# Stable high order methods for time-domain wave propagation in complex geometries and heterogeneous media

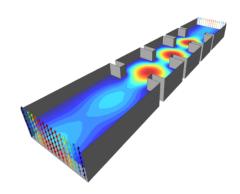
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## Numerical simulation of wave propagation

Many procedures requires **accurately** and **efficiently** solving time-dependent wave equations in realistic settings.

- Imaging (seismic, medical)
- Engineering design (scattering, design)
- Computational physics (aeroacoustics, astrophysics)



- Unstructured (tetrahedral) meshes for geometric flexibility.
- High order: low numerical dissipation and dispersion.
- High order approximations: more accurate per unknown
- Explicit time stepping: high performance on many-core.

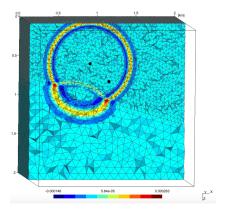
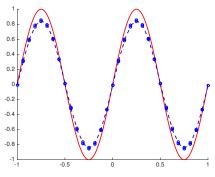


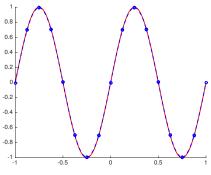
Figure courtesy of Axel Modave.

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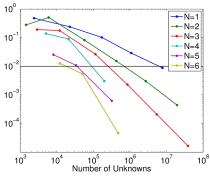
Fine linear approximation.

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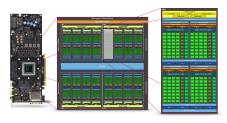
Coarse quadratic approximation.

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Max errors vs. dofs.

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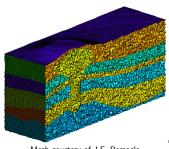
Graphics processing units (GPU).

## Assume $u(\mathbf{x},t) = \sum \mathbf{u}_j \phi_j(\mathbf{x})$ on $D^k$

- Compute numerical flux at face nodes (non-local).
- Compute RHS of (local) ODE.
- Evolve (local) solution using explicitime integration (RK, AB, etc).

$$\frac{\partial u}{\partial t} = \frac{\partial u}{\partial x}$$

Example: advection equation.



Mesh courtesy of J.F. Remacle

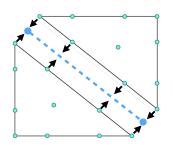
$$\mathbf{M}_{ij} = \int_{D^k} \phi_j(\mathbf{x}) \phi_i(\mathbf{x})$$

$$\mathbf{L}_f = \mathbf{M}^{-1} \mathbf{M}_f.$$

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$$\frac{\mathrm{d}\mathbf{u}}{\mathrm{d}t} = \mathbf{D}_{x}\mathbf{u} + \sum_{\mathrm{faces}} \mathbf{L}_{f} \left( \mathrm{flux} \right).$$

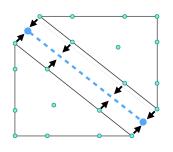


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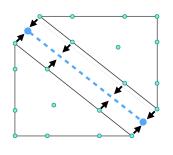
$$\frac{\mathrm{d}\mathbf{u}}{\mathrm{d}t} = \underbrace{\mathbf{D}_{x}\mathbf{u}}_{\text{Volume kernel}} + \underbrace{\sum_{\text{faces}} \mathbf{L}_{f} \left(\text{flux}\right)}_{\text{Surface kernel}}.$$



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$$\underbrace{\frac{\mathrm{d}\mathbf{u}}{\mathrm{d}t}}_{\text{Update kernel}} = \underbrace{\mathbf{D}_{\mathbf{x}}\mathbf{u}}_{\text{Volume kernel}} + \underbrace{\sum_{\text{faces}} \mathbf{L}_f\left(\text{flux}\right)}_{\text{Surface kernel}}.$$

$$\mathbf{M}_{ij} = \int_{D^k} \phi_j(\mathbf{x}) \phi_i(\mathbf{x})$$
  
 $\mathbf{L}_f = \mathbf{M}^{-1} \mathbf{M}_f.$ 

#### Outline

Weight-adjusted DG (WADG): high order heterogeneous media

Elastic-acoustic coupled media

3 Bernstein-Bezier WADG: high order efficiency

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## Energy stable discontinuous Galerkin formulations

■ Model problem: acoustic wave equation (pressure-velocity system)

$$\frac{1}{c^2} \frac{\partial p}{\partial t} = \nabla \cdot \boldsymbol{u}, \qquad \frac{\partial \boldsymbol{u}}{\partial t} = \nabla p$$

■ Local formulation

$$\int_{D^k} \frac{1}{c^2} \frac{\partial p}{\partial t} q = \int_{D^k} \nabla \cdot \boldsymbol{u} q + \frac{1}{2} \int_{\partial D^k} (\llbracket \boldsymbol{u} \rrbracket \cdot \boldsymbol{n} + \tau_p \llbracket p \rrbracket) q$$

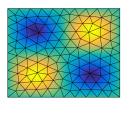
$$\int_{D^k} \frac{\partial \boldsymbol{u}}{\partial t} \boldsymbol{v} = \int_{D^k} \nabla p \cdot \boldsymbol{v} + \frac{1}{2} \int_{\partial D^k} (\llbracket p \rrbracket + \tau_u \llbracket \boldsymbol{u} \rrbracket \cdot \boldsymbol{n}) \boldsymbol{v}$$

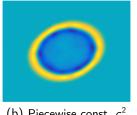
■ High order accuracy, semi-discrete energy stability

$$\frac{\partial}{\partial t} \left( \sum_{k} \int_{D^{k}} \frac{p^{2}}{c^{2}} + |\boldsymbol{u}|^{2} \right) = -\sum_{k} \int_{\partial D^{k}} \tau_{p} \left[\!\!\left[ \boldsymbol{p} \right]\!\!\right]^{2} + \tau_{u} \left[\!\!\left[ \boldsymbol{u} \cdot \boldsymbol{n} \right]\!\!\right]^{2} \leq 0.$$

Guo (CAAM)

## High order approximation of smoothly varying media







- (a) Mesh and exact  $c^2$  (b) Piecewise const.  $c^2$  (c) High order  $c^2$

■ Piecewise const.  $c^2$ : energy stable and efficient, but inaccurate.

$$\frac{1}{c^2(\mathbf{x})}\frac{\partial p}{\partial t} + \nabla \cdot \mathbf{u} = 0, \qquad \frac{\partial \mathbf{u}}{\partial t} + \nabla p = 0.$$

■ High order wavespeeds: weighted mass matrices. Stable, but expensive (pre-computation + storage of matrix inverses)!

$$oldsymbol{M}_{1/c^2}rac{\mathrm{d}oldsymbol{p}}{\mathrm{d}t}=oldsymbol{A}_holdsymbol{U}, \qquad ig(oldsymbol{M}_{1/c^2}ig)_{ij}=\int_{D^k}rac{1}{c^2(oldsymbol{x})}\phi_j(oldsymbol{x})\phi_i(oldsymbol{x}).$$

Guo (CAAM)

## Weight-adjusted DG (WADG)

lacktriangle Weight-adjusted DG: provably energy stable approx. of  $\emph{\textbf{M}}_{1/c^2}$ 

$$\mathbf{M}_{1/c^2} rac{\mathrm{d} \mathbf{p}}{\mathrm{d} t} pprox \mathbf{M} \left(\mathbf{M}_{c^2}
ight)^{-1} \mathbf{M} rac{\mathrm{d} \mathbf{p}}{\mathrm{d} t} = \mathbf{A}_h \mathbf{U}.$$

New evaluation reuses implementation for constant wavespeed

$$\frac{\mathrm{d} oldsymbol{p}}{\mathrm{d} t} = \underbrace{oldsymbol{\mathcal{M}}^{-1}(oldsymbol{\mathcal{M}}_{c^2})}_{\mathrm{modified update}} \quad \underbrace{oldsymbol{\mathcal{M}}^{-1}oldsymbol{\mathcal{A}}_h oldsymbol{U}}_{\mathrm{constant wavespeed RHS}}$$

■ Low storage matrix-free application of  $M^{-1}M_{c^2}$  using quadrature-based interpolation and  $L^2$  projection matrices  $V_q$ ,  $P_q$ .

$$(\boldsymbol{M})^{-1} \boldsymbol{M}_{c^2} = \underbrace{\boldsymbol{M}^{-1} \boldsymbol{V}_q^T \boldsymbol{W}}_{\boldsymbol{P}_q} \operatorname{diag}(c^2) \boldsymbol{V}_q.$$

Chan, Hewett, Warburton (2017). Weight-adjusted DG methods: wave propagation in heterogeneous media.

### WADG: nearly identical to DG w/weighted mass matrices

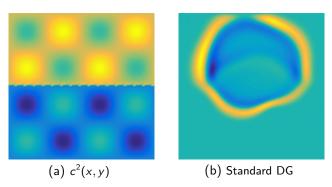


Figure: Standard vs. weight-adjusted DG with spatially varying  $c^2$ .

■ The  $L^2$  error is  $O(h^{N+1})$ , but the difference between the DG and WADG solutions is  $O(h^{N+2})$ !

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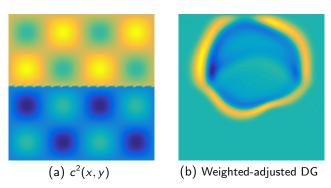
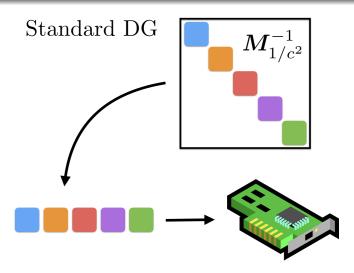


Figure: Standard vs. weight-adjusted DG with spatially varying  $c^2$ .

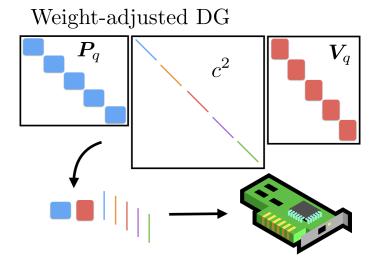
■ The  $L^2$  error is  $O(h^{N+1})$ , but the difference between the DG and WADG solutions is  $O(h^{N+2})$ !

# WADG: more efficient than storing $M_{1/c^2}^{-1}$ on GPUs



Efficiency on GPUs: reduce memory accesses and data movement!

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1 Weight-adjusted DG (WADG): high order heterogeneous media

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# Matrix-valued weights and elastic wave propagation

■ Symmetric velocity-stress formulation (entries of  $\boldsymbol{A}_i$  are  $\pm 1$  or 0)

$$\rho \frac{\partial \mathbf{v}}{\partial t} = \sum_{i=1}^d \mathbf{A}_i^T \frac{\partial \mathbf{\sigma}}{\partial \mathbf{x}_i}, \qquad \mathbf{C}^{-1} \frac{\partial \mathbf{\sigma}}{\partial t} = \sum_{i=1}^d \mathbf{A}_i \frac{\partial \mathbf{v}}{\partial \mathbf{x}_i}.$$

■ DG formulation: *simple* penalty fluxes, matrix-weighted mass matrix

■ Weight-adjusted approx. to  $(M_{C^{-1}})^{-1}$  decouples each component

$$extbf{M}_{ extbf{C}^{-1}}^{-1}pprox \left( extbf{I}\otimes extbf{M}^{-1}
ight) extbf{M}_{ extbf{C}}\left( extbf{I}\otimes extbf{M}^{-1}
ight).$$

#### Simple to incorporate anisotropic media

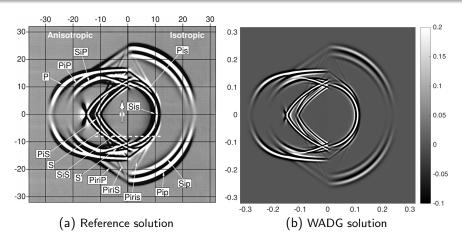


Figure: Anisotropic media simply involves modifying the definition of  $\boldsymbol{C}$ .

Komatitsch, Barnes, Tromp (2000). Simulation of anisotropic wave propagation based upon a spectral element method. Chan (2018). Weight-adjusted DG methods: matrix-valued weights and elastic wave prop. in heterogeneous media.

$$oldsymbol{\sigma}, oldsymbol{v}$$
 (Elastic)

$$oldsymbol{u} \cdot oldsymbol{n} = oldsymbol{v} \cdot oldsymbol{n} \ oldsymbol{A}_n^T oldsymbol{\sigma} = p oldsymbol{n}$$

 $p, \boldsymbol{u}$  (Acoustic)

$$(\mathfrak{F}\boldsymbol{q})^* = \mathfrak{F}^-\boldsymbol{q}^- + \frac{\boldsymbol{n} \cdot \llbracket \boldsymbol{S} \rrbracket + \rho^+ c_p^+ \llbracket \boldsymbol{\upsilon} \rrbracket}{\rho^+ c_p^+ + \rho^- c_p^-} \begin{pmatrix} \boldsymbol{n} \otimes \boldsymbol{n} \\ \rho^- c_p^- \boldsymbol{n} \end{pmatrix}.$$

$$\begin{split} (\mathfrak{F}\mathbf{q})^* &= \mathfrak{F}^-\mathbf{q}^- + \frac{c_p^-c_p^+\mathbf{n} \cdot \llbracket \mathbf{S} \rrbracket + c_p^-(\lambda^+ + 2\mu^+) \llbracket \boldsymbol{v} \rrbracket}{c_p^+(\lambda^+ + 2\mu^-) + c_p^-(\lambda^+ + 2\mu^+)} \begin{pmatrix} \mathbf{n} \otimes \mathbf{n} \\ \rho^-c_p^-\mathbf{n} \end{pmatrix} + \begin{pmatrix} \frac{c_r^-c_s^+}{\mu^+c_s^- + \mu^-c_s^+} \mathbf{s} \cdot \llbracket \mathbf{S} \rrbracket + \frac{c_r^-\mu^+}{\mu^+c_s^- + \mu^-c_s^+} \mathbf{s} \cdot \llbracket \boldsymbol{v} \rrbracket \end{pmatrix} \begin{pmatrix} \operatorname{sym}(\mathbf{s} \otimes \mathbf{n}) \\ \rho^-c_s^-\mathbf{s} \end{pmatrix} \\ &+ \begin{pmatrix} \frac{c_s^-c_s^+}{\mu^+c_s^- + \mu^-c_s^+} \mathbf{t} \cdot \llbracket \mathbf{S} \rrbracket + \frac{c_r^-\mu^+}{\mu^+c_s^- + \mu^-c_s^+} \mathbf{t} \cdot \llbracket \boldsymbol{v} \rrbracket \end{pmatrix} \begin{pmatrix} \operatorname{sym}(\mathbf{s} \otimes \mathbf{n}) \\ \rho^-c_s^-\mathbf{t} \end{pmatrix} = \mathfrak{F}^-\mathbf{q}^- + \frac{c_p^-c_p^+\mathbf{n} \cdot \llbracket \mathbf{S} \rrbracket + c_p^-(\lambda^+ + 2\mu^+) \llbracket \boldsymbol{v} \rrbracket}{c_p^+(\lambda^+ + 2\mu^+) \llbracket \boldsymbol{v} \rrbracket} \begin{pmatrix} \mathbf{n} \otimes \mathbf{n} \\ \rho^-c_p^-\mathbf{n} \end{pmatrix} \\ &- \frac{c_s^-c_s^+}{\mu^+c_s^- + \mu^-c_s^+} \begin{pmatrix} \operatorname{sym}(\mathbf{n} \otimes (\mathbf{n} \times (\mathbf{n} \times \llbracket \mathbf{S} \rrbracket))) \\ \rho^-c_s^-\mathbf{n} \times (\mathbf{n} \times \llbracket \mathbf{S} \rrbracket) \end{pmatrix} - \frac{c_s^-\mu^+}{\mu^+c_s^- + \mu^-c_s^+} \begin{pmatrix} \operatorname{sym}(\mathbf{n} \otimes (\mathbf{n} \times (\mathbf{n} \times \llbracket \mathbf{v} \rrbracket))) \\ \rho^-c_s^-\mathbf{n} \times (\mathbf{n} \times \llbracket \mathbf{v} \rrbracket) \end{pmatrix}, \end{split}$$

$$(\mathfrak{F}\boldsymbol{q})^* = \mathfrak{F}^-\boldsymbol{q}^- + \frac{\boldsymbol{n} \cdot \llbracket \boldsymbol{S} \rrbracket + \rho^+ c_p^+ \llbracket \boldsymbol{v} \rrbracket}{\rho^+ c_p^+ + \rho^- c_p^-} \begin{pmatrix} \boldsymbol{n} \otimes \boldsymbol{n} \\ \rho^- c_p^- \boldsymbol{n} \end{pmatrix} - \frac{1}{\rho^- c_s^-} \begin{pmatrix} \operatorname{sym}(\boldsymbol{n} \otimes (\boldsymbol{n} \times (\boldsymbol{n} \times \llbracket \boldsymbol{S} \rrbracket))) \\ \rho^- c_s^- \boldsymbol{n} \times (\boldsymbol{n} \times \llbracket \boldsymbol{S} \rrbracket) \end{pmatrix}.$$

- Traditional upwind acoustic-elastic fluxes are complex to derive.
- Cannot prove energy stability in the case of heterogeneous media.

Wilcox, Stadler, Burstedde, Ghattas (2010). A high-order discontinuous Galerkin method for wave propagation through coupled elastic-acoustic media.

$$A_{n} = A_{1}n_{x} + A_{2}n_{y} + A_{3}n_{z} \qquad \text{(Elastic)}$$

$$\frac{1}{2} \left( \mathbf{A}_{n}^{T}(\boldsymbol{\sigma}^{+} - \boldsymbol{\sigma}) + \tau_{v} \mathbf{A}_{n}^{T} \mathbf{A}_{n}(\boldsymbol{v}^{+} - \boldsymbol{v}) \right)$$

$$\frac{1}{2} \left( \mathbf{A}_{n}(\boldsymbol{v}^{+} - \boldsymbol{v}) + \tau_{\sigma} \mathbf{A}_{n} \mathbf{A}_{n}^{T}(\boldsymbol{\sigma}^{+} - \boldsymbol{\sigma}) \cdot \boldsymbol{n} \right) \boldsymbol{n}$$

$$\frac{1}{2} \left( (\mathbf{u}^{+} - \mathbf{u}) \cdot \mathbf{n} + \tau_{p} (p^{+} - p) \right)$$

$$\frac{1}{2} \left( (p^{+} - p) + \tau_{\mathbf{u}} (\mathbf{u}^{+} - \mathbf{u}) \cdot \mathbf{n} \right) \mathbf{n}$$
(Acoustic)

Guo (CAAM)

(Elastic)
$$\frac{1}{2} A_{n} \left( \boldsymbol{n} \boldsymbol{n}^{T} (\boldsymbol{u} - \boldsymbol{v}) + \tau_{\boldsymbol{\sigma}} (\boldsymbol{p} \boldsymbol{n} - \boldsymbol{A}_{n}^{T} \boldsymbol{\sigma}) \right)$$

$$\frac{1}{2} \boldsymbol{n}^{T} \left( \boldsymbol{p} \boldsymbol{n} - \boldsymbol{A}_{n}^{T} \boldsymbol{\sigma} + (\boldsymbol{I} - \boldsymbol{n} \boldsymbol{n}^{T}) \boldsymbol{A}_{n}^{T} \boldsymbol{\sigma} + \tau_{\boldsymbol{v}} (\boldsymbol{u} - \boldsymbol{v}) \right)$$

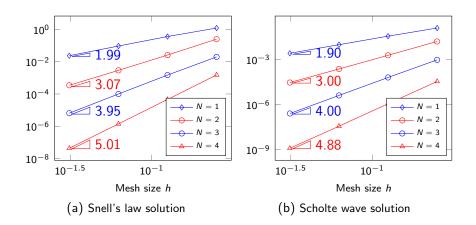
$$\boldsymbol{u} \cdot \boldsymbol{n} = \boldsymbol{v} \cdot \boldsymbol{n}$$

$$\boldsymbol{A}_{n}^{T} \boldsymbol{\sigma} = \boldsymbol{p} \boldsymbol{n}$$

$$\frac{1}{2} \boldsymbol{n}^{T} \left( \boldsymbol{v} - \boldsymbol{u} + \tau_{\boldsymbol{p}} (\boldsymbol{A}_{n}^{T} \boldsymbol{\sigma} - \boldsymbol{p} \boldsymbol{n}) \right)$$

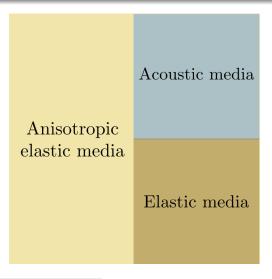
$$\frac{1}{2} \boldsymbol{n} \boldsymbol{n}^{T} \left( \boldsymbol{A}_{n}^{T} \boldsymbol{\sigma} - \boldsymbol{p} \boldsymbol{n} + \tau_{\boldsymbol{u}} (\boldsymbol{v} - \boldsymbol{u}) \right)$$
(Acoustic)

## Numerical results: coupled acoustic-elastic media



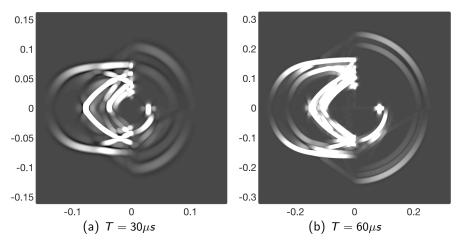
High order convergence of  $L^2$  error for acoustic-elastic media.

## Example with isotropic-anisotropic acoustic-elastic media



Komatitsch, Barnes, Tromp (2000). Simulation of anisotropic wave propagation based upon a spectral element method. Guo. Acosta. Chan (2019). A weight-adjusted DG method for wave propagation in coupled elastic-acoustic media.

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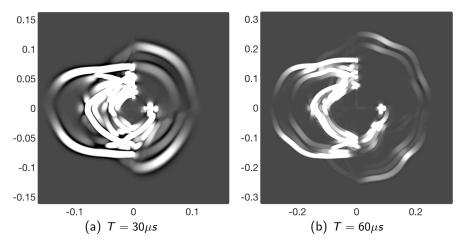


Piecewise constant anisotropic-isotropic acoustic-elastic media.

Komatitsch, Barnes, Tromp (2000). Simulation of anisotropic wave propagation based upon a spectral element method.

Guo. Acosta. Chan (2019). A weight-adjusted DG method for wave propagation in coupled elastic-acoustic media.

## Example with isotropic-anisotropic acoustic-elastic media



Piecewise smoothly varying anisotropic-isotropic acoustic-elastic media.

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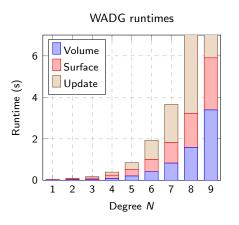
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#### Computational costs at high orders of approximation

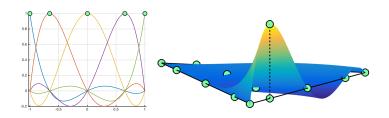
Problem: WADG at high orders becomes **expensive**!



- Large **dense** matrices:  $O(N^6)$  work per element.
- Idea: choose basis such that matrices are sparse.

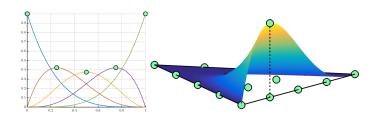
WADG runtimes for 50 timesteps, 98304 elements.

- Nodal DG:  $O(N^6)$  cost in 3D vs  $O(N^3)$  degrees of freedom.
- Switch to Bernstein basis: sparse and structured matrices.
- lacksquare Optimal  $O(N^3)$  application of differentiation and lifting matrices



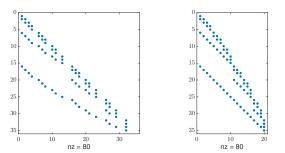
Nodal bases in one, two, and three dimensions.

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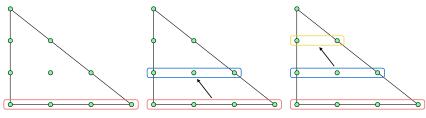
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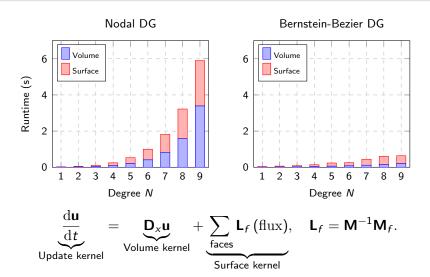
Tetrahedral Bernstein differentiation and degree elevation matrices.

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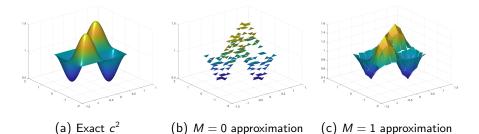


Optimal  $O(N^3)$  complexity "slice-by-slice" application of Bernstein lift.

#### BBDG: efficient volume, surface kernels



## A faster BBWADG update kernel

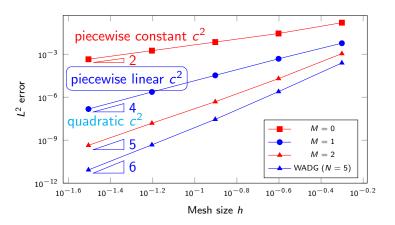


**Exploit** continuous WADG steps: given u(x), compute

$$P_N\left(u(\mathbf{x})c^2(\mathbf{x})\right), \qquad P_N=L^2 \text{ projection operator.}$$

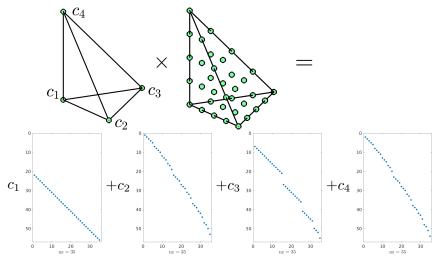
- Our approach: approx.  $c^2(x)$  with degree M polynomial, use fast Bernstein algorithms for polynomial multiplication and projection.
- Can reuse fast  $O(N^3)$  Bernstein-based volume and surface kernels.

# BBWADG: effect of approximating $c^2$ on accuracy



Approximating smooth  $c^2(x)$  using  $L^2$  projection:  $O(h^2)$  for M=0,  $O(h^4)$  for M=1,  $O(h^{M+3})$  for  $0 < M \le N-2$ .

## Fast Bernstein polynomial multiplication



Bernstein polynomial multiplication (M = 1 shown),  $O(N^3)$  cost for fixed M.

Guo (CAAM)

## Fast Bernstein polynomial projection

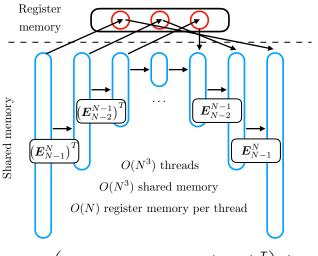
- Given  $c^2(x)u(x)$  as a degree (N+M) polynomial, apply  $L^2$  projection matrix  $P_N^{N+M}$  to reduce to degree N.
- Polynomial  $L^2$  projection matrix  $P_N^{N+M}$  under Bernstein basis:

$$\boldsymbol{P}_{N}^{N+M} = \underbrace{\sum_{j=0}^{N} c_{j} \boldsymbol{E}_{N-j}^{N} \left(\boldsymbol{E}_{N-j}^{N}\right)^{T}}_{\widetilde{\boldsymbol{P}}_{N}} \left(\boldsymbol{E}_{N}^{N+M}\right)^{T}$$

• "Telescoping" form of  $\tilde{P}_N$ :  $O(N^4)$  complexity, more GPU-friendly.

$$\left(c_0 \mathbf{I} + \mathbf{E}_{N-1}^{N} \left(c_1 \mathbf{I} + \mathbf{E}_{N-2}^{N-1} \left(c_2 \mathbf{I} + \cdots\right) \left(\mathbf{E}_{N-2}^{N-1}\right)^{T}\right) \left(\mathbf{E}_{N-1}^{N}\right)^{T}\right)$$

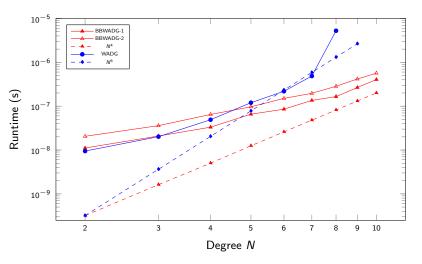
# Sketch of GPU algorithm for $\tilde{P}_N$



$$\left(c_0 \boldsymbol{\mathit{I}} + \boldsymbol{\mathit{E}}_{N-1}^{N} \left(c_1 \boldsymbol{\mathit{I}} + \boldsymbol{\mathit{E}}_{N-2}^{N-1} \left(c_2 \boldsymbol{\mathit{I}} + \cdots\right) \left(\boldsymbol{\mathit{E}}_{N-2}^{N-1}\right)^T\right) \left(\boldsymbol{\mathit{E}}_{N-1}^{N}\right)^T\right)$$

Guo (CAAM)

# BBWADG: computational runtime (3D acoustics)



Per-element runtimes of update kernels for BBWADG vs WADG (acoustic). We observe an asymptotic complexity of  $O(N^4)$  per element for  $N \gg 1$ .

## Summary and future work

- Weight-adjusted DG: high order accuracy, provable stability, and efficiency in heterogeneous acoustic-elastic media.
- Current work: stable WADG for moving curved meshes (*r*-adaptivity) and extension to nonlinear conservation laws.
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