

Lecture 1

Finite volume methods

Course *Block-structured Adaptive Finite Volume Methods for Shock-Induced Combustion Simulation*

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Outline

Conservation laws

- Mathematical background

- Examples

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Finite volume methods

- Basics of finite difference methods

- Splitting methods, second derivatives

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Upwind schemes

- Flux-difference splitting
- Flux-vector splitting
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- Elements of adaptive algorithms
- Adaptivity on unstructured meshes
- Structured mesh refinement techniques

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Hyperbolic Conservation Laws

$$\frac{\partial}{\partial t} \mathbf{q}(\mathbf{x}, t) + \sum_{n=1}^d \frac{\partial}{\partial x_n} \mathbf{f}_n(\mathbf{q}(\mathbf{x}, t)) = 0, \quad D \subset \{(\mathbf{x}, t) \in \mathbb{R}^d \times \mathbb{R}_0^+\}$$

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$\mathbf{q} = \mathbf{q}(\mathbf{x}, t) \in S \subset \mathbb{R}^M$ - vector of state, $\mathbf{f}_n(\mathbf{q}) \in C^1(S, \mathbb{R}^M)$ - flux functions,

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Definition (Hyperbolicity)

$\mathbf{A}(\mathbf{q}, \nu) = \nu_1 \mathbf{A}_1(\mathbf{q}) + \dots + \nu_d \mathbf{A}_d(\mathbf{q})$ with $\mathbf{A}_n(\mathbf{q}) = \partial \mathbf{f}_n(\mathbf{q}) / \partial \mathbf{q}$ has M real eigenvalues $\lambda_1(\mathbf{q}, \nu) \leq \dots \leq \lambda_M(\mathbf{q}, \nu)$ and M linear independent right eigenvectors $\mathbf{r}_m(\mathbf{q}, \nu)$.

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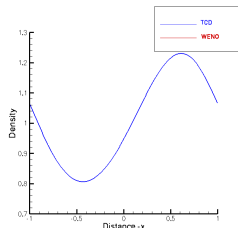
$\mathbf{q} = \mathbf{q}(\mathbf{x}, t) \in S \subset \mathbb{R}^M$ - vector of state, $\mathbf{f}_n(\mathbf{q}) \in C^1(S, \mathbb{R}^M)$ - flux functions, $\mathbf{s}(\mathbf{q}) \in C^1(S, \mathbb{R}^M)$ - source term

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If $\mathbf{f}_n(\mathbf{q})$ is nonlinear, classical solutions $\mathbf{q}(\mathbf{x}, t) \in C^1(D, S)$ do not generally exist, not even for $\mathbf{q}_0(\mathbf{x}) \in C^1(\mathbb{R}^d, S)$ [Majda, 1984], [Godlewski and Raviart, 1996], [Kröner, 1997]

Example: Euler equations



Weak solutions

Integral form (Gauss's theorem):

$$\int_{\Omega} \mathbf{q}(\mathbf{x}, t + \Delta t) d\mathbf{x} - \int_{\Omega} \mathbf{q}(\mathbf{x}, t) d\mathbf{x} + \sum_{n=1}^d \int_t^{t+\Delta t} \int_{\partial\Omega} \mathbf{f}_n(\mathbf{q}(\mathbf{o}, t)) \sigma_n(\mathbf{o}) d\mathbf{o} dt = \int_t^{t+\Delta t} \int_{\Omega} \mathbf{s}(\mathbf{q}(\mathbf{x}, t)) d\mathbf{x}$$

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Theorem (Weak solution)

$\mathbf{q}_0 \in L_{loc}^{\infty}(\mathbb{R}^d, S)$. $\mathbf{q} \in L_{loc}^{\infty}(D, S)$ is weak solution if \mathbf{q} satisfies

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for any test function $\varphi \in C_0^1(D, S)$

Entropy solutions

Select physical weak solution as $\lim_{\varepsilon \rightarrow 0} \mathbf{q}_\varepsilon = \mathbf{q}$ almost everywhere in D of

$$\frac{\partial \mathbf{q}_\varepsilon}{\partial t} + \sum_{n=1}^d \frac{\partial \mathbf{f}_n(\mathbf{q}_\varepsilon)}{\partial x_n} - \varepsilon \sum_{n=1}^d \frac{\partial^2 \mathbf{q}_\varepsilon}{\partial x_n^2} = \mathbf{s}(\mathbf{q}_\varepsilon), \quad \mathbf{x} \in \mathbb{R}^d, \quad t > 0$$

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Assume existence of entropy $\eta \in C^2(S, \mathbb{R})$ and entropy fluxes $\psi_n \in C^1(S, \mathbb{R})$ that satisfy

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then $\lim_{\varepsilon \rightarrow 0} \mathbf{q}_\varepsilon = \mathbf{q}$ almost everywhere in D is weak solution and satisfies

$$\frac{\partial \eta(\mathbf{q})}{\partial t} + \sum_{n=1}^d \frac{\partial \psi_n(\mathbf{q})}{\partial x_n} \leq \frac{\partial \eta(\mathbf{q})}{\partial \mathbf{q}} \cdot \mathbf{s}(\mathbf{q})$$

in the sense of distributions. Proof: [Godlewski and Raviart, 1996]

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for all entropy functions $\eta(\mathbf{q})$ and all test functions $\varphi \in C_0^1(D, \mathbb{R}_0^+)$, $\varphi \geq 0$

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An entropy solution \mathbf{q} is a classical solution $\mathbf{q} \in C^1(D, S)$ almost everywhere and satisfies the Rankine-Hugoniot (RH) jump condition

$$(\mathbf{q}^+ - \mathbf{q}^-) \sigma_t + \sum_{n=1}^d (\mathbf{f}_n(\mathbf{q}^+) - \mathbf{f}_n(\mathbf{q}^-)) \sigma_n = \mathbf{0}$$

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and the jump inequality

$$(\eta(\mathbf{q}^+) - \eta(\mathbf{q}^-)) \sigma_t + \sum_{n=1}^d (\psi_n(\mathbf{q}^+) - \psi_n(\mathbf{q}^-)) \sigma_n \leq 0$$

along discontinuities. Proof: [Godlewski and Raviart, 1996]

Examples

Euler equations

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_n} (\rho u_n) = 0$$

$$\frac{\partial}{\partial t} (\rho u_k) + \frac{\partial}{\partial x_n} (\rho u_k u_n + \delta_{kn} p) = 0, \quad k = 1, \dots, d$$

$$\frac{\partial}{\partial t} (\rho E) + \frac{\partial}{\partial x_n} (u_n (\rho E + p)) = 0$$

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with polytrope gas equation of state

$$p = (\gamma - 1)\left(\rho E - \frac{1}{2}\rho u_n u_n\right)$$

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have structure

$$\partial_t \mathbf{q}(\mathbf{x}, t) + \nabla \cdot \mathbf{f}(\mathbf{q}(\mathbf{x}, t)) = 0$$

Examples II

Navier-Stokes equations

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with stress tensor

$$\tau_{kn} = \mu \left(\frac{\partial u_n}{\partial x_k} + \frac{\partial u_k}{\partial x_n} \right) - \frac{2}{3} \mu \frac{\partial u_j}{\partial x_j} \delta_{kn}$$

and heat conduction

$$q_n = -\lambda \frac{\partial T}{\partial x_n}$$

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Type can be either hyperbolic or parabolic

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Time discretization $t_n = n\Delta t$, discrete volumes

$I_j = [x_j - \frac{1}{2}\Delta x, x_j + \frac{1}{2}\Delta x[=: [x_{j-1/2}, x_{j+1/2}[$

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Using approximations $\mathbf{Q}_j(t) \approx \frac{1}{|I_j|} \int_{I_j} \mathbf{q}(\mathbf{x}, t) dx$, $\mathbf{s}(\mathbf{Q}_j(t)) \approx \frac{1}{|I_j|} \int_{I_j} \mathbf{s}(\mathbf{q}(\mathbf{x}, t)) dx$

and numerical fluxes

$$\mathbf{F}(\mathbf{Q}_j(t), \mathbf{Q}_{j+1}(t)) \approx \mathbf{f}(\mathbf{q}(x_{j+1/2}, t)), \quad \mathbf{H}(\mathbf{Q}_j(t), \mathbf{Q}_{j+1}(t)) \approx \mathbf{h}(\mathbf{q}(x_{j+1/2}, t), \nabla \mathbf{q}(x_{j+1/2}, t))$$

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yields after integration (Gauss theorem)

$$\begin{aligned} \mathbf{Q}_j(t_{n+1}) = \mathbf{Q}_j(t_n) - \frac{1}{\Delta x} \int_{t_n}^{t_{n+1}} [\mathbf{F}(\mathbf{Q}_j(t), \mathbf{Q}_{j+1}(t)) - \mathbf{F}(\mathbf{Q}_{j-1}(t), \mathbf{Q}_j(t))] dt - \\ \frac{1}{\Delta x} \int_{t_n}^{t_{n+1}} [\mathbf{H}(\mathbf{Q}_j(t), \mathbf{Q}_{j+1}(t)) - \mathbf{H}(\mathbf{Q}_{j-1}(t), \mathbf{Q}_j(t))] dt + \int_{t_n}^{t_{n+1}} \mathbf{s}(\mathbf{Q}_j(t)) dt \end{aligned}$$

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For instance:

$$\begin{aligned} \mathbf{Q}_j^{n+1} = \mathbf{Q}_j^n - \frac{\Delta t}{\Delta x} [\mathbf{F}(\mathbf{Q}_j^n, \mathbf{Q}_{j+1}^n) - \mathbf{F}(\mathbf{Q}_{j-1}^n, \mathbf{Q}_j^n)] - \\ \frac{\Delta t}{\Delta x} [\mathbf{H}(\mathbf{Q}_j^n, \mathbf{Q}_{j+1}^n) - \mathbf{H}(\mathbf{Q}_{j-1}^n, \mathbf{Q}_j^n)] + \Delta t \mathbf{s}(\mathbf{Q}_j^n) \end{aligned}$$

Some classical definitions

$(2s + 1)$ -point difference scheme of the form

$$\mathbf{Q}_j^{n+1} = \mathcal{H}^{(\Delta t)}(\mathbf{Q}_{j-s}^n, \dots, \mathbf{Q}_{j+s}^n)$$

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For each time τ there is a constant C_S and a value $n_0 \in \mathbb{N}$ such that

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If the local truncation error

$$\mathcal{L}^{(\Delta t)}(\mathbf{x}, t) := \frac{1}{\Delta t} \left[\mathbf{q}(\mathbf{x}, t + \Delta t) - \mathcal{H}^{(\Delta t)}(\mathbf{q}(\cdot, t)) \right]$$

satisfies $\|\mathcal{L}^{(\Delta t)}(\cdot, t)\| \rightarrow 0$ as $\Delta t \rightarrow 0$

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Definition (Convergence)

If the global error $\mathcal{E}^{(\Delta t)}(\mathbf{x}, t) := \mathbf{Q}(\mathbf{x}, t) - \mathbf{q}(\mathbf{x}, t)$ satisfies $\|\mathcal{E}^{(\Delta t)}(\cdot, t)\| \rightarrow 0$ as $\Delta t \rightarrow 0$ for all admissible initial data $\mathbf{q}_0(\mathbf{x})$

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$\mathcal{H}(\cdot)$ is accurate of order o if for all sufficiently smooth initial data $\mathbf{q}_0(\mathbf{x})$, there is a constant C_L , such that the local truncation error satisfies

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Definition (Consistency of a conservative method)

If the numerical flux satisfies $\mathbf{F}(\mathbf{q}, \dots, \mathbf{q}) = \mathbf{f}(\mathbf{q})$ for all $\mathbf{q} \in S$

Splitting methods

Solve homogeneous PDE and ODE successively!

$$\mathcal{H}^{(\Delta t)} : \quad \partial_t \mathbf{q} + \nabla \cdot \mathbf{f}(\mathbf{q}) = 0, \quad \text{IC: } \mathbf{Q}(t_m) \xrightarrow{\Delta t} \tilde{\mathbf{Q}}$$

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1st-order dimensional splitting for $\mathcal{H}^{(\cdot)}$:

$$\mathcal{X}_1^{(\Delta t)} : \quad \partial_t \mathbf{q} + \partial_{x_1} \mathbf{f}_1(\mathbf{q}) = 0, \quad \text{IC: } \mathbf{Q}(t_m) \xrightarrow{\Delta t} \tilde{\mathbf{Q}}^{1/2}$$

$$\mathcal{X}_2^{(\Delta t)} : \quad \partial_t \mathbf{q} + \partial_{x_2} \mathbf{f}_2(\mathbf{q}) = 0, \quad \text{IC: } \tilde{\mathbf{Q}}^{1/2} \xrightarrow{\Delta t} \tilde{\mathbf{Q}}$$

[Toro, 1999]

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i.e.

$$|1 - 4C_1 - 4C_2| \leq 1$$

from which we derive the stability condition

$$0 \leq c \left(\frac{\Delta t}{\Delta x_1^2} + \frac{\Delta t}{\Delta x_2^2} \right) \leq \frac{1}{2}$$

Linear upwind schemes

Consider Riemann problem

$$\frac{\partial}{\partial t} \mathbf{q}(x, t) + \mathbf{A} \frac{\partial}{\partial x} \mathbf{q}(x, t) = \mathbf{0}, \quad x \in \mathbb{R}, \quad t > 0$$

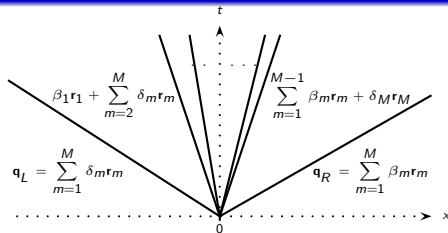
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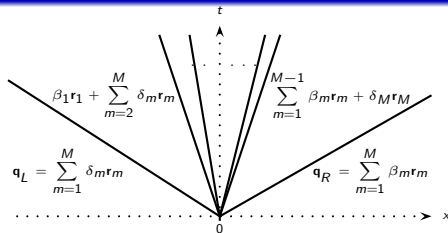
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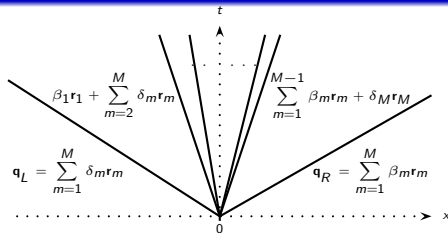
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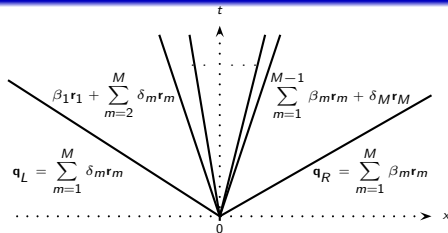


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with $\Delta \mathbf{q} = \mathbf{q}_R - \mathbf{q}_L$

Flux difference splitting

Godunov-type scheme with $\Delta \mathbf{Q}_{j+1/2}^n = \mathbf{Q}_{j+1}^n - \mathbf{Q}_j^n$

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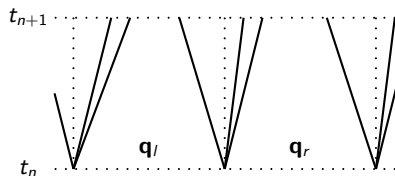
stability condition

$$\max_{j \in \mathbb{Z}} |\hat{\lambda}_{m,j+\frac{1}{2}}| \frac{\Delta t}{\Delta x} \leq 1, \quad \text{for all } m = 1, \dots, M$$

[LeVeque, 1992]

Roe's approximate Riemann solver

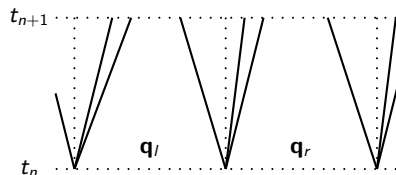
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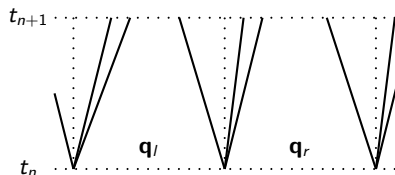
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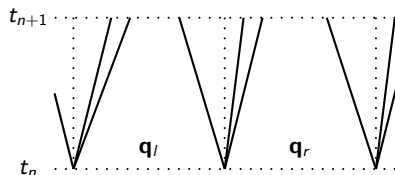
- (i) $\hat{\mathbf{A}}(\mathbf{q}_L, \mathbf{q}_R)$ has real eigenvalues
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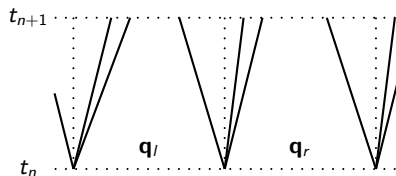
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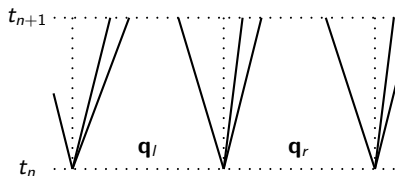
For Euler equations:

$$\hat{\rho} = \frac{\sqrt{\rho_L \rho_R} + \sqrt{\rho_R \rho_L}}{\sqrt{\rho_L} + \sqrt{\rho_R}} = \sqrt{\rho_L \rho_R} \quad \text{and} \quad \hat{v} = \frac{\sqrt{\rho_L} v_L + \sqrt{\rho_R} v_R}{\sqrt{\rho_L} + \sqrt{\rho_R}} \quad \text{for } v = u_n, H$$

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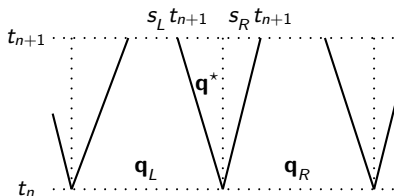
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Wave decomposition: $\Delta \mathbf{q} = \mathbf{q}_r - \mathbf{q}_l = \sum_m a_m \hat{\mathbf{r}}_m$

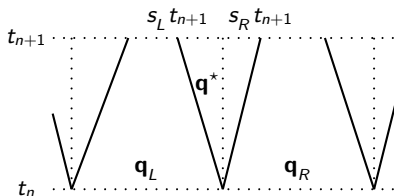
$$\begin{aligned} \mathbf{F}(\mathbf{q}_L, \mathbf{q}_R) &= \mathbf{f}(\mathbf{q}_L) + \sum_{\hat{\lambda}_m < 0} \hat{\lambda}_m a_m \hat{\mathbf{r}}_m = \mathbf{f}(\mathbf{q}_R) - \sum_{\hat{\lambda}_m \geq 0} \hat{\lambda}_m a_m \hat{\mathbf{r}}_m \\ &= \frac{1}{2} \left(\mathbf{f}(\mathbf{q}_L) + \mathbf{f}(\mathbf{q}_R) - \sum_m |\hat{\lambda}_m| a_m \hat{\mathbf{r}}_m \right) \end{aligned}$$

Harten-Lax-Van Leer (HLL) approximate Riemann solver



$$\bar{q}(x, t) = \begin{cases} q_L, & x < s_L t \\ q^*, & s_L t \leq x \leq s_R t \\ q_R, & x > s_R t \end{cases}$$

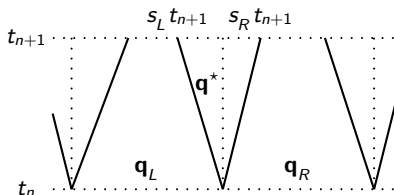
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Euler equations:

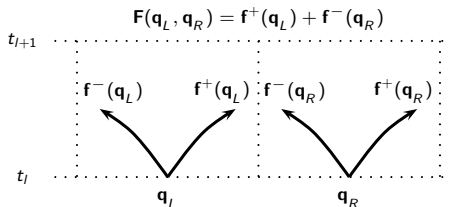
$$s_L = \min(u_{1,L} - c_L, u_{1,R} - c_R), \quad s_R = \max(u_{1,L} + c_L, u_{1,R} + c_R)$$

[Toro, 1999], HLLC: [Toro et al., 1994]

Flux vector splitting

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$$\mathbf{f}(\mathbf{q}) = \mathbf{f}^+(\mathbf{q}) + \mathbf{f}^-(\mathbf{q})$$



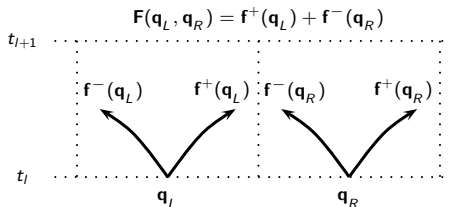
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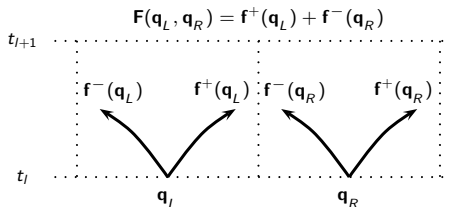
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plus reproduction of regular upwinding

$$\begin{aligned} \mathbf{f}^+(\mathbf{q}) &= \mathbf{f}(\mathbf{q}), & \mathbf{f}^-(\mathbf{q}) &= \mathbf{0} & \text{if } \lambda_m \geq 0 & \text{for all } m = 1, \dots, M \\ \mathbf{f}^+(\mathbf{q}) &= \mathbf{0}, & \mathbf{f}^-(\mathbf{q}) &= \mathbf{f}(\mathbf{q}) & \text{if } \lambda_m \leq 0 & \text{for all } m = 1, \dots, M \end{aligned}$$

Then use

$$\mathbf{F}(\mathbf{q}_L, \mathbf{q}_R) = \mathbf{f}^+(\mathbf{q}_L) + \mathbf{f}^-(\mathbf{q}_R)$$



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Jacobians of the split fluxes are identical to $\mathbf{A}^\pm(\mathbf{q})$ only in linear case

$$\frac{\partial \mathbf{f}^\pm(\mathbf{q})}{\partial \mathbf{q}} = \frac{\partial (\mathbf{A}^\pm(\mathbf{q}) \mathbf{q})}{\partial \mathbf{q}} = \mathbf{A}^\pm(\mathbf{q}) + \frac{\partial \mathbf{A}^\pm(\mathbf{q})}{\partial \mathbf{q}} \mathbf{q}$$

Further methods: Van Leer FVS [Toro, 1999], AUSM [Wada and Liou, 1997]

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Objective: Higher-order accuracy in smooth solution regions but no spurious oscillations near large gradients

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TVD schemes: no new extrema, local minima are non-decreasing, local maxima are non-increasing (termed *monotonicity-preserving*). *Monotonicity-preserving* higher-order schemes are at least 5-point methods. Proofs: [Harten, 1983]

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TVD concept is proven [Godlewski and Raviart, 1996] for scalar schemes only but nevertheless used to construct *high resolution* schemes.

Monotonicity-preserving scheme can converge toward non-physical weak solutions.

MUSCL slope limiting

Monotone Upwind Schemes for Conservation Laws [van Leer, 1979]

$$\tilde{Q}_{j+\frac{1}{2}}^L = Q_j^n + \frac{1}{4} \left[(1 - \omega) \Phi_{j-\frac{1}{2}}^+ \Delta_{j-\frac{1}{2}} + (1 + \omega) \Phi_{j+\frac{1}{2}}^- \Delta_{j+\frac{1}{2}} \right],$$

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and *slope limiters*, e.g., *Minmod*

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Using a midpoint rule for temporal integration, e.g.,

$$Q_j^* = Q_j^n - \frac{1}{2} \frac{\Delta t}{\Delta x} \left(F(Q_{j+1}^n, Q_j^n) - F(Q_j^n, Q_{j-1}^n) \right)$$

and constructing limited values from Q^* to be used in FV scheme gives a TVD method if

$$\frac{1}{2} \left[(1 - \omega) \Phi(r) + (1 + \omega) r \Phi \left(\frac{1}{r} \right) \right] < \min(2, 2r)$$

is satisfied for $r > 0$. Proof: [Hirsch, 1988]

Wave Propagation with flux limiting

Wave Propagation Method [LeVeque, 1997] is built on the flux differencing approach $\mathcal{A}^\pm \Delta := \hat{\mathbf{A}}^\pm(\mathbf{q}_L, \mathbf{q}_R) \Delta \mathbf{q}$ and the waves $\mathcal{W}_m := a_m \hat{\mathbf{r}}_m$, i.e.

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with

$$\tilde{\mathbf{F}}_{j+\frac{1}{2}} = \frac{1}{2} |\mathcal{A}| \left(1 - \frac{\Delta t}{\Delta x} |\mathcal{A}| \right) \Delta_{j+\frac{1}{2}} = \frac{1}{2} \sum_{m=1}^M |\hat{\lambda}_{j+\frac{1}{2}}^m| \left(1 - \frac{\Delta t}{\Delta x} |\hat{\lambda}_{j+\frac{1}{2}}^m| \right) \tilde{\mathcal{W}}_{j+\frac{1}{2}}^m$$

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$$\mathbf{Q}^{n+1} = \mathbf{Q}_j^n - \frac{\Delta t}{\Delta x} \left(\mathcal{A}^- \Delta_{j+\frac{1}{2}} + \mathcal{A}^+ \Delta_{j-\frac{1}{2}} \right) - \frac{\Delta t}{\Delta x} \left(\tilde{\mathbf{F}}_{j+\frac{1}{2}} - \tilde{\mathbf{F}}_{j-\frac{1}{2}} \right)$$

with

$$\tilde{\mathbf{F}}_{j+\frac{1}{2}} = \frac{1}{2} |\mathcal{A}| \left(1 - \frac{\Delta t}{\Delta x} |\mathcal{A}| \right) \Delta_{j+\frac{1}{2}} = \frac{1}{2} \sum_{m=1}^M |\hat{\lambda}_{j+\frac{1}{2}}^m| \left(1 - \frac{\Delta t}{\Delta x} |\hat{\lambda}_{j+\frac{1}{2}}^m| \right) \tilde{\mathcal{W}}_{j+\frac{1}{2}}^m$$

and wave limiter

$$\tilde{\mathcal{W}}_{j+\frac{1}{2}}^m = \Phi(\Theta_{j+\frac{1}{2}}^m) \mathcal{W}_{j+\frac{1}{2}}^m$$

with

$$\Theta_{j+\frac{1}{2}}^m = \begin{cases} a_{j-\frac{1}{2}}^m / a_{j+\frac{1}{2}}^m, & \hat{\lambda}_{j+\frac{1}{2}}^m \geq 0, \\ a_{j+\frac{3}{2}}^m / a_{j+\frac{1}{2}}^m, & \hat{\lambda}_{j+\frac{1}{2}}^m < 0 \end{cases}$$

Wave Propagation Method in 2D

Writing $\tilde{\mathcal{A}}^{\pm} \Delta_{j \pm 1/2} := \mathcal{A}^+ \Delta_{j \pm 1/2} + \tilde{\mathbf{F}}_{j \pm 1/2}$ one can develop a truly two-dimensional one-step method [Langseth and LeVeque, 2000]

$$\begin{aligned} \mathbf{Q}_{jk}^{n+1} = \mathbf{Q}_{jk}^n &- \frac{\Delta t}{\Delta x_1} \left(\tilde{\mathcal{A}}^- \Delta_{j+\frac{1}{2},k} - \frac{1}{2} \frac{\Delta t}{\Delta x_2} \left[\mathcal{A}^- \tilde{\mathcal{B}}^- \Delta_{j+1,k+\frac{1}{2}} + \mathcal{A}^- \tilde{\mathcal{B}}^+ \Delta_{j+1,k-\frac{1}{2}} \right] + \right. \\ &\quad \left. \tilde{\mathcal{A}}^+ \Delta_{j-\frac{1}{2},k} - \frac{1}{2} \frac{\Delta t}{\Delta x_2} \left[\mathcal{A}^+ \tilde{\mathcal{B}}^- \Delta_{j-1,k+\frac{1}{2}} + \mathcal{A}^+ \tilde{\mathcal{B}}^+ \Delta_{j-1,k-\frac{1}{2}} \right] \right) \\ &- \frac{\Delta t}{\Delta x_2} \left(\tilde{\mathcal{B}}^- \Delta_{j,k+\frac{1}{2}} - \frac{1}{2} \frac{\Delta t}{\Delta x_1} \left[\mathcal{B}^- \tilde{\mathcal{A}}^- \Delta_{j+\frac{1}{2},k+1} + \mathcal{B}^- \tilde{\mathcal{A}}^+ \Delta_{j-\frac{1}{2},k+1} \right] + \right. \\ &\quad \left. \tilde{\mathcal{B}}^+ \Delta_{j,k-\frac{1}{2}} - \frac{1}{2} \frac{\Delta t}{\Delta x_1} \left[\mathcal{B}^+ \tilde{\mathcal{A}}^- \Delta_{j+\frac{1}{2},k-1} + \mathcal{B}^+ \tilde{\mathcal{A}}^+ \Delta_{j-\frac{1}{2},k-1} \right] \right) \end{aligned}$$

that is stable for

$$\left\{ \max_{j \in \mathbb{Z}} |\hat{\lambda}_{m,j+\frac{1}{2}}| \frac{\Delta t}{\Delta x_1}, \max_{k \in \mathbb{Z}} |\hat{\lambda}_{m,k+\frac{1}{2}}| \frac{\Delta t}{\Delta x_2} \right\} \leq 1, \quad \text{for all } m = 1, \dots, M$$

Further high-resolution methods

Some further high-resolution methods (good overview in [Laney, 1998]):

- ▶ FCT: 2nd order [Oran and Boris, 2001]

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3rd order methods must make use of strong-stability preserving Runge-Kutta methods [Gottlieb et al., 2001] for time integration that use a multi-step update

$$\tilde{\mathbf{Q}}_j^v = \alpha_v \mathbf{Q}_j^n + \beta_v \tilde{\mathbf{Q}}_j^{v-1} + \gamma_v \frac{\Delta t}{\Delta x} \left(\mathbf{F}_{j+\frac{1}{2}}(\tilde{\mathbf{Q}}^{v-1}) - \mathbf{F}_{j-\frac{1}{2}}(\tilde{\mathbf{Q}}^{v-1}) \right)$$

with $\tilde{\mathbf{Q}}^0 := \mathbf{Q}^n$, $\alpha_1 = 1$, $\beta_1 = 0$; and $\mathbf{Q}^{n+1} := \tilde{\mathbf{Q}}^\Upsilon$ after final stage Υ

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Typical storage-efficient SSPRK(3,3):

$$\begin{aligned} \tilde{\mathbf{Q}}^1 &= \mathbf{Q}^n + \Delta t \mathcal{F}(\mathbf{Q}^n), & \tilde{\mathbf{Q}}^2 &= \frac{3}{4} \mathbf{Q}^n + \frac{1}{4} \tilde{\mathbf{Q}}^1 + \frac{1}{4} \Delta t \mathcal{F}(\tilde{\mathbf{Q}}^1), \\ \mathbf{Q}^{n+1} &= \frac{1}{3} \mathbf{Q}^n + \frac{2}{3} \tilde{\mathbf{Q}}^2 + \frac{2}{3} \Delta t \mathcal{F}(\tilde{\mathbf{Q}}^2) \end{aligned}$$

Outline

Conservation laws

- Mathematical background
- Examples

Finite volume methods

- Basics of finite difference methods
- Splitting methods, second derivatives

Upwind schemes

- Flux-difference splitting
- Flux-vector splitting
- High-resolution methods

Meshes and adaptation

- Elements of adaptive algorithms
- Adaptivity on unstructured meshes
- Structured mesh refinement techniques

Elements of adaptive algorithms

- ▶ Base grid

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- ▶ Solver

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Elements of adaptive algorithms

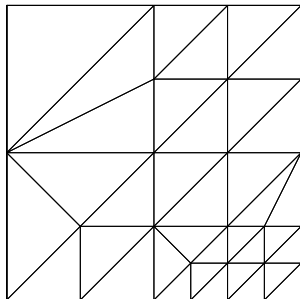
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Elements of adaptive algorithms

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- ▶ Interpolation (restriction and prolongation)
- ▶ Load-balancing

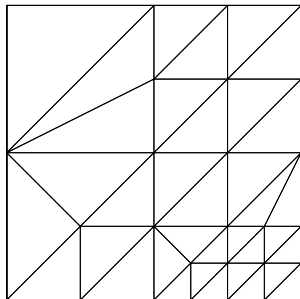
Adaptivity on unstructured meshes

- Coarse cells replaced by finer ones



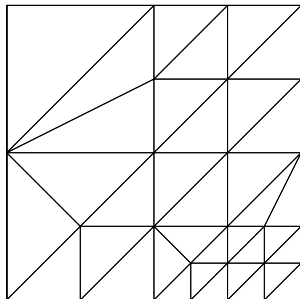
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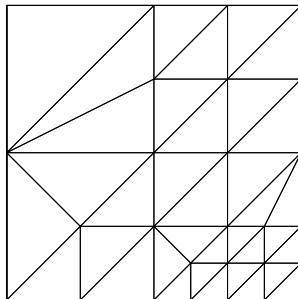
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- ▶ Coarse cells replaced by finer ones
- ▶ Global time-step
- ▶ Cell-based data structures



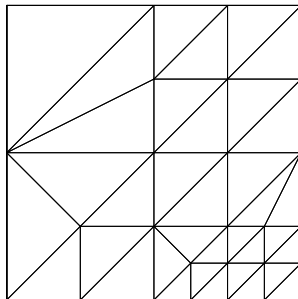
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- ▶ Neighborhoods have to stored



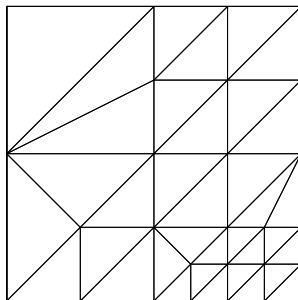
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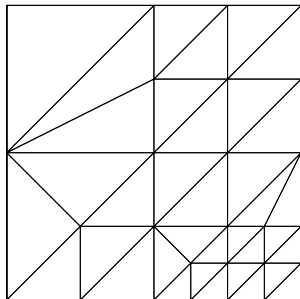
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- + **No hanging nodes**



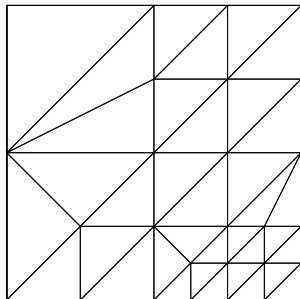
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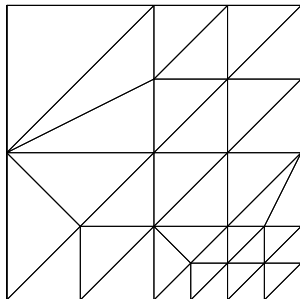
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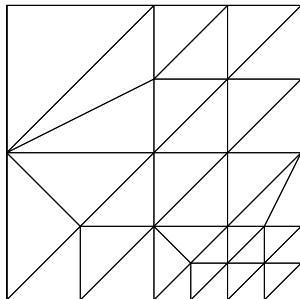
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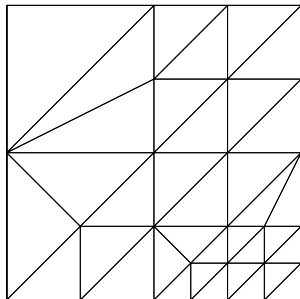
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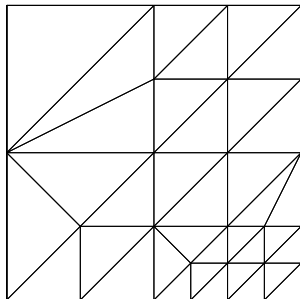
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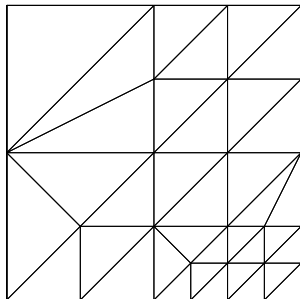
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Structured mesh refinement techniques

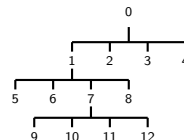
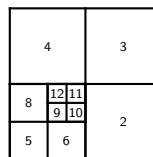
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Structured mesh refinement techniques

- ▶ Block-based data of equal size
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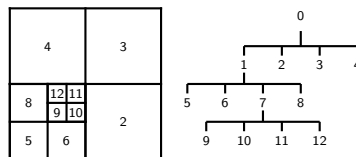
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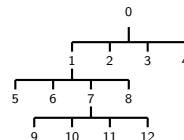
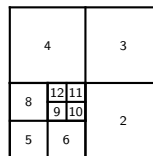
Structured mesh refinement techniques

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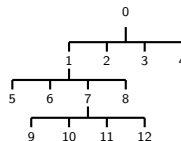
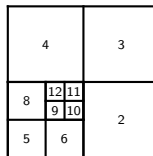
Structured mesh refinement techniques

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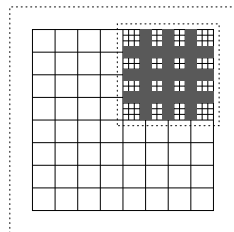
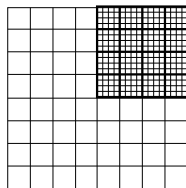
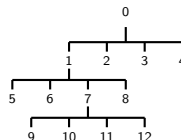
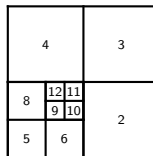
Structured mesh refinement techniques

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Structured mesh refinement techniques

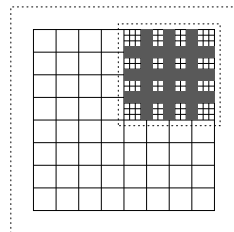
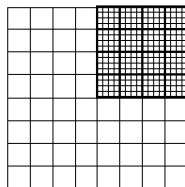
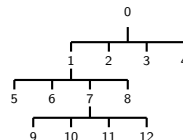
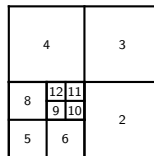
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Wasted boundary space in a quad-tree

Structured mesh refinement techniques

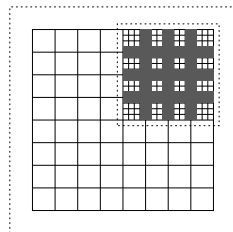
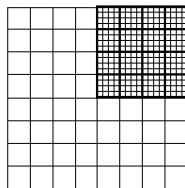
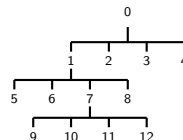
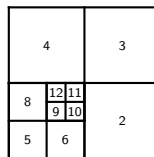
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Wasted boundary space in a quad-tree

Structured mesh refinement techniques

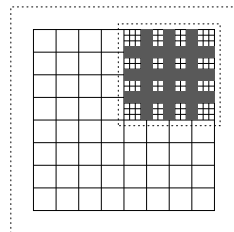
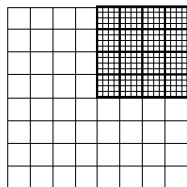
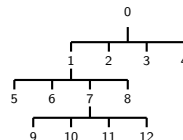
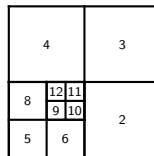
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Structured mesh refinement techniques

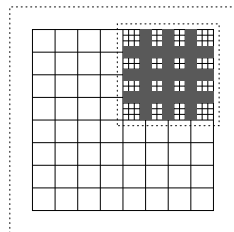
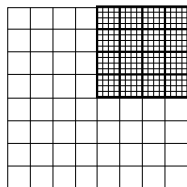
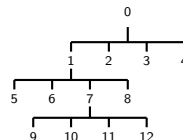
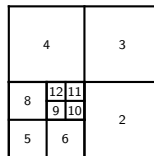
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Wasted boundary space in a quad-tree

Structured mesh refinement techniques

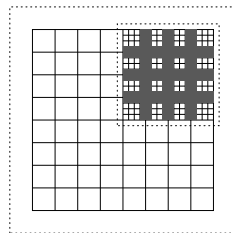
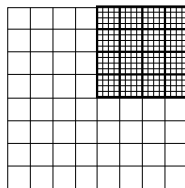
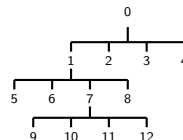
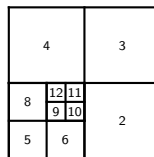
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Wasted boundary space in a quad-tree

Structured mesh refinement techniques

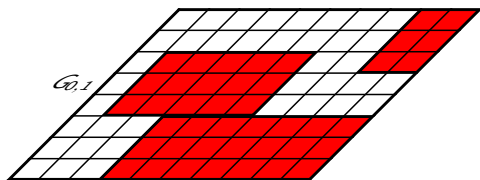
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Wasted boundary space in a quad-tree

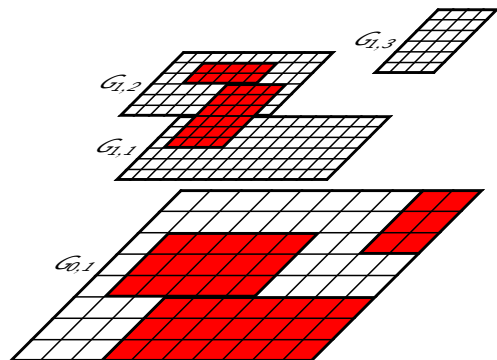
Block-structured adaptive mesh refinement (SAMR)

- Refined block overlay coarser ones



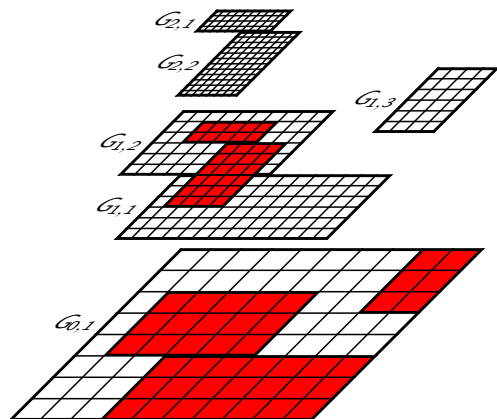
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- Refined block overlay coarser ones



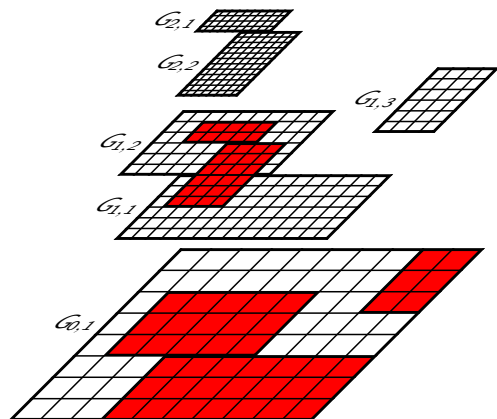
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones



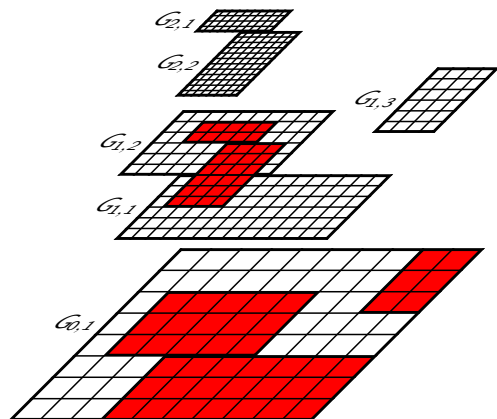
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement



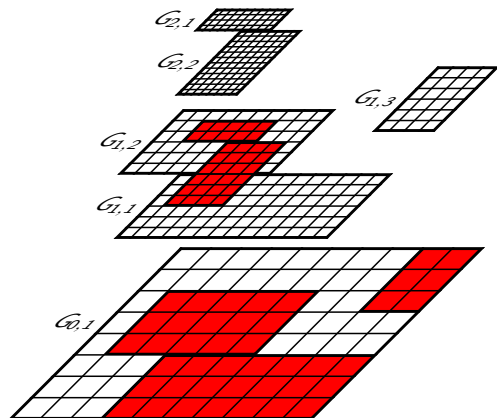
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures



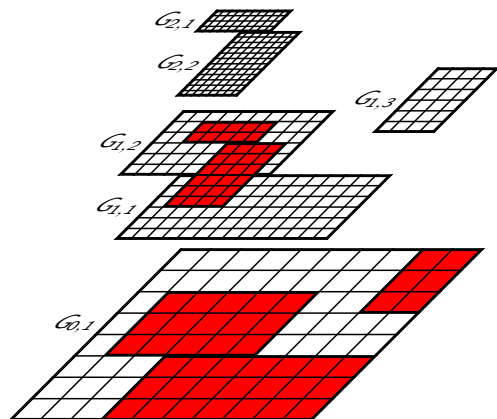
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system



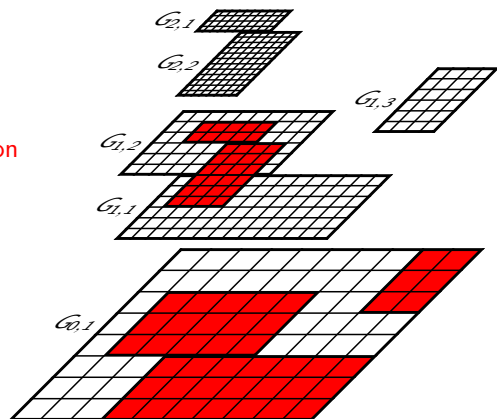
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary



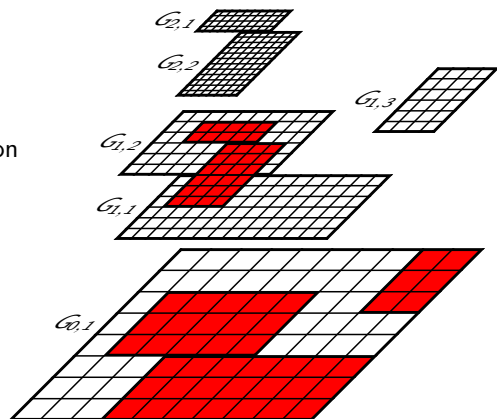
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary
- + Efficient cache-reuse / vectorization possible



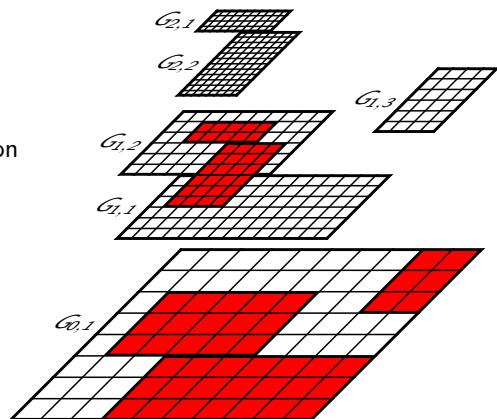
Block-structured adaptive mesh refinement (SAMR)

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- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary
- + Efficient cache-reuse / vectorization possible
- + Simple load-balancing



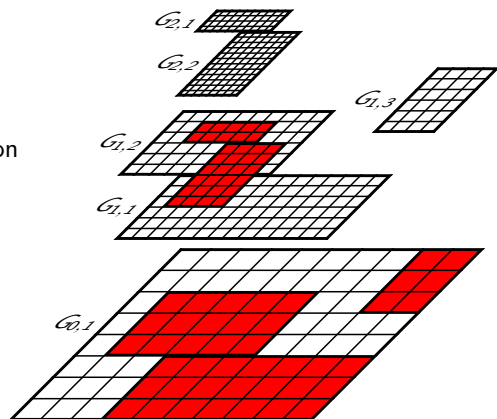
Block-structured adaptive mesh refinement (SAMR)

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- ▶ Global index coordinate system
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- + Efficient cache-reuse / vectorization possible
- + Simple load-balancing
- + Minimal synchronization overhead



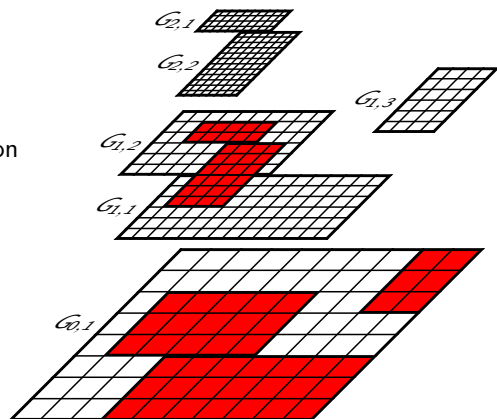
Block-structured adaptive mesh refinement (SAMR)

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- + Simple load-balancing
- + Minimal synchronization overhead
- Cells without mark are refined



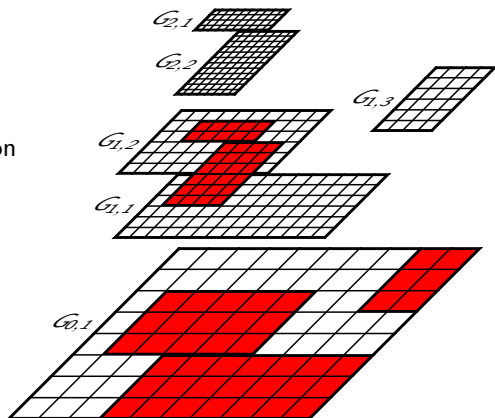
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary
- + Efficient cache-reuse / vectorization possible
- + Simple load-balancing
- + Minimal synchronization overhead
 - Cells without mark are refined
 - **Hanging nodes unavoidable**



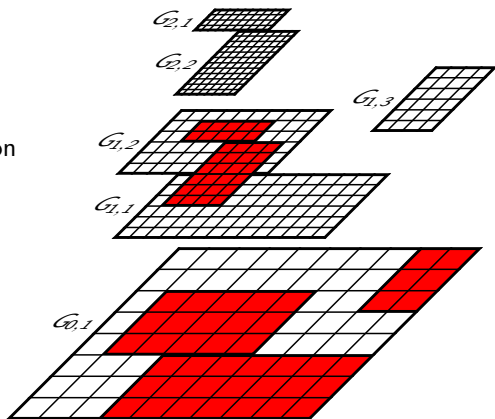
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary
- + Efficient cache-reuse / vectorization possible
- + Simple load-balancing
- + Minimal synchronization overhead
 - Cells without mark are refined
 - Hanging nodes unavoidable
 - Cluster-algorithm necessary



Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined block overlay coarser ones
- ▶ Time-step refinement
- ▶ Block (aka patch) based data structures
- ▶ Global index coordinate system
- + Numerical scheme only for single patch necessary
- + Efficient cache-reuse / vectorization possible
- + Simple load-balancing
- + Minimal synchronization overhead
 - Cells without mark are refined
 - Hanging nodes unavoidable
 - Cluster-algorithm necessary
 - **Difficult to implement**



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