# Monte Carlo approximation

### An intoduction for practical use in R

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#### Aim

Students will be able to

- apply Monte Carlo approximation with R
- approximate integrals, expected values, modes, quantiles, etc... with R

### Briefly ...

*Monte Carlo approximation* is a stochastic procedure for the evaluation of intractable quantities. It involves:

- 1. properly generating a large random sample from a suitable distribution, and
- 2. computing a statistic asymptotically consistent to the intractable quantity of interest.

This statistic is the Monte Carlo approximate of the intractable quantity.

## Approximating expected values, and integrals...

#### Notation

Assume a random variable  $\theta \in \Theta$  following a distribution  $\Pi(\cdot)$ .

Let  $h(\cdot): \Theta \to \mathbb{R}^d$ ,  $d \ge 1$ , be a measurable function.

Assume that there is interest in approximating the expected value

$$E_{\pi}(h(\theta)) = \int h(\theta) d\Pi(\theta) = \begin{cases} \int h(\theta)\pi(\theta) d\theta, \text{ cont.} \\ \sum h(\theta)\pi(\theta), \text{ discr.} \end{cases}$$

### Monte Carlo approximation procedure:

1. Draw a random sample

$$\theta^{(j)} \sim \Pi(\cdot), \text{ for } j = 1, ..., N$$

2. Compute

$$\bar{h}_N = \frac{1}{N} \sum_{j=1}^{N} h(\theta^{(j)})$$

We say that  $\bar{h}_N$  is a Monte Carlo approximate of  $\mathbf{E}_{\pi}(h(\theta))$ , where

$$E_{\pi}(h(\theta)) \approx \bar{h}_N$$
 when N is large enough.

which can be justified by aw of Large Number (LLN) arguments:  $\bar{h}_N \to E_{\pi}(h(\theta))$  as  $N \to \infty$ .

### Example

Consider the integral

$$I = \int_0^\infty x^5 \exp(-3x) \mathrm{d}x$$

Compute a Monte Carlo approximate of integral I.

The Exact value is I = 0.1646091.

#### Solution

It is

$$\begin{split} I &= \int_0^\infty x^5 \exp(-3x) \mathrm{d}x &= \int x^5 \frac{1}{3} 3 \exp(-3x) \mathbf{1}(x \in (0, \infty)) \mathrm{d}x \\ &= \frac{1}{3} \int x^5 \underbrace{3 \exp(-3x) \mathbf{1}(x \in (0, \infty)) \mathrm{d}x}_{=\mathrm{d}\Pi_{\mathrm{Exp}(3)}(x)} &= \frac{1}{3} \int x^5 \mathrm{d}\Pi_{\mathrm{Exp}(3)}(x) = \frac{1}{3} \mathrm{E}_{\mathrm{Exp}(3)}(x^5) \\ &\approx \frac{1}{3} \frac{1}{N} \sum_{j=1}^N \left( x^{(j)} \right)^5, \quad \text{where} \quad x^{(j)} \sim \mathrm{Exp}(3) \end{split}$$

we compute the Monte Carlo approximate for N = 1000.

```
x = rexp(n = 10^5, rate = 3)
I_mc = (1/3)*mean(x^5)
print(I_mc)
```

## [1] 0.1674372

# Approximating other quantities...

Monte Carlo approximation can approximate quantities other than expected values, and the theoretical justification is out of the scope.

Let  $\theta \in \Theta$  be a random variable with distribution  $\Pi(\cdot)$ .

• Assume there is a random sample generated:

$$\theta^{(j)} \sim \Pi(\cdot)$$
, for  $j = 1, ..., N$ 

for j = 1, ..., N

• The characteristic of the distribution  $\Pi(\cdot)$  such as population mode, quantiles, etc... can be approximated by their sample analogues such as sample mode, sample quantiles, etc...