# Acceleration methods: Reducing variance

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PSTAT 194CS

### **Acceleration methods**

Clever methods to reduce the variance of a Monte Carlo estimate of an integral.

- Antithetic variables
- Control variates

#### **Antithetic variables**

This is another variance reduction method. The idea is that it may be helpful to generate samples of correlated variables in estimating integrals

Recall, covariance of X and Y is

$$Cov(X,Y) = E((X - EX)(Y - EY)) = E(XY) - (EX)(EY)$$

Suppose  $X_1$  and  $X_2$  are identically distributed then

$$Var(\frac{X_1 + X_2}{2}) = \frac{1}{4}(Var(X_1) + Var(X_2) + 2Cov(X_1, X_2))$$

So, variance is reduced if  $X_1$  and  $X_2$  are negatively correlated.

## Antithetic Variables approach

Given two samples  $X_1, X_2, \ldots, X_m \sim f$  and  $Y_1, Y_2, \ldots, Y_m \sim f$  for estimating

$$I = \int h(x)f(x)dx$$

where  $X_i$  and  $Y_i$  are **negatively correlated** and h is a monotone(increasing or decreasing) function.

Define an antithetic pair

$$\frac{h(X_i)+h(Y_i)}{2}$$

then the antithetic estimator

$$\hat{I} = \frac{1}{m} \sum_{i=1}^{m} \frac{h(X_i) + h(Y_i)}{2}$$

is more efficient than the estimate based only on X's and Y's.

### **Question:**

How to get negatively correlated  $X_i$  and  $Y_i$ 

- Using inversion
  - $U \sim \text{Unif}(0,1)$
  - U and 1-U are negatively correlated
  - So,  $X = F^{-1}(U)$  and  $Y = F^{-1}(1 U)$  are negatively correlated.
- Use symmetry
  - $X \sim f(x)$  which is symmetric about  $\mu$
  - $X \mu$  has the same distribution as  $\mu X$
  - So X has the same distribution as  $2\mu X$

# **Example: Want to compute**

$$\int_0^\infty x^2 e^{-x} dx$$

 $g(x) = x^2$  is monotone on 0 to  $\infty$ , so could use antithetic approach.

- Be careful,  $g(x) = x^2$  is not monotone on  $(-\infty, \infty)$ .
- 1. Boring way
  - Generate  $X_1, X_2, \ldots, X_m \sim \operatorname{Exp}(1)$

$$\hat{I} = \frac{1}{m} \sum_{i=1}^{m} X_i^2$$

- 2. Better way
  - Generate  $U_1, U_2, \ldots, U_m \sim \text{Unif}(0, 1)$
  - Let  $X_i = -\ln(U_i)$  and  $Y_i = -\ln(1-U_i)$

$$\hat{I} = \frac{1}{m} \sum_{i=1}^{m} \frac{X_i^2 + Y_i^2}{2}$$

### **Activity:**

For the above example, compare antithetic variable approach with simple monte-carlo. What is the reduction in variance?

#### **Control Variates**

Another variance reduction technique.

**Problem:** We are looking to estimate *EY*.

The idea here is that you have a bi-variate random variable (C, Y) and a simulated sample of size m:

$$(c_1, y_1), (c_2, y_2), \cdots, (c_m, y_m)$$

Suppose we know *EC*. We can use this to speed up convergence of *EY*.

When C and Y are strongly positively correlated then variance is improved over the naive estimator  $\bar{y}$ . C is called the control variable.

Notation:  $EC = \mu_c$  is known in advance.  $\sigma_C^2$  is the variance of C,  $\sigma_Y^2$  is the variance of Y and  $\rho_{cY}$  is the correlation between C and Y