



Four-dimensional anisotropic mesh adaptation for spacetime numerical simulations

Philip Claude Caplan

Thesis committee: Prof. David L. Darmofal (chair), Bob Haimes, Prof. Jaime Peraire

PhD Thesis Defense

17 May 2019

Introduction



Introduction

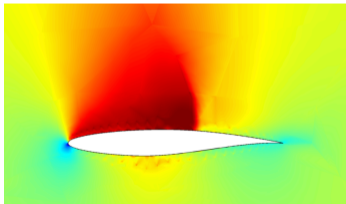
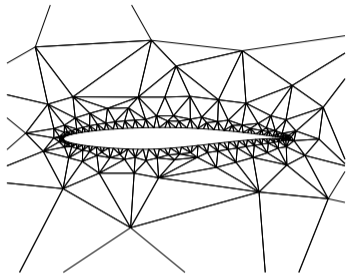
Mesh adaptation
algorithm

Demonstration with
analytic metrics

Demonstration
within adaptive
framework

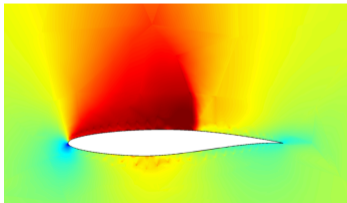
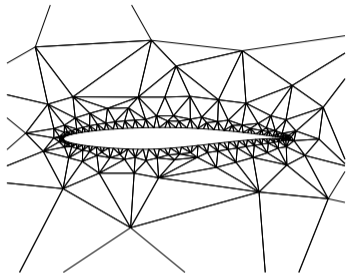
Conclusions

We want accurate answers to PDEs using numerical simulations.

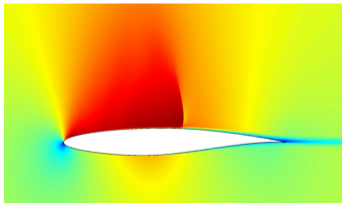
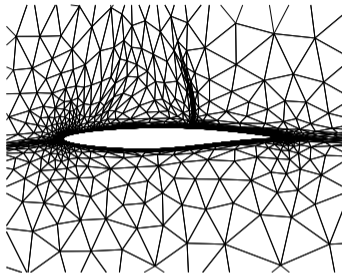


783% error in drag

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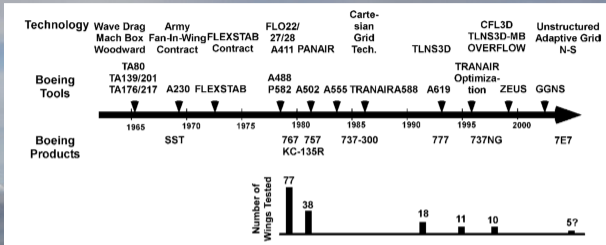
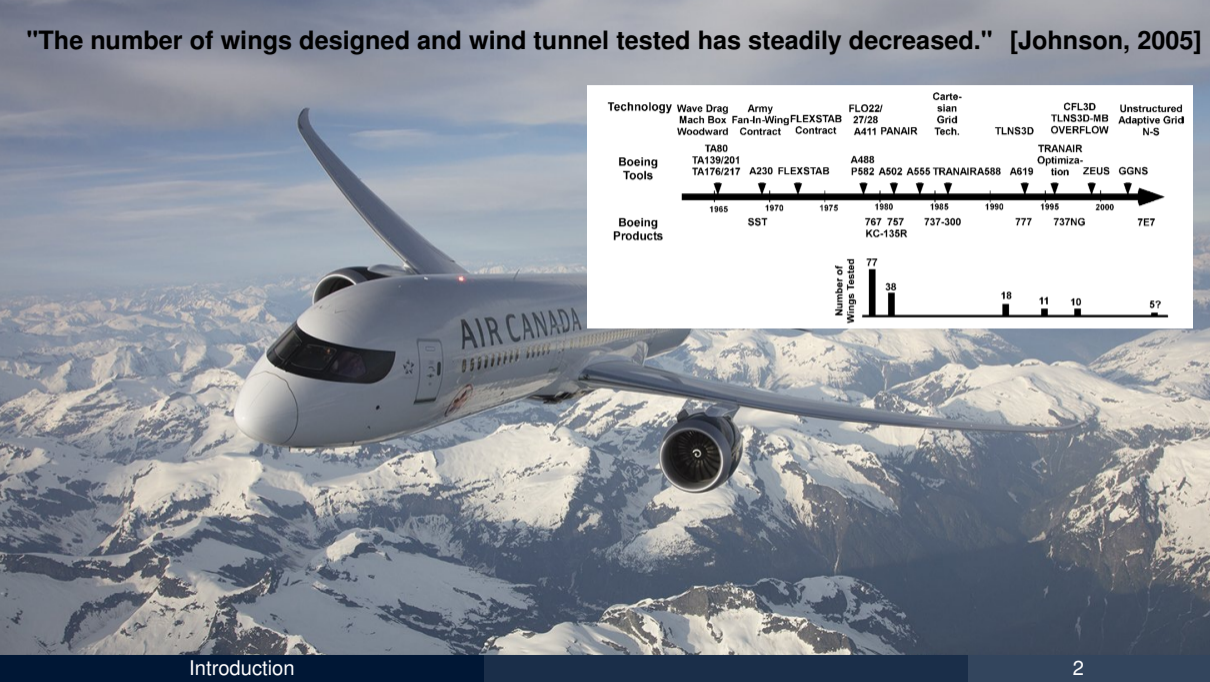


0.5% error in drag

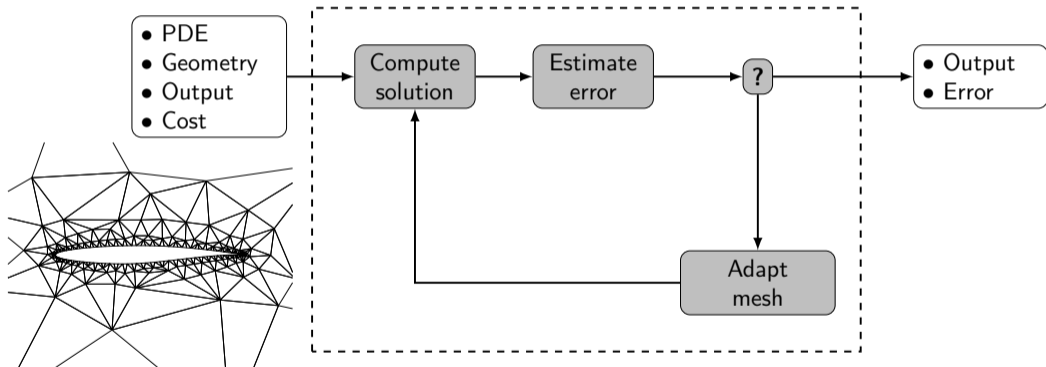
"The number of wings designed and wind tunnel tested has steadily decreased." [Johnson, 2005]



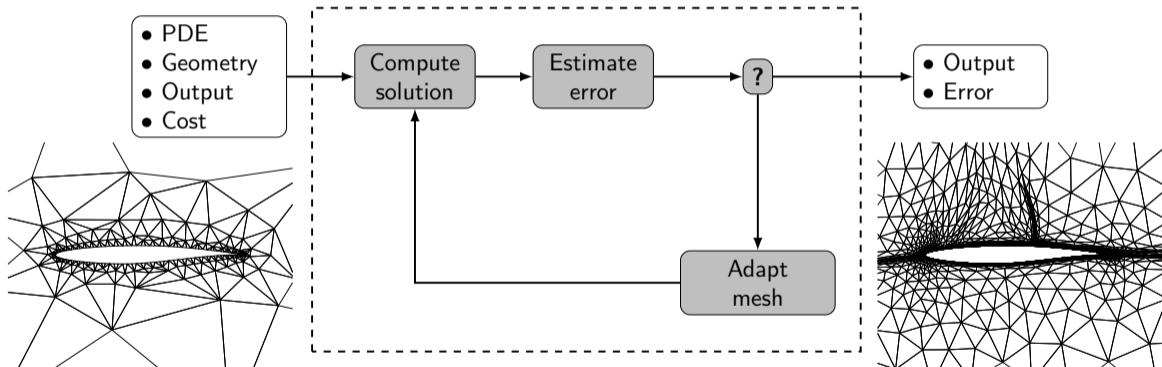
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Mesh adaptation is useful for obtaining accurate solutions.



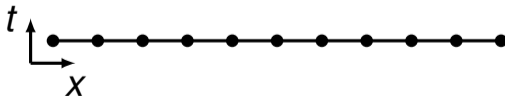
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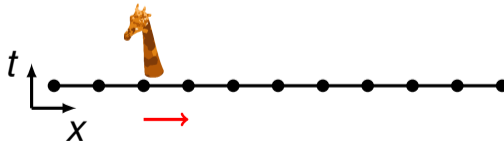
How to solve problems with time-dependent features?

[Song, 2014]

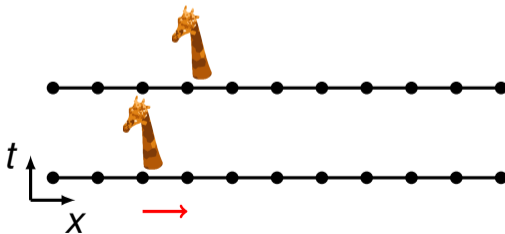
Time-marching approach is computationally expensive.



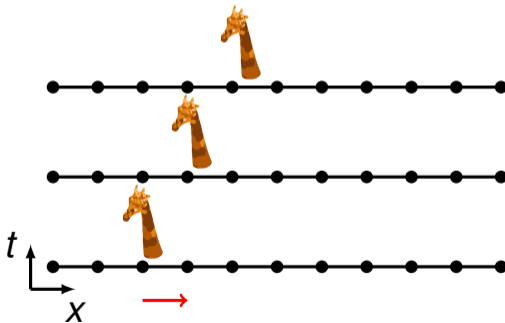
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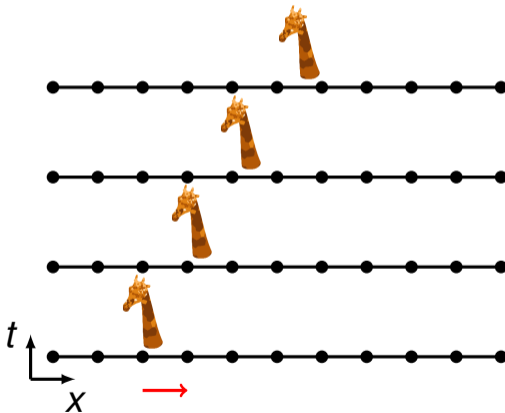
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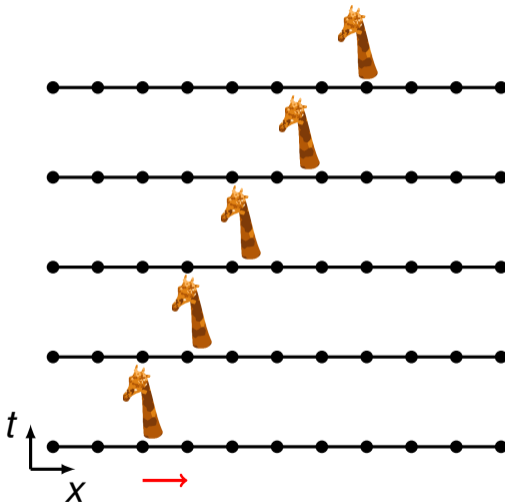
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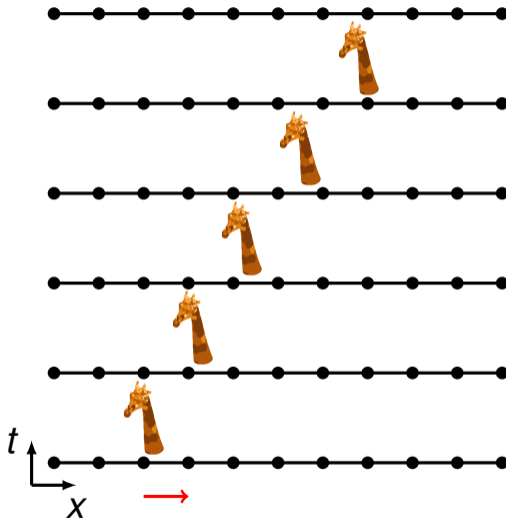
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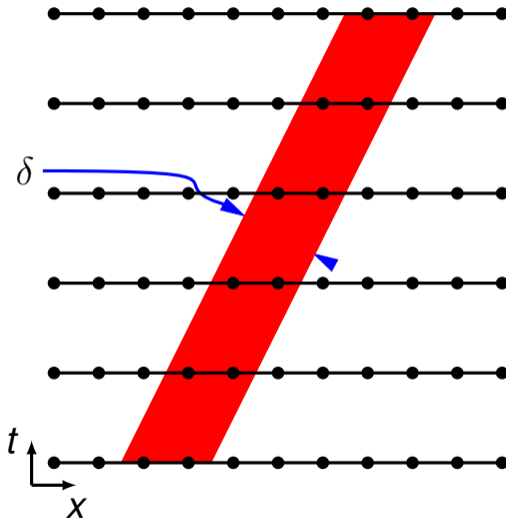
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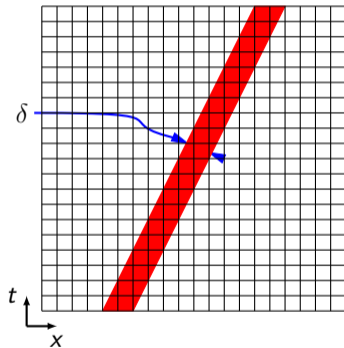
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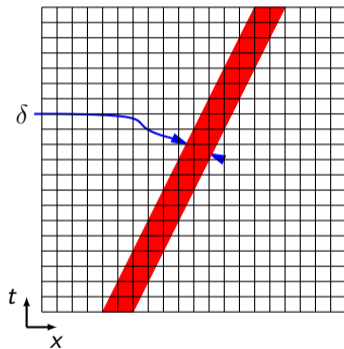
Spacetime uniform refinement requires $\mathcal{O}(\delta^{-2})$ elements.



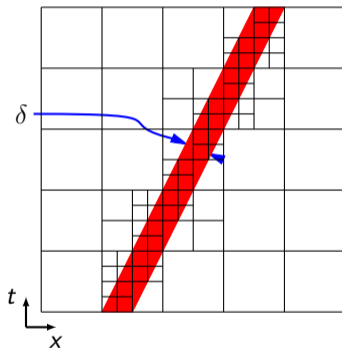
Uniform $\mathcal{O}(\delta^{-2})$

Spacetime tensor-product approach requires $\mathcal{O}(\delta^{-1})$ elements.

[Bangerth, 1999]
[Hartmann, 2001]



Uniform $\mathcal{O}(\delta^{-2})$

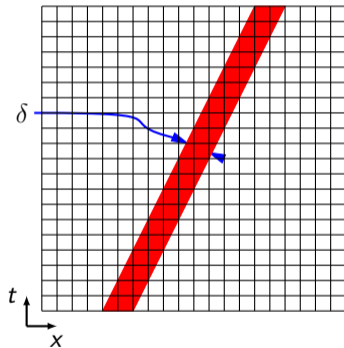


Tensor-product $\mathcal{O}(\delta^{-1})$

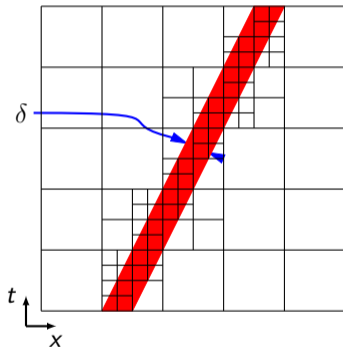
Spacetime unstructured approach requires $\mathcal{O}(1)$ elements.

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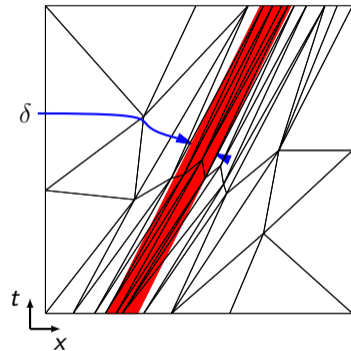
[Yano, 2012]



Uniform $\mathcal{O}(\delta^{-2})$



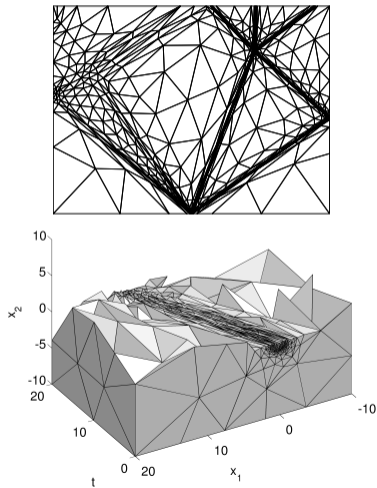
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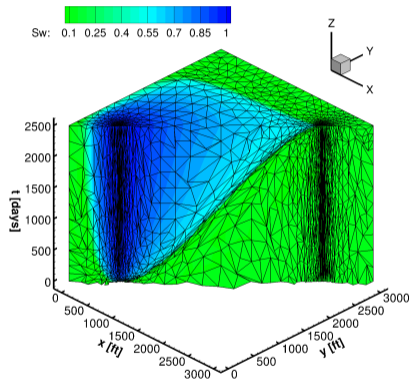
Unstructured $\mathcal{O}(1)$

Spacetime unstructured approach demonstrated in $1d + t$ and $2d + t$.

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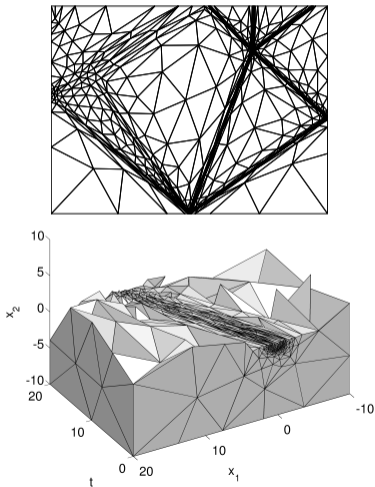


[Jayasinghe, 2018]

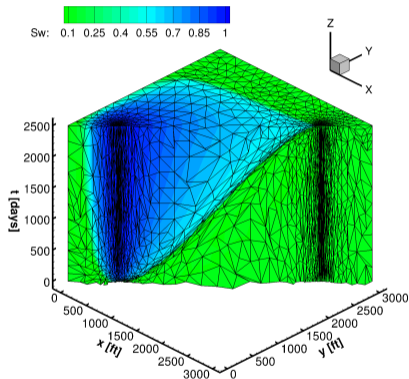


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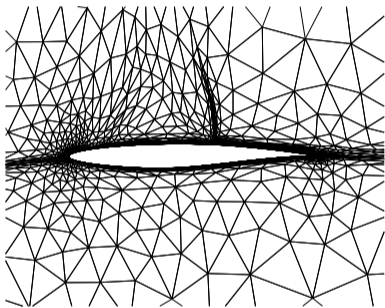
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We need unstructured anisotropic $4d$ meshes for unsteady $3d$ problems.

Anisotropic meshes can be obtained using a metric field.

Optimize mesh?

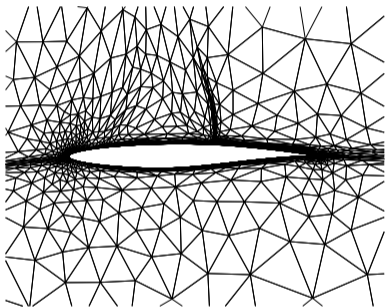


\mathcal{M} : mesh

$$\mathcal{M}^* = \arg \min_{\mathcal{M}} \underbrace{E(\mathcal{M})}_{\text{error}} \quad \text{such that} \quad \underbrace{C(\mathcal{M})}_{\text{cost}} \leq c_t$$

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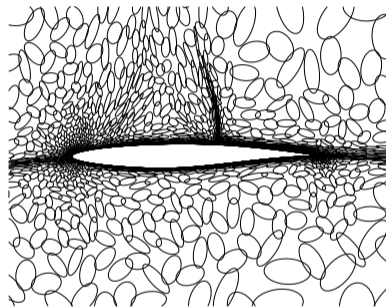
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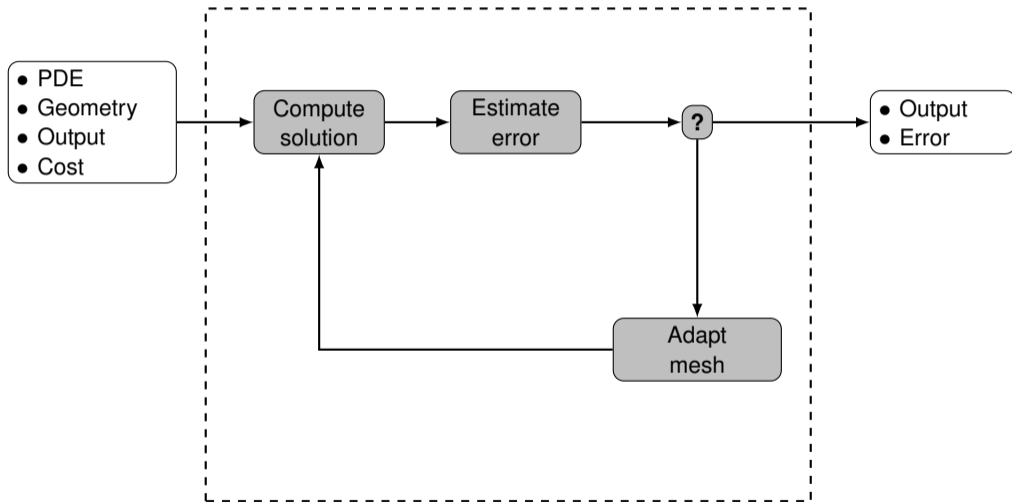
Optimize metric field? [Loseille, 2011]



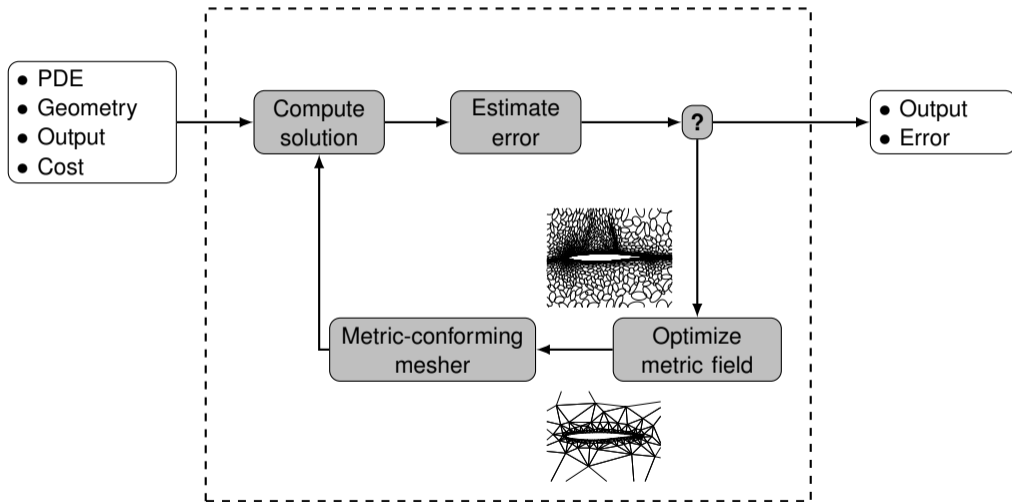
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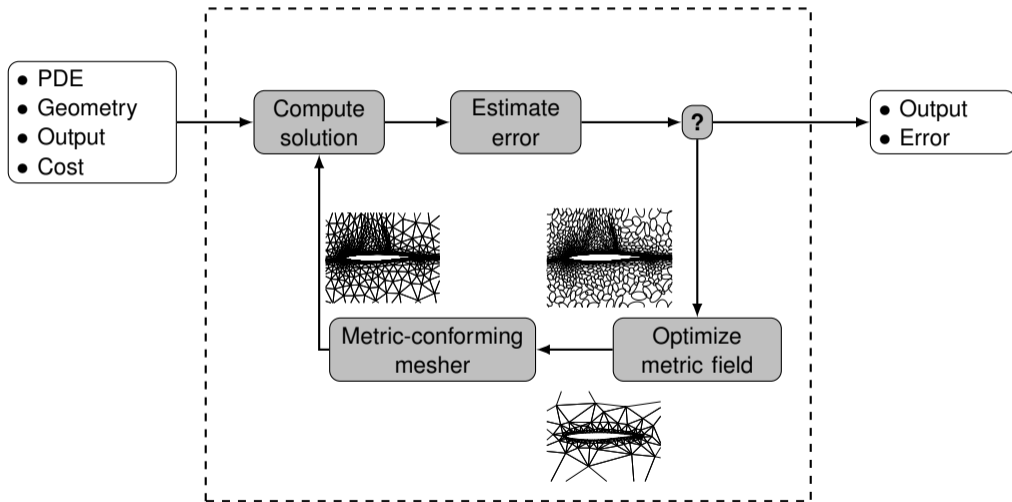
Anisotropic meshes can be obtained from a metric-conforming mesher.



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Previous attempts at anisotropic 4d meshing were not successful.

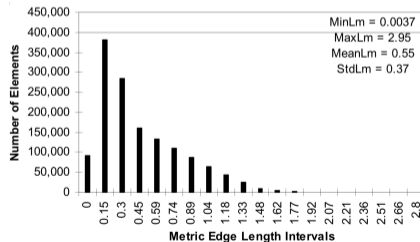
[Coupez, 2000] & [Gruau, 2005]

- Star operator for local mesh operations,
- Minimum volume principle,
- Uniform metric fields.

h	time (s)	# vertices	# elements
1/2	4	126	946
1/3	41	451	4573
1/4	179	1192	14887
1/5	547	2588	35894

[Tremblay, 2007]

- *Simulated* edge swapping,
- Poor metric-conformity (aspect ratio 10:1),
- Heat equation in $3d+t$ with isotropic meshes.



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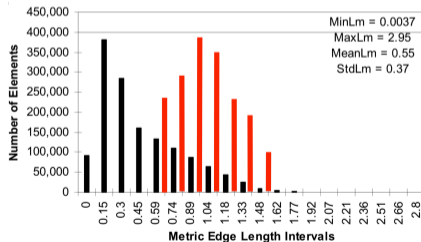
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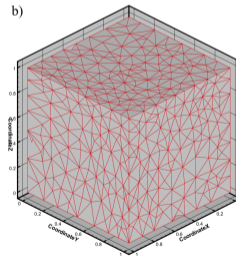
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Main objective:

Develop an anisotropic four-dimensional meshing capability for adaptive numerical simulations.

Approach:

- Simplex meshes (triangles, tetrahedra, pentatopes).
- Local cavity operator framework.
- Mesh Optimization via Error Sampling and Synthesis.

Contributions:

- (1) Develop an algorithm and software for $4d$ **metric-conforming mesh adaptation**.
- (2) **Validate the adaptive algorithm** on $4d$ problems.
- (3) Demonstrate first **PDE-driven anisotropic** unstructured adaptation for **unsteady $3d$ problems**.

Four-dimensional anisotropic mesh adaptation algorithm



Introduction

Mesh adaptation
algorithm

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Conclusions

Metric-conforming mesher strives to create unit n -simplices.

Goals for a mesh $\mathcal{M} = (\mathcal{V}, \mathcal{T})$ of $\Omega \subset \mathbb{R}^n$

- Edge lengths are 1:

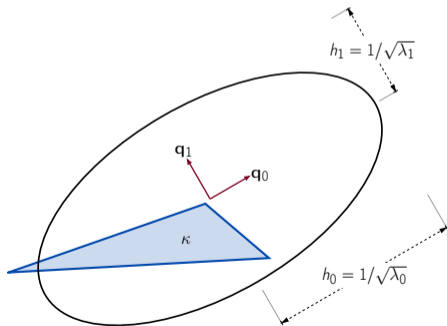
$$\ell_{\mathbf{m}}(e) = 1, \quad \forall e \in \mathcal{E}(\mathcal{T})$$

- Quality is that of equilateral simplex:

$$q_{\mathbf{m}}(\kappa) = \frac{1}{q_{\Delta}} \frac{v_{\mathbf{m}}^{2/n}(\kappa)}{\sum_{e \in \mathcal{E}(\kappa)} \ell_{\mathbf{m}}^2(e)} = 1, \quad \forall \kappa \in \mathcal{T}$$

- # simplices matches metric field complexity:

$$n_S v_{\Delta} = \int_{\mathcal{M}} \sqrt{\det \mathbf{m}} \, dx$$



Metric-conforming mesher strives to create unit n -simplices.

Goals for a mesh $\mathcal{M} = (\mathcal{V}, \mathcal{T})$ of $\Omega \subset \mathbb{R}^n$

- Edge lengths are close to 1:

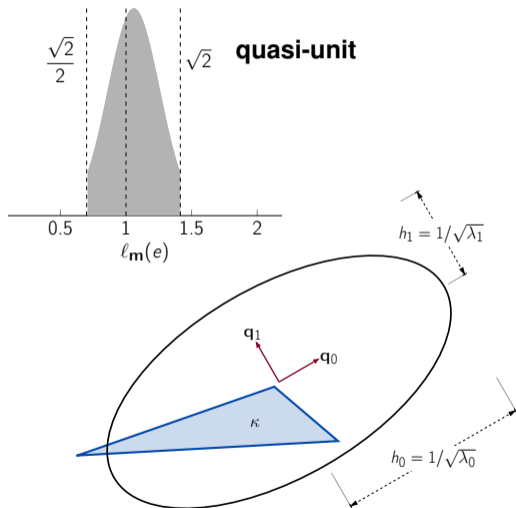
$$1/\sqrt{2} \leq \ell_{\mathbf{m}}(e) \leq \sqrt{2}, \quad \forall e \in \mathcal{E}(\mathcal{T})$$

- Quality is close to that of equilateral simplex:

$$q_{\mathbf{m}}(\kappa) = \frac{1}{q_{\Delta}} \frac{v_{\mathbf{m}}^{2/n}(\kappa)}{\sum_{e \in \mathcal{E}(\kappa)} \ell_{\mathbf{m}}^2(e)} \in [0.8, 1], \quad \forall \kappa \in \mathcal{T}$$

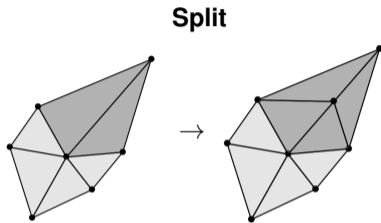
- # simplices matches metric field complexity:

$$n_S v_{\Delta} \approx \int_{\mathcal{M}} \sqrt{\det \mathbf{m}} \, dx$$



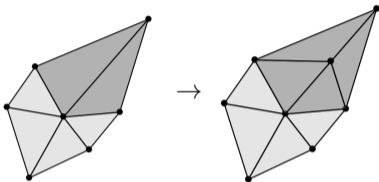
Local operators modify an existing mesh to meet target criteria.

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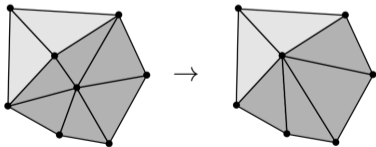


Local operators modify an existing mesh to meet target criteria.

Split

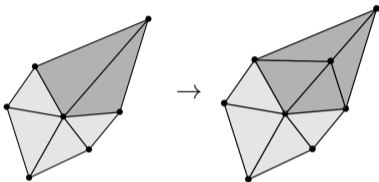


Collapse

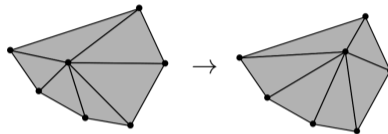


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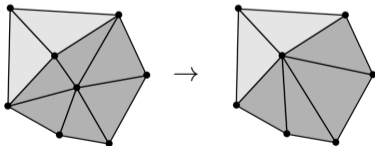
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Smoothing

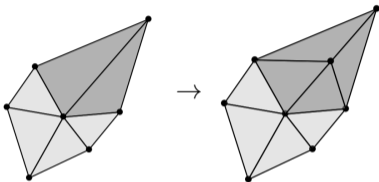


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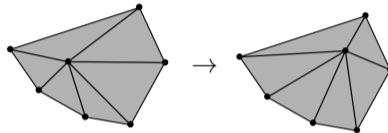


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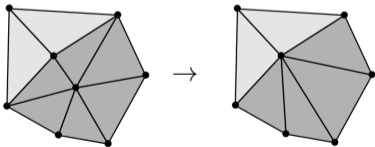
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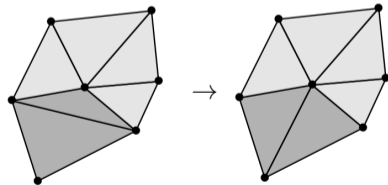
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Collapse



Swap

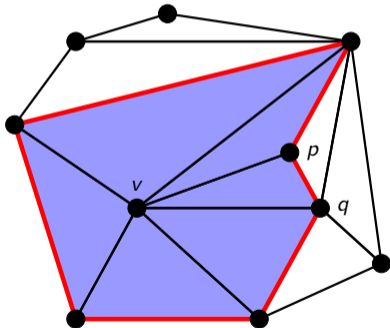


Local operators can be viewed in a dimension-independent way.

Application of mesh modification operator:

[Coupez, 2000],[Loseille, 2017]

$$\mathcal{T}^{k+1} = \mathcal{T}^k \setminus \underbrace{\mathcal{C}(f)}_{\text{cavity}} \cup \underbrace{\mathcal{B}(p, \partial\mathcal{C}^k)}_{\text{insertion}}$$

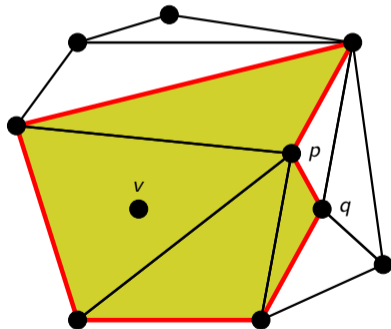
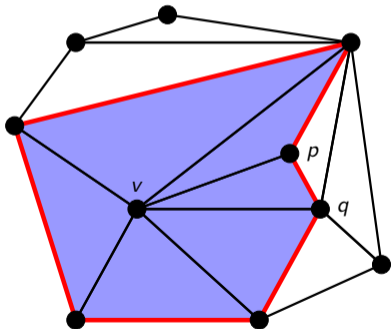


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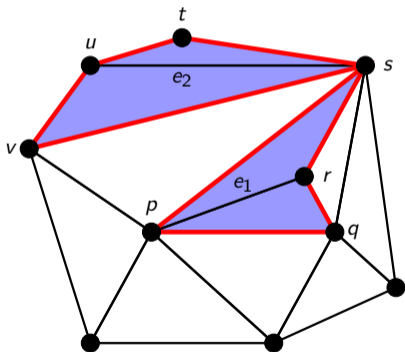


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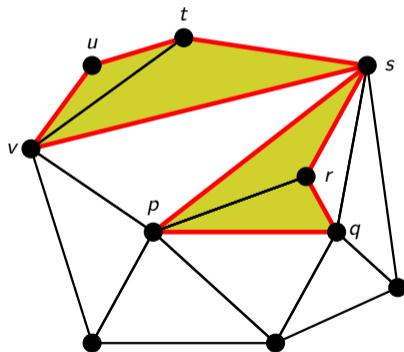
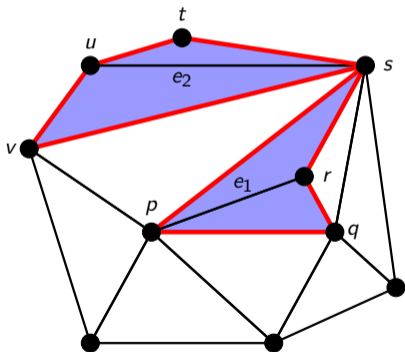


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adaptMesh

input: $\mathcal{M}_{in} = (\mathcal{V}_{in}, \mathcal{T}_{in}), \mathbf{m}$

output: \mathcal{M}_{out} (modified)

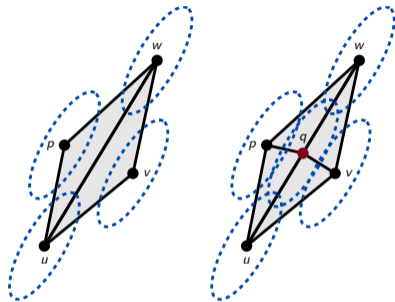
▷ **stage 1:** target edges longer than 2

$\mathcal{M} \leftarrow \text{collapseEdges}(\mathcal{M}, \mathbf{m})$

$\mathcal{M} \leftarrow \text{splitEdges}(\mathcal{M}, \mathbf{m}, 2)$

$\mathcal{M} \leftarrow \text{swapEdges}(\mathcal{M}, \mathbf{m})$

$\mathcal{M} \leftarrow \text{smoothVertices}(\mathcal{M}, \mathbf{m})$



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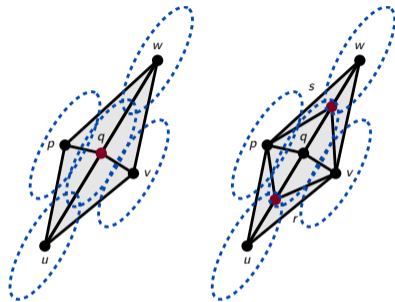
▷ **stage 2:** target edges longer than $\sqrt{2}$

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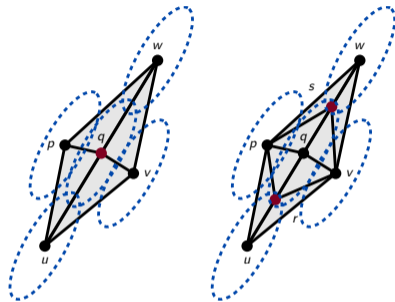
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Highlights:

- do not create short edges during splits
- check number of pentatopes matches metric volume in $4d$

Demonstration with analytic metrics



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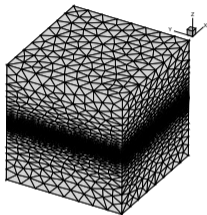
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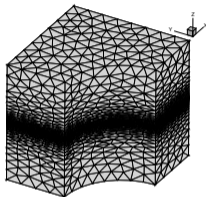
Conclusions

The meshing algorithm performs well on 3d benchmark cases.

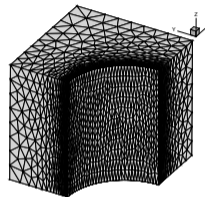
Benchmarks of the Unstructured Grid Adaptation Working Group (UGAWG) [Ibanez et al., 2017]:



Cube Linear



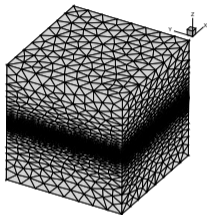
Cube-Cylinder Linear



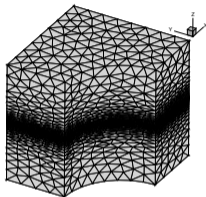
Cube-Cylinder Polar 2

The meshing algorithm performs well on 3d benchmark cases.

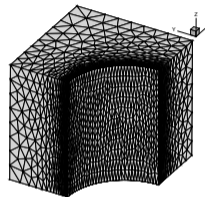
Benchmarks of the Unstructured Grid Adaptation Working Group (UGAWG) [Ibanez et al., 2017]:



Cube Linear



Cube-Cylinder Linear



Cube-Cylinder Polar 2

feflo.a



EPIC-ICSM

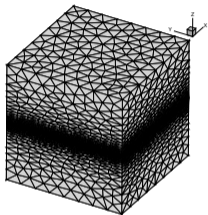


Omega_h

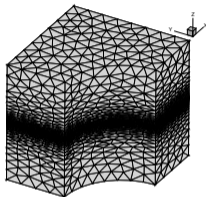


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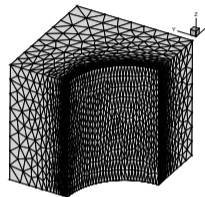
Benchmarks of the Unstructured Grid Adaptation Working Group (UGAWG) [Ibanez et al., 2017]:



Cube Linear



Cube-Cylinder Linear



Cube-Cylinder Polar 2

feflo.a



EPIC-ICSM



Omega_h



avro





Assessment of metric-conformity.

% quasi-unit edge lengths

$$1/\sqrt{2} \leq l_{\mathbf{m}}(e) \leq \sqrt{2}$$

% quasi-unit simplices

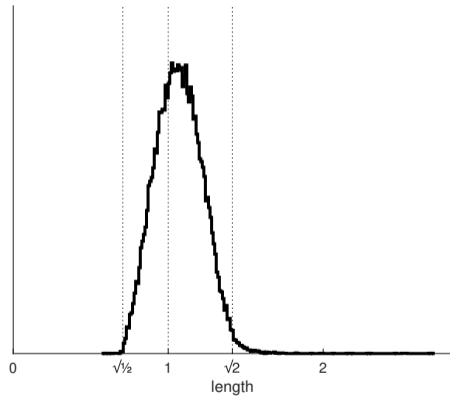
$$q_{\mathbf{m}}(\kappa) > 0.8$$

simplices

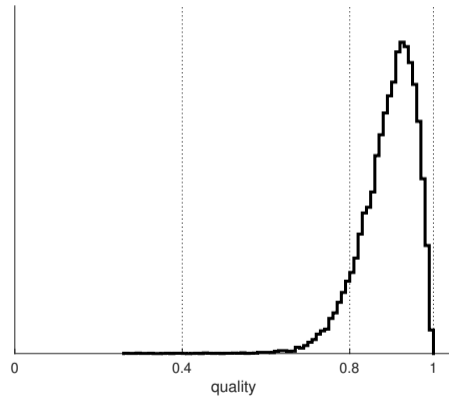
$$n_s v_{\Delta} \approx \int \sqrt{\det \mathbf{m}} \, dx$$

Length and quality histograms will be plotted on logarithmic scales.

% edges

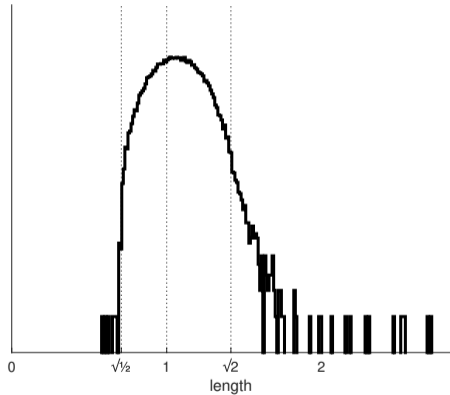


% simplices

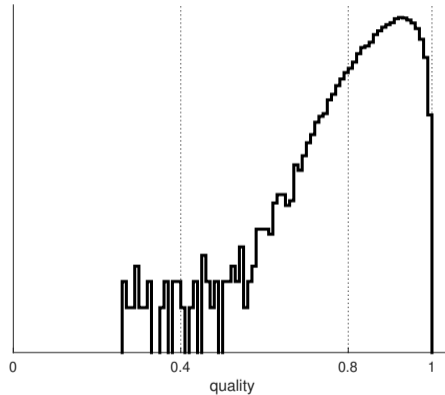


Length and quality histograms will be plotted on logarithmic scales.

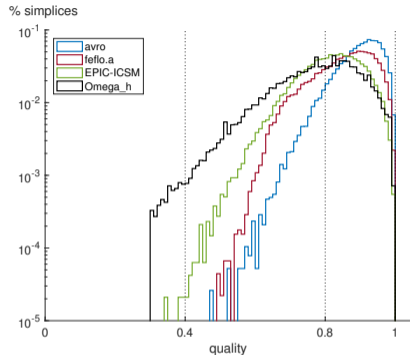
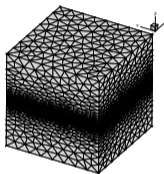
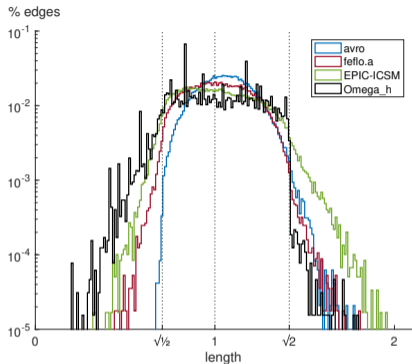
% edges



% simplices

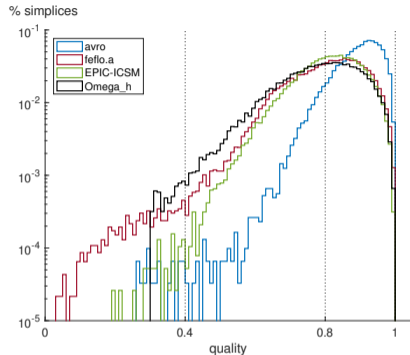
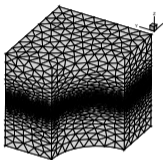
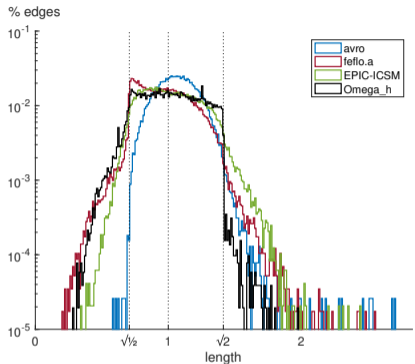


Expect 39k tetrahedra for the Cube Linear case.



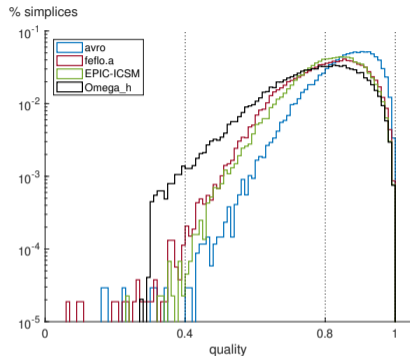
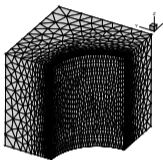
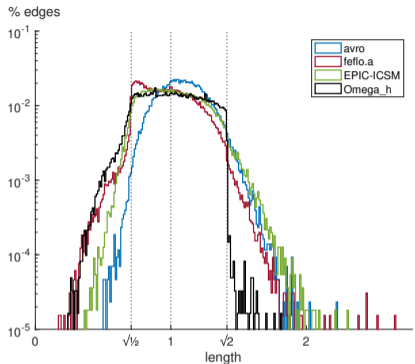
	% l_{unit}	% q_{unit}	# simplices
avro	99.10 %	92.15 %	38.30k
feflo.a	98.28 %	74.28 %	45.16k
EPIC-ICSM	93.04 %	59.52 %	47.55k
Omega_h	93.00 %	47.15 %	51.67k

Expect 31.7k tetrahedra for the Cube-Cylinder Linear case.



	$\%l_{unit}$	$\%q_{unit}$	# simplices
avro	98.79 %	90.88 %	30.45k
feflo.a	93.73 %	55.65 %	46.29k
EPIC-ICSM	92.07 %	56.71 %	38.30k
Omega_h	92.79 %	47.30 %	40.96k

Expect 36.4k tetrahedra for the Cube-Cylinder Polar 2 case.



	$\%l_{unit}$	$\%q_{unit}$	# simplices
avro	95.76 %	78.76 %	34.20k
feflo.a	93.83 %	55.92 %	53.12k
EPIC-ICSM	91.77 %	58.52 %	44.28k
Omega_h	92.19 %	44.96 %	49.15k

Let's look at some 4d metric-conforming cases in a tesseract.

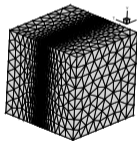
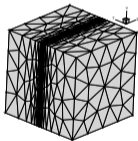
Tesseract Linear

$$\mathbf{m}(\mathbf{x}) = \text{diag} \left(h_x^{-2}, h_y^{-2}, h_z^{-2}, h_t^{-2} \right)$$

$h_x = h_y = h_z = h = \text{constant}$

h_t increases away from $t = 0.5$.

TL1 ($h = 0.25$) **TL2** ($h = 0.125$)



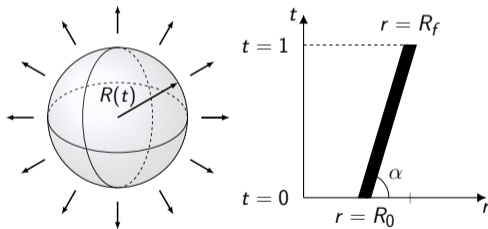
Tesseract Wave

$$\mathbf{m}(\mathbf{x}) = \mathbf{Q} \text{diag} \left(h_r^{-2}, h_\theta^{-2}, h_\phi^{-2}, h_t^{-2} \right) \mathbf{Q}^t$$

Wave radius increases at $R(t) = R_0 + v_w t$

h_r increases away from $R(t)$, $h_t = \text{constant}$

h_ϕ, h_θ similar to Cube-Cylinder Polar 2 case



Let's look at some 4d metric-conforming cases in a tesseract.

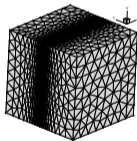
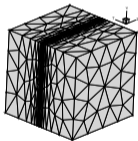
Tesseract Linear

$$\mathbf{m}(\mathbf{x}) = \text{diag} \left(h_x^{-2}, h_y^{-2}, h_z^{-2}, h_t^{-2} \right)$$

$h_x = h_y = h_z = h = \text{constant}$

h_t increases away from $t = 0.5$.

TL1 ($h = 0.25$) TL2 ($h = 0.125$)



Refine **hyperplanes** at non-constant t .

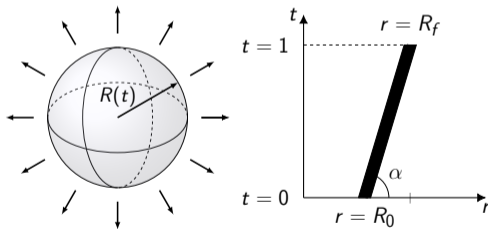
Tesseract Wave

$$\mathbf{m}(\mathbf{x}) = \mathbf{Q} \text{diag} \left(h_r^{-2}, h_\theta^{-2}, h_\phi^{-2}, h_t^{-2} \right) \mathbf{Q}^t$$

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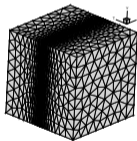
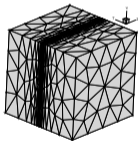
Tesseract Linear

$$\mathbf{m}(\mathbf{x}) = \text{diag} \left(h_x^{-2}, h_y^{-2}, h_z^{-2}, h_t^{-2} \right)$$

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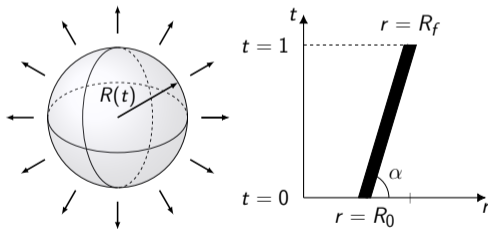
Tesseract Wave

$$\mathbf{m}(\mathbf{x}) = \mathbf{Q} \text{diag} \left(h_r^{-2}, h_\theta^{-2}, h_\phi^{-2}, h_t^{-2} \right) \mathbf{Q}^t$$

Wave radius increases at $R(t) = R_0 + v_w t$

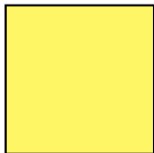
h_r increases away from $R(t)$, $h_t = \text{constant}$

h_ϕ, h_θ similar to Cube-Cylinder Polar 2 case

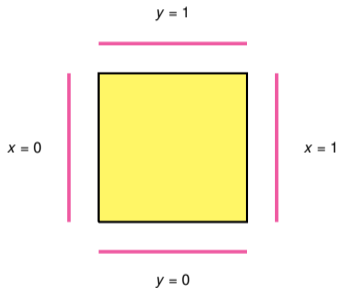


Refine **spheres** at constant t and
cones at non-constant t .

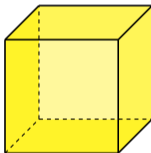
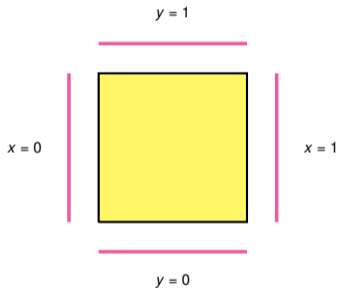
We will look at *boundaries* of the tesseract meshes.



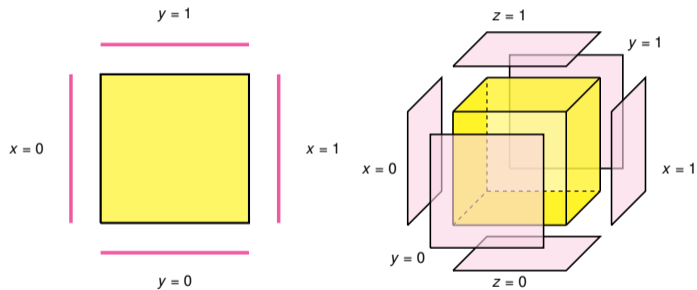
We will look at *boundaries* of the tesseract meshes.



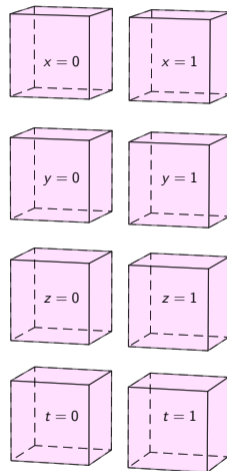
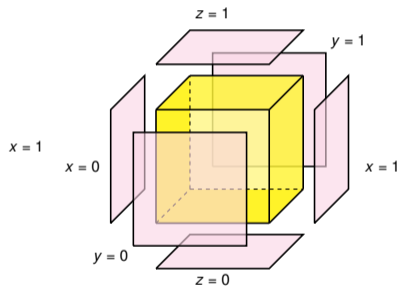
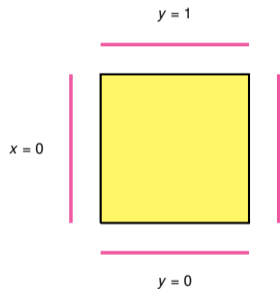
We will look at *boundaries* of the tesseract meshes.



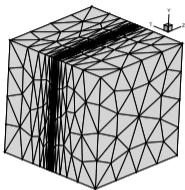
We will look at *boundaries* of the tesseract meshes.



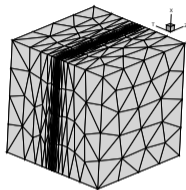
We will look at *boundaries* of the tesseract meshes.



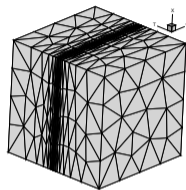
Expected hyperplanes are refined for the Tesseract Linear cases.



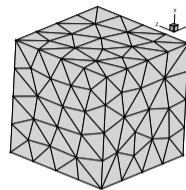
$x = 0$



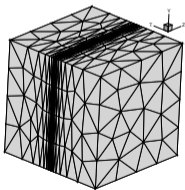
$y = 0$



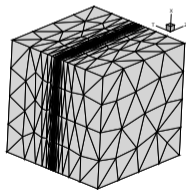
$z = 0$



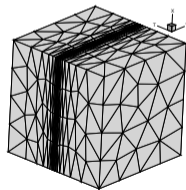
$t = 0$



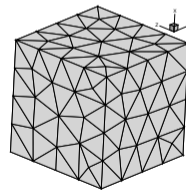
$x = 1$



$y = 1$

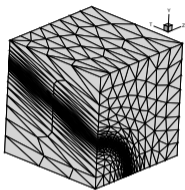


$z = 1$

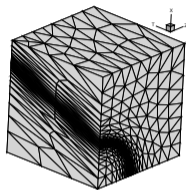


$t = 1$

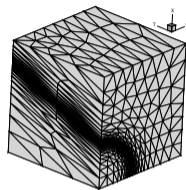
Expected spheres and cones are refined for the Tesseract Wave case.



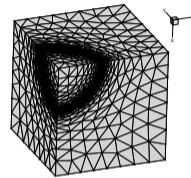
$x = 0$



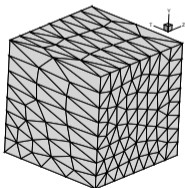
$y = 0$



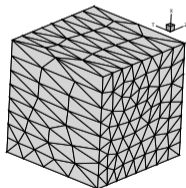
$z = 0$



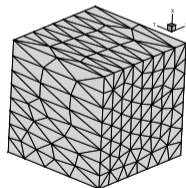
$t = 0$



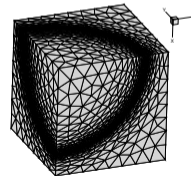
$x = 1$



$y = 1$

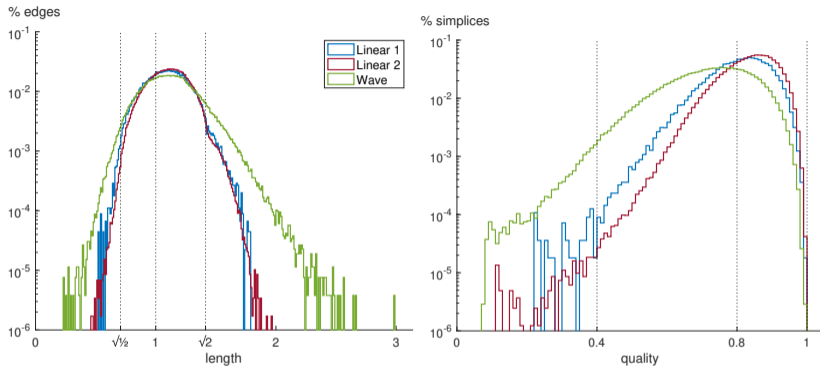


$z = 1$



$t = 1$

Good metric-conformity is observed for the 4d benchmark cases.



	$\%l_{\text{unit}}$	$\%q_{\text{unit}}$	# simplices	expected
Linear 1	96.46 %	56.35 %	55.66k	51k
Linear 2	97.27 %	70.00 %	814.50k	818k
Wave	88.96 %	26.89 %	347.19k	n/a

Demonstration within adaptive framework



Introduction

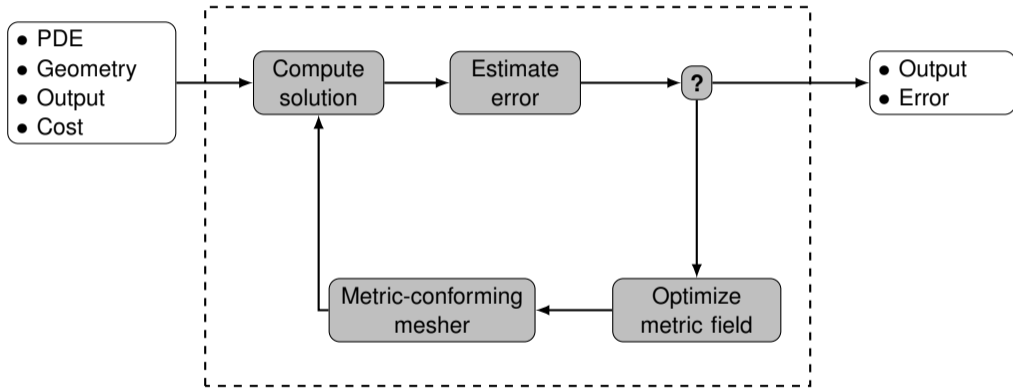
Mesh adaptation
algorithm

Demonstration with
analytic metrics

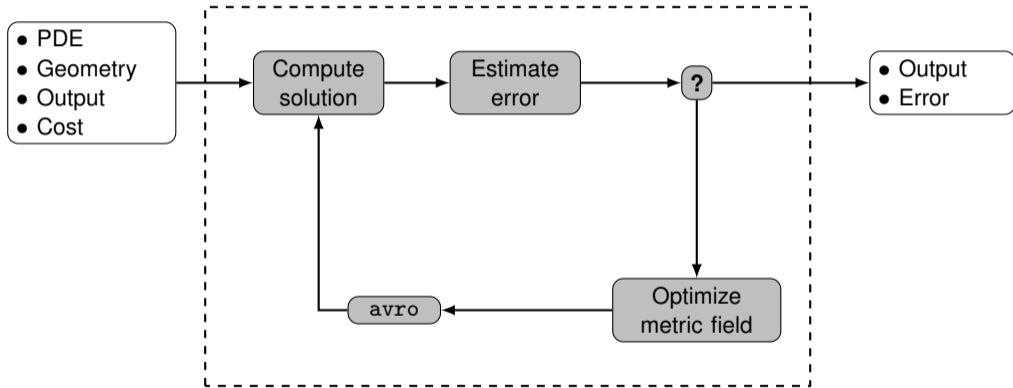
Demonstration
within adaptive
framework

Conclusions

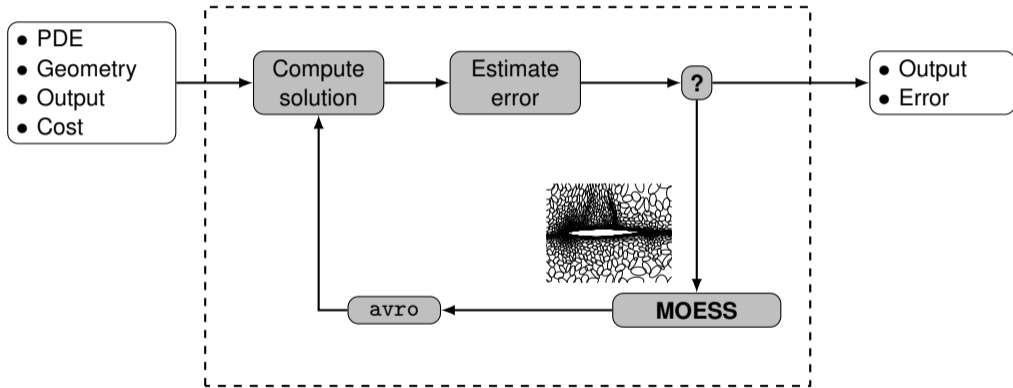
Metric fields obtained from MOESS [Yano, 2012].



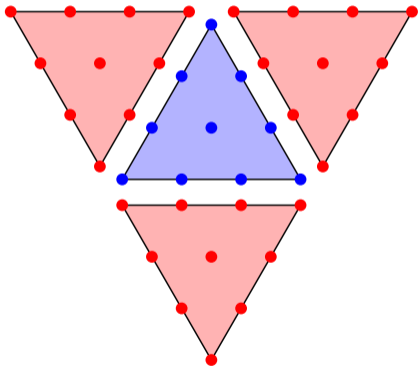
Metric fields obtained from MOESS [Yano, 2012].



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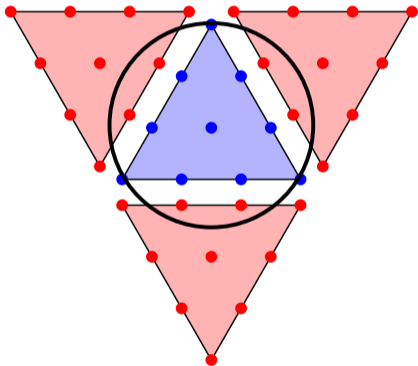


Mesh Optimization via Error Sampling and Synthesis



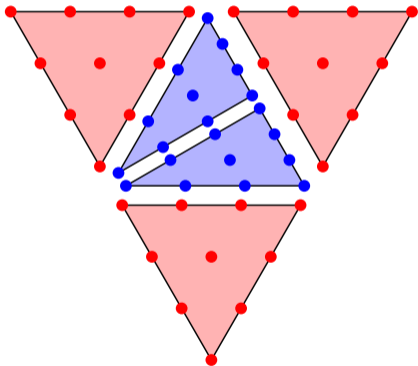
Original

Mesh Optimization via Error Sampling and Synthesis



Original m_0 η_0

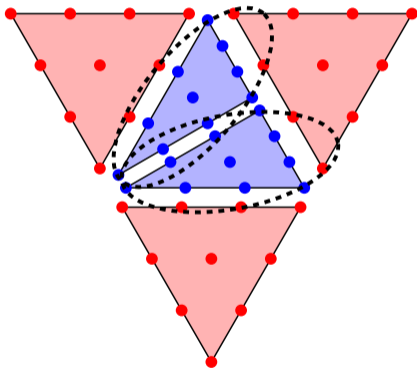
Mesh Optimization via **Error Sampling** and Synthesis



Original
Edge split 1

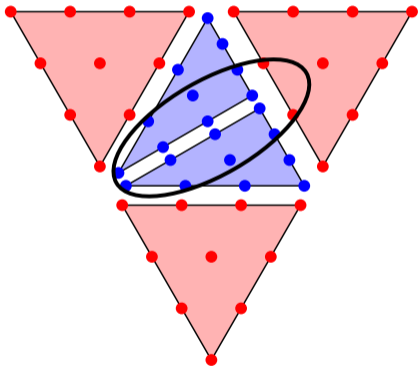
m_0 η_0

Mesh Optimization via **Error Sampling** and Synthesis



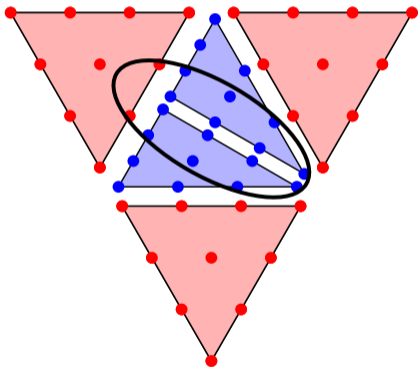
Original m_0 η_0
Edge split 1

Mesh Optimization via **Error Sampling** and Synthesis



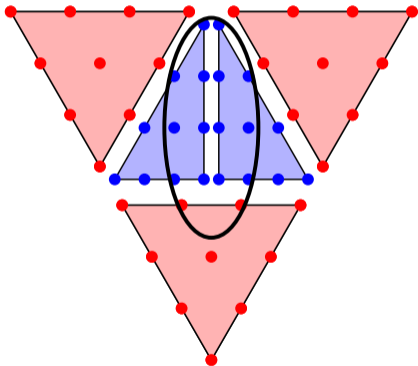
Original	m_0	η_0
Edge split 1	m_1	η_1

Mesh Optimization via **Error Sampling** and Synthesis



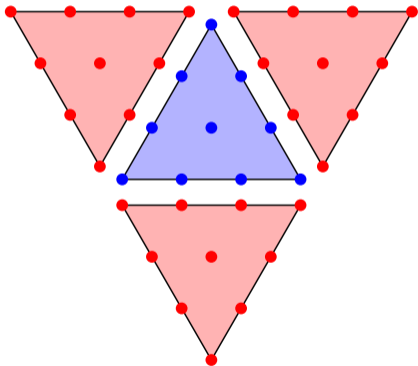
Original	m_0	η_0
Edge split 1	m_1	η_1
Edge split 2	m_2	η_2

Mesh Optimization via **Error Sampling** and Synthesis



Original	m_0	η_0
Edge split 1	m_1	η_1
Edge split 2	m_2	η_2
Edge split 3	m_3	η_3

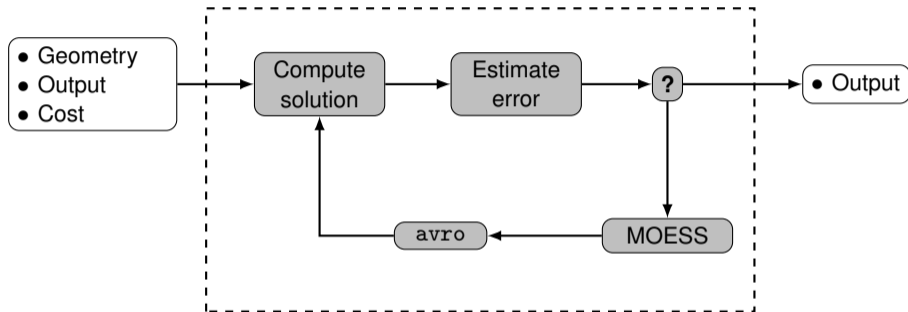
Mesh Optimization via Error Sampling and **Synthesis**



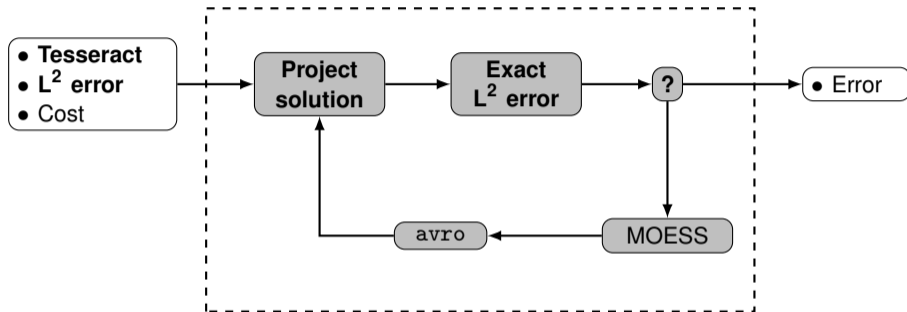
Original	\mathbf{m}_0	η_0
Edge split 1	\mathbf{m}_1	η_1
Edge split 2	\mathbf{m}_2	η_2
Edge split 3	\mathbf{m}_3	η_3

Error model: $\eta(\mathbf{s}) = \eta_0 \exp(\text{tr}(\mathbf{r}\mathbf{s}))$

Adapt to the exact L^2 error.



Adapt to the exact L^2 error of a boundary layer function.



$$u(x, y, z, t) = \exp(-x/\epsilon) + \frac{(2y)^{p+1}}{(p+1)!} + \frac{(4z)^{p+1}}{(p+1)!} + \frac{(6t)^{p+1}}{(p+1)!} \quad \epsilon = 0.01$$

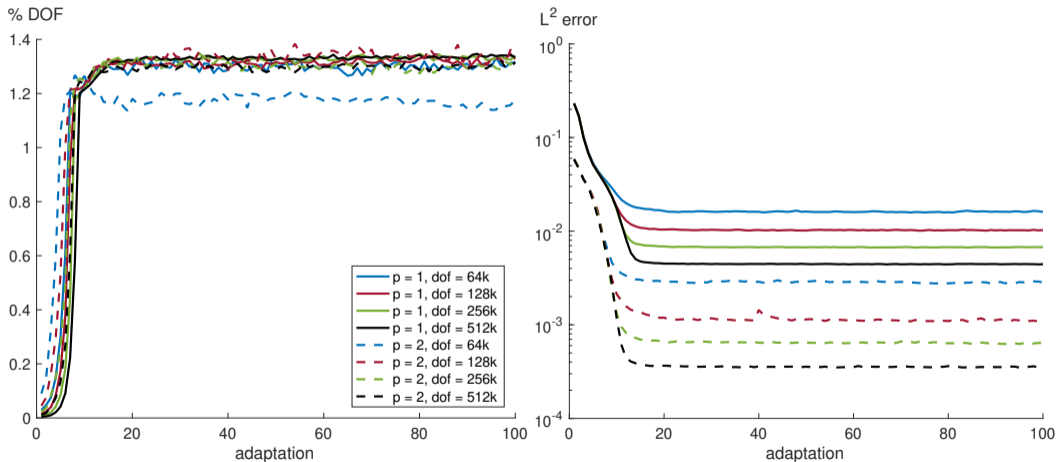
using $p = 1$ and $p = 2$ discontinuous Galerkin solution spaces

Optimal mesh size & aspect ratios
[Yano, 2012]

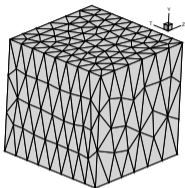
$$h_x = h_{x,0} \exp(k_{h_x} x)$$

$$a_i = a_{i,0} \exp(k_a x), \quad i = y, z, t$$

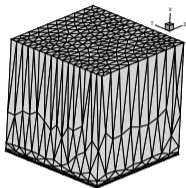
L^2 error control for boundary layer: DOF overshoot $\approx 35\%$.



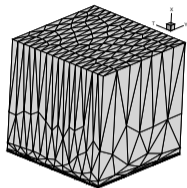
$\rho = 1$ 512k DOF optimized meshes show expected refinement.



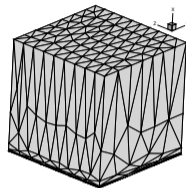
$x = 0$



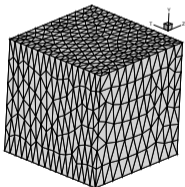
$y = 0$



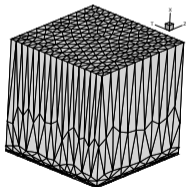
$z = 0$



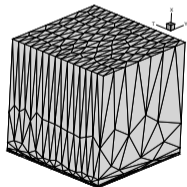
$t = 0$



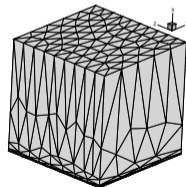
$x = 1$



$y = 1$

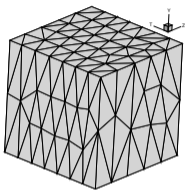


$z = 1$

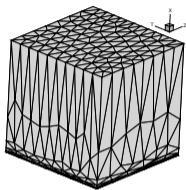


$t = 1$

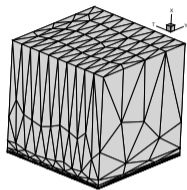
$p = 2$ 512k DOF optimized meshes show expected refinement.



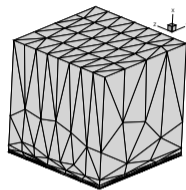
$x = 0$



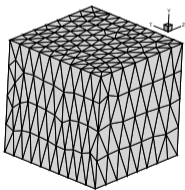
$y = 0$



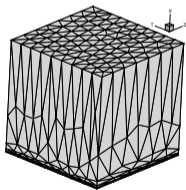
$z = 0$



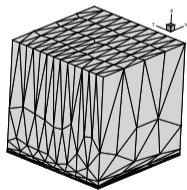
$t = 0$



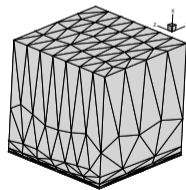
$x = 1$



$y = 1$

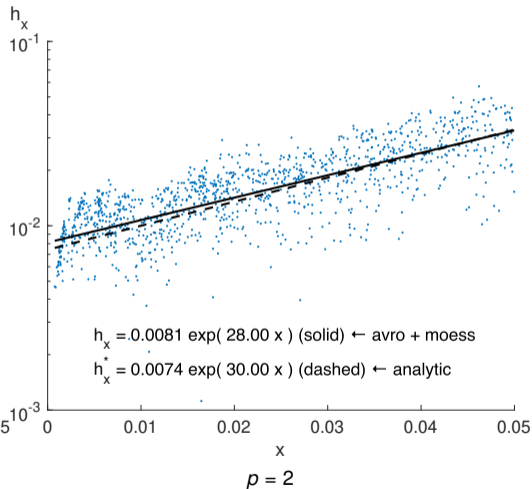
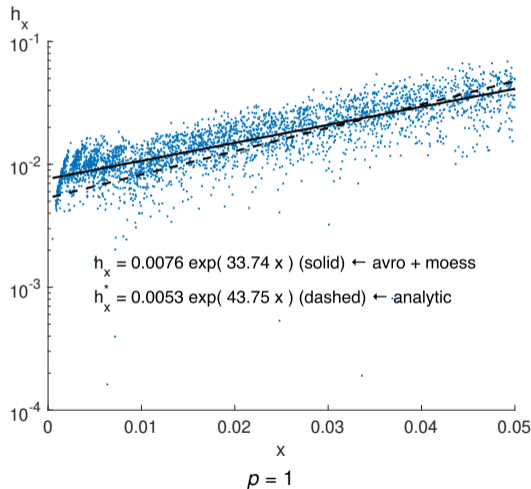


$z = 1$

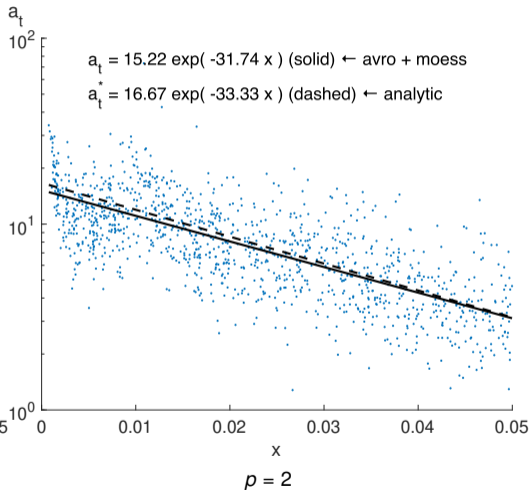
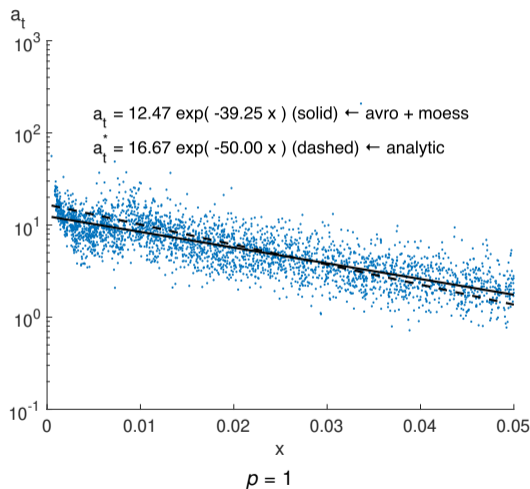


$t = 1$

Analytic mesh distributions achieved.

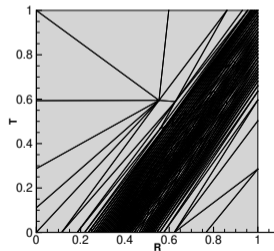
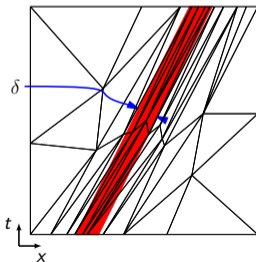
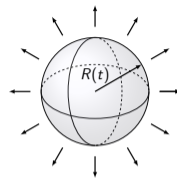


Analytic mesh distributions achieved.

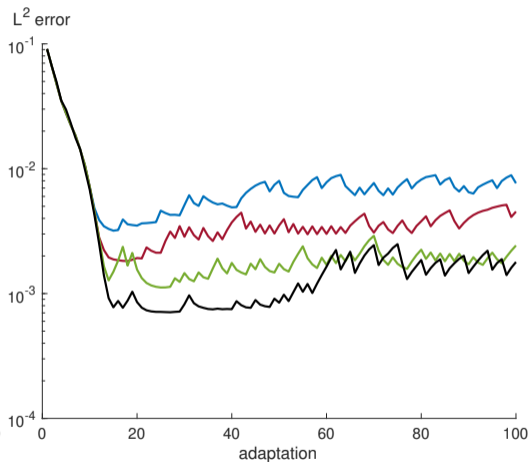
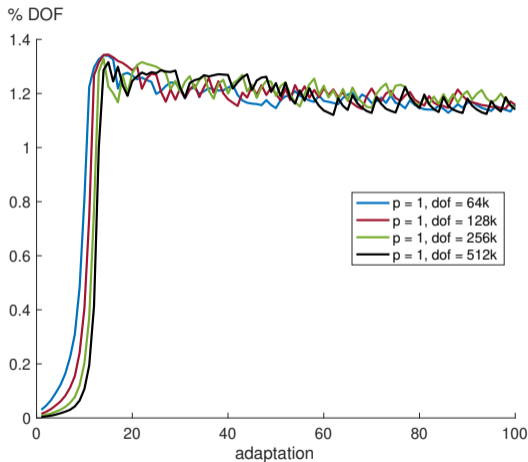


Adapt to the exact L^2 error of a spherical wave function.

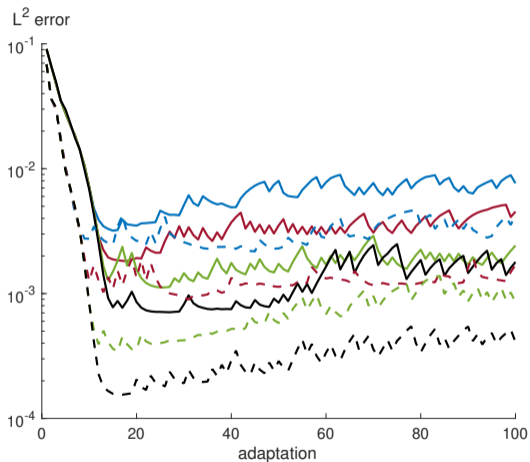
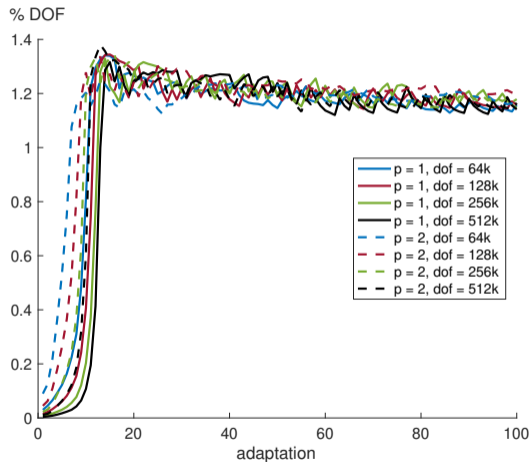
$$u(\mathbf{x}, t) = \exp(-t) \exp\left(-200(R(t) - \|\mathbf{x}\|)^2\right) \quad \text{with } R(t) = 0.4 + 0.7t$$



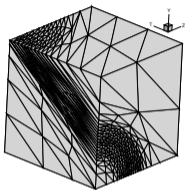
DOF overshoot complemented by slight rise in error.



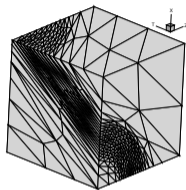
DOF overshoot complemented by slight rise in error.



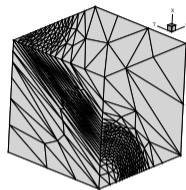
$\rho = 1$ 512k DOF optimized meshes show expected refinement.



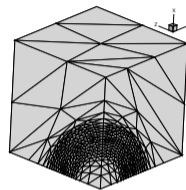
$x = 0$



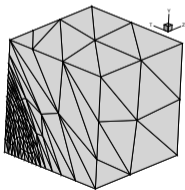
$y = 0$



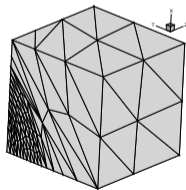
$z = 0$



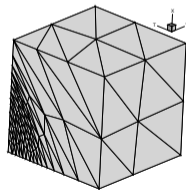
$t = 0$



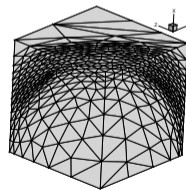
$x = 1$



$y = 1$

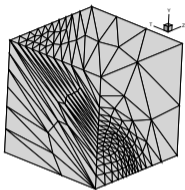


$z = 1$

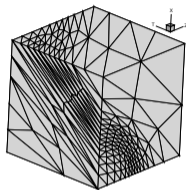


$t = 1$

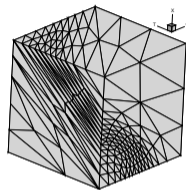
$p = 2$ 512k DOF optimized meshes show expected refinement.



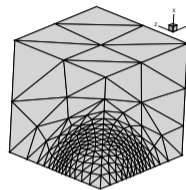
$x = 0$



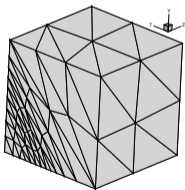
$y = 0$



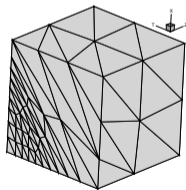
$z = 0$



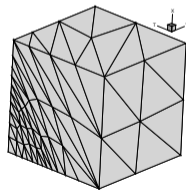
$t = 0$



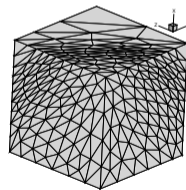
$x = 1$



$y = 1$

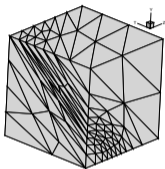


$z = 1$

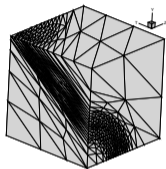


$t = 1$

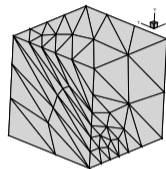
Metric conformity is good.



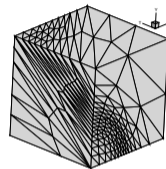
$p = 1$, 64k DOF



$p = 1$, 512k DOF



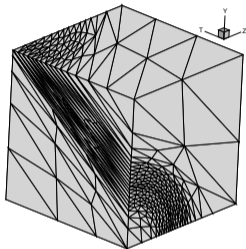
$p = 2$, 64k DOF



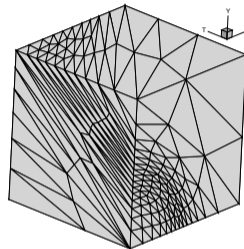
$p = 2$, 512k DOF

	$\%l_{\text{unit}}$	$\%q_{\text{unit}}$	# simplices	% overshoot
$p = 1$, 64k	97.72 %	53.31 %	14.83k	15.84 %
$p = 1$, 512k	97.57 %	51.76 %	116.86k	14.12 %
$p = 2$, 64k	98.71 %	50.15 %	4.87k	14.12 %
$p = 2$, 512k	97.40 %	49.63 %	39.76k	16.47 %

Large aspect ratios are obtained.



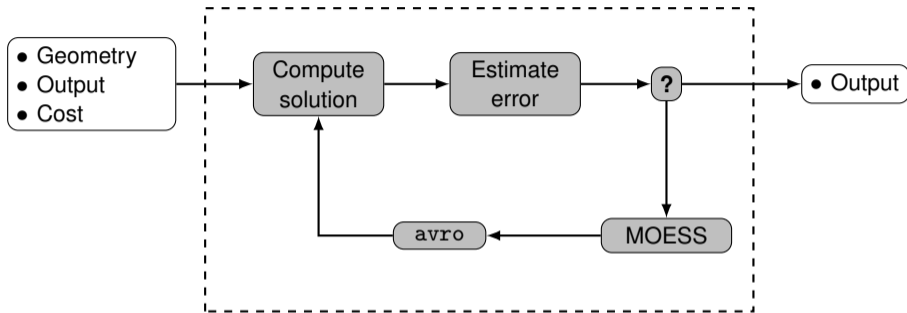
$p = 1$, 512k DOF



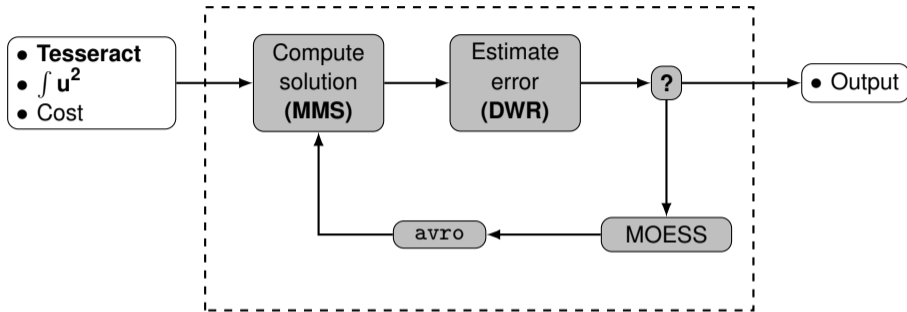
$p = 2$, 512k DOF

Order / DOF	64k	128k	256k	512k
$p = 1$	2.07e+02	5.07e+02	8.32e+02	2.12e+03
$p = 2$	6.53e+01	1.26e+02	2.60e+02	5.88e+02

Unsteady advection-diffusion with a boundary layer.

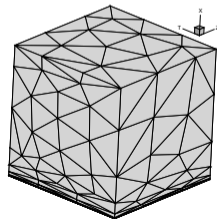


Unsteady advection-diffusion with a boundary layer.

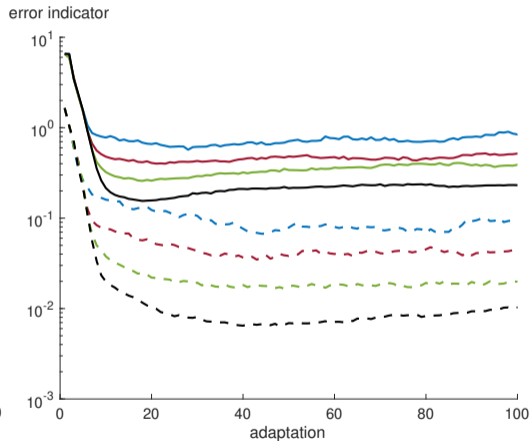
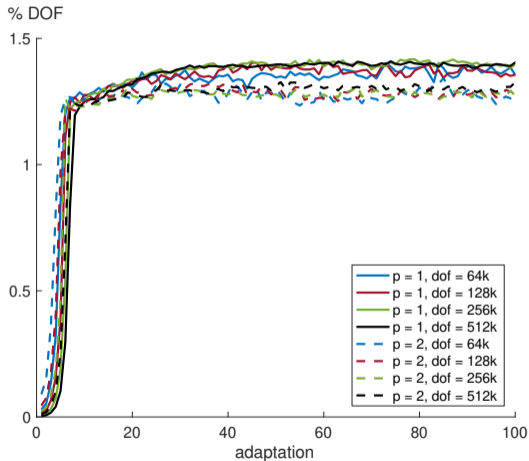


$$\frac{\partial u}{\partial t} + \nabla \cdot (\mathbf{c} u - \nabla u) = s(\mathbf{x}, t) \quad \mathbf{c} = (0.5, 0.5, 0.5)^t$$

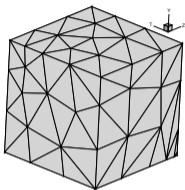
$$u(x, y, z, t) = \exp(-x/\epsilon) + \frac{(2y)^{\rho+1}}{(\rho+1)!} + \frac{(4z)^{\rho+1}}{(\rho+1)!} + \frac{(6t)^{\rho+1}}{(\rho+1)!} \quad \epsilon = 0.01$$



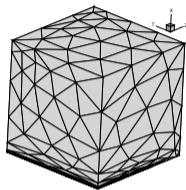
DOF overshoot as high as 43%.



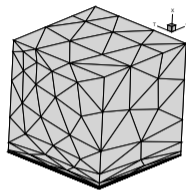
$p = 2$ 512k DOF optimized meshes show expected refinement.



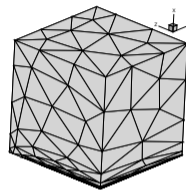
$x = 0$



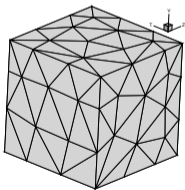
$y = 0$



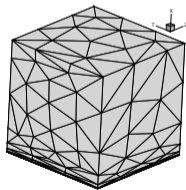
$z = 0$



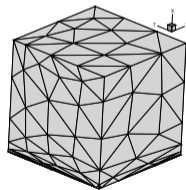
$t = 0$



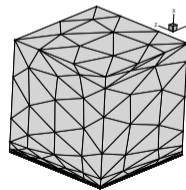
$x = 1$



$y = 1$

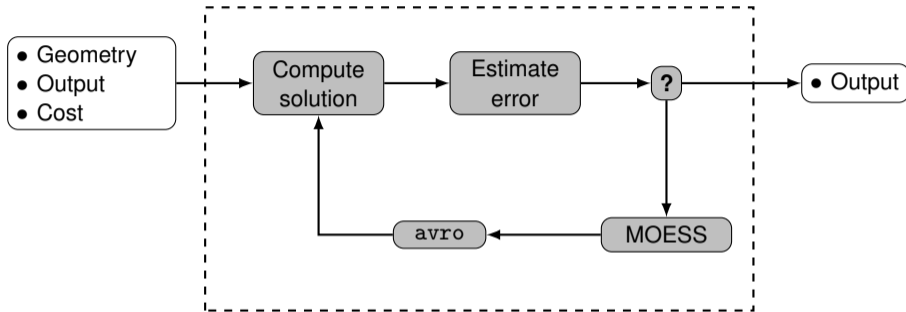


$z = 1$

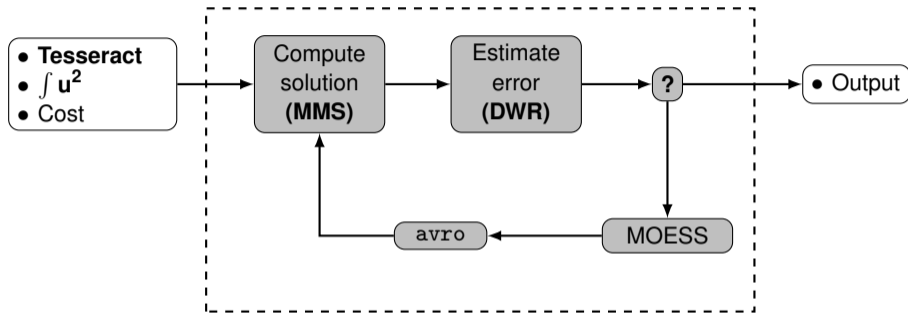


$t = 1$

Unsteady advection-diffusion with an expanding spherical wave.

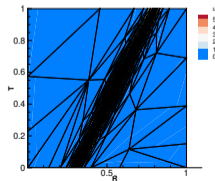
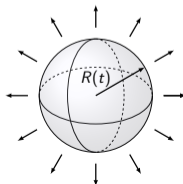


Unsteady advection-diffusion with an expanding spherical wave.

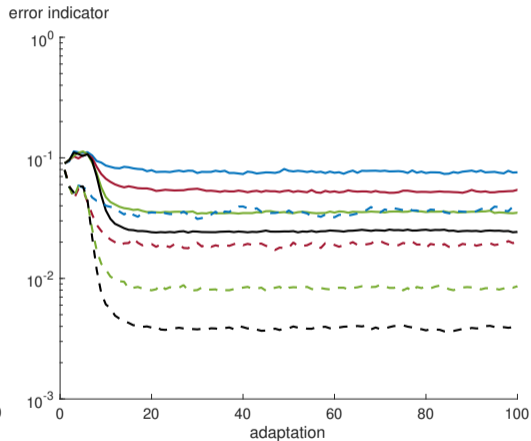
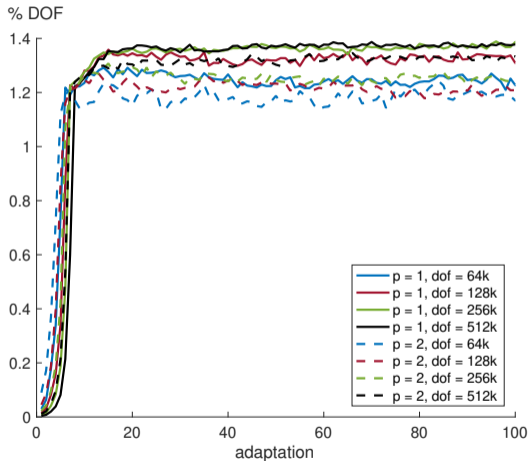


$$\frac{\partial u}{\partial t} + \nabla \cdot (\mathbf{c} u - \nu \nabla u) = s(\mathbf{x}, t) \quad \mathbf{c} = 0.5\mathbf{e}_r, \quad \nu = 0.01$$

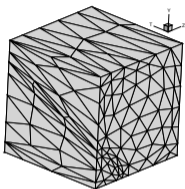
$$u(\mathbf{x}, t) = 5 \exp(-0.1t) \exp\left(-1000(R(t) - \|\mathbf{x}\|)^2\right) \quad R(t) = 0.3 + 0.5t$$



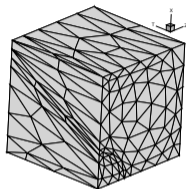
DOF overshoot as high as 38%.



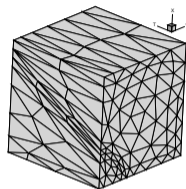
$\rho = 1$ 512k DOF optimized meshes show expected refinement.



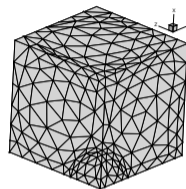
$x = 0$



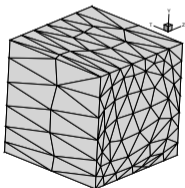
$y = 0$



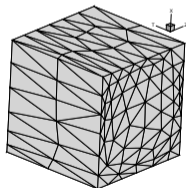
$z = 0$



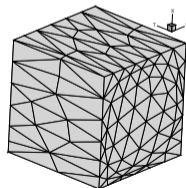
$t = 0$



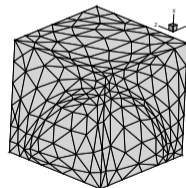
$x = 1$



$y = 1$



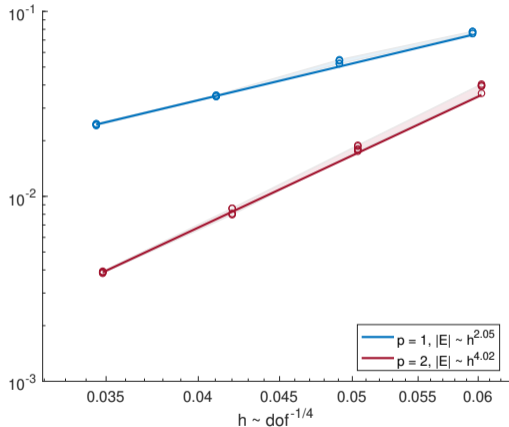
$z = 1$



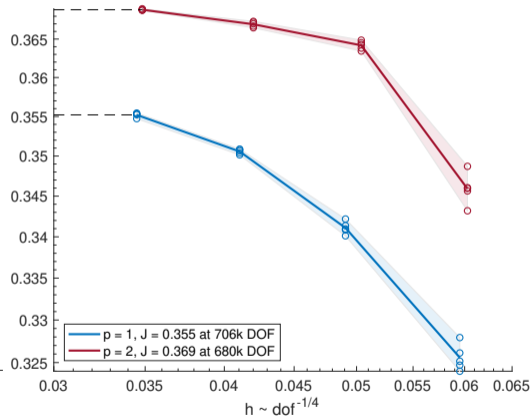
$t = 1$

Need more DOF to fully refine solution.

error indicator



$J(u)$



Spherical wave at $t = 1$ is captured.



Conclusions



Introduction

Mesh adaptation
algorithm

Demonstration with
analytic metrics

Demonstration
within adaptive
framework

Conclusions

Main objective:

Develop an anisotropic four-dimensional meshing capability for adaptive numerical simulations.

Contributions:

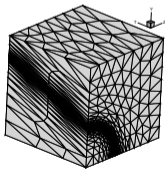
- (1) Develop an algorithm and software for $4d$ **metric-conforming mesh adaptation**.
- (2) **Validate the adaptive algorithm** on $4d$ problems.
- (3) Demonstrate first **PDE-driven anisotropic** unstructured adaptation for **unsteady $3d$ problems**.

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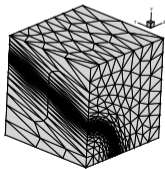
metric-conforming

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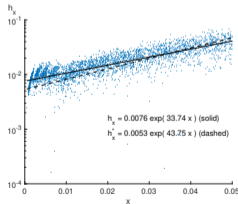
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metric-conforming



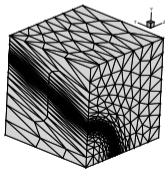
validation

Main objective:

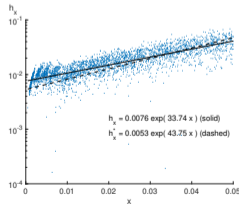
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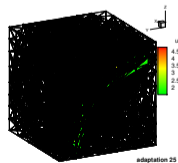
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metric-conforming



validation



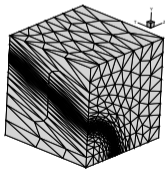
PDE-driven

Main objective:

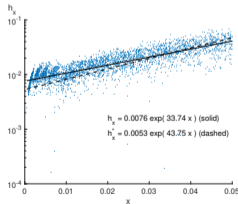
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Contributions:

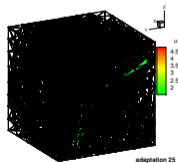
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metric-conforming



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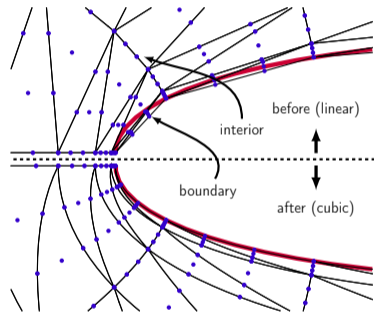


PDE-driven



avro

- **Parallelization** of the mesh adaptation algorithm,
- Adaptation of **curvilinear** meshes,
- Applications of the framework to **other PDEs**,
- Investigating **continuous Galerkin** discretization,
- Mesh adaptation for **higher-dimensional parameter spaces**.



Acknowledgements

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CAPS: Computational Aircraft Prototype Syntheses with Dean Bryson as technical monitor.

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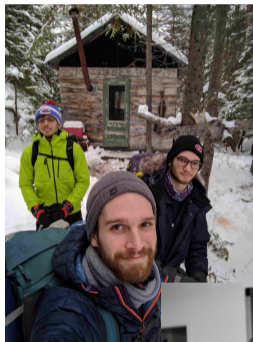
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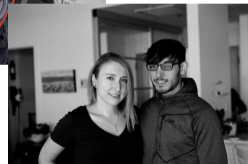
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Maman

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Ferran

Irina

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Questions?



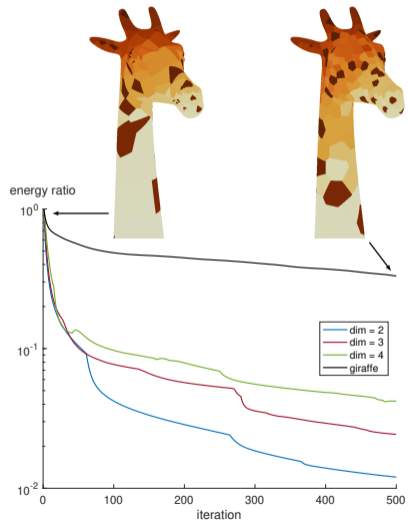
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[49\(1\):38–60.](#)
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[Computer-Aided Design, 85:53 – 67.](#)
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2-D, 3-D and 4-D Anisotropic Mesh Adaptation for the Time-Continuous Space-Time Finite Element Method with Applications to the Incompressible Navier-Stokes Equations.
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An Optimization Framework for Adaptive Higher-Order Discretizations of Partial Differential Equations on Anisotropic Simplex Meshes.
[PhD thesis, Massachusetts Institute of Technology, Department of Aeronautics and Astronautics.](#)

Why the giraffe?



Why the giraffe?



Why the $\sqrt{2}$?

[Frey and George, 2008]: *The coefficient $\sqrt{2}$ is related to the fact that an edge can be split if the lengths of the two sub-edges minimize the error distance to the unit length as compared with the initial length.*

In other words:

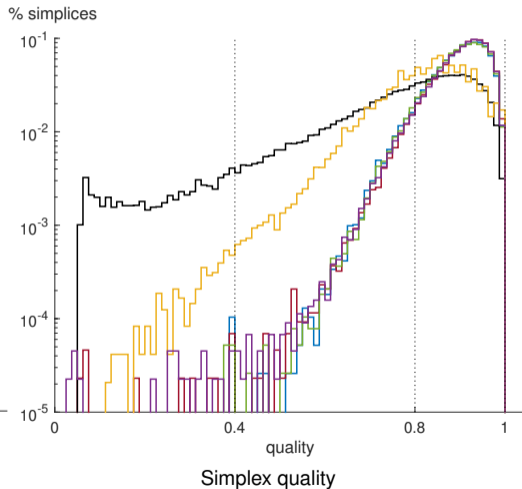
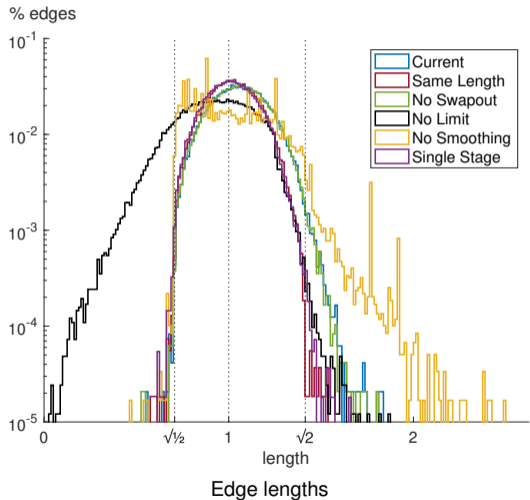
- We want to minimize ℓ_{split} :

$$\ell_{\text{split}} = \arg \min_{\alpha} \alpha \quad \text{such that } g(\alpha) = 2 \left(\frac{\alpha}{2} - 1 \right)^2 - (\alpha - 1)^2 \leq 0$$

Applying the KKT conditions leads to $g(\alpha) = 0 \rightarrow \ell_{\text{split}} = \sqrt{2}$.

- We want to maximize ℓ_{collapse} but if we make it too big, then we will cycle between splitting and collapsing because splitting can create edges with length $\ell_{\text{split}}/2$. We can avoid this by setting $\ell_{\text{collapse}} = \sqrt{2}/2$.

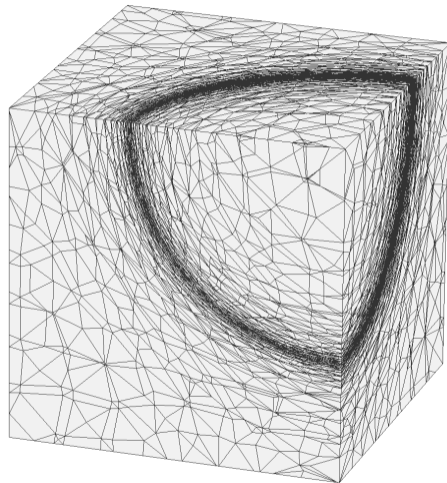
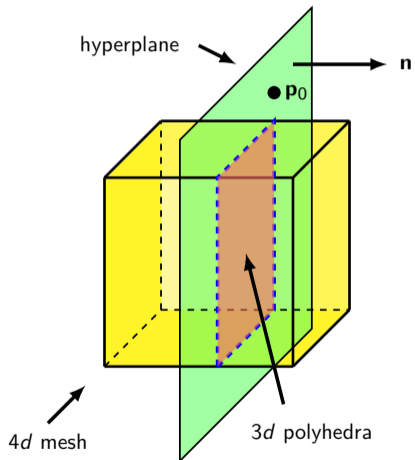
Study of mesh adaptation components



Study of mesh adaptation components

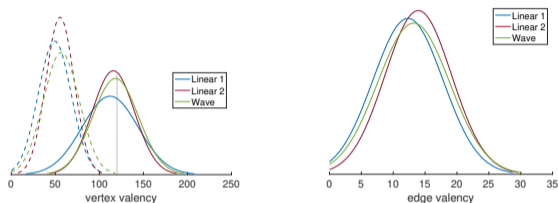
Property	Current	Variant 1 Same Length	Variant 2 No Swapout	Variant 3 No Limit	Variant 4 No Smoothing	Variant 5 Single Stage
l_{\min}	0.67	0.57	0.64	0.06	0.58	0.55
l_{\max}	1.77	1.66	1.69	1.62	2.50	1.72
l_{avg}	1.06	1.03	1.06	0.95	1.03	1.02
$\%l_{\text{unit}}$	99.10 %	99.92 %	99.06 %	89.60 %	96.48 %	99.74 %
q_{\min}	0.47	0.09	0.46	0.10	0.09	0.17
q_{avg}	0.90	0.90	0.90	0.80	0.83	0.90
$\%q_{\text{unit}}$	92.15 %	93.46 %	92.08 %	62.45 %	67.44 %	92.25 %
# simplices	38.30k	42.67k	38.20k	58.62k	47.65k	43.60k

Visualizing slices of a 4d mesh



Why so much overshoot?

Valency statistics



Packing fraction

Packing fraction (ϕ) defined as volume fraction of space covered by particles.

Best known packing fractions of unit-length equilateral simplices **[Kallus, 2011]**:

$$\phi_3 = 100/117 \approx 85.47\%$$

$$\phi_4 = 128/219 \approx 58.45\%$$

Maintaining a valid mesh

Theory

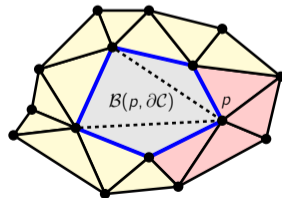
Definition of a mesh topology:

Let \mathcal{V} be a set of vertices in some domain Ω and \mathcal{T} be a set of n -polytopes with vertices in \mathcal{V} . Let \mathcal{F} be the set of faces ($n-1$)-facets of \mathcal{T} . \mathcal{T} is called a *mesh topology* if

- (1) $\text{card}(f \cap \mathcal{T}) \leq 2, \forall f \in \mathcal{F}$,
- (2) $(\mathcal{V}, \partial\mathcal{T})$ is a mesh of $\partial\Omega$.

Practice

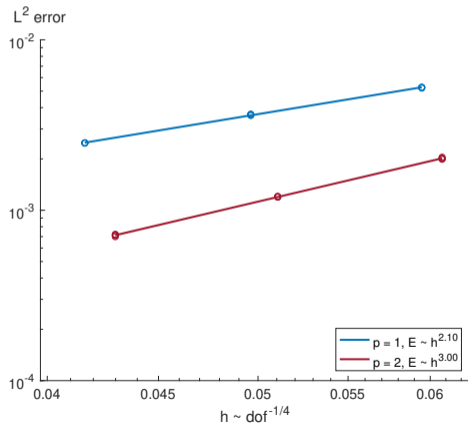
- (1) Close the mesh by connecting boundary to a ghost vertex.
- (2) Check inserted elements do not already exist in \mathcal{T} .
- (3) Use neighbour relations to ensure every facet touches two elements.



Convergence of the L^2 error for simple problem.

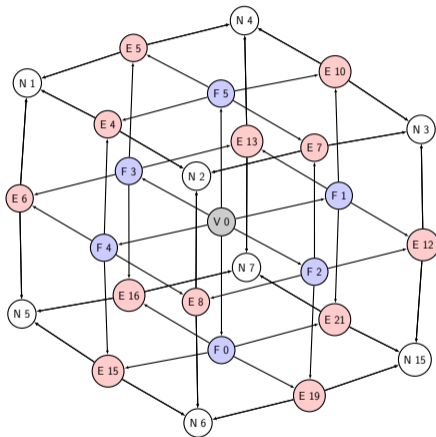
Adapting to L^2 error in solution:

$$u(x, y, z, t) = \exp(-5t) \sin 2\pi x \sin 2\pi y \sin 2\pi z$$



Geometry metadata is important to determine validity of operators.

- **Collapse edge** $e = (p, q)$: $g_q \preceq g_p$,
- **Insert vertex** p along edge e : $g_p \leftarrow g_e$,
- **Swap edge** e with re-insertion vertex p :
 $g_p \preceq g_e$,
- **Smooth vertex** p : driven by lengths of edges
 (p, q) surrounding p such that $g_q \preceq g_p$.



Difference in 4d benchmarks with or without DOF control.

Property	Linear 1 (no control)	Linear 2 (no control)	Wave (no control)	Linear 1 (control)	Linear 2 (control)	Wave (control)
l_{\min}	0.50	0.40	0.20	0.53	0.47	0.23
l_{\max}	1.91	1.86	2.71	1.78	1.96	2.99
l_{avg}	1.08	1.08	1.08	1.10	1.11	1.12
$\%l_{\text{unit}}$	97.29 %	98.01 %	92.48 %	96.46 %	97.27 %	88.96 %
q_{\min}	0.16	0.02	0.02	0.23	0.11	0.08
q_{avg}	0.80	0.83	0.72	0.80	0.83	0.72
$\%q_{\text{unit}}$	56.67 %	67.13 %	28.75 %	56.35 %	70.00 %	26.89 %
# simplices	59.56k	915.30k	394.07k	55.66k	814.50k	347.19k
expected	51.00k	818.00k	n/a	51.00k	818.00k	n/a

Mesh size and aspect ratio distributions for L^2 boundary layer case

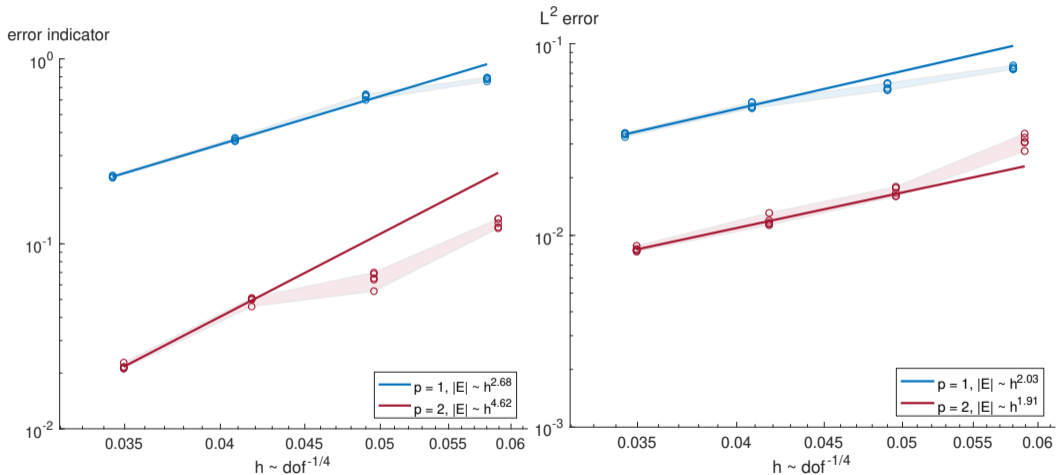
Coefficient	Analytic	64k	128k	256k	512k
$h_{x,0}$	$h_{x,0}^*$	0.0155	0.0121	0.0095	0.0076
$h_{x,0}^*$	-	0.0090	0.0075	0.0063	0.0053
k_{h_x}	43.75	25.14	28.39	31.29	33.74
$a_{y,0}$	50.00	25.52	29.46	32.15	33.76
k_{a_y}	-50.00	-26.57	-31.14	-35.22	-38.05
$a_{z,0}$	25.00	14.32	16.02	17.09	18.18
k_{a_z}	-50.00	-26.84	-31.79	-35.92	-38.92
$a_{t,0}$	16.67	9.59	10.75	11.76	12.47
k_{a_t}	-50.00	-26.11	-31.44	-35.58	-39.25

$p = 1$

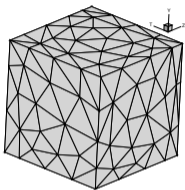
Coefficient	Analytic	64k	128k	256k	512k
$h_{x,0}$	$h_{x,0}^*$	0.0146	0.0119	0.0100	0.0081
$h_{x,0}^*$	-	0.0125	0.0105	0.0088	0.0074
k_{h_x}	30.00	25.46	24.56	27.36	28.00
$a_{y,0}$	50.00	38.06	42.92	40.45	44.13
k_{a_y}	-33.33	-24.69	-28.12	-29.80	-31.04
$a_{z,0}$	25.00	22.41	21.51	22.11	22.06
k_{a_z}	-33.33	-25.96	-27.69	-30.55	-30.68
$a_{t,0}$	16.67	15.39	14.55	14.64	15.22
k_{a_t}	-33.33	-26.60	-28.50	-30.14	-31.74

$p = 2$

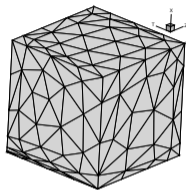
Advection-diffusion boundary layer convergence rates



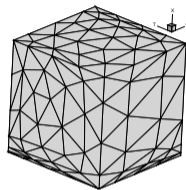
Boundary layer (advection-diffusion) $p = 2$ meshes.



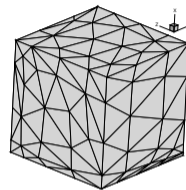
$x = 0$



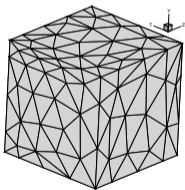
$y = 0$



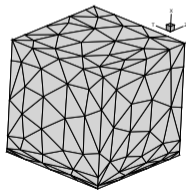
$z = 0$



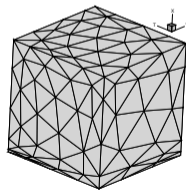
$t = 0$



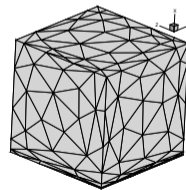
$x = 1$



$y = 1$

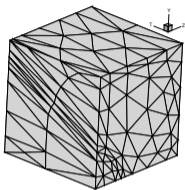


$z = 1$

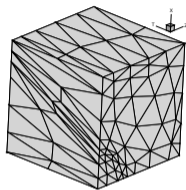


$t = 1$

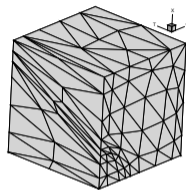
Spherical wave (advection-diffusion) $p = 2$ meshes.



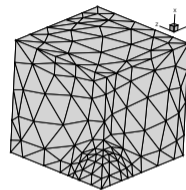
$x = 0$



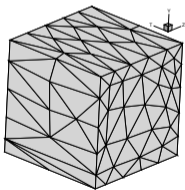
$y = 0$



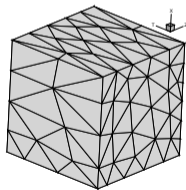
$z = 0$



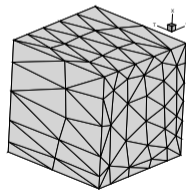
$t = 0$



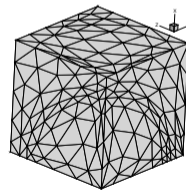
$x = 1$



$y = 1$



$z = 1$



$t = 1$