

Low-discrepancy sampling

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Laboratory Class Scientific Computing (WISM454)

Recap of last week

► Quadrature methods

- Trapezoidal rule
- Error depends on $\frac{1}{k^2} \max_x |f''(x)|$
- $n = k^d$ points in d dimensions
- To achieve error ϵ we need

$$n \propto \left(\frac{1}{\epsilon}\right)^{d/2}$$

► Monte Carlo methods

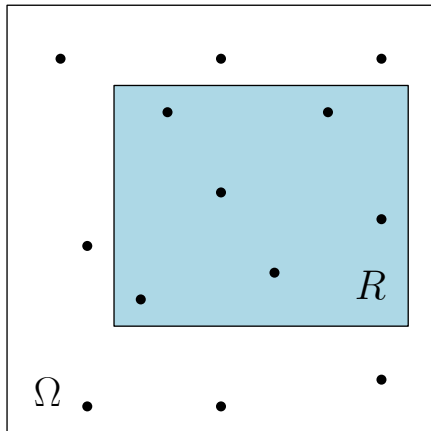
- Hit-or-miss
- Simple sampling
- To achieve error ϵ we need

$$n \propto \left(\frac{1}{\epsilon}\right)^2$$

Discrepancy

- ▶ With Monte Carlo methods, sample points are selected randomly, is this optimal?
- ▶ Intuitively, the *discrepancy* of a sequence is a measure of the gaps that a sequence leaves
- ▶ *Sampling for low discrepancy* is the subject of today

Discrepancy



- We estimate the area of R by hit-or-miss sampling with sequence of points

Discrepancy definition

- ▶ Let $\Omega = [0, 1]^d$. For some sampling sequence $\{\vec{x}_j\}$, what is the largest error in estimating rectangular volumes?
- ▶ $R = [a_1, b_1] \times \dots \times [a_d, b_d]$, volume is

$$V(R) = \prod_{i=1}^d (b_i - a_i).$$

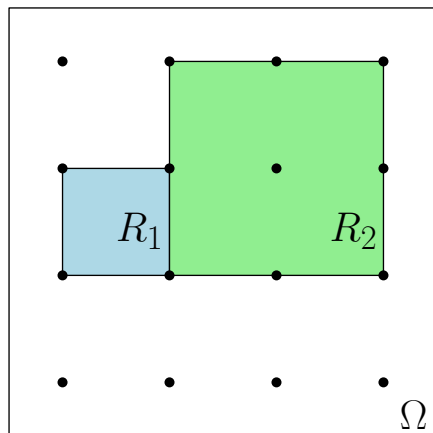
- ▶ Simple sampling with first n elements of the sequence gives:

$$\tilde{V}_n(R) = |\{j \leq n \mid \vec{x}_j \in R\}|.$$

- ▶ Discrepancy D defined as

$$D_n = \sup_{\text{rectangles } R} |\tilde{V}_n(R) - V(R)|.$$

Discrepancy Example (uniform)



- $V(R_1) = \frac{1}{16}$, $V(R_2) = \frac{1}{4}$, $\tilde{V}(R_1) = 0$, $\tilde{V}(R_2) = \frac{1}{16}$.

Discrepancy for first n points

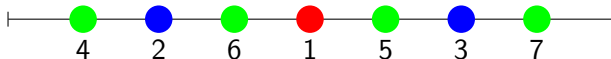
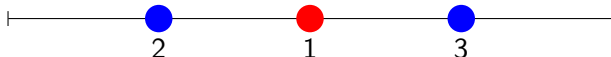
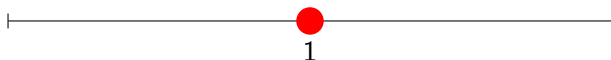
- ▶ We want a sequence that has low discrepancy for all n
- ▶ Instead of a random sequence, we can start with something uniform, and then start filling in the gaps
- ▶ There are various deterministic sequences that obtain low discrepancy

Van der Corput sequence

- ▶ Exercise 2.13
- ▶ $\pi(b_{n-1} \dots b_0) = 0.b_0 b_1 \dots b_{n-1}$.
- ▶ The sequence $\{\pi(1), \pi(2), \pi(3), \dots\}$ is the van der Corput sequence.
- ▶ Example of a deterministic uniform distribution
- ▶ This coincides with the 'uniform distribution then fill up gaps' for $d = 1$!

Example van der Corput sequence

- First elements are $\frac{1}{2}, \frac{1}{4}, \frac{3}{4}, \frac{1}{8}, \frac{5}{8}, \frac{3}{8}, \frac{7}{8}, \dots$



Sampling for low discrepancy

- ▶ We want to extend this idea to $d > 1$.
- ▶ Prime number p , base- p expansions. Change of notation:

$$\pi_2((b_{n-1} \dots b_0)_2) = (0.b_0 b_1 \dots b_{n-1})_2.$$

- ▶ This is for binary representation, but we can do this for arbitrary base p :

$$\pi_p((a_{n-1} \dots a_0)_p) = (0.a_0 a_1 \dots a_{n-1})_p.$$

- ▶ More explicitly:

$$\pi_p \left(\sum_{i=0}^{n-1} a_i p^i \right) = \sum_{i=0}^{n-1} a_i p^{-i-1}.$$

Halton sequence

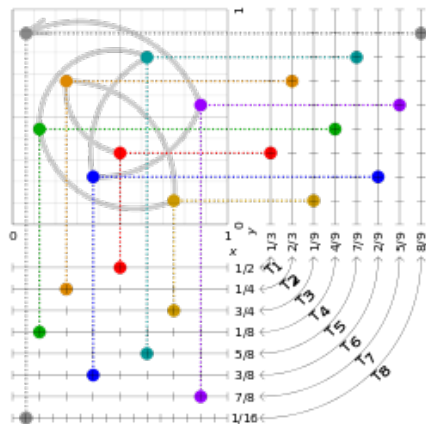
- ▶ Let p_1, \dots, p_d be the first d primes (i.e. 2, 3, 5, 7, 11, ...).
- ▶ *Halton* sequence is:

$$\vec{x}_j = \left(\pi_{p_1}(j), \pi_{p_2}(j), \dots, \pi_{p_d}(j) \right)^T.$$

- ▶ Note that this is different from the 'uniform then fill gaps' idea!

Example Halton sequence

$$\left\{ \left(\frac{1}{2}, \frac{1}{3} \right), \left(\frac{1}{4}, \frac{2}{3} \right), \left(\frac{3}{4}, \frac{1}{9} \right), \left(\frac{1}{8}, \frac{4}{9} \right), \left(\frac{5}{8}, \frac{7}{9} \right), \right. \\ \left. \left(\frac{3}{8}, \frac{2}{9} \right), \left(\frac{7}{8}, \frac{5}{9} \right), \left(\frac{1}{16}, \frac{8}{9} \right), \left(\frac{9}{16}, \frac{1}{27} \right), \dots \right\}$$



Halton discrepancy

- ▶ As we have seen, for Monte Carlo the (expected) error (and discrepancy) is of

$$\mathcal{O}\left(\frac{1}{\sqrt{n}}\right).$$

- ▶ For Halton we instead have (deterministically)

$$\mathcal{O}\left(\frac{\log^d(n)}{n}\right).$$

- ▶ This is almost a quadratic improvement!

Exercise 3.8

- ▶ Last week, we already needed code to generate d -dimensional points
- ▶ Implement the Halton sequence in d -dimensions:
 - ▶ How does this tie into your RNG code?
- ▶ Find the volume of the d -dimensional sphere using
 1. Random sequence
 2. Halton sequence
- ▶ Plot the error for both methods