

# 1 Abstract

You will learn

- taking definite integral by ordinary Monte Carlo (OMC)
- exact sampling with python provided random number generators

# 2 Problem

**Example 1** *Our goal is to compute, using OMC by exact sampling*

$$\alpha = \int_0^1 h(x)dx$$

where

$$h(x) = 100 \cdot I_{(0,1/100]}(x) + 1 \cdot I_{(1/100,1)}(x).$$

The exact value shall be

$$\alpha = 1.99.$$

# 3 Analysis

## 3.1 OMC by exact sampling

The objective is to

- estimate

$$\alpha = \mathbb{E}[X], \quad X \sim p(x)$$

one can use random number generator by computer (if possible)

$$\{\text{iid } X_i \sim p(x) : i = 1, 2, \dots, n, \}.$$

Then, one can compute the approximation of  $\alpha$  by

$$\hat{\alpha}_n = \frac{1}{n} \sum_{i=1}^n X_i.$$

We say  $\hat{\alpha}_n$  as OMC by *exact sampling*, since the sample  $X_i$  produced by random generator has the same distribution as true distribution  $X$ , i.e.

$$X_i \sim X, \quad \forall i.$$

**Example 2** *Prove the properties of the OMC by exact sampling below:*

- $X_1$  itself can be treated as an unbiased MC, because

$$\mathbb{E}[X_1] = \alpha.$$

However, MSE is big, ie.

$$MSE(X_1) = Var(X) = \int x^2 p(x) dx.$$

- $\hat{\alpha}_n$  is consistent almost surely due to LLN, i.e.

$$\hat{\alpha}_n \rightarrow \alpha, \text{ almost surely as } n \rightarrow \infty.$$

Moreover,  $\hat{\alpha}_n$  is unbiased too, and

$$MSE(\hat{\alpha}_n) = Var(\hat{\alpha}_n) = \frac{1}{n} Var(X) \rightarrow 0.$$

### 3.2 Evaluation of integral

Back to our Example 1, we write

$$\alpha = \mathbb{E}[X] = \mathbb{E}[h(Y)],$$

where  $X = h(Y)$  and  $Y \sim U(0, 1)$ . In other words, although  $X$ -sampling is not directly available in python, one can use  $U(0, 1)$  random generator (see `numpy.random.uniform`) to produce  $Y_i$ , then compute  $h(Y_i)$  for the sample  $X_i$ .

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#### Algorithm 1 Integral by MC - Example 1

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<pre> 1: <b>procedure</b> MCINTEGRAL(<math>N</math>) 2:   <math>s \leftarrow 0</math> 3:   <b>for</b> <math>i = 1 \dots N</math> <b>do</b> 4:     generate two numbers <math>Y</math> from <math>U(0, 1)</math> 5:     <math>s \leftarrow s + h(Y)</math> 6:   <b>return</b> <math>\frac{s}{N}</math> </pre>	<pre> ▷ <math>N</math> is total number of samples ▷ <math>s</math> is the sum of samples ▷ use <code>numpy.random.uniform</code> ▷ return the average </pre>
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