

# MONTE CARLO METHODS

## Worksheet 11: Convergence test and quasi-Monte Carlo quadrature

**Exercise 23** (Rayleigh–Ritz principle). Let  $\mathcal{H}$  be a finite-dimensional Hilbert space with inner product  $\langle \cdot, \cdot \rangle$ . Let  $A : \mathcal{H} \rightarrow \mathcal{H}$  be a linear selfadjoint operator. Deduce from the spectral theorem that

$$\min \sigma(H) = \min \left\{ \frac{\langle x, Hv \rangle}{\langle v, v \rangle} : v \in \mathcal{H} \setminus \{0\} \right\}.$$

**Exercise 24** (Gelman–Rubin test). Assume that each of the  $m$  Markov chains satisfies the strong law of large numbers. Prove that  $\hat{V}_n$  is a strongly consistent estimator for  $\mathbb{V}_\mu(f)$ :

$$\lim_{n \rightarrow \infty} \hat{V}_n = \mathbb{V}_\mu(f) \quad \text{a.e.}$$

**Exercise 25** (Quasi-Monte Carlo quadrature). Consider the standard normal distribution  $\mu$  on  $\mathbb{R}^{10}$  and its mean

$$a = \int_{\mathbb{R}^d} x d\mu(x) = 0.$$

- Compute  $a$  by direct simulation using up to  $n = 10^4$  i.i.d. samples.
- Compute  $a$  by quasi-Monte Carlo quadrature using up to  $n = 10^4$  Halton points.
- Compute  $a$  by quasi-Monte Carlo quadrature using up to  $n = 10^4$  Sobol points.

Plot the error of all three simulations as a function of  $n$  and compare with the theoretically expected rate.