

# Weight-adjusted Bernstein-Bezier DG methods for wave propagation in heterogeneous media

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# High order DG methods for wave propagation

- 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.

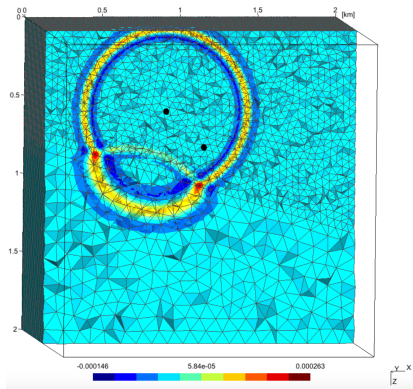
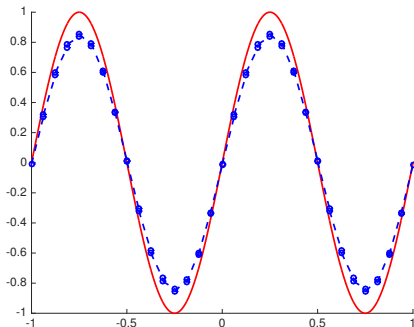


Figure courtesy of Axel Modave.

Goal: accuracy **and** efficiency for heterogeneous media.

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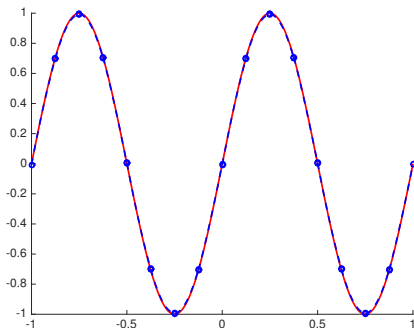


**Fine** linear approximation.

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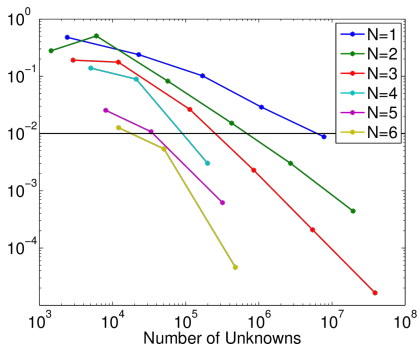


**Coarse** quadratic approximation.

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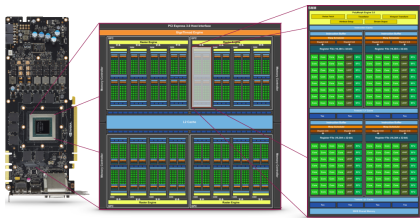


Max errors vs. dofs.

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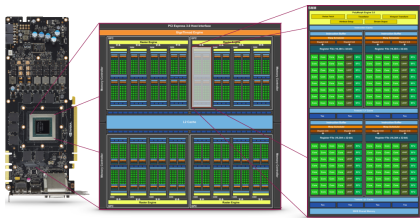


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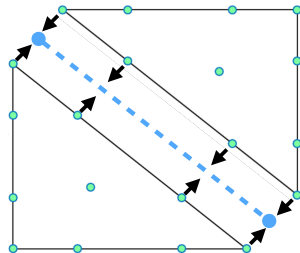
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# Time-domain nodal DG methods

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 explicit time integration (RK, AB, etc).



$$\frac{d\mathbf{u}}{dt} = \mathbf{D}_x \mathbf{u} + \sum_{\text{faces}} \mathbf{L}_f (\text{flux}).$$

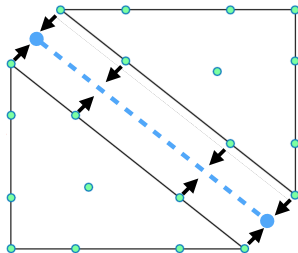
$$\mathbf{M}_{ij} = \int_{D^k} \phi_j(\mathbf{x}) \phi_i(\mathbf{x})$$
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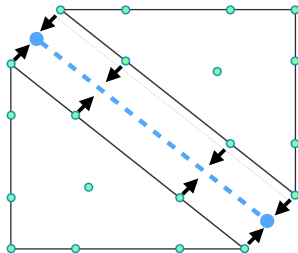
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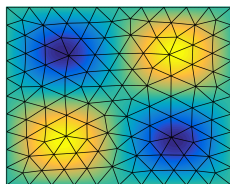
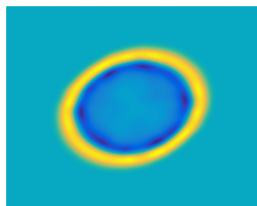
# Outline

- 1 Weight-adjusted DG (WADG): arbitrary heterogeneous media
- 2 Bernstein-Bezier WADG: high order efficiency

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# High order approximation of media and geometry

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

- Piecewise constant wavespeed  $c^2$ : efficient, but spurious reflections.

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

- High order wavespeeds: weighted mass matrices. Stable, but requires pre-computation/storage of inverses or factorizations!

$$\mathbf{M}_{1/c^2} \frac{d\mathbf{p}}{dt} = \mathbf{A}_h \mathbf{U}, \quad (\mathbf{M}_{1/c^2})_{ij} = \int_{D^k} \frac{1}{c^2(\mathbf{x})} \phi_j(\mathbf{x}) \phi_i(\mathbf{x}).$$

# Weight-adjusted DG: stable, accurate, non-invasive

- **Weight-adjusted DG (WADG)**: energy stable approx. of  $\mathbf{M}_{1/c^2}$

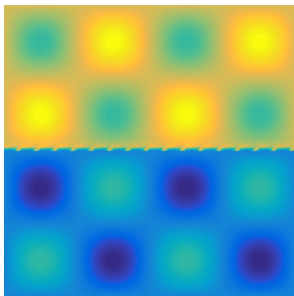
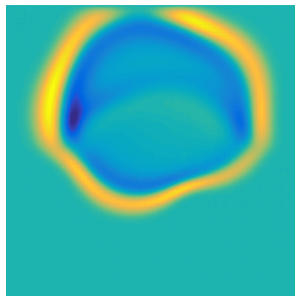
$$\mathbf{M}_{1/c^2} \frac{d\mathbf{p}}{dt} \approx \mathbf{M} (\mathbf{M}_{c^2})^{-1} \mathbf{M} \frac{d\mathbf{p}}{dt} = \mathbf{A}_h \mathbf{U}.$$

- New evaluation reuses implementation for constant wavespeed

$$\frac{d\mathbf{p}}{dt} = \underbrace{\mathbf{M}^{-1} (\mathbf{M}_{c^2})}_{\text{modified update}} \quad \underbrace{\mathbf{M}^{-1} \mathbf{A}_h \mathbf{U}}_{\text{constant wavespeed RHS}}$$

- Low storage matrix-free application of  $\mathbf{M}^{-1} \mathbf{M}_{c^2}$  using **quadrature**-based interpolation and  $L^2$  projection matrices  $\mathbf{V}_q, \mathbf{P}_q$ .

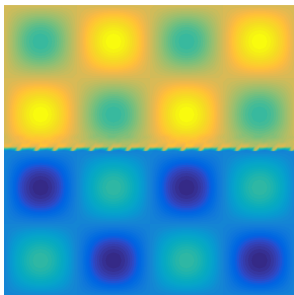
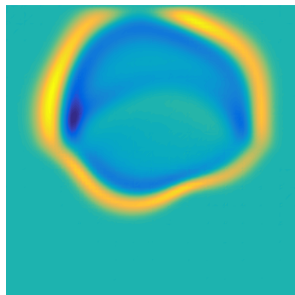
$$(\mathbf{M})^{-1} \mathbf{M}_{c^2} \text{RHS} = \underbrace{\mathbf{M}^{-1} \mathbf{V}_q^T \mathbf{W} \text{diag}(c^2) \mathbf{V}_q}_{\mathbf{P}_q} (\text{RHS}).$$

WADG: nearly identical to using  $M_{1/c^2}^{-1}$ (a)  $c^2(x, y)$ 

(b) Standard DG

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

- $L^2$  error is  $O(h^{N+1})$ ; standard DG and WADG difference is  $O(h^{N+2})$ .
- Can generalize to matrix weights (elastic wave propagation).

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(b) Weighted-adjusted DG

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

	$N = 1$	$N = 2$	$N = 3$	$N = 4$	$N = 5$	$N = 6$	$N = 7$
$M_{1/c^2}^{-1}$	.66	2.79	9.90	29.4	73.9	170.5	329.4
WADG	0.59	1.44	4.30	13.9	43.0	107.8	227.7
Speedup	1.11	1.94	2.30	2.16	1.72	1.58	1.45

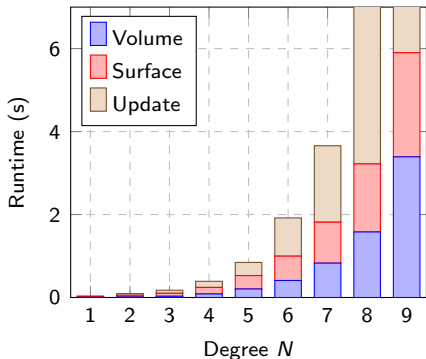
Time (ns) per element: storing/applying  $M_{1/c^2}^{-1}$  vs WADG (deg.  $2N$  quadrature).

- Efficiency on GPUs: reduce memory accesses and data movement.
- (Tuned) low storage WADG faster than storing and applying  $M_{1/c^2}^{-1}$ !

# Computational costs at high orders of approximation

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

WADG runtimes

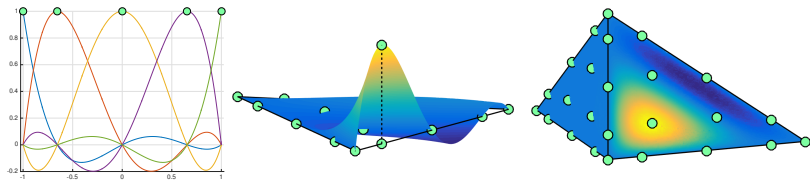


- Large **dense** matrices:  
 $O(N^6)$  work per tet.
- High orders usually use tensor-product elements:  
 $O(N^4)$  vs  $O(N^6)$  cost, but less geometric flexibility.
- Idea: choose basis such that matrices are **sparse**.

WADG runtimes for 50 timesteps, 98304 elements.

# BBDG: Bernstein-Bezier DG methods

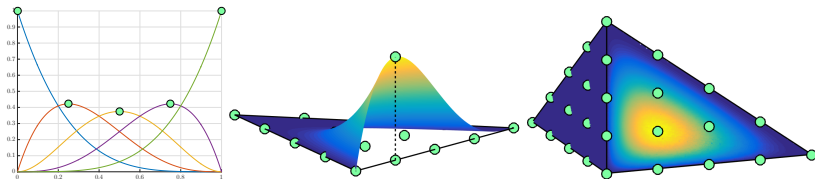
- Nodal DG:  $O(N^6)$  cost in 3D vs  $O(N^3)$  degrees of freedom.
- Switch to Bernstein basis: sparse and structured matrices.
- Optimal  $O(N^3)$  application of differentiation and lifting matrices.



Nodal bases in one, two, and three dimensions.

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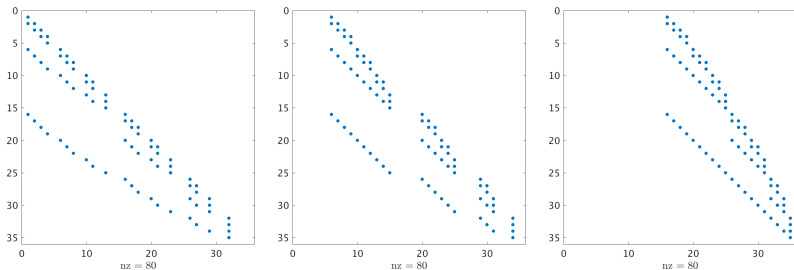
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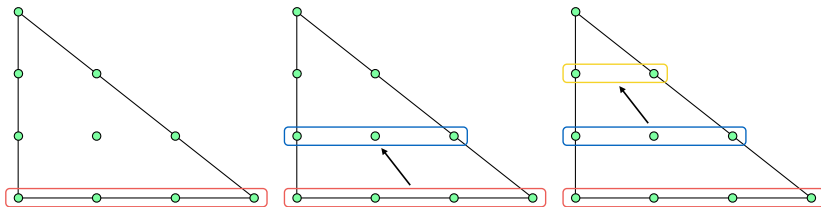
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Sparse Bernstein differentiation matrices for the reference tetrahedron.

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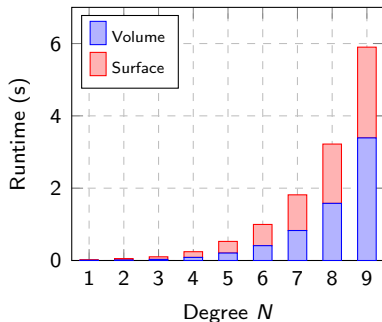
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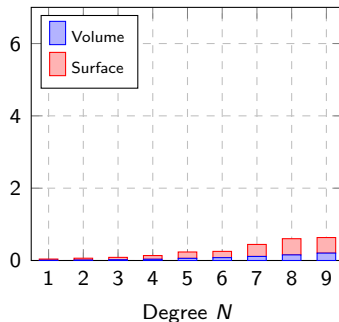
Optimal  $O(N^3)$  complexity “slice-by-slice” application of Bernstein lift.

## BBDG: efficient volume, surface kernels

Nodal DG

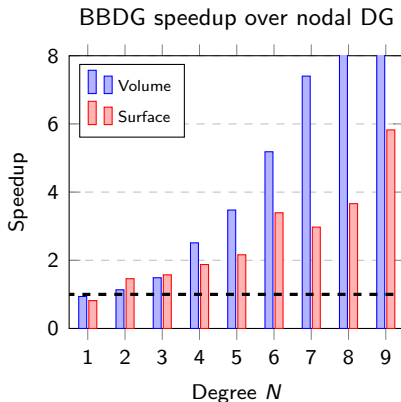


Bernstein-Bezier DG



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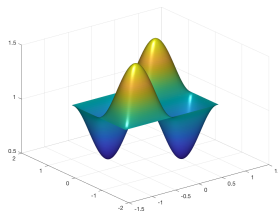
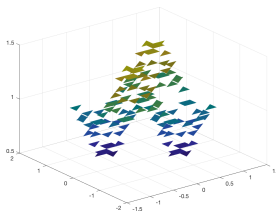
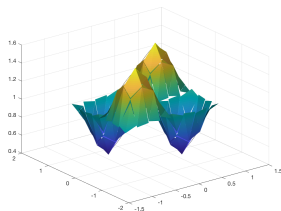
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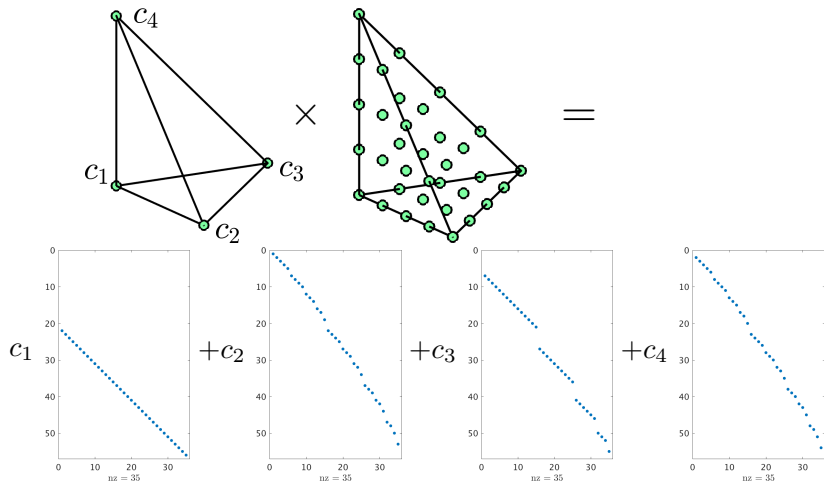


# BBWADG: polynomial multiplication and projection

(a) Exact  $c^2$ (b)  $M = 0$  approximation(c)  $M = 1$  approximation

- WADG: can reuse fast Bernstein volume and surface kernels.
- $O(N^6)$  update kernel:  $\mathbf{V}_q$  interpolates  $u(\mathbf{x})$  to quadrature points, scale by  $c^2(\mathbf{x})$  at quadrature points, apply  $\mathbf{P}_q$  to project back to  $P^N$ .
- New approach: approx.  $c^2(\mathbf{x})$  with degree  $M$  polynomial, use fast Bernstein algorithms for polynomial multiplication and projection.

# Fast Bernstein polynomial multiplication



Bernstein polynomial multiplication: for fixed  $M$ ,  $O(N^3)$  complexity.

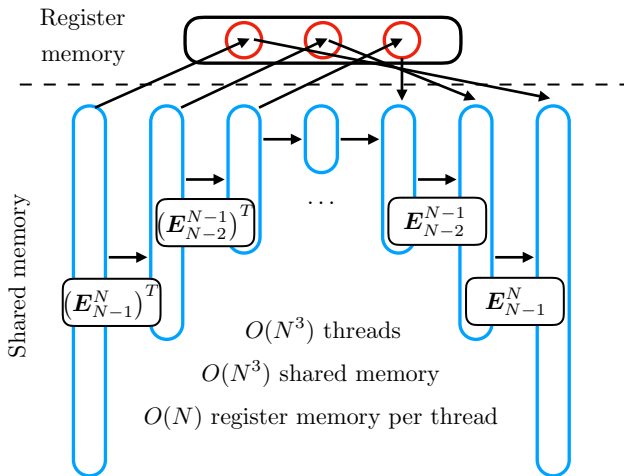
# Fast Bernstein polynomial projection

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

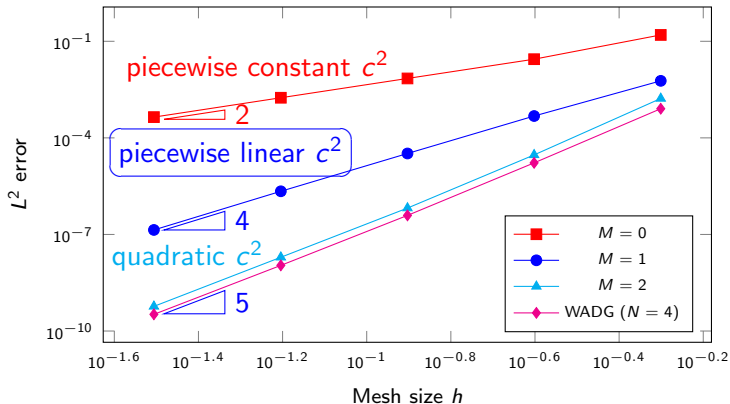
$$\mathbf{P}_N^{N+M} = \underbrace{\sum_{j=0}^N c_j \mathbf{E}_{N-j}^N \left( \mathbf{E}_{N-j}^N \right)^T}_{\tilde{\mathbf{P}}_N} \left( \mathbf{E}_N^{N+M} \right)^T$$

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

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

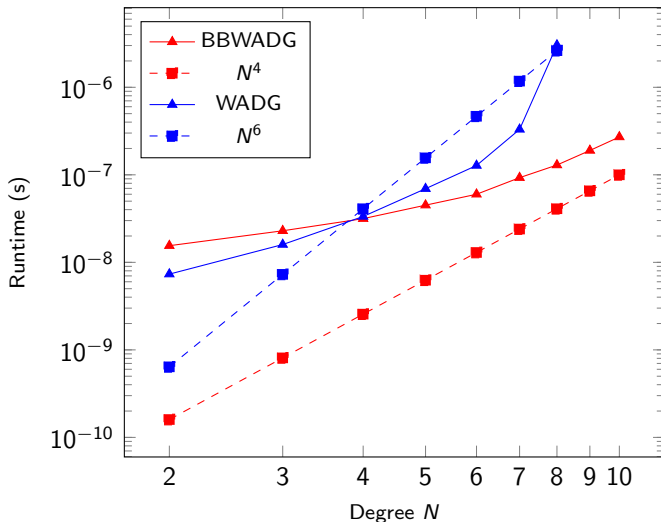
Sketch of GPU algorithm for  $\tilde{P}_N$ 

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

BBWADG: approximating  $c^2$  and accuracy

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

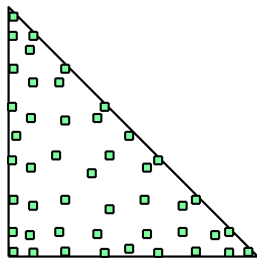
## BBWADG: computational runtime (acoustics)

Update kernel for  $M = 1$ : runtime per element

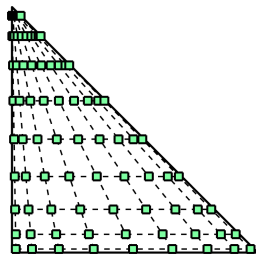
# BBWADG: update kernel speedup over WADG (acoustics)

	$N = 3$	$N = 4$	$N = 5$	$N = 6$	$N = 7$	$N = 8$
WADG	1.60e-8	3.34e-8	6.94e-8	1.28e-7	3.31e-7	3.03e-6
BBWADG	2.20e-8	3.30e-8	4.42e-8	6.01e-8	9.46e-8	1.31e-7
Speedup	0.7260	1.0127	1.5706	2.1258	3.4938	23.1591

For  $N \geq 8$ , quadrature (and WADG) becomes much more expensive.



(a)  $N = 7$  quadrature



(b)  $N = 8$  quadrature

# Summary and acknowledgements

- Weight-adjusted DG: stability and efficiency for heterogeneous media.
- BBWADG: improved complexity for approximate wavespeeds.
- This work is supported by the National Science Foundation under DMS-1712639 and DMS-1719818.

Thank you! Questions?



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Chan, Hewett, Warburton. 2016. Weight-adjusted DG methods: wave propagation in heterogeneous media (SISC).

Chan 2017. Weight-adjusted DG methods: matrix-valued weights and elastic wave prop. in heterogeneous media (IJNME).

Chan, Warburton 2015. GPU-accelerated Bernstein-Bezier discontinuous Galerkin methods for wave propagation (SISC).