD of 0.1, so the proposed values are taking much smaller jumps. Repeat 1b,c,d. Explain what happened.

3. Let
$$f(x) = \begin{cases} 1 & 0 \le x \le 1 \\ 3 & 10 \le x \le 11 \\ 0 & \text{otherwise} \end{cases}$$

- (a) In R, you can use dunif(x,0,1)+3*dunif(x,10,11) to implement f (why?) Run the M-H sampler for f.
- (b) How do the starting value of t and the proposal sd affect the results?