

# High-order Solution Transfer between Curved Meshes and Ill-conditioned Bézier Curve Intersection

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# Outline

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1. Introduction and motivation
2. Curved Elements
3. Solution Transfer
4. Ill-conditioned Bézier Curve Intersection
5. Compensated Evaluation
6. Modified Newton's for Intersection

## Introduction and motivation

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# Method of Characteristics

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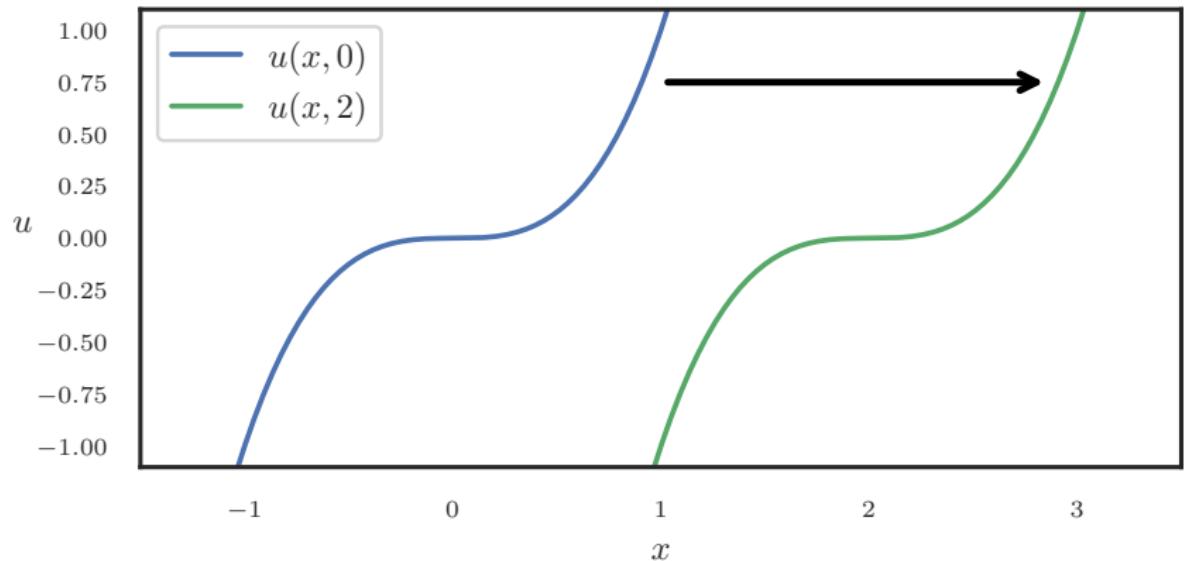
Divide physical domain

$$x(t) = x_0 + ct$$

PDE becomes a (trivial) ODE

$$\frac{d}{dt}u(x(t), t) = 0.$$

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- Transform PDE to family of ODEs

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  - Resolve sensitive features

## Remeshing Example

Consider

$$u_t + \begin{bmatrix} y^2 \\ 1 \end{bmatrix} \cdot \nabla u + F(u, \nabla u) = 0$$

## Remeshing Example

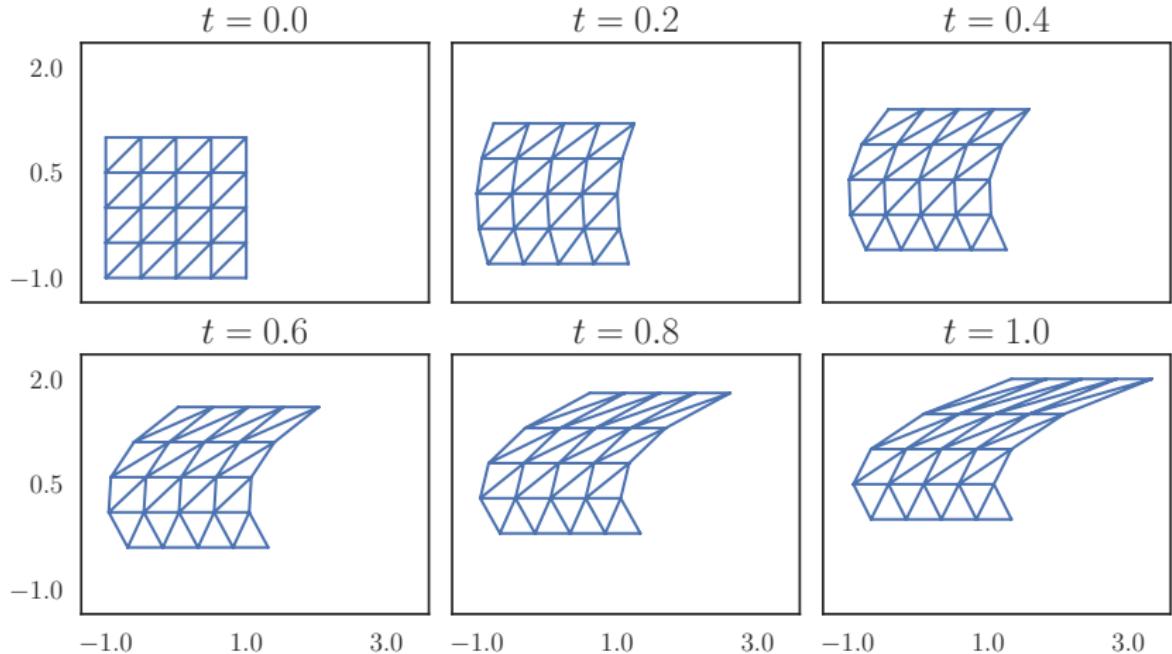
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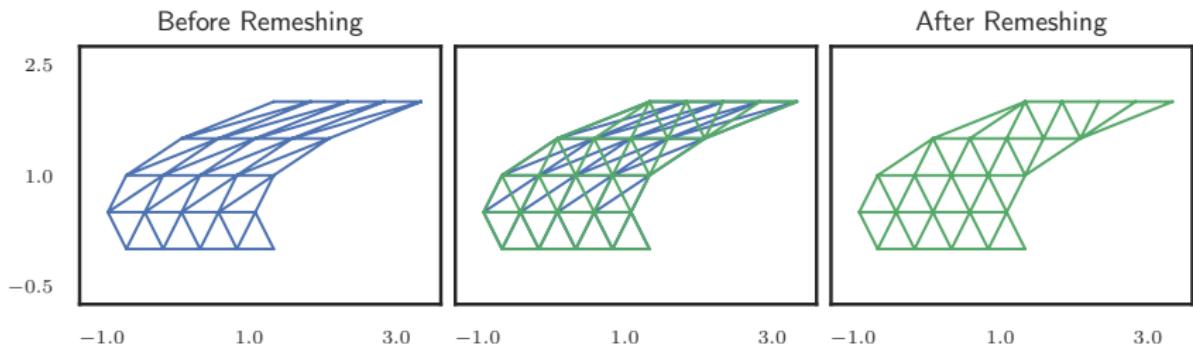
with cubic characteristics

$$\begin{bmatrix} x(t) \\ y(t) \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} + \frac{1}{3} \begin{bmatrix} (y_0 + t)^3 - y_0^3 \\ 3t \end{bmatrix}.$$

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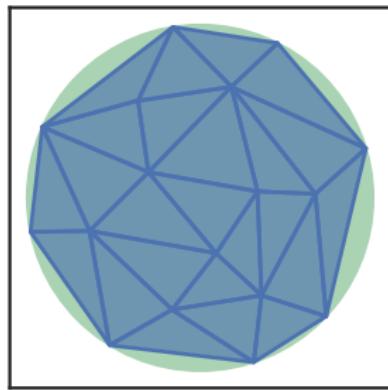
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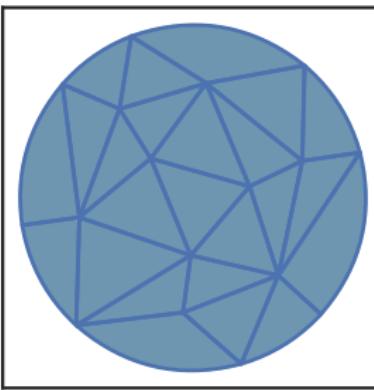
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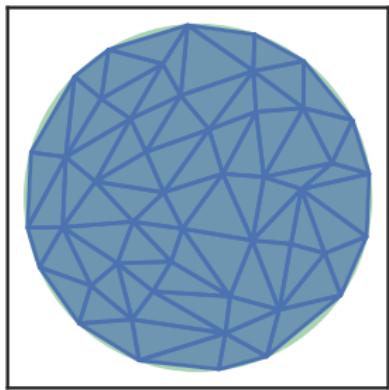
Linear Mesh, 24 elements



Quadratic Mesh, 24 elements



Linear Mesh, 74 elements



-1 0 1 -1 0 1 -1 0 1

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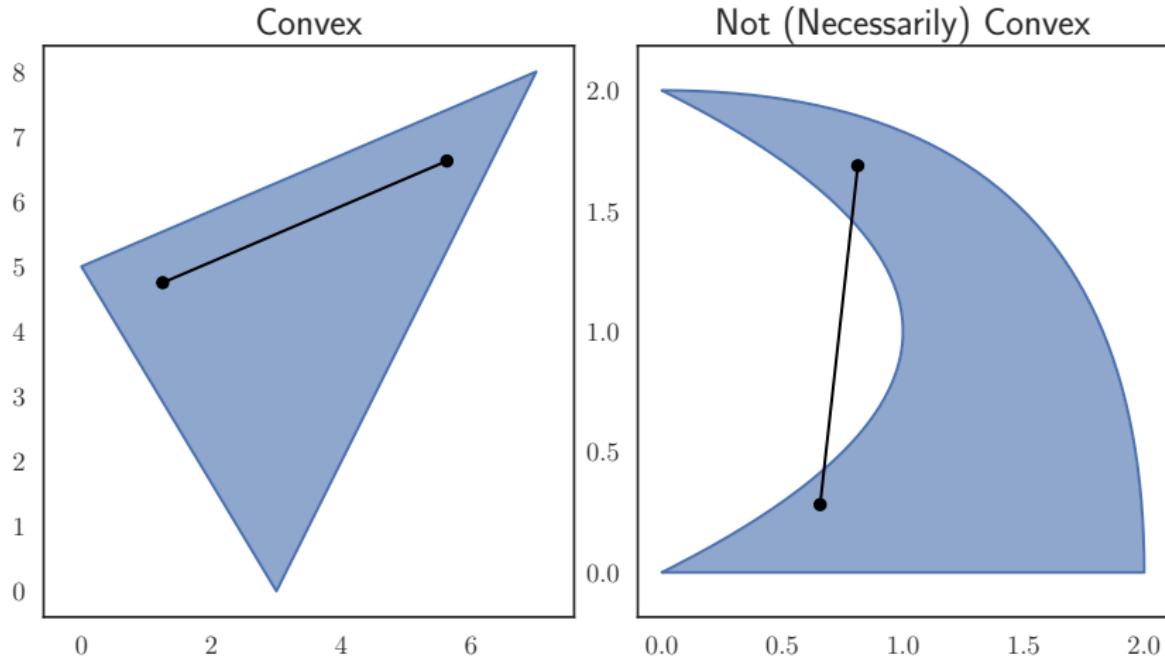
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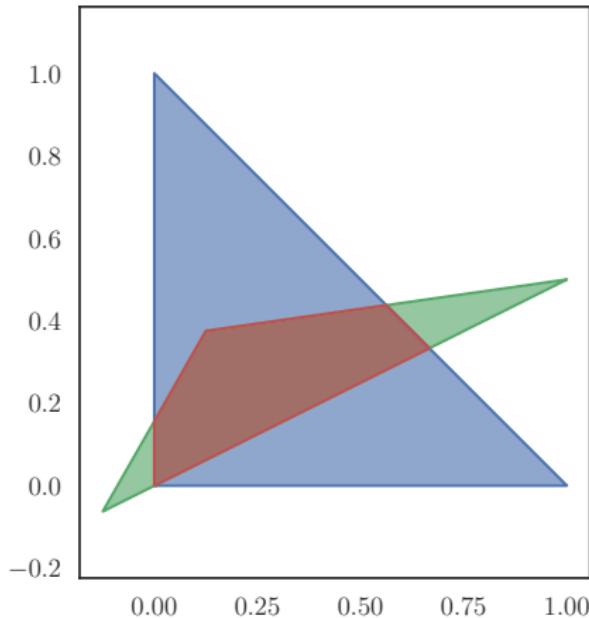
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  - Loss of accuracy in high degree (e.g. Runge's phenomenon)
  - More challenging geometry

# Curved Meshes

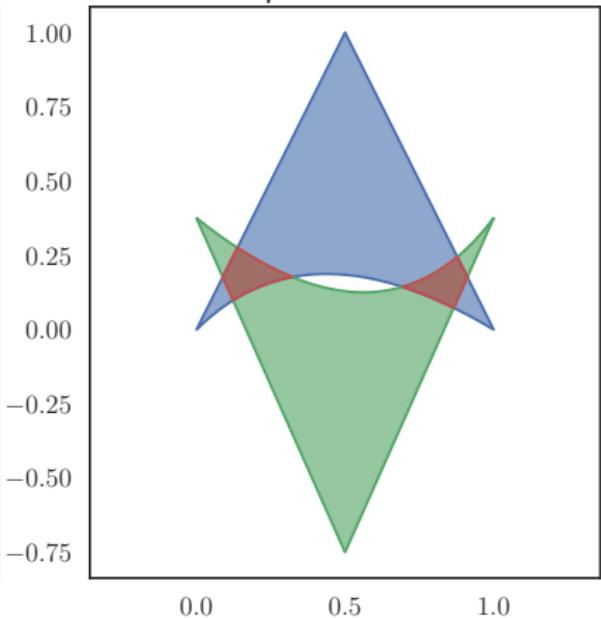


# Curved Meshes

Convex Intersection



Multiple Intersections



# Curved Elements

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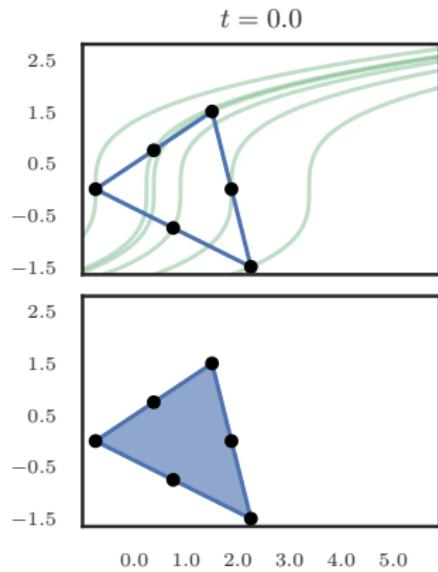
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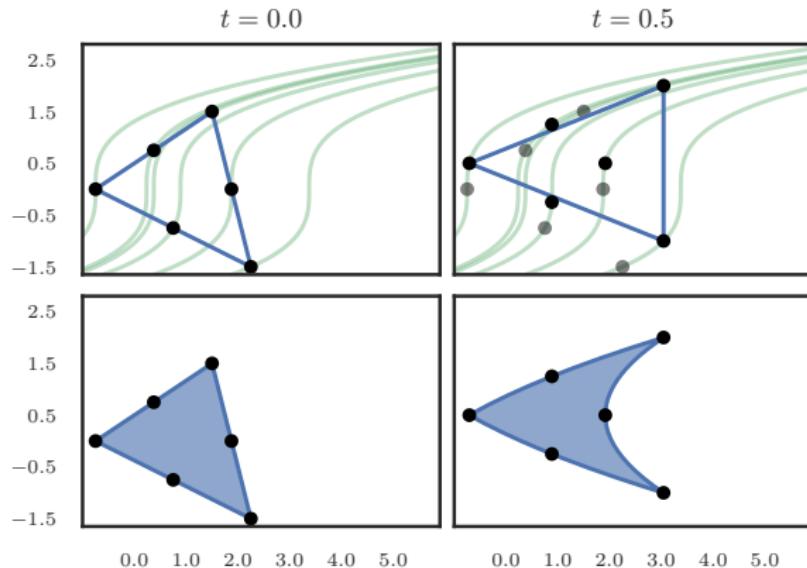
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- Lagrangian method must either curve mesh or information about flow of geometry will be lost

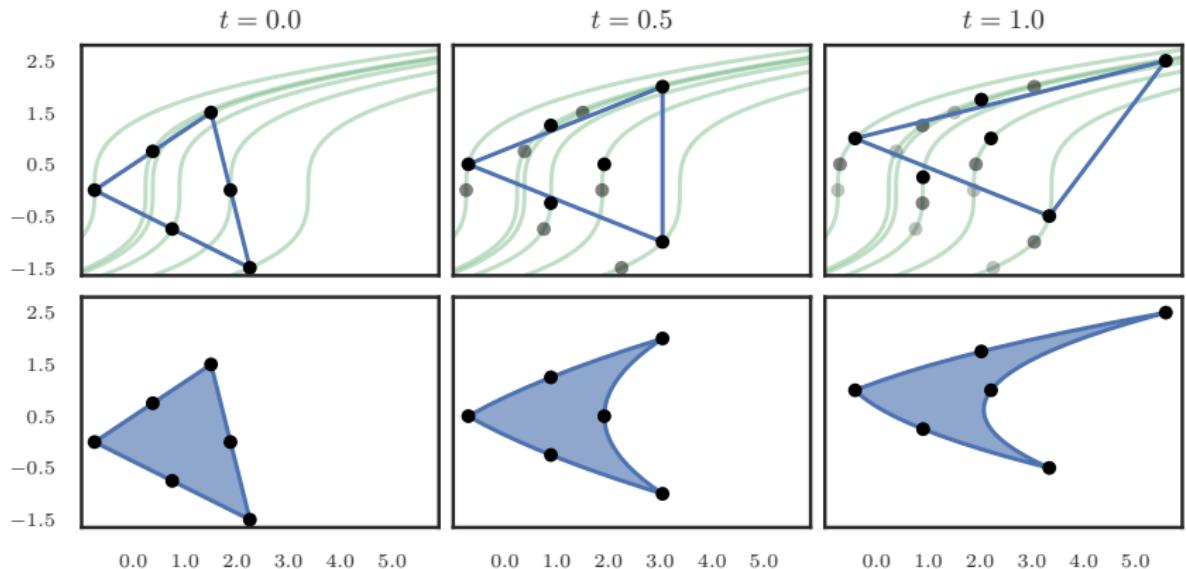
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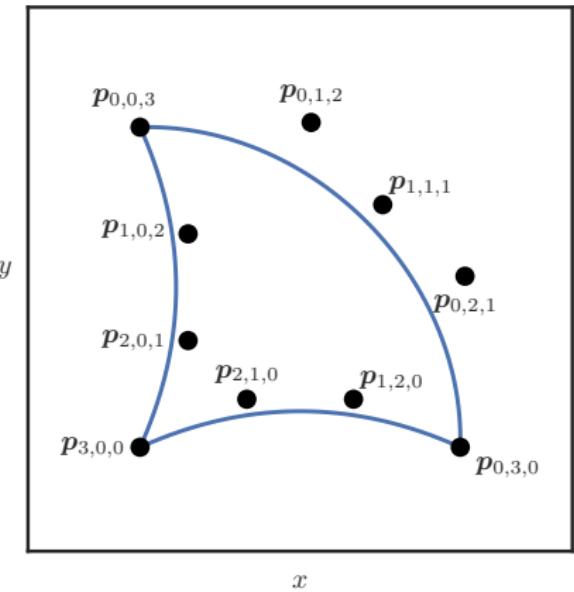
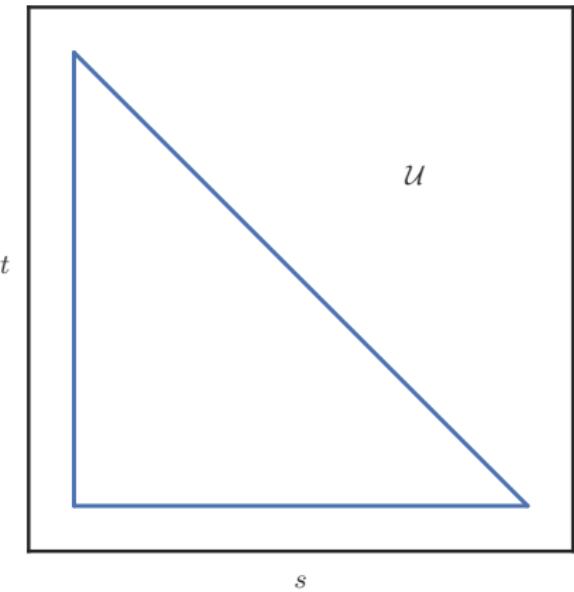
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- Convex combination of control points

$$b(s, t) = \sum_{\substack{i+j+k=n \\ i,j,k \geq 0}} \binom{n}{i,j,k} \lambda_1^i \lambda_2^j \lambda_3^k \mathbf{p}_{i,j,k}$$

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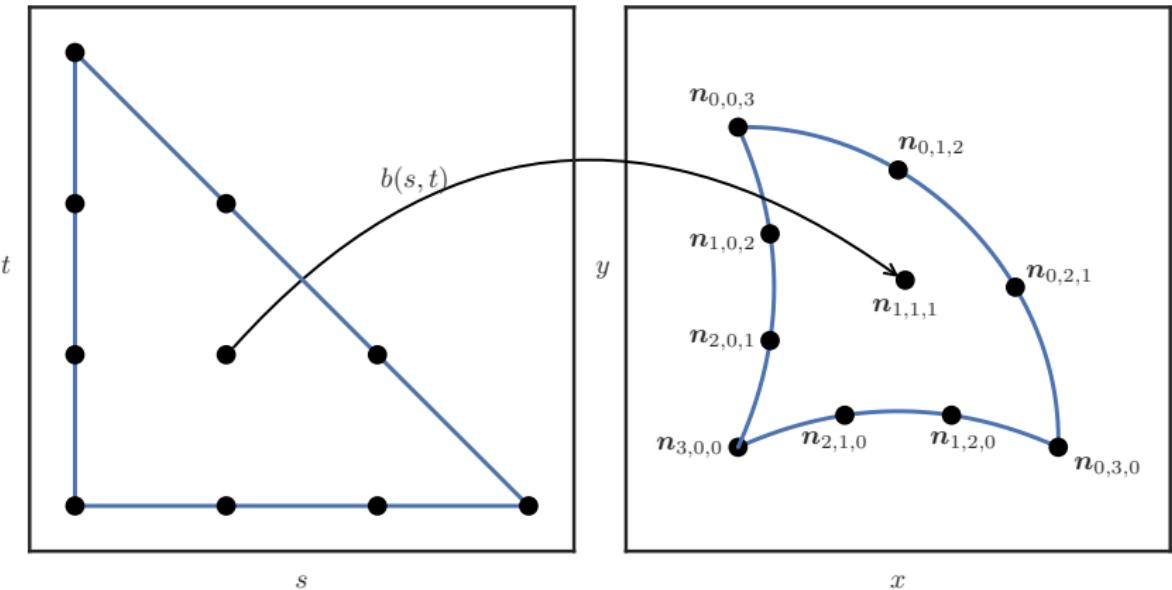
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- Conversion between  $\mathbf{n}_{i,j,k}$  and  $\mathbf{p}_{i,j,k}$  has condition number exponential in  $n$

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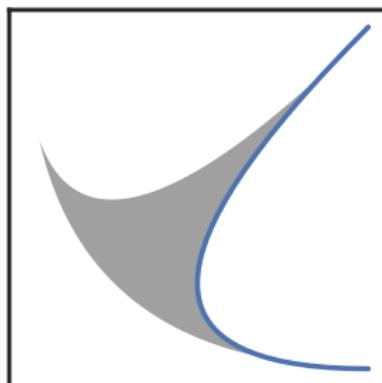
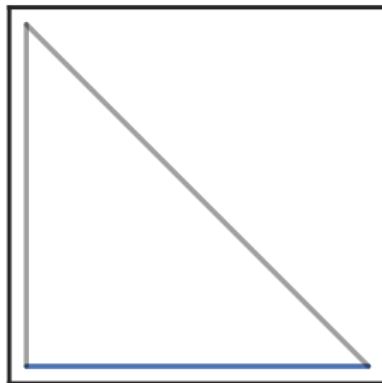
- Element  $\mathcal{T}$  is **valid** if diffeomorphic to  $\mathcal{U}$
- $b(s, t)$  bijective, i.e. Jacobian  $Db$  is everywhere invertible
- $\det(Db)$  positive, preserves orientation

## Inverted Element

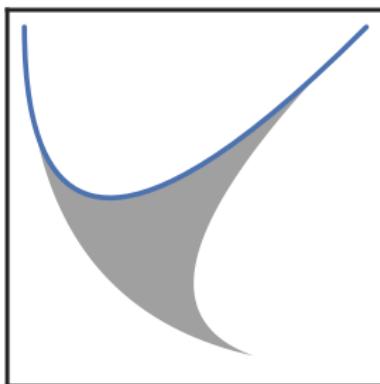
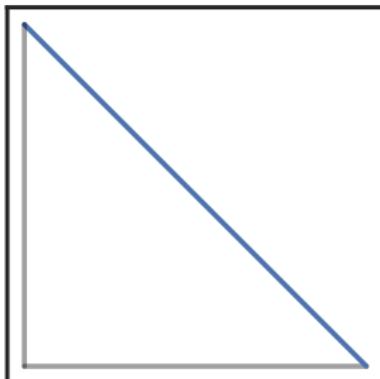
Consider element given by map

$$b(s, t) = \lambda_1^2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \lambda_2^2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \lambda_3^2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

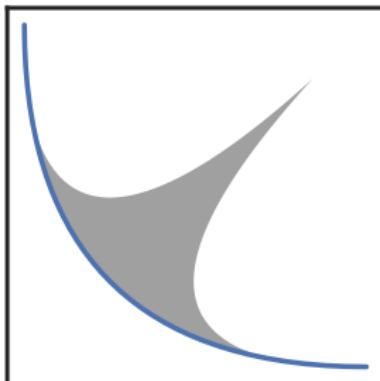
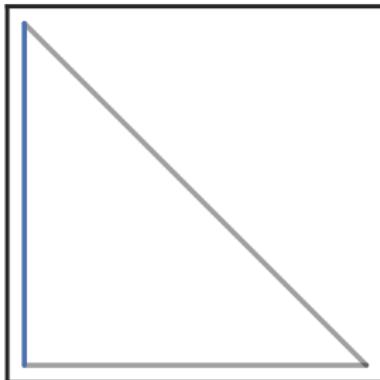
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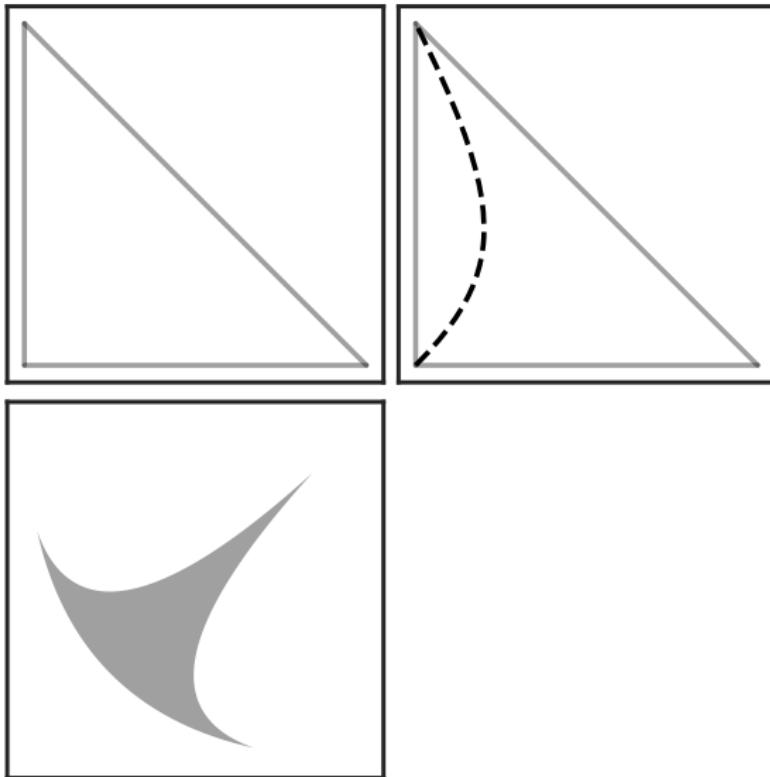
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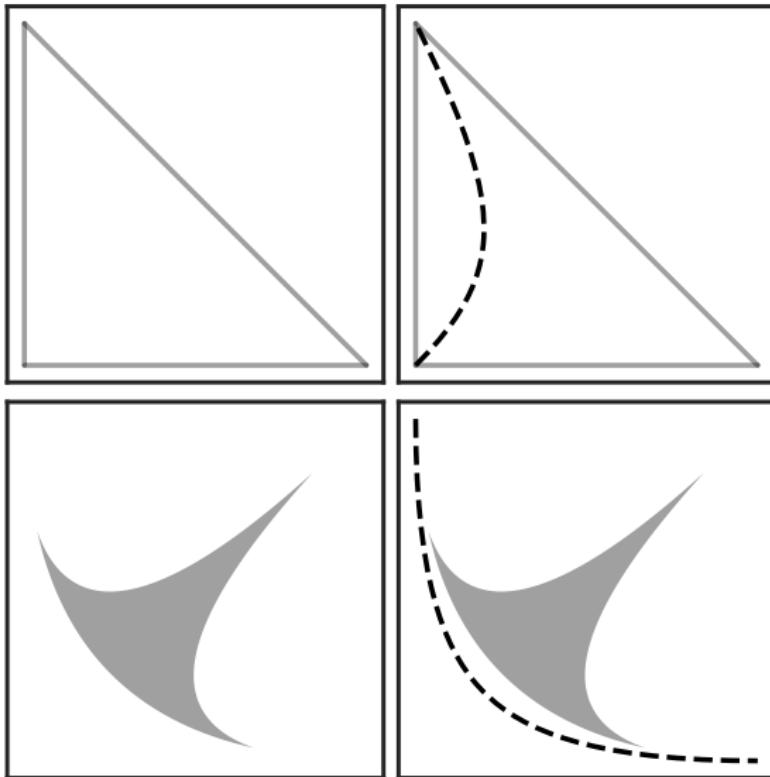
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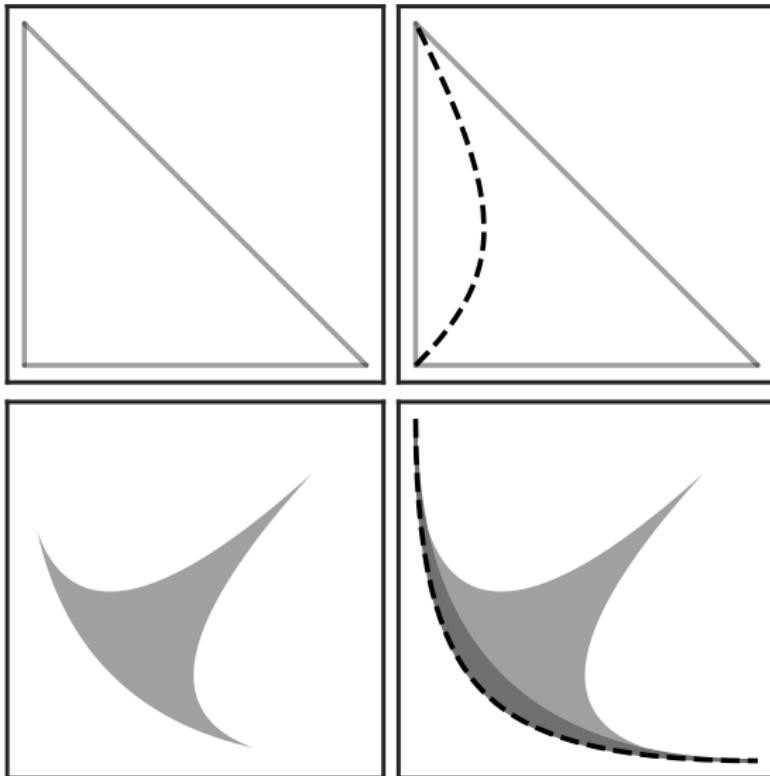
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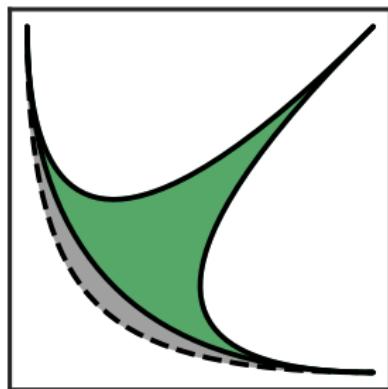
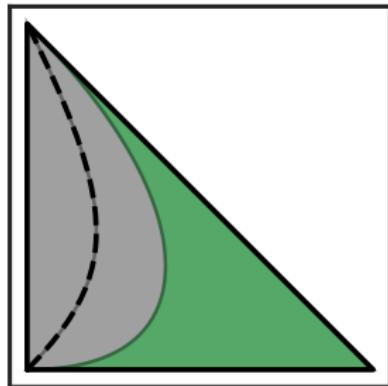
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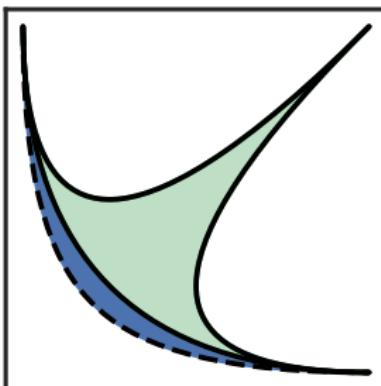
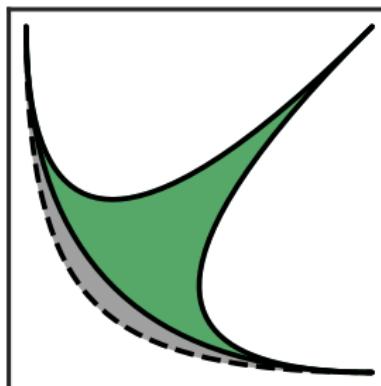
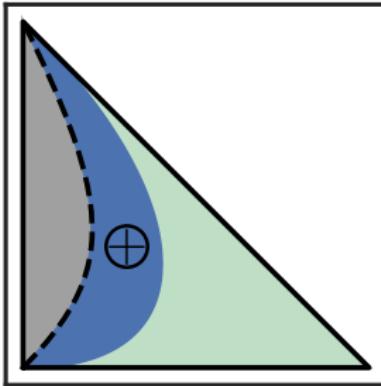
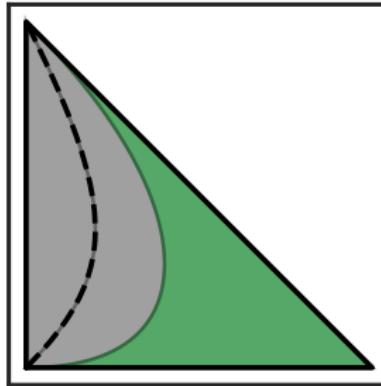
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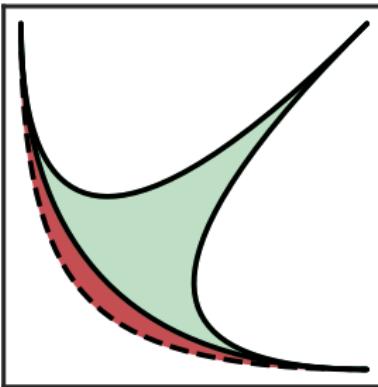
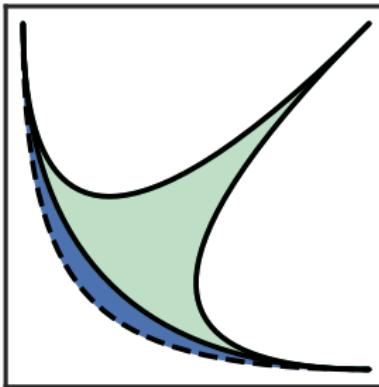
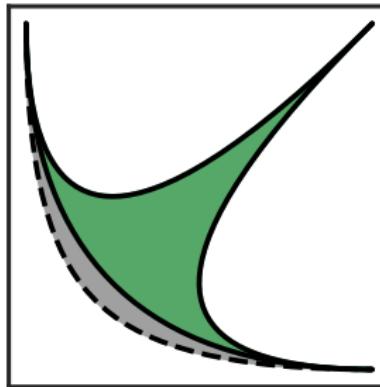
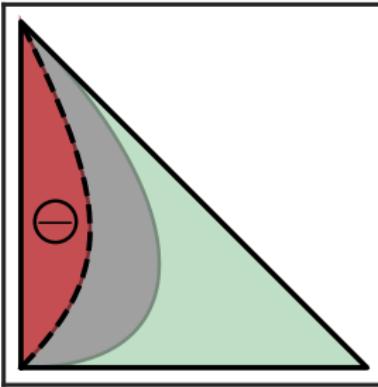
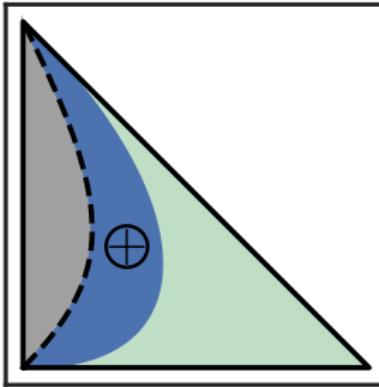
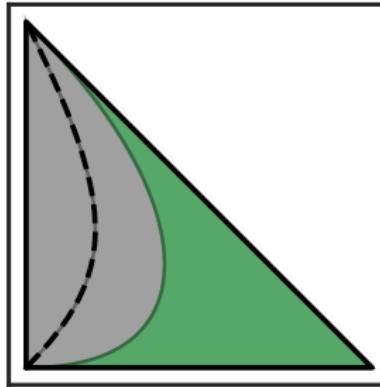
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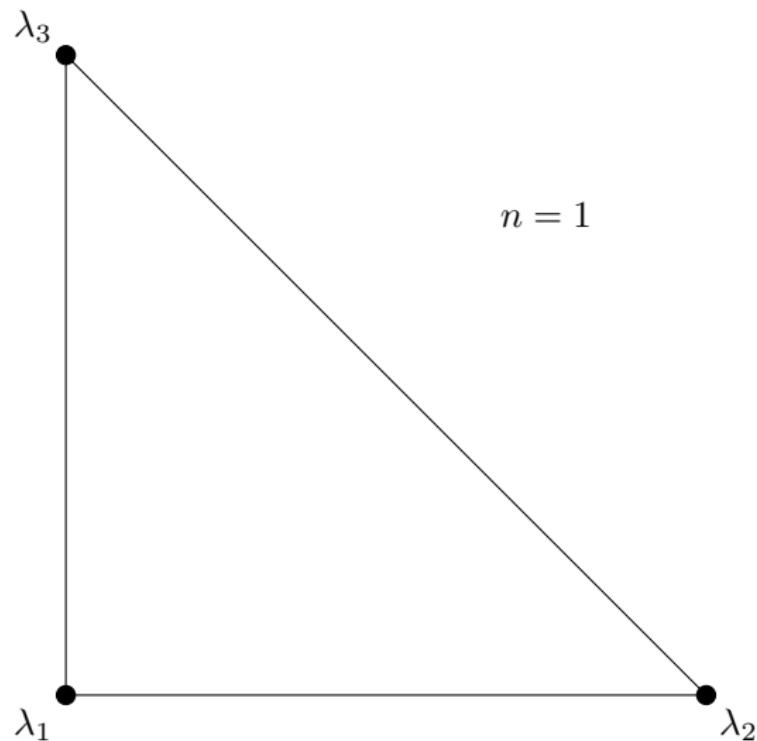
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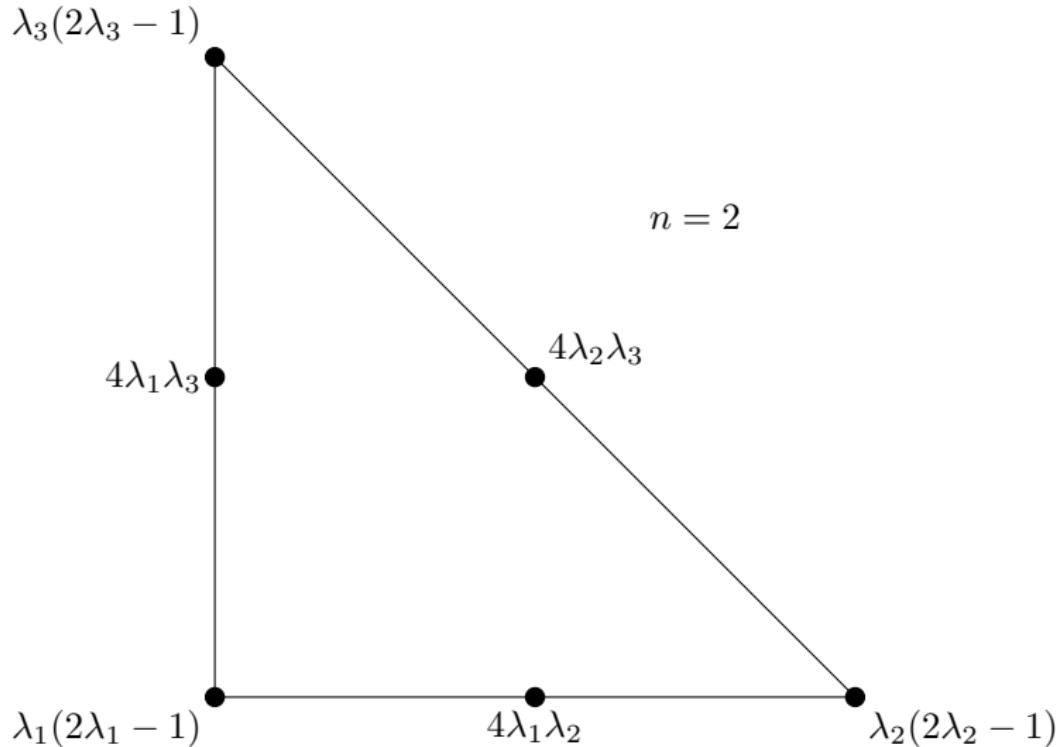
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- $\text{supp}(\phi) = \mathcal{T}$

## Solution Transfer

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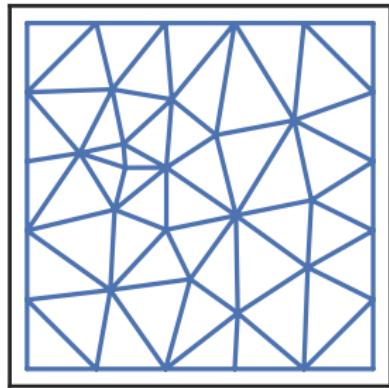
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  - Shape function bases  $\phi_D^{(j)}$  and  $\phi_T^{(j)}$
  - Known discrete field  $\mathbf{q}_D = \sum_j d_j \phi_D^{(j)}$
- Want:  $L_2$ -optimal interpolant  $\mathbf{q}_T = \sum_j t_j \phi_T^{(j)}$ :

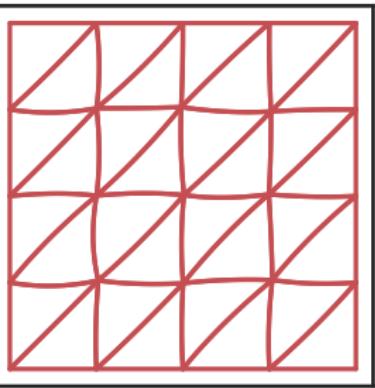
$$\|\mathbf{q}_T - \mathbf{q}_D\|_2 = \min_{\mathbf{q} \in \mathcal{V}_T} \|\mathbf{q} - \mathbf{q}_D\|_2$$

# Galerkin Projection

$\mathcal{M}_T$



$\mathcal{M}_D$



## Galerkin Projection

Differentiating w.r.t. each  $t_j$  in  $\mathbf{q}_T = \sum_j t_j \phi_T^{(j)}$  gives **weak form**

$$\int_{\Omega} \mathbf{q}_D \phi_T^{(j)} dV = \int_{\Omega} \mathbf{q}_T \phi_T^{(j)} dV, \quad \text{for all } j.$$

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If  $(x \mapsto 1) \in \mathcal{V}_T$ , then  $\mathbf{q}_T$  is globally **conservative**

$$\int_{\Omega} \mathbf{q}_D dV = \int_{\Omega} \mathbf{q}_T dV.$$

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Weak form gives rise to a linear system in coefficients  $\mathbf{d}$  and  $t$ .

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$$M_T \mathbf{t} = M_{TD} \mathbf{d}.$$

# Linear System

$M_T$  is (symmetric) mass matrix for target mesh

$$(M_T)_{ij} = \int_{\Omega} \phi_T^{(i)} \phi_T^{(j)} dV.$$

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$$(M_T)_{ij} = \int_{\Omega} \phi_T^{(i)} \phi_T^{(j)} dV.$$

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Integrate via substitution for  $F = \phi_T^{(i)} \phi_T^{(j)}$

$$\int_{b(\mathcal{U})} F(x, y) dx dy = \int_{\mathcal{U}} \det(Db) F(x(s, t), y(s, t)) ds dt$$

and then use quadrature rule on  $\mathcal{U}$ .

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Instead, compute entire RHS

$$(M_{TD}\mathbf{d})_j = \int_{\Omega} \phi_T^{(j)} \mathbf{q}_D dV.$$

## Common Refinement

Given  $\phi$  supported on  $\mathcal{T}$

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## Common Refinement

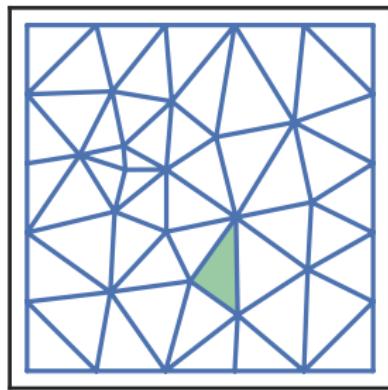
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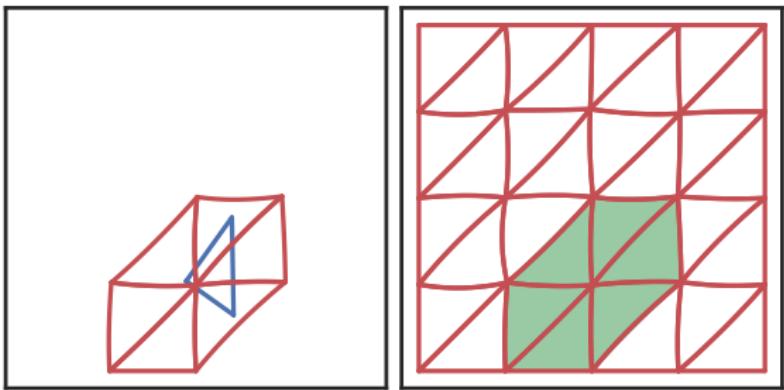
In CG,  $\mathbf{q}_D$  need not be differentiable across elements and in DG  $\mathbf{q}_D$  need not even be continuous

# Common Refinement

$\mathcal{M}_T$



$\mathcal{M}_D$



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- $\mathcal{T}_1 = \textcolor{green}{b}_1(\mathcal{U})$ ,  $\partial\mathcal{T}_1 = \textcolor{green}{E}_3 \cup \textcolor{green}{E}_4 \cup \textcolor{green}{E}_5$

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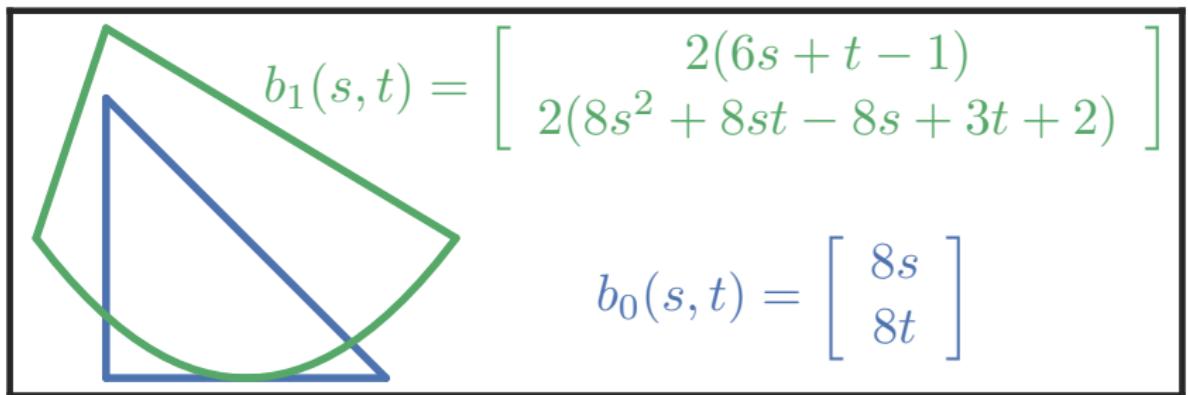
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- $\mathcal{P} = \mathcal{T}_0 \cap \mathcal{T}_1$ ,  $\partial\mathcal{P}$  defined by segments of edges from  $\mathcal{T}_0$  and  $\mathcal{T}_1$

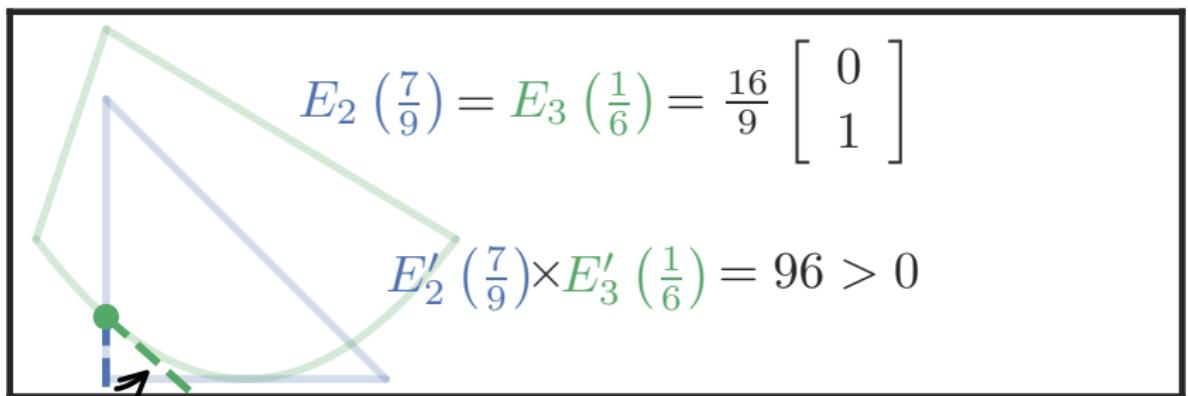
## Intersecting Curved Elements



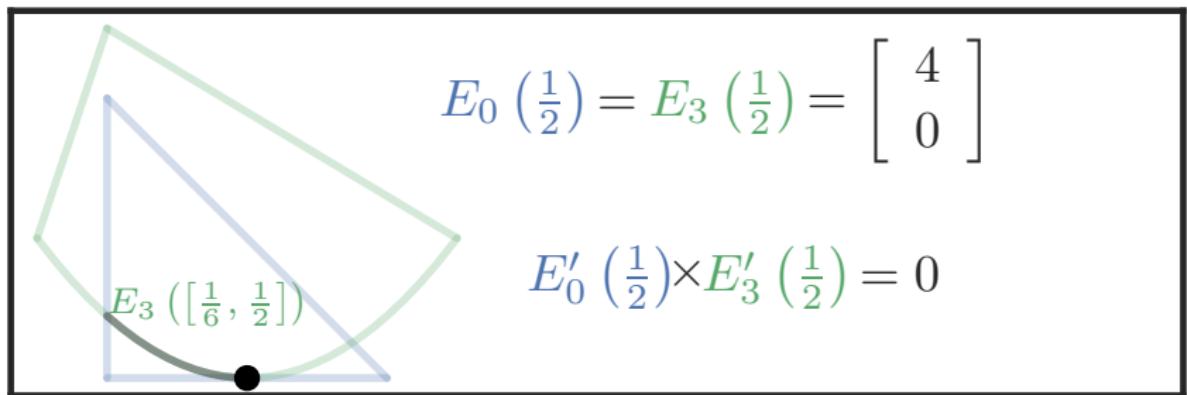
# Intersecting Curved Elements

$$E_2 \left( \frac{7}{9} \right) = E_3 \left( \frac{1}{6} \right) = \frac{16}{9} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

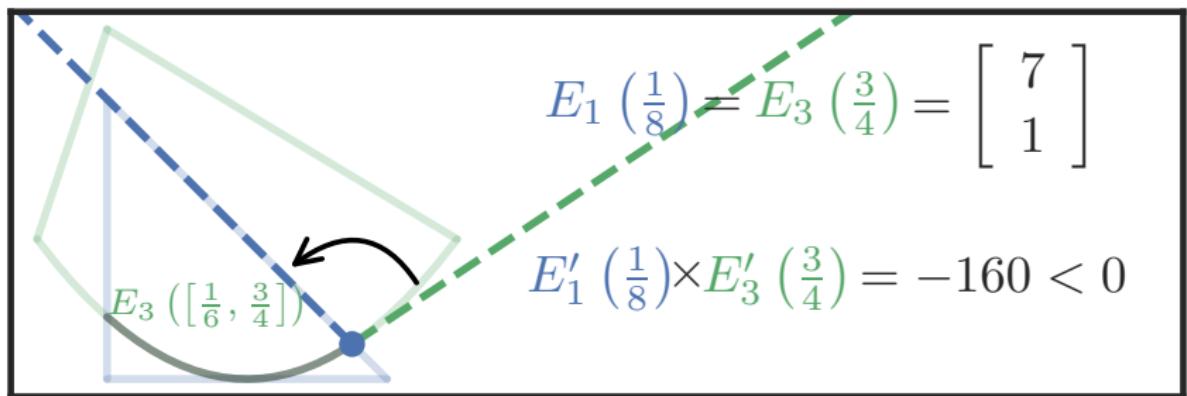
$$E'_2 \left( \frac{7}{9} \right) \times E'_3 \left( \frac{1}{6} \right) = 96 > 0$$



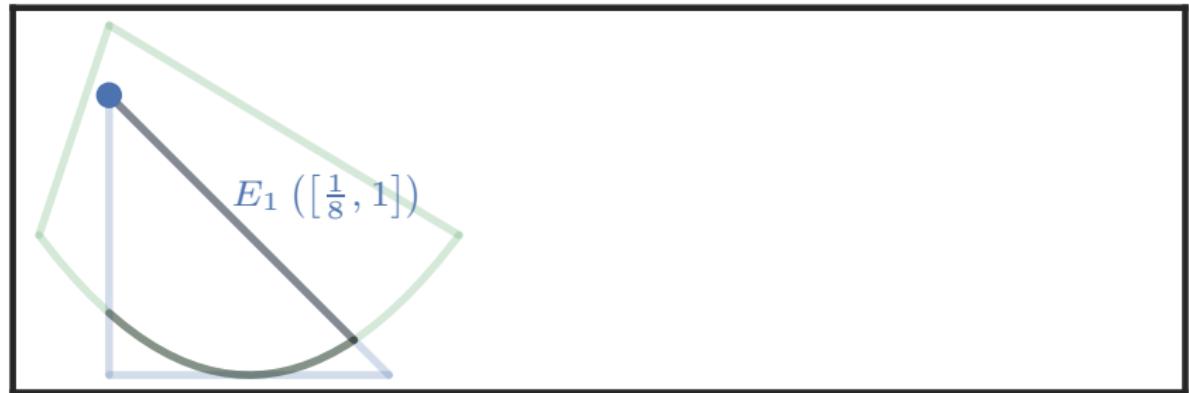
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- $\mathcal{P} = \mathcal{T}_0 \cap \mathcal{T}_1$
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## Advancing Front

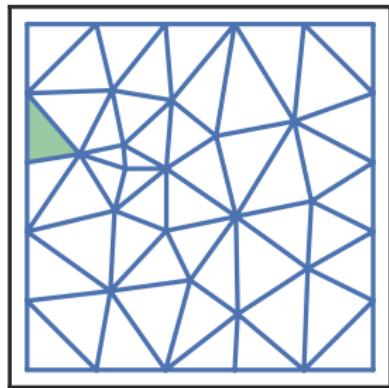
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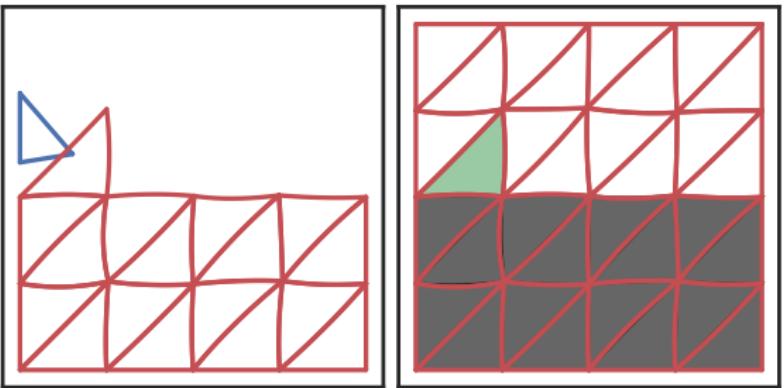
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# Advancing Front

$\mathcal{M}_T$



$\mathcal{M}_D$



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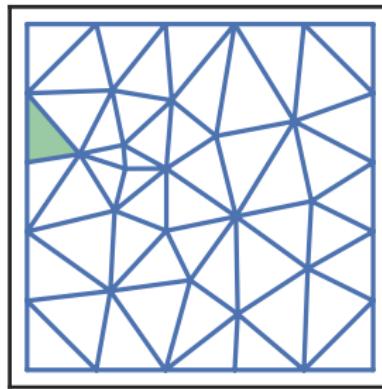
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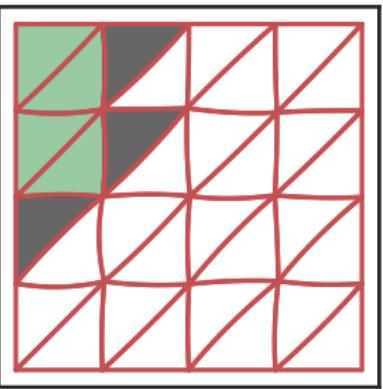
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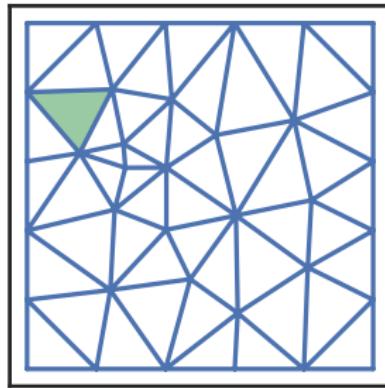
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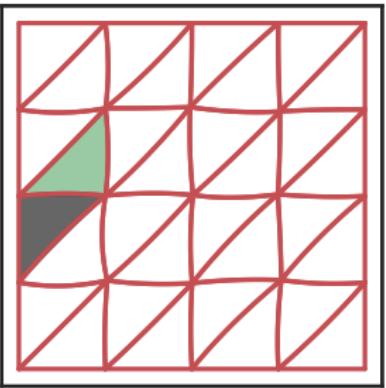
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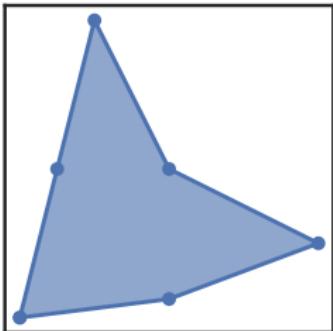
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- Total  $\mathcal{O}(|\mathcal{M}_D| + |\mathcal{M}_T|)$

# Integration over Curved Polygons

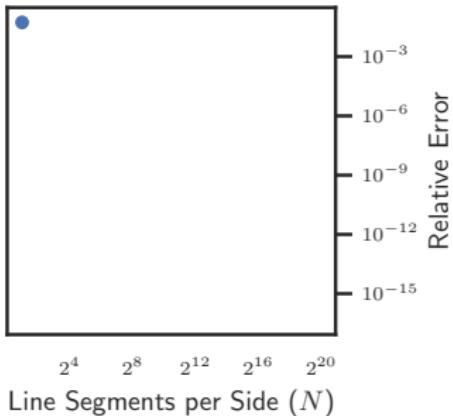
$$\int_{\mathcal{P}} F(x, y) \, dV, \quad \mathcal{P} = \mathcal{T}_0 \cap \mathcal{T}_1, \quad F = \phi_0 \phi_1$$

# Integrate via Polygonal Approximation

$$N = 2$$

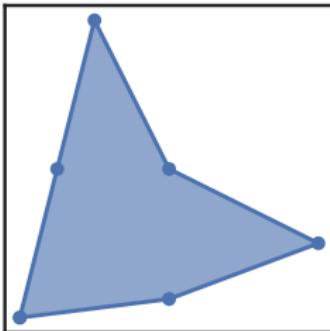


Area Estimates

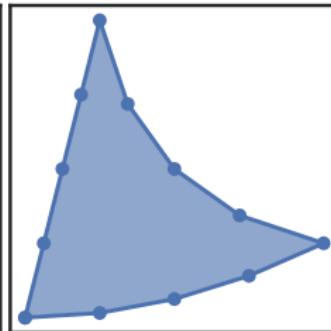


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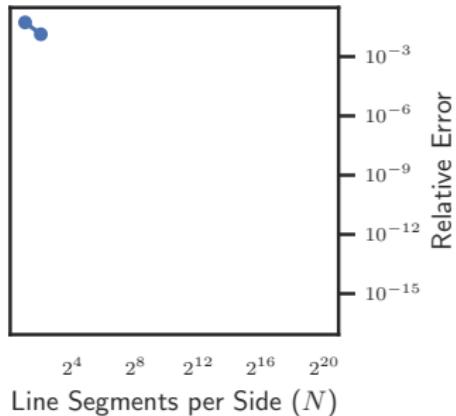
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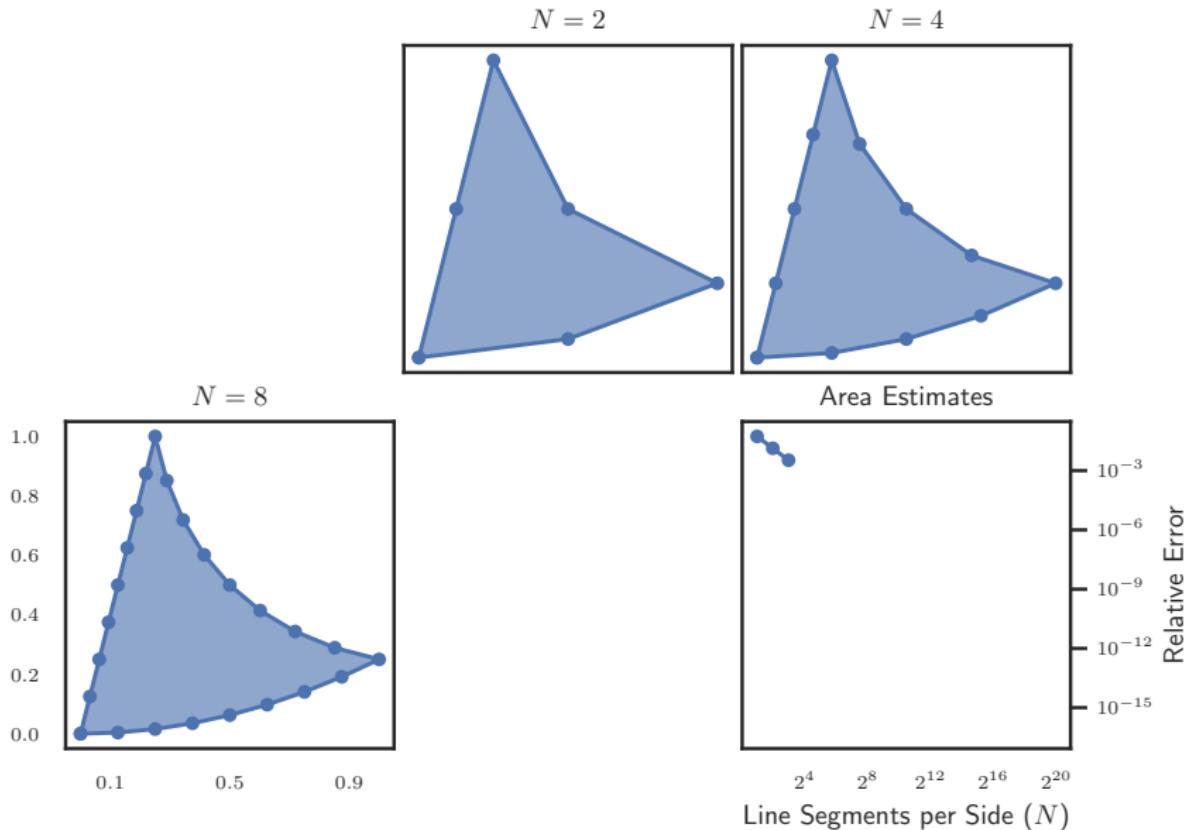
$N = 4$



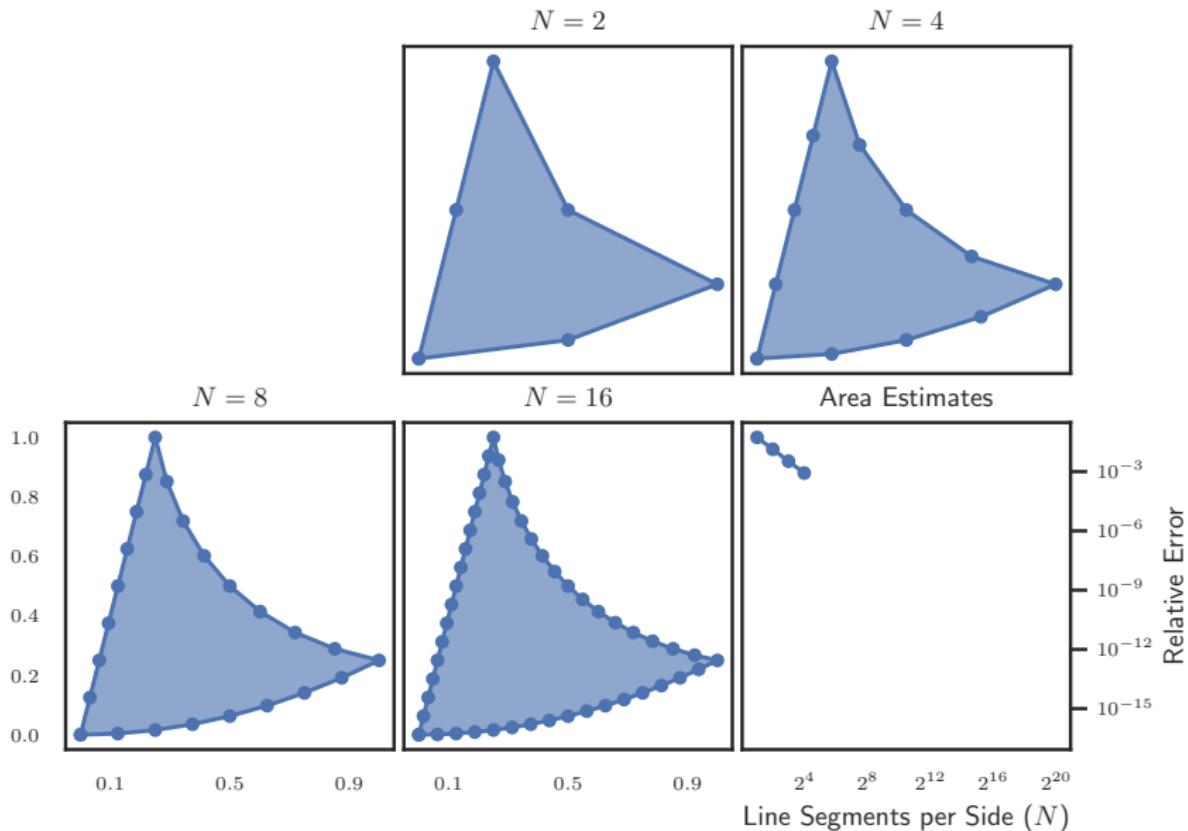
Area Estimates



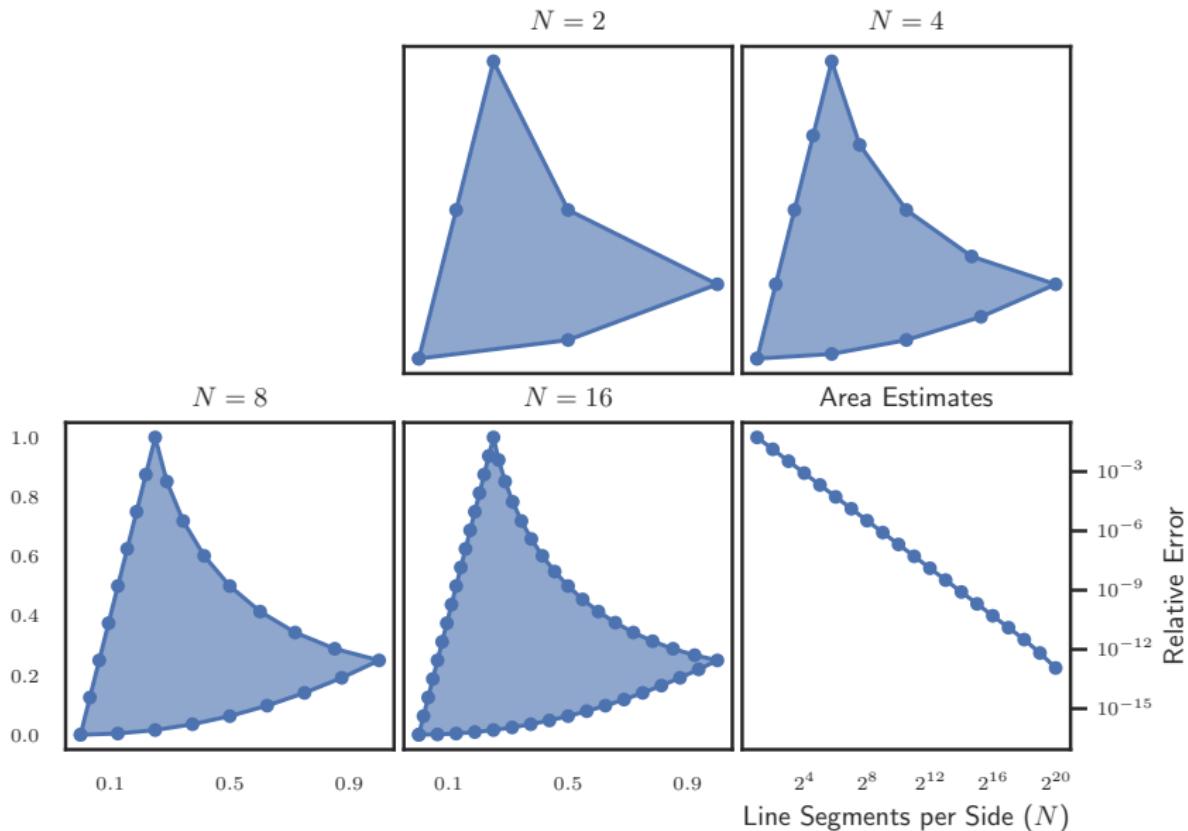
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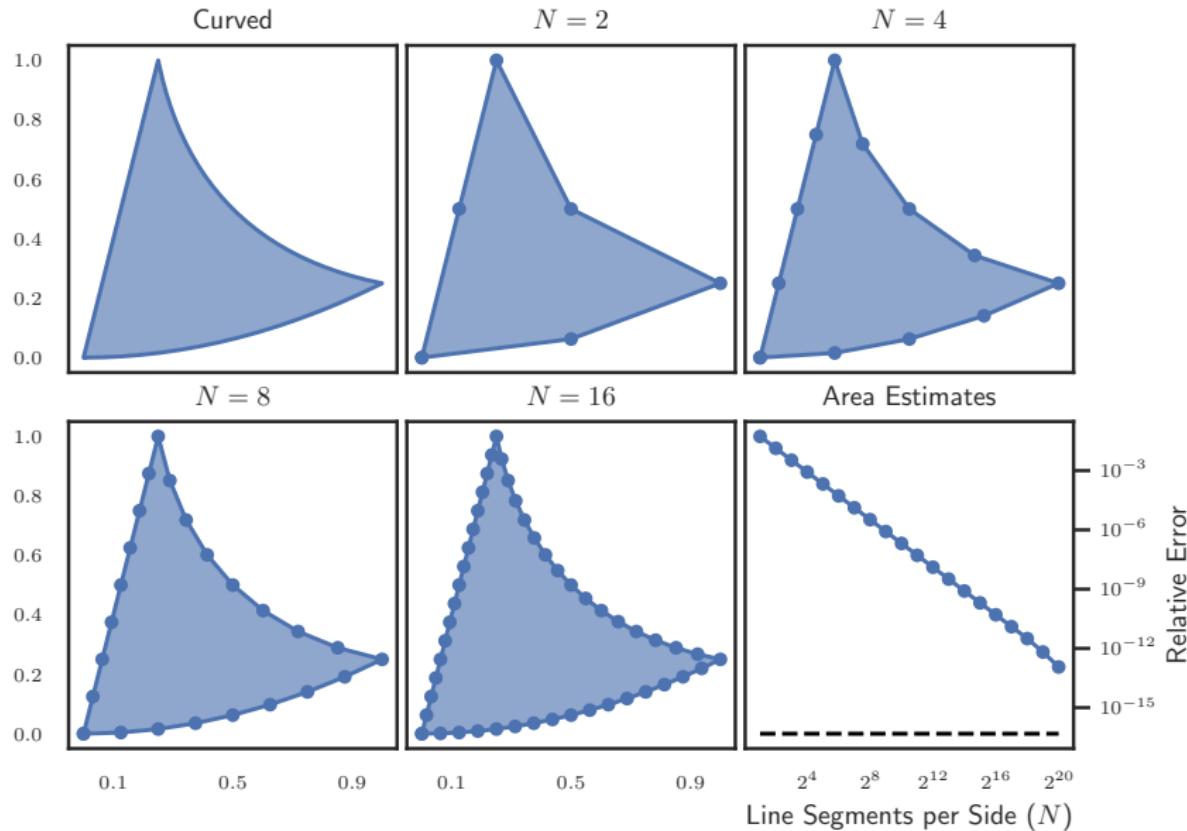
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- Transfinite interpolation or mean value coordinates
  - Maps from (straight sided) reference domain, but increase degree or are not bijective

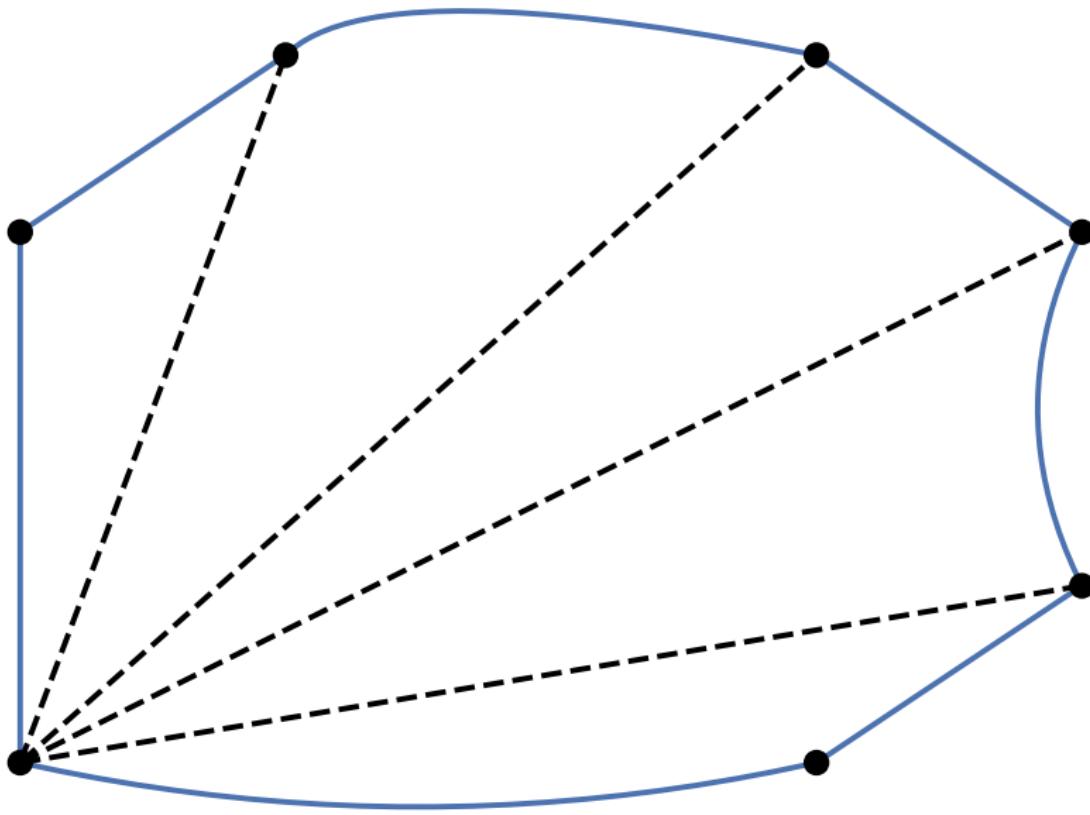
## Integrate via Tessellation

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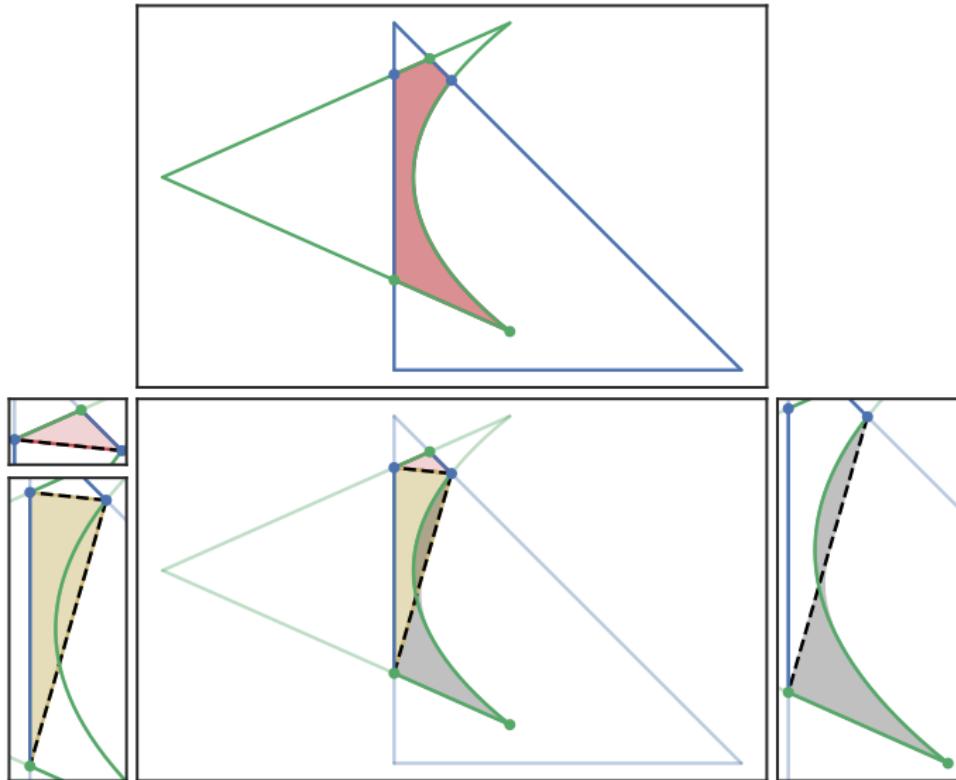
## Integrate via Tessellation

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- If diagonals are valid, high degree Bézier triangles need interior control points introduced that don't cause triangle to invert

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- Each  $C$  given by  $x(r), y(r)$ , 1D quadrature on unit interval of

$$G(r) = H(x(r), y(r))y'(r) - V(x(r), y(r))x'(r)$$

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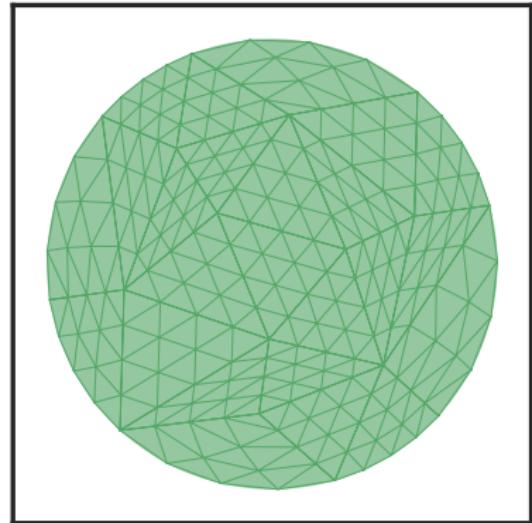
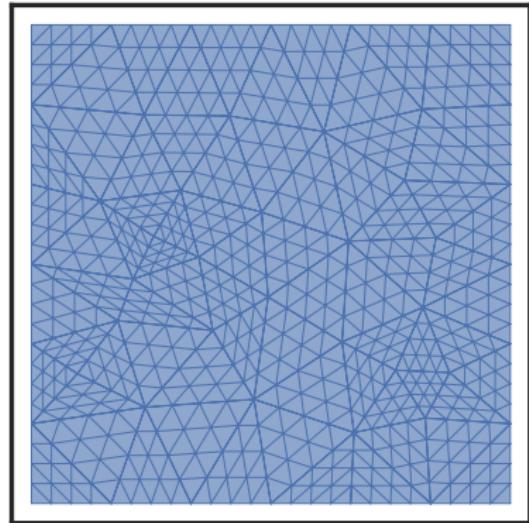
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## Numerical Experiments

- Three meshes ( $p = 1, 2, 3$ )

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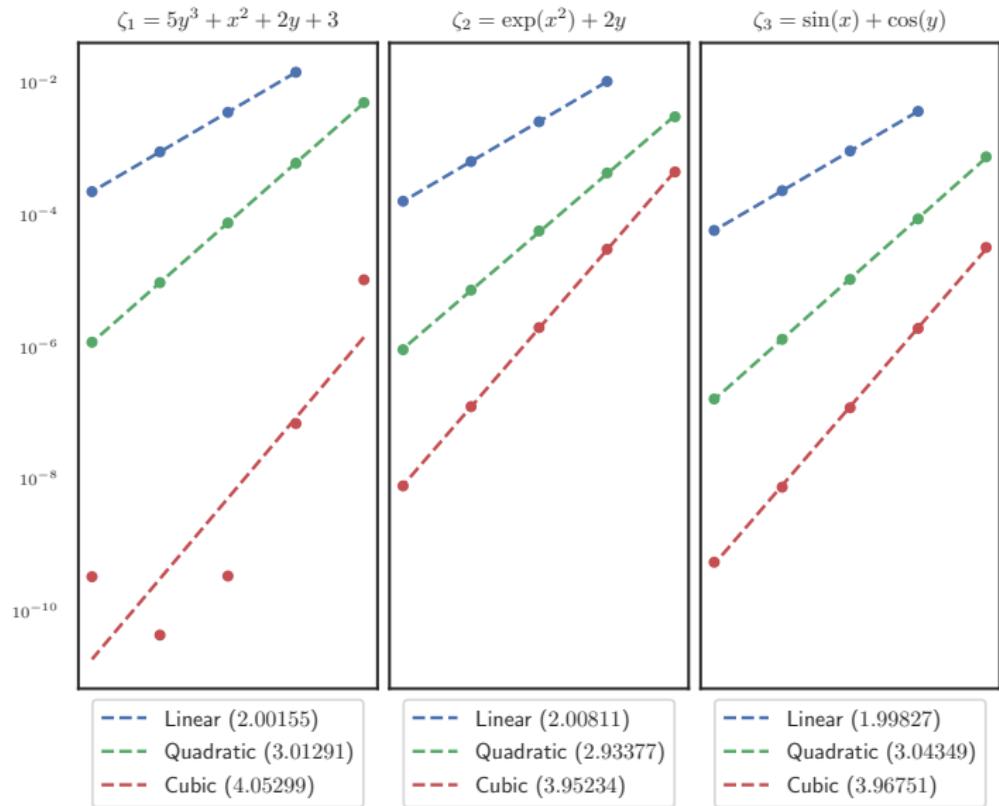
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  - $\zeta_2(x, y) = \exp(x^2) + 2y$
  - $\zeta_3(x, y) = \sin(x) + \cos(y)$
- Nodal interpolant  $\mathbf{q}_D = \sum_j \zeta(\mathbf{n}_j) \phi_D^{(j)}$

# Numerical Experiments

- Three meshes ( $p = 1, 2, 3$ )
- Three functions
  - $\zeta_1(x, y) = 5y^3 + x^2 + 2y + 3$
  - $\zeta_2(x, y) = \exp(x^2) + 2y$
  - $\zeta_3(x, y) = \sin(x) + \cos(y)$
- Nodal interpolant  $\mathbf{q}_D = \sum_j \zeta(\mathbf{n}_j) \phi_D^{(j)}$
- Expect  $\mathcal{O}(h^{p+1})$  errors

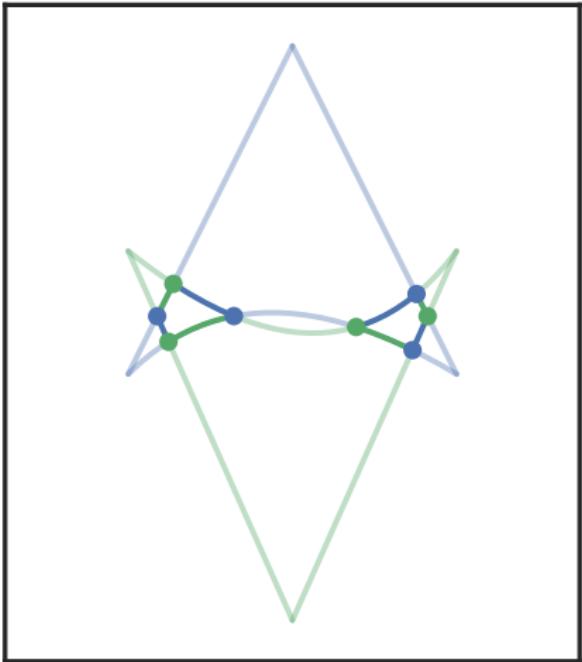
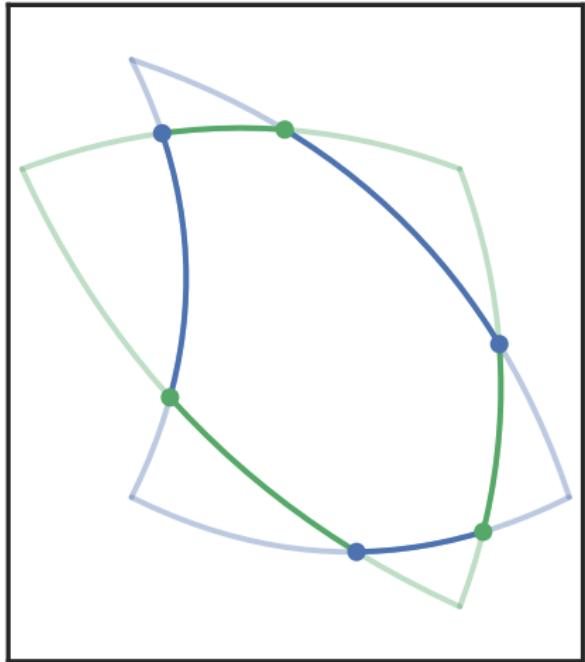
# Numerical Experiments



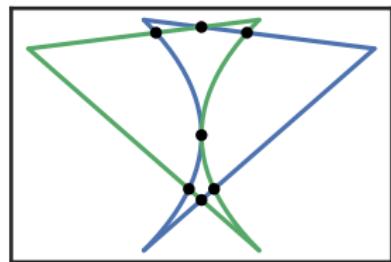
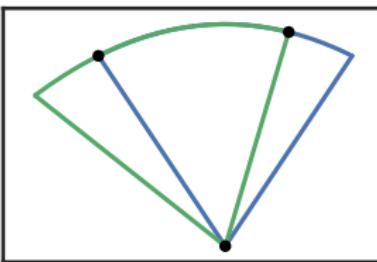
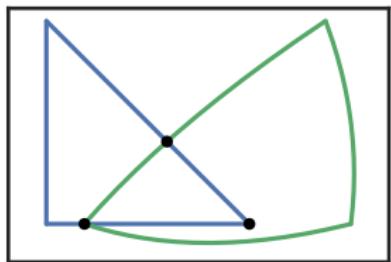
## Ill-conditioned Bézier Curve Intersection

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## III-Conditioned Intersections



# Ill-Conditioned Intersections



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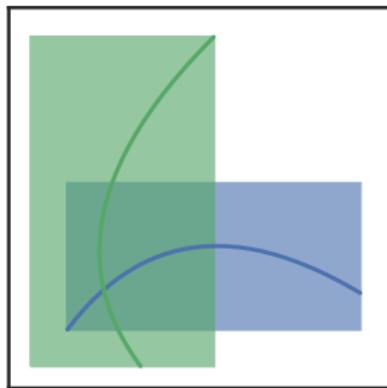
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- Random pair of meshes, “almost tangent” intersections increasingly frequent as  $h \rightarrow 0^+$

# Intersection Algorithm

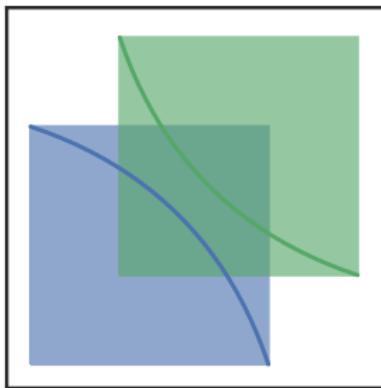
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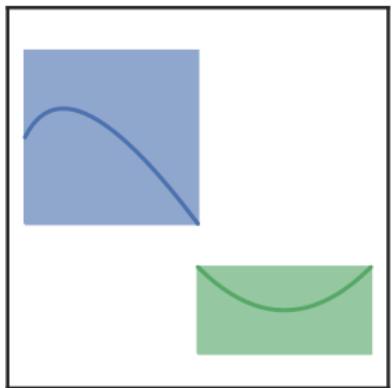
MAYBE



MAYBE



NO



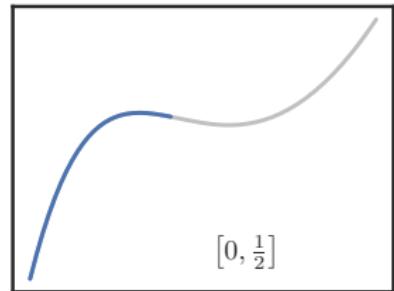
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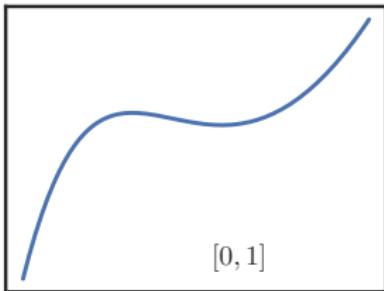
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- Bounding box check
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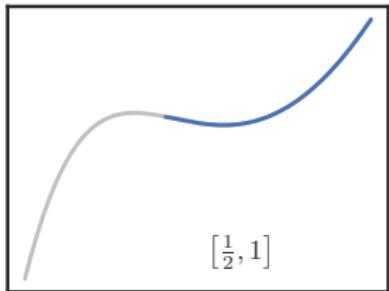
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$$\left[0, \frac{1}{2}\right]$$



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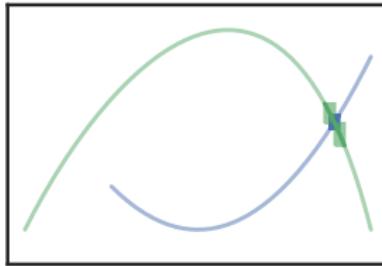
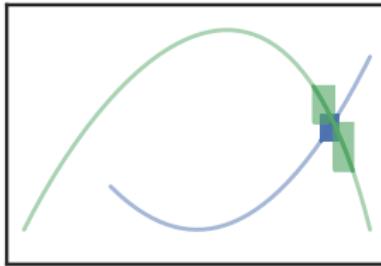
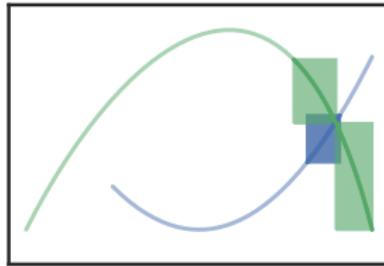
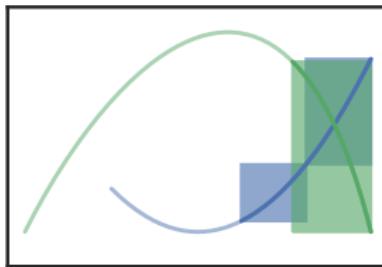
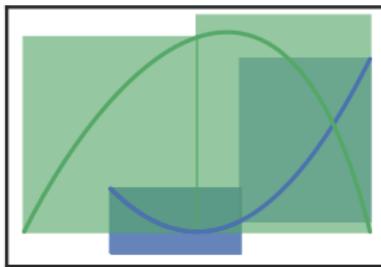
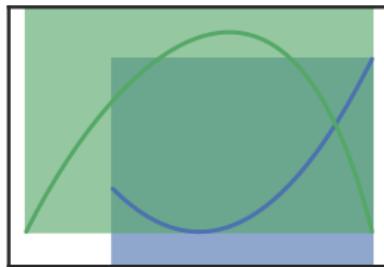
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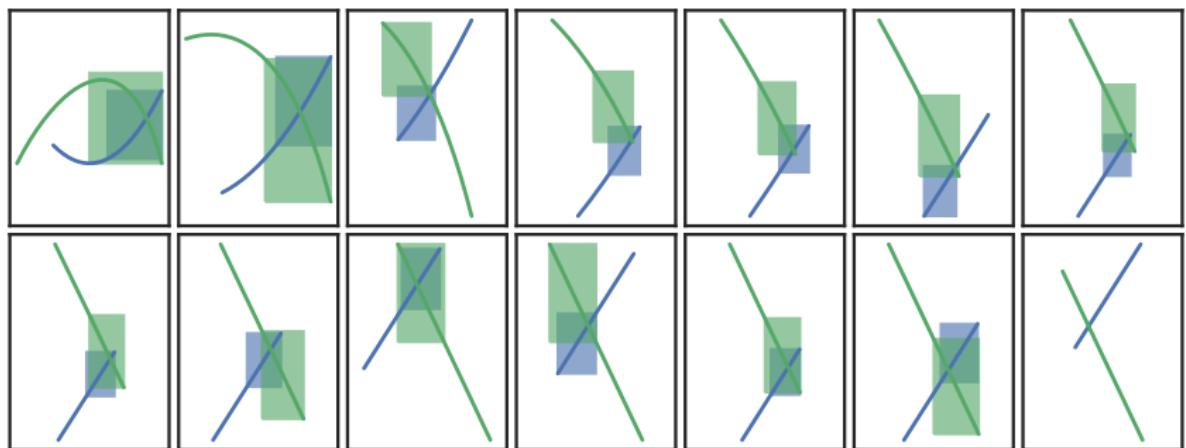
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- At ill-conditioned intersections,  $J$  is almost singular and evaluation of  $F$  is typically ill-conditioned as well

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  - Tisseur (2001) showed this generically and applied to iterative refinement for generalized eigenvalue problem

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- **Compensated** algorithm uses  $\hat{f} \oplus \hat{e}$ ; better approximation of  $f(x)$  than  $\hat{f}$  (usually)

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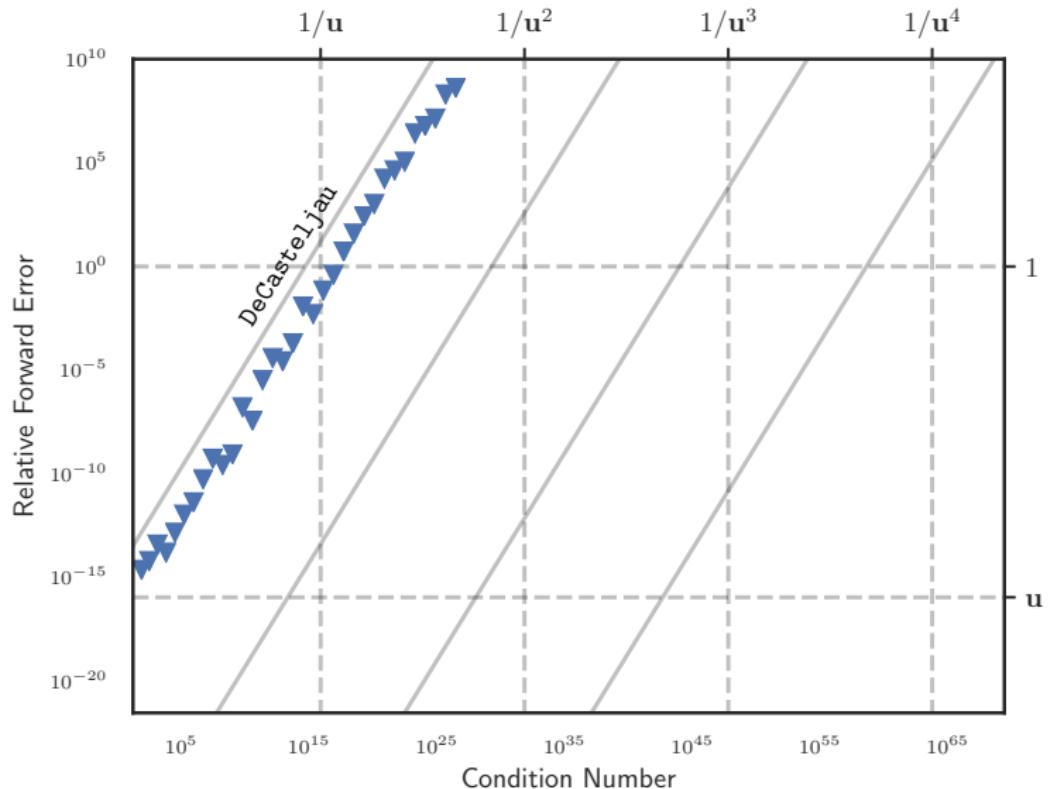
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- Track errors and combine them progressively
- Test polynomial  $p(s) = (s - 1) \left(s - \frac{3}{4}\right)^7$ ; evaluation at  $s = \frac{3}{4}$  has infinite condition, very ill-conditioned nearby

## Compensated de Castlejau

$$\frac{|p(s) - \text{DeCasteljau}(p, s)|}{|p(s)|} \leq \text{cond}(p, s) \cdot \mathcal{O}(\mathbf{u})$$

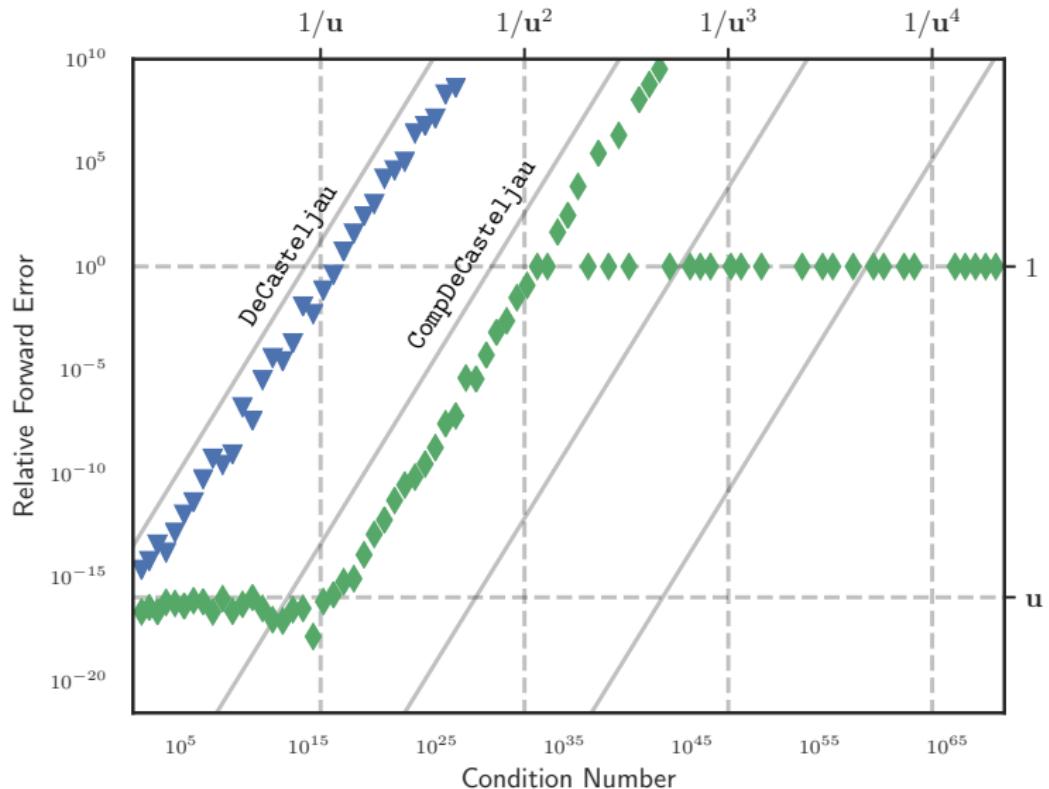
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$$\frac{|p(s) - \text{CompDeCasteljau}(p, s)|}{|p(s)|} \leq \mathcal{O}(\mathbf{u}) + \text{cond}(p, s) \cdot \mathcal{O}(\mathbf{u}^2)$$

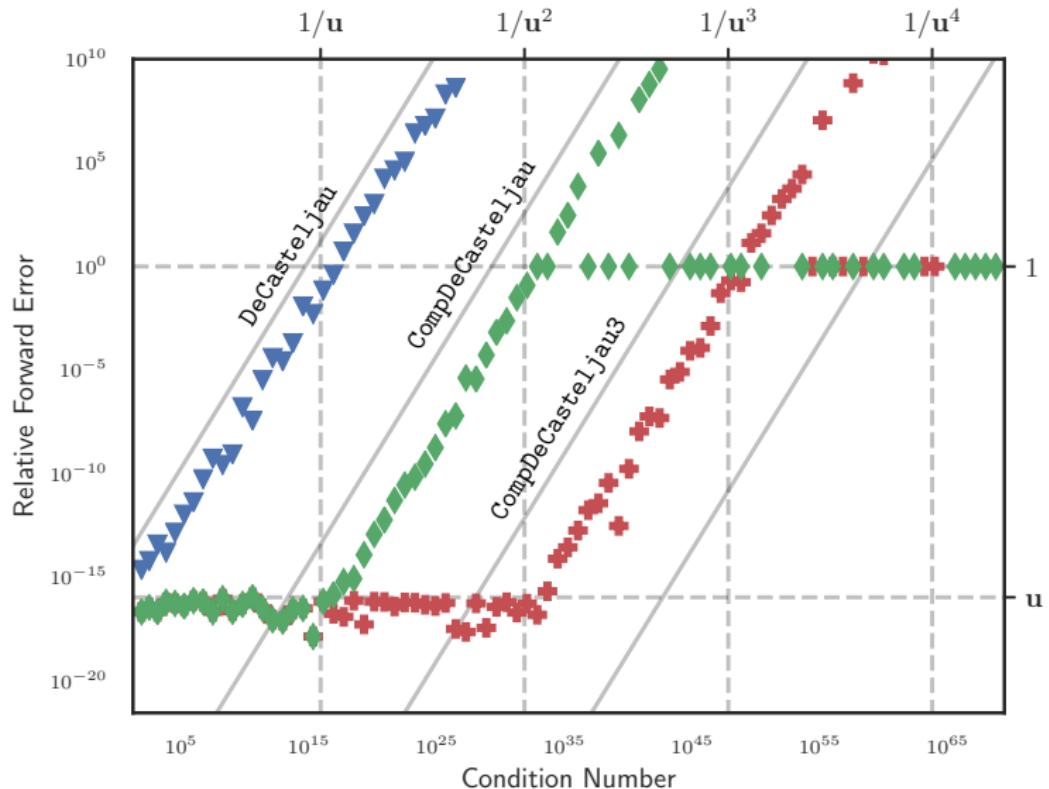
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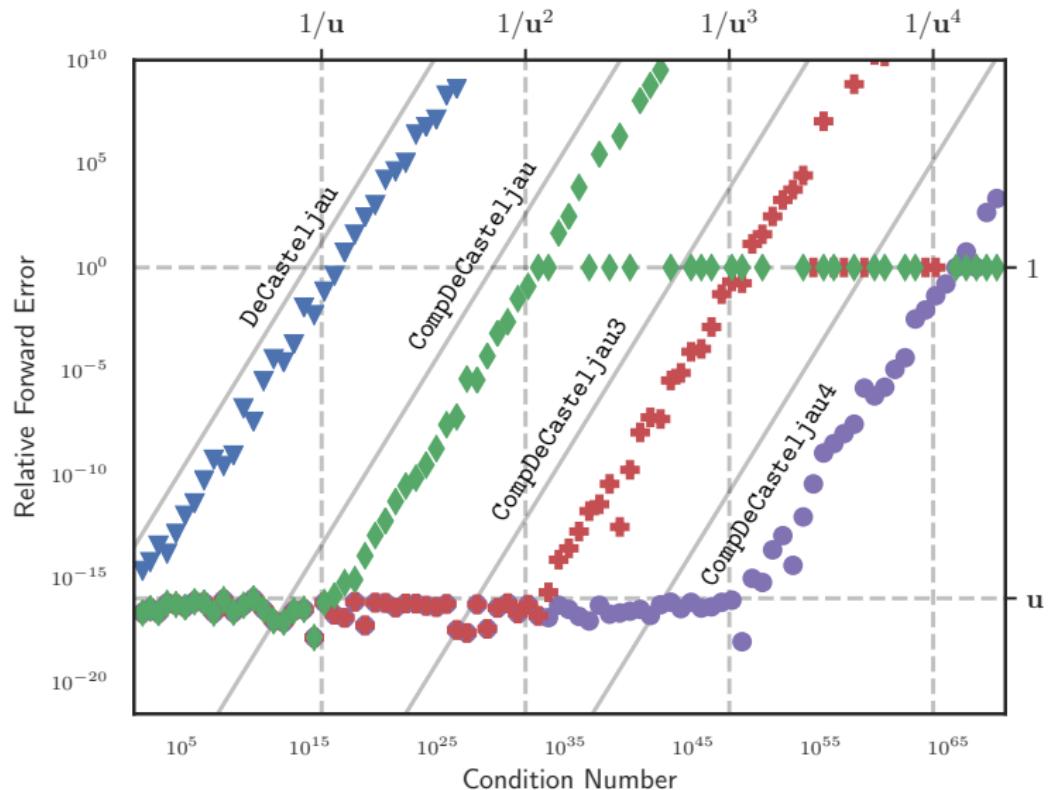
## Compensated de Castlejau

$$\frac{|p(s) - \text{CompDeCasteljauK}(p, s)|}{|p(s)|} \leq \mathcal{O}(\mathbf{u}) + \text{cond}(p, s) \cdot \mathcal{O}(\mathbf{u}^K)$$

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## Modified Newton's for Intersection

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  - **DNewtonFull**: CompDeCasteljau for residual and Jacobian

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- Test problem; need coefficients that can be exactly represented

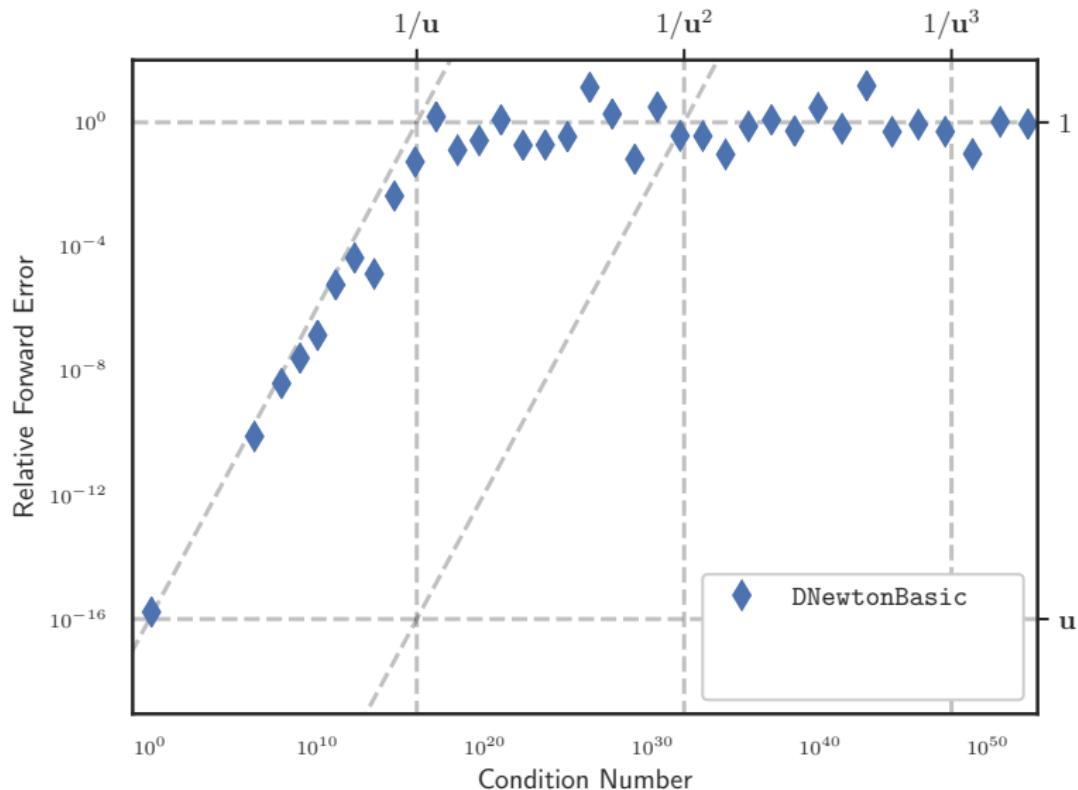
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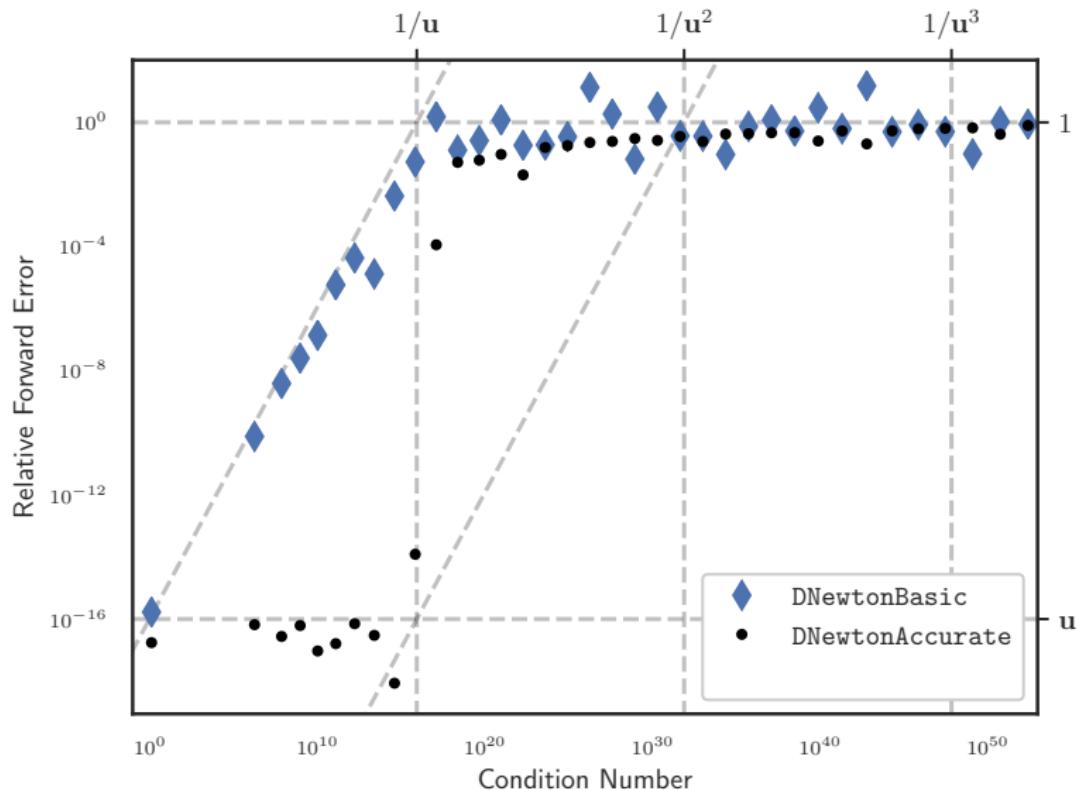
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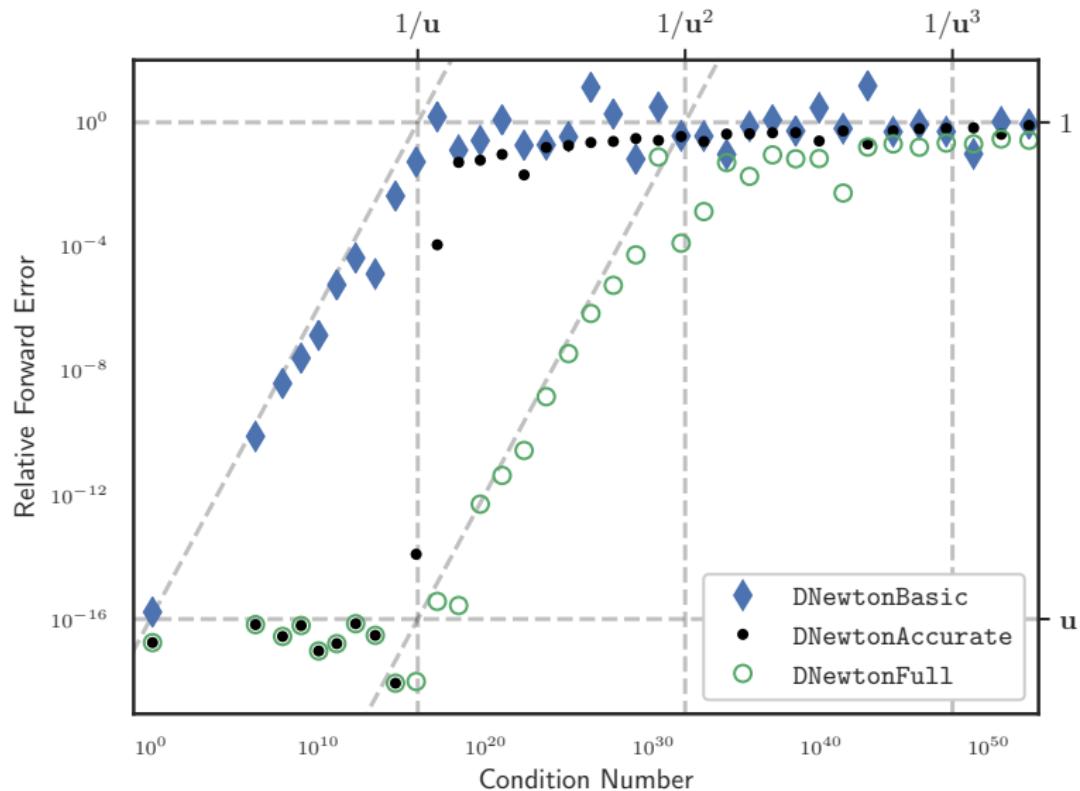
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- Modify Newton's method by using “extended precision” (i.e. compensated method) for  $\hat{F}$

## Compensated Residual

$$\cdot \hat{F} = \begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix}$$

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- Compensation term  $\tau = (\widehat{e}_0 \ominus \widehat{e}_1) \oplus \widehat{e}_2$

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- Compensated addition (via TwoSum)  $\widehat{x}_0 - \widehat{x}_1 = D + \widehat{e}_2$
- Compensation term  $\tau = (\widehat{e}_0 \ominus \widehat{e}_1) \oplus \widehat{e}_2$
- Computed residual  $D \oplus \tau$

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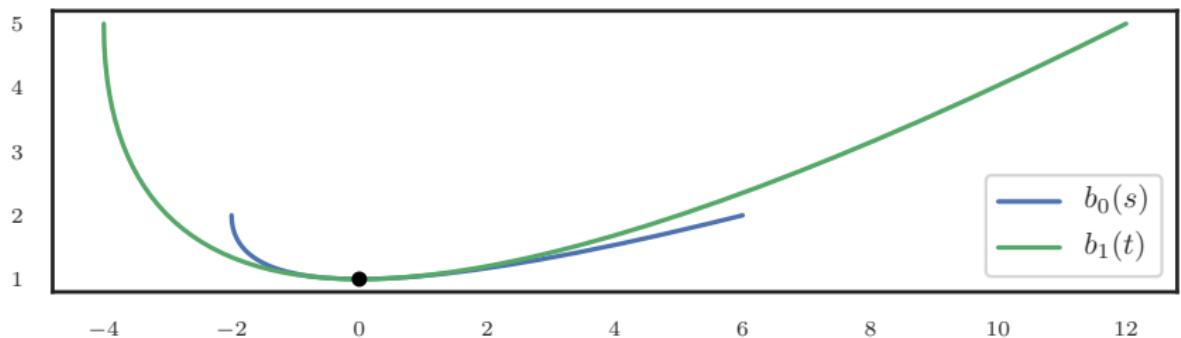
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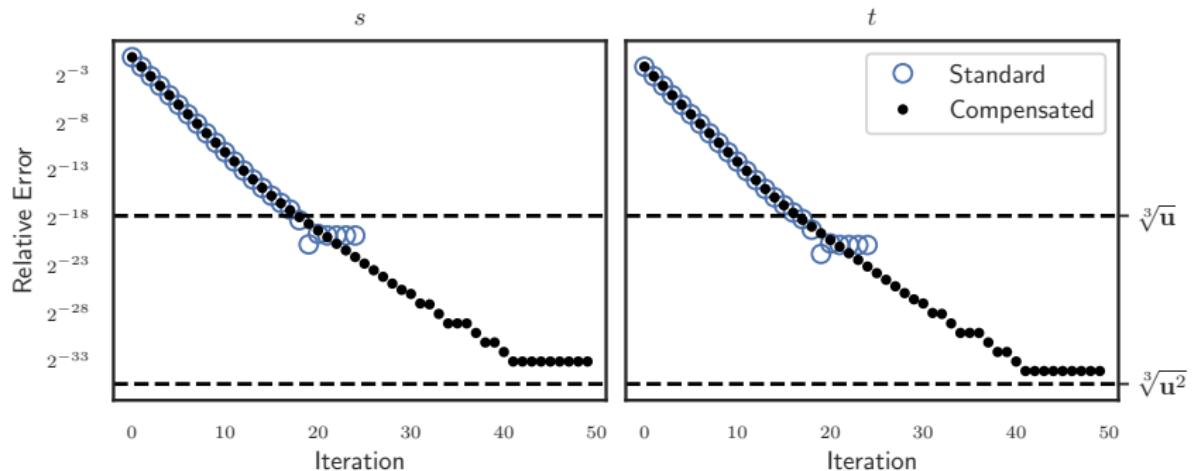
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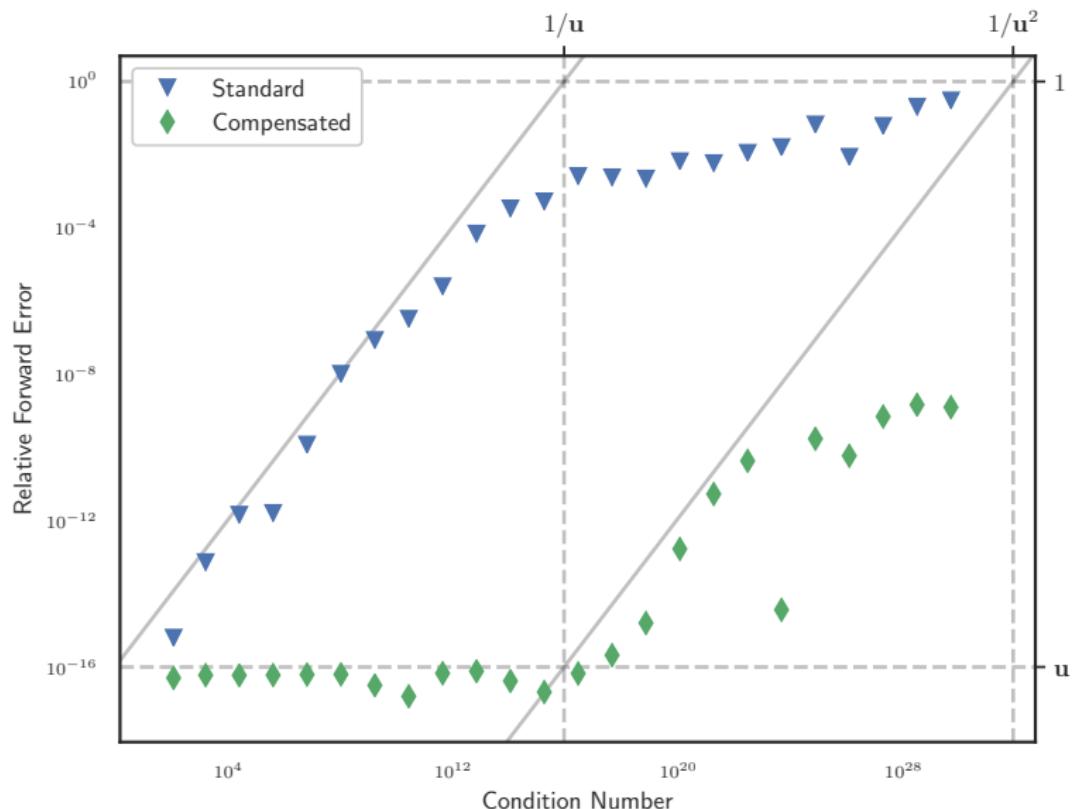
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Approaches triple root as  $r \rightarrow 0^+$ ,  $F\left(\frac{1+\sqrt{r}}{2}, \frac{2+\sqrt{r}}{4}\right)$

# Modified Newton's in Practice



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  - Modified Newton's method; computes residual  $F(s, t)$  as if in extended precision

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- Solution transfer in  $\mathbf{R}^3$ ; integrate via Stoke's theorem, geometry is greatest challenge