

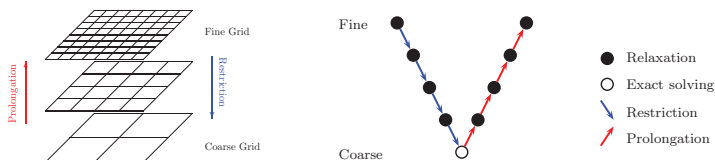
## Section 03. Basic Ideas of Multigrid Methods

# Multilevel Iterative Methods

## Examples of multilevel algorithms

- Quick Sort, FFT, FMM, GMG, AMG, H-Matrix,  $H^2$ -Matrix, ...

## Multigrid V-cycle



## Key ingredients for multilevel iterative methods

- Construct multilevel hierarchy in an efficient way (setup)
- Find effective (yet cheap) smoothers for each level
- Find good coarse-grid correction (CGC) algorithms

Need complementary **smoothing** and **CGC** steps to get better convergence.



## Finite Difference Methods

In one-dimensional case, we can assume  $\Omega = (0, 1)$  and it is divided into  $N + 1$  equally spaced pieces. So we get a uniform mesh with meshsize  $h = \frac{1}{N+1}$ ; see the following figure for illustration.

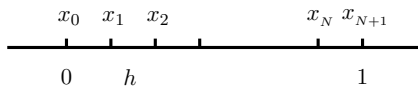


Figure: Uniform mesh in 1D.

For the right-hand side, we can use an approximation:  $\vec{f} := (f_i)_{i=1}^N = (f(x_i))_{i=1}^N$ . For the left-hand side, using the Taylor's expansion, we can easily obtain that

$$\begin{aligned} u''(x_i) &= \frac{1}{h^2} \left[ u(x_{i-1}) - 2u(x_i) + u(x_{i+1}) \right] + O(h^2) \\ &\approx \frac{1}{h^2} \left[ u_{i-1} - 2u_i + u_{i+1} \right], \end{aligned}$$

where  $u_i \approx u(x_i)$  is an approximate solution (**finite difference solution**).



## Nested Grids

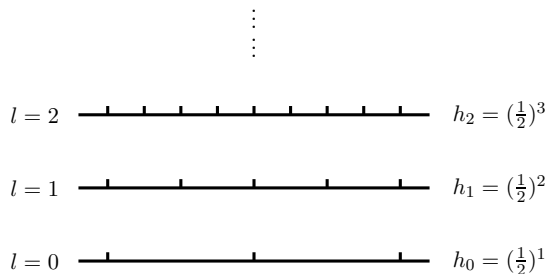
Solve the 1D Poisson's equation:

$$A\vec{u} = \vec{f} \quad \text{with } A = \frac{1}{h^2} \text{tridiag}(-1, 2, -1), \quad f_i = f(x_i).$$

Suppose there are a hierarchy of  $L + 1$  grids with  $h_l = (\frac{1}{2})^{l+1}$  ( $l = 0, 1, \dots, L$ ). It is clear that

$$h_0 > h_1 > h_2 > \dots > h_L =: h$$

and  $N = 2^{L+1} - 1$ . We call level  $L$  the finest level and level 0 the coarsest level.



**Figure:** Hierarchical grids for 1D multigrid method.





# Multigrid Algorithm

**Error correction for linear problems:** Suppose that  $\vec{u}^{(m)}$  is an approximate solution. Then we have

$$A(\vec{u} - \vec{u}^{(m)}) = \vec{r}^{(m)} := \vec{f} - A\vec{u}^{(m)}$$

and the error equation can be written

$$A\vec{e}^{(m)} = \vec{r}^{(m)}.$$

We then update the iterative solution by  $\vec{u}^{(m+1)} = \vec{u}^{(m)} + \vec{e}^{(m)}$  to obtain a new approximation of  $\vec{u}$ .

Then we have the following recursively-defined algorithm:

**Algorithm (One iteration of multigrid method  $\vec{u}_l = MG(l, \vec{f}_l, \vec{u}_l)$ )**

- ① Pre-smoothing:  $\vec{u}_l \leftarrow \vec{u}_l + \frac{1}{2}D_l^{-1}(\vec{f}_l - A_l\vec{u}_l)$ .
- ② Restriction:  $\vec{r}_{l-1} \leftarrow R_{l,l-1}(\vec{f}_l - A_l\vec{u}_l)$ .
- ③ Coarse-grid correction: If  $l = 1$ ,  $\vec{e}_{l-1} \leftarrow A_{l-1}^{-1}\vec{r}_{l-1}$ ;  $\vec{e}_{l-1} \leftarrow MG(l-1, \vec{r}_{l-1}, \vec{0}_{l-1})$ , otherwise.
- ④ Prolongation:  $\vec{u}_l \leftarrow \vec{u}_l + P_{l-1,l}\vec{e}_{l-1}$ .
- ⑤ Post-smoothing:  $\vec{u}_l \leftarrow \vec{u}_l + \frac{1}{2}D_l^{-1}(\vec{f}_l - A_l\vec{u}_l)$ .



## A Simple Numerical Experiment

In the following table, we give the numerical results of the above algorithm for the 1D Poisson's equation (using three G-S iterations as smoother). From the table, we find that, unlike the classical Jacobi and G-S methods, this multigrid method converges uniformly with respect to the meshsize  $h$ . This is, of course, a very desirable feature of the multilevel iterative methods, which will be investigated in this course.

#Levels	#DOF	#Iter	Contract factor
5	31	4	0.0257
6	63	4	0.0259
7	127	4	0.0260
8	255	4	0.0260
9	511	4	0.0261
10	1023	4	0.0262

**Table:** Convergence behavior of 1D geometric multigrid method.

Textbook multigrid efficiency: "TME means solving a discrete PDE problem in a computational work which is only a small (less than 10) multiple of the operation count in the discretized system of equations itself."



# Computational Cost

## Assumptions:

- Denote the work needed by  $\mathcal{B}_l$  is  $W_l$ .
- Assume the **each smoothing sweep** costs  $O(N_l)$  operations and  $N_l \sim h_l^{-d} \sim \gamma^{-ld}$ . Then it requires  $2m O(N_l)$  operations for the pre- and post-smoothing ( $m$ -steps) on level  $l$ .
- The **prolongation and restriction** also requires  $O(N_l)$  operations.

## Work of multilevel cycles:

$$\begin{aligned}
 W_{k+1} &= O(N_{k+1}) + W_k \\
 &= \dots\dots \\
 &= O\left(\sum_{j=0}^{k+1} N_j\right) = O(N_{k+1}) \sum_{j=0}^{k+1} \gamma^{jd}.
 \end{aligned}$$

Let  $N = N_L$  be the number of unknowns on the finest grid. The V-cycle costs  $O(N)$  operations in each cycle. Apparently, this analysis also yields computational complexity of the W-cycle, if we choose an appropriate  $\mu_1$  such that  $\mu_1 \gamma^d < 1$ .

One question remains: How many iterations (cycles) needed to reach certain accuracy?





# General Multilevel Iterative Methods

Generally speaking, multilevel iterations: **Setup Phase (fixed)**  $\implies$  **Solve Phase (variable)**

## Setup phase

- Constructing transfer operators, coarse problems, etc
- Using geometric information or algebraic information or both
- Needed only once in each solution procedure
- Sometimes even shared by multiple solution procedures

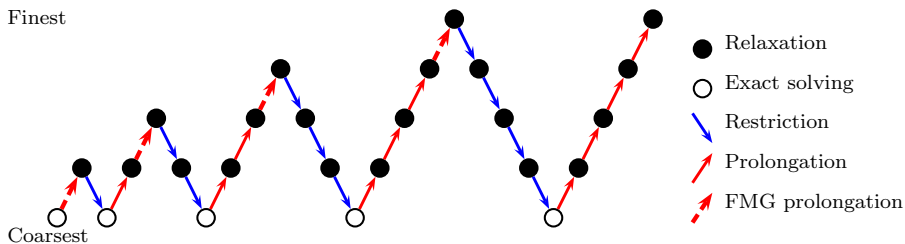
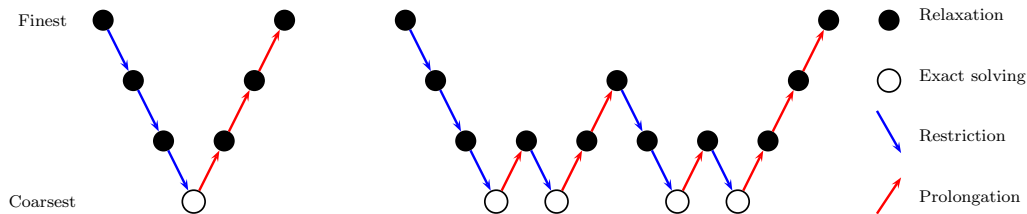
## Solve phase

- Applying relaxation (simple iterative methods) on different levels and putting components together
- Needed many times as iteration or precondition step, but hopeful not too many
- Main concern: How to approximate coarse solution accurately without costing too much

## Two-level method

- Simplest case: Two-level method (solve the coarse level exactly or approximately)
- Easier to implement and analyze, provide insight for design multilevel methods

# Examples of Multilevel Cycles



V, VV, VW, W, AMLI, K, N, H, ... , Nested Iterations

# A General Workflow

