Multidimensiontal PDE with Value Iteration

WPI

Problem: nd BVP

Domain

$$O = \{x \in \mathbb{R}^d : 0 < x_i < 1, i = 1, 2, \dots d\}.$$



► Equation on *O*:

$$(\frac{1}{2}\Delta - \lambda)v(x) + \sum_{i=1}^{d} b_i(x) \frac{\partial v(x)}{\partial x_i} + \ell(x) = 0.$$

$$\Delta v(x) = \sum_{i=1}^{d} \partial_{i} v(x)$$
lata on ∂O :

▶ Dirichlet data on ∂O :

$$v(x)=g(x).$$

A class of benchmarks

Set.
$$d=2$$
 $k=1$, $b=0$
 $1=-2+(x_1-\frac{1}{2})^2+(x_2-\frac{1}{2})^2$
 $=+x_1^2+x_2^2-x_1-x_2-\frac{3}{2}$

Consider a class of PDE with coefficients satisfying,

$$d - \lambda ||x - \frac{1}{2}\mathbf{1}||_2^2 + b(x) \cdot (2x - \mathbf{1}) + \ell(x) = 0,$$

where $\mathbf{1}$ is an \mathbb{R}^d -vector with each element being 1. The exact solution is

$$\mathbf{\hat{1}} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \qquad v(x) = \|x - \frac{1}{2}\mathbf{1}\|_2^2 = \sum_{i=1}^d (x_i - \frac{1}{2})^2.$$

Verify the above exact solution.
$$2iV(x) = \frac{\partial}{\partial x_i}(x_i - \frac{1}{2})^2$$

$$= 2X_i - 1$$

$$= 2X_i - 1$$

$$= 2X_i - 1$$

A benchmark with 2-d

Equation on
$$O = (0,1)^2$$
:= $(0,1) \times (0,1) = \frac{1}{2} \Delta v(x) - v(x) + x_1^2 + x_2^2 - x_1 - x_2 - \frac{3}{2} = 0.$

▶ Dirichlet data on ∂O :

$$v(x) = (x_1 - \frac{1}{2})^2 + (x_2 - \frac{1}{2})^2.$$

Exact solution is

$$v(x) = (x_1 - \frac{1}{2})^2 + (x_2 - \frac{1}{2})^2.$$

Finite difference operators

FFD
$$\frac{\partial}{\partial x_i} v(x) \approx \delta_{he_i} v(x) := \frac{v(x + he_i) - v(x)}{h}.$$

▶ BFD
$$\frac{\partial}{\partial x_i} v(x) \approx \delta_{-he_i} v(x) := \frac{v(x - he_i) - v(x)}{-h}$$

FD
$$\frac{\partial}{\partial x_{i}}v(x) \approx \delta_{-he_{i}}v(x) := \frac{v(x - he_{i}) - v(x)}{-h}.$$
FD
$$\frac{\partial}{\partial x_{i}}v(x) \approx \bar{\delta}_{he_{i}}v(x) = \frac{1}{2}(\delta_{-he_{i}} + \delta_{he_{i}})v(x) = \frac{\sqrt{(x - he_{i})}}{2-h}$$
CFD
$$\frac{\partial}{\partial x_{i}}v(x) \approx \frac{1}{2}(\delta_{-he_{i}} + \delta_{he_{i}})v(x) = \frac{\sqrt{(x - he_{i})}}{2-h}$$

$$\frac{\partial^{2}}{\partial x_{i}\partial x_{j}}v(x) \approx \frac{1}{2}\left(\delta_{\underline{he_{i}}}\delta_{-\underline{he_{j}}}v(x) + \delta_{\underline{he_{j}}}\delta_{-\underline{he_{i}}}v(x)\right)$$

$$\frac{1}{2}$$

$$\frac{\partial^{2}}{\partial x_{j}\partial x_{j}}V(x)$$

Explicit forms of FD operators

CFD

$$\frac{\partial}{\partial x_i}v(x) \approx \bar{\delta}_{he_i}v(x) = \frac{v(x+he_i)-v(x-he_i)}{2h}$$
 = 0(h²)

▶ SCFD with (i = j:)

$$\frac{\partial^2}{\partial x_i \partial x_j} v(x) \approx \delta_{he_i} \delta_{-he_i} v(x) = \frac{v(x + he_i) - 2v(x) + v(x - he_i)}{h^2} = O(h^2)$$

- What is the convergence rate of the above FD operators?
- For 2-d Benchmark, derive CFD with h = 1/4, i.e. write the linear system of the form $L^h u^h = f^h$.

$$\frac{1}{2}\Delta V(x) = \frac{1}{2}\sum_{i=1}^{d} \left(\frac{\partial i}{\partial i}V(x)\right) \frac{\partial^2}{\partial x_i \partial x_i}$$

Sketch of CFD on PDE

- \triangleright In multidimensional case, L^h is not simply tridiagonal and it's hard for programing.

One can rewrite
$$L^h v^h = f^h$$
 by
$$v^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h = \underbrace{I \, v^h - \underbrace{L^h \, v^h + f^h}}_{\text{one can rewrite } V^h$$

Implement value iteration:

$$v^{(n+1)} = F^h v^{(n)}.$$

Then, we expect $v^{(n)} \rightarrow v^h$.

I want to solve, for a, bEIR. $\alpha \times = b$ $0 \quad x = b/a = a^{-1} \cdot b$ @Init Xo JUE,

@ Repeat $\chi_{n+1} = g(\chi_n)$ Nn->X

CFD solution



For simplicity, if we set

$$\gamma = \frac{d}{d+h^2\lambda}, \ p^h(x\pm he_i|x) = \frac{1}{2d}(1\pm hb_i(x)), \ \ell^h(x) = \frac{h^2\ell(x)}{d},$$

then CFD solution $v^h: G^h \mapsto \mathbb{R}$ is



► For
$$x \notin \partial O^h$$
,

$$v(x)=g(x).$$



▶ For $x \in O^h = O \cap G^h$

$$v(x) = \gamma \Big\{ \ell^{h}(x) + \sum_{i=1}^{d} p^{h}(x + he_{i}|x)v(x + he_{i}) + p^{h}(x - he_{i}|x)v(x - he_{i}) \Big\}.$$

which we write as

$$v(x) = F^h v(x).$$

$$(\frac{1}{2}\Delta - \lambda)v(x) + \sum_{i=1}^{d} b_i(x) \frac{\partial v(x)}{\partial x_i} + \ell(x) = 0.$$

$$\frac{1}{2}\sum_{i=1}^{d} \frac{V_i^{\dagger} - 2V + V_i}{h^2} - \lambda V + \sum_{i=1}^{d} b_i \frac{V_i^{\dagger} - V_i^{\dagger}}{2h} + \frac{2}{2h}$$

$$\frac{1}{2}\sum_{i=1}^{d} \frac{V_i^{\dagger} - 2V + V_i}{h^2} - \lambda V + \sum_{i=1}^{d} b_i \frac{V_i^{\dagger} - V_i^{\dagger}}{2h} + \frac{2}{2h}$$

 $V\left(\frac{d}{h^2} + \lambda\right) = \sum_{i=1}^{d} V_i^{\dagger} \left(\frac{1}{2h^2} + \frac{b_i}{2h}\right) + \sum_{i=1}^{d} V_i^{\dagger} \left(\frac{1}{2h^2} - \frac{b_i}{2h}\right) + \lambda$

$$V\left(d+\chi h^{2}\right)=\sum_{i=1}^{d}V_{i}^{\dagger}\left(\frac{1}{2}+\frac{hb_{i}}{2}\right)+\sum_{i}V_{i}^{\dagger}\left(\frac{1}{2}-\frac{b_{i}h}{2}\right)+Qh^{2}$$

$$V(d+\chi h^2) = \sum_{i=1}^{d} V_i^{\dagger} \left(\frac{1}{2} + \frac{hb_i}{2}\right) + \sum_{i=1}^{d} V_i^{\dagger} \left(\frac{1}{2} - \frac{b_i h}{2}\right) + Qh^2$$

 $V = \frac{d+\lambda h^2}{d} = \frac{1}{d} \sum_{i=1}^{d} V_i^{\dagger} \left(\frac{1}{2} + \frac{hb_i}{2} \right) + \sum_{i} V_i^{\dagger} \left(\frac{1}{2} - \frac{b_i h}{2} \right) + Q h^2$

$$\frac{d+\ln d}{d} = \frac{1}{d}\sum_{i=1}^{\infty} V_i^{\dagger} \left(\frac{1}{2} + \frac{hb_i}{2}\right) + \sum_{i} V_i^{\dagger} \left(\frac{1}{2} - \frac{b_ih}{2}\right) + Q$$

Value iteration - Jacobi

Algo.

```
1: procedure VI(\hat{\epsilon}, \hat{n})
                                                                    \triangleright \hat{n}: max iteration, \hat{\epsilon}: tolerance
            Initial guess: \{v(x): x \in O^h\}.
 3:
            flag \leftarrow 1, n \leftarrow 0
                                                                              \triangleright n is number of iteration
           while flag do
 4.
 5:
                  \epsilon \leftarrow 0; n \leftarrow n + 1
                                                                         1(m) < + hn/m)

for x < 0, yr

n(m) < n(m)

for x < 0, yr

                 for x \in O^h do
 6:
 7:
                      u(x) \leftarrow v(x)
                       v(x) \leftarrow F^h u(x).
 8:
                       \epsilon \leftarrow \max\{\epsilon, |u(x) - v(x)|
 9:
                        if \epsilon < \hat{\epsilon} then
10:
                             flag = 0
11:
            return \{v(x): x \in O^h\}.
12:
```

▶ If line 8 is replaced by $v(x) \leftarrow F^h v(x)$, then it becomes Gauss-Seidel iteration.

Ex on Benchmark with 2d

- ▶ Write explicit form of $v^h = F^h v^h$ for CFD.
- ▶ Prove that, for any $v: O^h \mapsto \mathbb{R}$, iteration has contraction, i.e. there exists $\hat{\gamma} \in (0,1)$ such that

$$||F^h v||_{\infty} \le \hat{\gamma} ||v||_{\infty}.$$

- ▶ Prove that, value iteration converges, i.e. $v^{(n)} \rightarrow v^h$ as $n \rightarrow \infty$.
- ▶ Use value iteration (Jacobi) to compute $||v^h R^h v||_{\infty}$ for h = 1/8.
- ▶ Use Gauss-Seidel value iteration to compute $||v^h R^h v||_{\infty}$ for h = 1/8.
- (optional) What is the highest dimension can you handle with value iteration?
- ▶ Write explicit form of $v^h = F^h v^h$ for UFD.