

# Take-home quiz 5

15-468/668/868, Physics-based Rendering, Spring 2021  
Carnegie Mellon University

Due Wednesday, Mar. 24, at 11:59 pm ET

## Question 1

We discussed in class the *multi-sample* version of multiple importance sampling. It is worth exploring the mathematical properties of this procedure in more detail. Let's assume that we want to estimate an integral of the form:

$$F = \int_D f(x) \, d\mu(x). \quad (1)$$

To this end, we are given  $T$  sampling distributions, each with a probability density function  $p_t(x), x \in D, t = 1, \dots, T$ . We assume that each of these sampling distributions can be used to form an unbiased Monte Carlo estimate of the target integral of Equation (1). We also assume that we have a total sample budget of  $N$ .

Instead of using just one of these distributions, in multi-sample multiple importance sampling we proceed as follows: In pre-processing, we allocate  $n_t$  samples to each sampling distribution  $t$ , so that  $\sum_{t=1}^T n_t = N$ . At runtime, first, we independently draw  $n_t$  samples  $X_{t,n}, n = 1, \dots, n_t$ , from each sampling distribution  $t$ , for a total of  $N$  independent samples. Second, we form an estimate of the target integral as:

$$\langle F \rangle = \sum_{t=1}^T \frac{1}{n_t} \sum_{n=1}^{n_t} \frac{w_t(X_{t,n}) f(X_{t,n})}{p_t(X_{t,n})}, \quad (2)$$

where we get to select the weights  $w_t(X)$  for each  $t = 1, \dots, T$  and  $X \in D$ .

1. Prove that the estimator of Equation (2) is unbiased if  $\sum_{t=1}^T w_t(X) = 1$ , for all  $X \in D$ .
2. Prove that the variance of the estimator of Equation (2) equals

$$\text{Var}(\langle F \rangle) = \underbrace{\sum_{t=1}^T \frac{1}{n_t^2} \sum_{n=1}^{n_t} \mathbb{E}[F_{n,t}^2]}_{\equiv \mathcal{S}} - \sum_{t=1}^T \frac{1}{n_t^2} \sum_{n=1}^{n_t} \mathbb{E}[F_{n,t}]^2, \quad (3)$$

where  $\mathbb{E}[\cdot]$  is expectation and we have defined

$$F_{n,t} \equiv \frac{w_t(X_{t,n}) f(X_{t,n})}{p_t(X_{t,n})}. \quad (4)$$

3. Prove that the term  $\mathcal{S}$  in Equation (3) is minimized by setting the weights  $w_t$  for every  $X \in D$  according to the *balance heuristic*:

$$w_t(X) = \frac{c_t p_t(X)}{\sum_{t=1}^T c_t p_t(X)}, \quad (5)$$

where we have defined  $c_t \equiv \frac{n_t}{N}$ ,  $t = 1, \dots, T$ . To prove this, you should first express  $\mathcal{S}$  as an integral over  $D$ ; then use the method of Lagrange multipliers to minimize the integrand this integral for each  $X \in D$ , subject to the constraint  $\sum_{t=1}^T w_t(X) = 1$ .

## Deliverables

As described on the course website, solutions are submitted through Canvas.

- **Write-up:** Your write-up should be typeset in L<sup>A</sup>T<sub>E</sub>X, and should consist of your answers to the quiz questions. **Please note that we do not accept handwritten scans for your write-up.**
- **Submission:** Your submission should be a PDF file, <andrew-id.pdf>, with your write-up. **Please do not submit ZIP files.**

## Credits

Inspiration for the questions in this quiz came from Veach and Guibas [1].

## References

- [1] E. Veach and L. J. Guibas. Optimally combining sampling techniques for monte carlo rendering. In *Proceedings of the 22nd annual conference on Computer graphics and interactive techniques*, pages 419–428, 1995.