

Let $U_s := [-1/2, 1/2]^s$ and $D = (0, 1)$. Consider the Dirichlet–Neumann problem

$$\begin{cases} -\frac{\partial}{\partial x} \left(a(x, \mathbf{y}) \frac{\partial}{\partial x} u(x, \mathbf{y}) \right) = f(x), & x \in D, \mathbf{y} \in U_s, \\ u(0, \mathbf{y}) = 0 = \frac{\partial}{\partial x} u(1, \mathbf{y}), & \mathbf{y} \in U_s, \end{cases}$$

endowed with the parametric coefficient

$$a(x, \mathbf{y}) := a_0(x) + \sum_{j=1}^s y_j \psi_j(x), \quad x \in D, \mathbf{y} = (y_j)_{j=1}^s \in U_s,$$

where $a_0 \in L^\infty(D)$ and $\psi_j \in L^\infty(D)$ for all $j \geq 1$. Furthermore, we assume that $f \in L^2(D)$ is given and there exist constants $a_{\min}, a_{\max} > 0$ such that $0 < a_{\min} \leq a(x, \mathbf{y}) \leq a_{\max} < \infty$ for all $x \in D$ and $\mathbf{y} \in U_s$.

The solution is given by

$$u(x, \mathbf{y}) = \int_0^x \left(\int_w^1 f(z) dz \right) \frac{1}{a(w, \mathbf{y})} dw, \quad x \in D, \mathbf{y} \in U_s. \quad (1)$$

1. Show that

$$|\partial_{\mathbf{y}}^{\boldsymbol{\nu}} u(x, \mathbf{y})| \leq \frac{\|f\|_{L^2(D)}}{a_{\min}} |\boldsymbol{\nu}|! \mathbf{b}^{\boldsymbol{\nu}} \quad \text{for all } x \in D, \mathbf{y} \in U_s, \boldsymbol{\nu} \in \mathbb{N}_0^s,$$

where $\mathbf{b} = (b_j)_{j=1}^s$ with $b_j = \frac{\|\psi_j\|_{L^\infty(D)}}{a_{\min}}$, $\mathbf{b}^{\boldsymbol{\nu}} = \prod_{j=1}^s b_j^{\nu_j}$, and $|\boldsymbol{\nu}| := \sum_{j=1}^s \nu_j$.

2. Let $\boldsymbol{\gamma} := (\gamma_{\mathbf{u}})_{\mathbf{u} \subseteq \{1, \dots, s\}}$ be a sequence of positive real numbers. During the lectures we considered an unanchored, weighted Sobolev space $H_{s, \boldsymbol{\gamma}}$ equipped with the norm

$$\|f\|_{s, \boldsymbol{\gamma}}^2 = \sum_{\mathbf{u} \subseteq \{1, \dots, s\}} \frac{1}{\gamma_{\mathbf{u}}} \int_{[0, 1]^{|\mathbf{u}|}} \left(\int_{[0, 1]^{s-|\mathbf{u}|}} \frac{\partial^{|\mathbf{u}|}}{\partial \mathbf{y}_{\mathbf{u}}} f(\mathbf{y}) d\mathbf{y}_{-\mathbf{u}} \right)^2 d\mathbf{y}_{\mathbf{u}}, \quad f \in H_{s, \boldsymbol{\gamma}},$$

where $d\mathbf{y}_{\mathbf{u}} := \prod_{j \in \mathbf{u}} dy_j$ and $d\mathbf{y}_{-\mathbf{u}} := \prod_{j \in \{1, \dots, s\} \setminus \mathbf{u}} dy_j$ for $\mathbf{u} \subseteq \{1, \dots, s\}$.

Fix $x \in D$ and define $F(\mathbf{y}) = u(x, \mathbf{y} - \frac{1}{2})$ for $\mathbf{y} \in [0, 1]^s$. Show that

$$\|F\|_{s, \boldsymbol{\gamma}}^2 \leq \frac{\|f\|_{L^2(D)}^2}{a_{\min}^2} \sum_{\mathbf{u} \subseteq \{1, \dots, s\}} \frac{(|\mathbf{u}|!)^2 \prod_{j \in \mathbf{u}} b_j^2}{\gamma_{\mathbf{u}}}. \quad (2)$$

3. Let us consider the QMC approximation of the integral $\int_{[0, 1]^s} F(\mathbf{y}) d\mathbf{y}$. Using s , n , and $\boldsymbol{\gamma}$ as inputs in a component-by-component (CBC) algorithm, it is possible to construct a QMC rule satisfying the error bound

$$\text{R.M.S. error} \leq \left(\frac{1}{\varphi(n)} \sum_{\emptyset \neq \mathbf{u} \subseteq \{1, \dots, s\}} \gamma_{\mathbf{u}}^{\lambda} \left(\frac{2\zeta(2\lambda)}{(2\pi^2)^{\lambda}} \right)^{|\mathbf{u}|} \right)^{\frac{1}{2\lambda}} \|F\|_{s, \boldsymbol{\gamma}} \quad \text{for all } \lambda \in (\tfrac{1}{2}, 1],$$

where $\varphi(n) = |\{k \in \mathbb{N} : 1 \leq k \leq n-1, \gcd(k, n) = 1\}|$ is the Euler totient function and $\zeta(x) = \sum_{k=1}^{\infty} k^{-x}$ is the Riemann zeta function for $x > 1$.

By plugging in (2), we obtain the error bound

R.M.S. error

$$\leq \frac{\|f\|_{L^2(D)}}{a_{\min}} \left(\frac{1}{\varphi(n)} \sum_{\emptyset \neq \mathbf{u} \subseteq \{1, \dots, s\}} \gamma_{\mathbf{u}}^{\lambda} \left(\frac{2\zeta(2\lambda)}{(2\pi^2)^{\lambda}} \right)^{|\mathbf{u}|} \right)^{\frac{1}{2\lambda}} \left(\sum_{\mathbf{u} \subseteq \{1, \dots, s\}} \frac{(|\mathbf{u}|!)^2 \prod_{j \in \mathbf{u}} b_j^2}{\gamma_{\mathbf{u}}} \right)^{\frac{1}{2}}. \quad (3)$$

Show that the upper bound (3) is minimized by choosing the weights

$$\gamma_{\mathbf{u}} = \left(|\mathbf{u}|! \prod_{j \in \mathbf{u}} \frac{b_j}{\sqrt{2\zeta(2\lambda)/(2\pi^2)^{\lambda}}} \right)^{\frac{2}{1+\lambda}}, \quad \mathbf{u} \subseteq \{1, \dots, s\}, \quad (4)$$

where we use the convention that an empty product is equal to 1.

Under which conditions can the R.M.S. error be bounded independently of the dimension s ?

4. Let us consider a simple numerical discretization of (1). Let $x_k = hk$, $h = \frac{1}{100}$, $k \in \{0, \dots, 100\}$. For simplicity, let $f(x) = 1$. The integral in (1) can be discretized, e.g., using the trapezoidal rule as

$$\int_0^{x_k} g(w, \mathbf{y}) dw \approx h \sum_{i=1}^k \frac{g(x_i, \mathbf{y}) + g(x_{i-1}, \mathbf{y})}{2} \text{ for } k \in \{1, \dots, n\} \text{ with } g(w, \mathbf{y}) := \frac{1-w}{a(w, \mathbf{y})}.$$

This leads to the discretized solution

$$\mathbf{u}(\mathbf{y}) = G \frac{1}{\mathbf{a}(\mathbf{y})}, \quad (5)$$

where $G \in \mathbb{R}^{100 \times 101}$, $\mathbf{u}(\mathbf{y}) = [u(x_1, \mathbf{y}), \dots, u(x_{100}, \mathbf{y})]^T$, $\mathbf{a}(\mathbf{y}) = [a(x_0, \mathbf{y}), \dots, a(x_{100}, \mathbf{y})]^T$, and $\frac{1}{\mathbf{a}(\mathbf{y})} = \left(\frac{1}{a(x_{i-1}, \mathbf{y})} \right)_{i=1}^{101}$ denotes the elementwise reciprocal vector of \mathbf{a} .

- (a) In tasks 1–3, we analyzed the use of QMC for the *non-discretized problem*. Are the conclusions still valid for the numerically discretized solution (5)?
- (b) Fix $x = 0.5$ ($= x_{50}$ in our discretization) and consider the function $F(\mathbf{y}) = u(0.5, \mathbf{y} - \frac{1}{2})$, $\mathbf{y} \in [0, 1]^s$. We can apply a randomly shifted rank-1 lattice rule by drawing R shifts $\Delta_1, \dots, \Delta_R$ from $\mathcal{U}([0, 1]^s)$ and computing the cubatures

$$Q_n^{(r)} F = \frac{1}{n} \sum_{k=1}^n F(\text{mod}(\mathbf{t}_k + \Delta_r, 1)) \quad \text{for } r \in \{1, \dots, R\},$$

where $\mathbf{t}_k = \text{mod}(\frac{k\mathbf{z}}{n}, 1)$. As our approximation of $\int_{[0, 1]^s} F(\mathbf{y}) d\mathbf{y}$, we take the average

$$\overline{Q}_{n,R} F = \frac{1}{R} \sum_{r=1}^R Q_n^{(r)} F.$$

We can estimate the root-mean-square error by computing

$$E_{n,R} = \sqrt{\frac{1}{R(R-1)} \sum_{r=1}^R (\bar{Q}_{n,R} F - Q_n^{(r)} F)^2}.$$

Fix a “reasonable” number of random shifts (e.g., you may choose $R = 4$ or $R = 8$ or $R = 16 \dots$) and compute $E_{n,R}$ for increasing n . As the parameterization of the diffusion coefficient, you can consider, e.g.,

$$a(x, \mathbf{y}) = 1 + \sum_{j=1}^s y_j j^{-2} \sin(\pi j x), \quad x \in D, \quad \mathbf{y} \in [-1/2, 1/2]^s.$$

What convergence rate do you observe?

For $s = 10$ and $n \in \{2^{10}, 2^{11}, \dots, 2^{20}\}$, you can use the following precomputed (“off-the-shelf”) generating vector:

$$\mathbf{z} = [1, 182667, 279195, 223491, 205755, 359329, 198937, 246491, 466233, 379083]^T.$$

Tailored lattices with arbitrary s and n for this integration problem can be obtained by using a CBC algorithm with the weights (4) as inputs. Some implementations and “off-the-shelf” lattice rules are available at [1, 2, 3].

Note: “Off-the-shelf” lattice rules tend to work pretty well in practice and it is quite common to simply use some of the precomputed ones available at [1, 3]. However, strictly speaking, one loses the convergence guarantee when the generating vector is not obtained using a CBC algorithm with weights appropriately tailored for the integration problem.

References

- [1] F. Y. Kuo. Lattice rule generating vectors. <https://web.maths.unsw.edu.au/~fkuo/lattice/>
- [2] F. Y. Kuo and D. Nuyens. *QMC4PDE*. <https://people.cs.kuleuven.be/~dirk.nuyens/qmc4pde/>
- [3] D. Nuyens. *Magic Point Shop*. <https://people.cs.kuleuven.be/~dirk.nuyens/qmc-generators/>