# Network Analysis and Simulation - AY 2020/2021 Homework 4

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## Exercise 1

We want to study the delay in function of the utilization factor of an M/M/1 queue with Poisson arrivals with rate  $\lambda$  and service time of one or two time units with same probability, i.e.,  $\mathbb{P}[\text{service time 1 time unit}] = \mathbb{P}[\text{service time 2 time units}] = a = 0.5$ . Hence, the mean service is  $\mu = 0.5 \cdot 1 + 0.5 \cdot 2 = 1.5$ , which means that the utilization factor is  $\rho = \lambda/\mu = 3\lambda/2$ . Now, if we let vary  $\lambda$  in the interval [0.1, 1.5] we will have  $\rho$  varying in the interval (0, 1]. In Fig. 1 is depicted the average delay, expressed in minuted, in function of  $\rho$ .

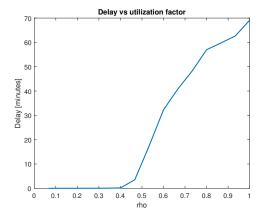


Figure 1: Delay vs  $\rho$ 

### Exercise 2

If we have  $X_1, X_2$  iid random variables, then

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

The key idea of the antithetic variables method is that, if we drop the assumption of  $X_1, X_2$  being independent, and instead we try to make them negatively correlated, then

$$Cov(X_1, X_2) \le 0 \tag{2}$$

and the variance of the estimator  $\frac{X_1+X_2}{2}$  is reduced. Now, suppose that we want to estimate

$$\theta = \mathbb{E}[\mathbf{U}] = \int_0^1 e^x dx, \quad with \ \mathbf{U} \sim \mathcal{U}[0, 1]$$
 (3)

With Monte Carlo simulation, using the classic method, we do

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Algorithm 1: Raw estimator
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i = 0;
n = 10000;
initialize empty array a;
while i \le n do
 draw u_1 \text{ and } u_2 \text{ from } \mathcal{U}[0, 1] \text{ independently;}
compute \frac{e^{u^1} + e^{u^2}}{2} \text{ and append the result to array a;}
i = i + 1;
end
```

Instead, with the antithetic variables method, we do

#### Algorithm 2: Antithetic variables

```
\begin{array}{l} \mathbf{i} = 0; \\ \mathbf{n} = 10000; \\ \mathbf{initialize} \ \mathbf{empty} \ \mathbf{array} \ \mathbf{b}; \\ \mathbf{while} \ i \leq n \ \mathbf{do} \\ & \ \  \  \, \mathbf{draw} \ \mathbf{u} \ \mathbf{from} \ \mathcal{U}[0,1]; \\ & \ \  \, \mathbf{compute} \ \frac{e^u + e^{1-u}}{2} \ \mathbf{and} \ \mathbf{append} \ \mathbf{the} \ \mathbf{result} \ \mathbf{to} \ \mathbf{array} \ \mathbf{b}; \\ & \ \  \, \mathbf{i} = \mathbf{i} + 1; \\ \mathbf{end} \end{array}
```

In both cases, we obtained as result  $\theta = e - 1$ , as expected. But the sample variance of the raw estimator is  $V_1 = 0.1212$ , while the sample variance of the estimator taking advantage of antithetic variables is  $V_2 = 0.0039$ . Hence, we experience a variance reduction of

$$\Delta_V = \frac{V_1 - V_2}{V_1} = 0.9674 = 96.74\% \tag{4}$$

This result is in compliance with the theory results in Example 8d in book [1, p. 144].

#### References

[1] Sheldon M. Ross. Simulation, Fourth Edition. Academic Press, Inc., USA, 2006.