## Lecture 26: Importance Sampling

Ciaran Evans

#### Last time

$$\theta = \int_0^1 \frac{e^{-x}}{1 + x^2} dx$$

#### Monte Carlo integration: write

$$\theta = \int_{0}^{1} \frac{e^{-x}}{1 + x^{2}} dx = \int_{0}^{1} g(x) f(x) dx$$

where f is a density function.

- ► Sample  $X_1, ..., X_n \stackrel{iid}{\sim} f$
- ▶ Monte Carlo estimate:  $\hat{\theta} = \frac{1}{n} \sum_{i=1}^{n} g(X_i)$

#### Last time

$$\theta = \int_{0}^{1} \frac{e^{-x}}{1 + x^{2}} dx = \int_{0}^{1} g(x) f(x) dx$$

A couple different options:

$$f_1(x) = 1, \ g_1(x) = \frac{e^{-x}}{1 + x^2}$$

$$f_2(x) = \frac{4}{\pi(1+x^2)}, \ g_2(x) = \frac{\pi}{4}e^{-x}$$

**Question:** How would we sample from the distribution with density  $f_2$ ?

### Inverse transform method

pdf: 
$$f_2(x) = \frac{4}{\pi(1+x^2)}$$
 cdf:  $F_2(x) = \frac{4}{\pi} atan(x)$ 

To sample  $X_1, ..., X_n \stackrel{iid}{\sim} f_2$ :

▶ Sample 
$$U_1, ..., U_n \stackrel{iid}{\sim} Uniform(0,1)$$

$$X_i = F_2^{-1}(U_i)$$

$$= +an \left( \ddot{q} u_i \right)$$

Variance

```
g2 \leftarrow function(x) \{ pi/4 * exp(-x) \}
                                                  transformi
n <- 10; nsim <- 1000
                                             xi= tan ( " Wi)
theta_hat2 <- rep(NA, nsim)
                                          ô= 1 2 g2(Xi)
for(i in 1:nsim){
  x \leftarrow tan(runif(n) * pi/4)
  theta_hat2[i] <- mean(g2(x))</pre>
var(theta hat2)
## [1] 0.001882555
sd(theta hat2)
```

## [1] 0.04338842

### Comparing variance

$$\theta = \int_{0}^{1} \frac{e^{-x}}{1 + x^{2}} dx = \int_{0}^{1} g(x) f(x) dx$$

$$f_1(x) = 1$$
,  $g_1(x) = \frac{e^{-x}}{1 + x^2}$ :  $Var(\hat{\theta}_1) \approx 0.006$ 

• 
$$f_2(x) = \frac{4}{\pi(1+x^2)}$$
,  $g_2(x) = \frac{\pi}{4}e^{-x}$ :  $Var(\widehat{\theta}_2) \approx 0.002$ 

Relative efficiency:  $\frac{Var(\widehat{\theta}_2)}{Var(\widehat{\theta}_1)} \approx \frac{1}{3}$ 

Reduction in variance:  $100 \cdot \frac{Var(\hat{\theta}_1) - Var(\hat{\theta}_2)}{Var(\hat{\theta}_1)} \approx 68\%$  reduction in variance

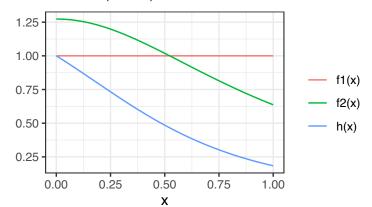
### Comparing options

$$\theta = \int_{0}^{1} h(x)dx = \int_{0}^{1} g(x)f(x)dx$$
  $h(x) = \frac{e^{-x}}{1+x^{2}}$ 

• 
$$f_1(x) = 1$$
,  $g_1(x) = \frac{e}{1 + x^2}$ :  $Var(\hat{\theta}_1) \approx 0.00e$ 

$$f_1(x) = 1, \ g_1(x) = \frac{e^{-x}}{1+x^2} : \ Var(\widehat{\theta}_1) \approx 0.006$$

$$f_2(x) = \frac{4}{\pi(1+x^2)}, \ g_2(x) = \frac{\pi}{4}e^{-x} : \ Var(\widehat{\theta}_2) \approx 0.002$$



### Importance sampling

For any density function f supported on the domain of integration  $\mathcal{X}$ :

$$\theta = \int_{\mathcal{X}} h(x)dx = \int_{\mathcal{X}} \frac{h(x)}{f(x)} f(x)dx$$

$$\downarrow f \qquad \downarrow f$$

**Question:** What choice of *f* would **minimize** this variance?

# Importance sampling

$$\theta = \int_{\mathcal{X}} h(x) dx = \int_{\mathcal{X}} \frac{h(x)}{f(x)} f(x) dx$$

- ▶  $Var(\widehat{\theta})$  is smaller if f(x) is "similar to" h(x)
- ▶ If  $f(x) = c \cdot h(x)$  for all x and some constant c, then

$$Var\left(\frac{h(X)}{f(X)}\right) = Var\left(\frac{1}{c}\right) = 0$$

- ►  $Var(\hat{\theta})$  is minimized if  $f(x) = \frac{|h(x)|}{\int |h(t)| dt}$ But, if we can't into
- possible to use this f(x)In practice: just try to get f "close to" h

#### Your turn

Experiment with using different densities for Monte Carlo integration with another integral:

https://sta379-s25.github.io/practice\_questions/pq\_26.html

- Start in class
- You are welcome and encouraged to work with your neighbors
- Solutions posted on course website