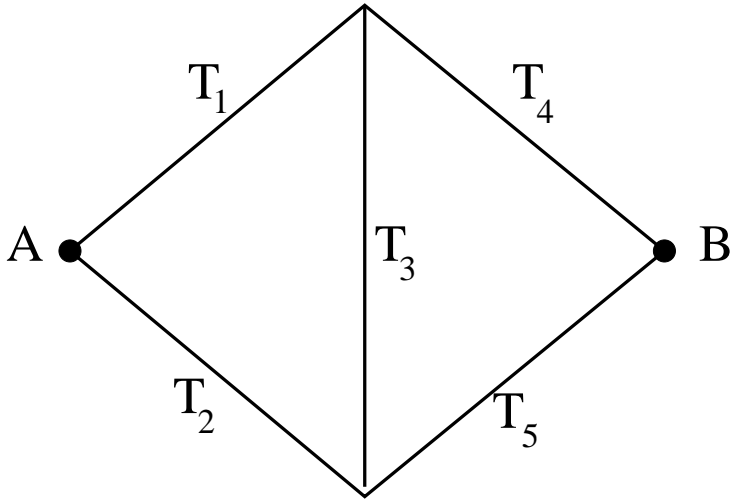


## Monte Carlo Methods - Spring 16 - Homework 1

1. This is based on an example in the review article by Kroese. Consider the small network shown in the figure.



The random variables  $T_i$  are the times it takes to traverse that bond. The times  $T_i$  are uniformly distributed but with different ranges:

$T_1$  uniform on  $[0, 1]$

$T_2$  uniform on  $[0, 2]$

$T_3$  uniform on  $[0, 3]$

$T_4$  uniform on  $[0, 1]$

$T_5$  uniform on  $[0, 2]$

We want to find the quickest path from A to B. Let  $X$  be the total time it takes to traverse this quickest path. We want to compute the mean  $\mu = E[X]$ .

(a) Using  $N = 10,000$  samples, use direct Monte Carlo to estimate  $\mu$ . Using your data estimate  $\sigma^2$  and find a 95% confidence interval for your estimate.

As a check on your simulation, the exact answer is known to be  $\frac{1339}{1440} \approx 0.92986$ .

(b) Repeat what you did in (a) 100 times and see how many times your confidence interval contains the exact value of  $\mu$ .

2. (corrected 2/9) In class and in the notes I claimed that if you apply the delta method to find the variance of the estimator  $\hat{\theta} = \bar{Y}/\bar{X}$  for  $\theta = E[Y]/E[X]$ , then the variance of  $\hat{\theta}$  is approximately

$$\text{var}(\hat{\theta}) \approx \frac{1}{n} \frac{\sum_{i=1}^n (Y_i - \hat{\theta}X_i)^2}{n\bar{X}^2}$$

where  $\bar{X}$  and  $\bar{Y}$  are the sample means for  $X$  and  $Y$  and  $\hat{\theta} = \bar{Y}/\bar{X}$ . For the general estimator  $\hat{\theta} = f(\hat{\mu}_1, \dots, \hat{\mu}_d)$  we found from the delta method that the variance is  $(\hat{\nabla}f, \hat{\Sigma}\hat{\nabla}f)/n$ . Use this general result to derive the equation above, i.e., show

$$(\hat{\nabla}f, \hat{\Sigma}\hat{\nabla}f) = \frac{\sum_{i=1}^n (Y_i - \hat{\theta}X_i)^2}{n\bar{X}^2}$$

3. Let  $X_1, \dots, X_n$  be our sample. Define

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i, \quad s^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

The kurtosis of  $X$  is

$$\kappa = \frac{E[(X - \mu)^4]}{\sigma^4}$$

For a normal RV the kurtosis is 3. So  $\kappa - 3$  is sometimes called the excess kurtosis. Unfortunately sometimes the excess kurtosis is just called the kurtosis. In particular Owen does this.

(a) Find the variance of  $s^2$  in terms of  $\sigma^2$ ,  $\kappa$  and  $n$ . I think the answer is

$$\frac{1}{n} \left( \kappa \sigma^4 - \frac{n-3}{n-1} \sigma^4 \right)$$

This is just a computation and rather tedious. You may want to skip this part and just do (b).

(b) Define

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

This is just  $s^2$  with the  $1/(n-1)$  replaced by  $1/n$ . Expanding out the square we can also write this as

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X}^2$$

Now let  $Y = X^2$ . The variance of  $X$  is  $E[Y] - E[X]^2$  which is  $f(E[X], E[Y])$  where  $f(x, y) = y - x^2$ . The estimator  $\hat{\sigma}^2$  is  $\bar{Y} - \bar{X}^2 = f(\bar{X}, \bar{Y})$ . So we can use the delta method to find the variance of  $\hat{\sigma}^2$ . Carry out this computation and compare with the result in part (a).