

Computing Bounds on the Geometrical Quality of Curvilinear Finite Elements

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Gmsh Workshop 2013

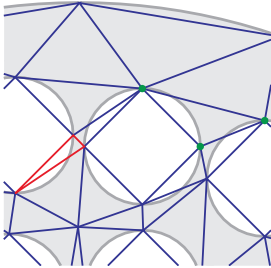
- ▶ Introduction
- ▶ Bounds for Geometrical Validity
- ▶ Bounds on Geometrical Quality

- ▶ Introduction

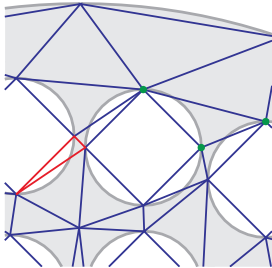
- ▶ Bounds for Geometrical Validity

- ▶ Bounds on Geometrical Quality

Way of Building a Curvilinear Mesh



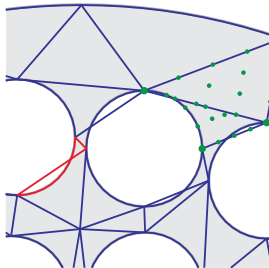
Way of Building a Curvilinear Mesh



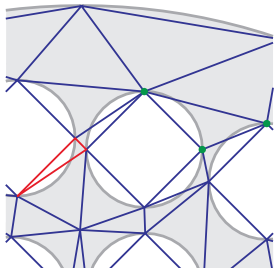
→ poor quality elements

→ invalid elements

Mesh entities
classified on the
boundaries are curved

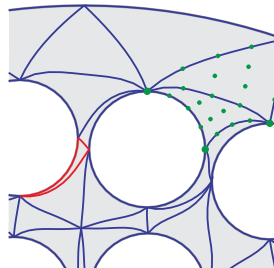


Way of Building a Curvilinear Mesh



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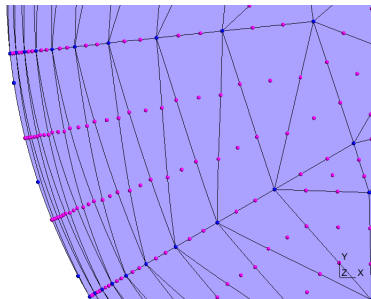
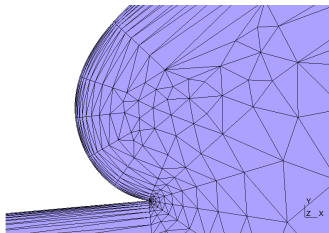
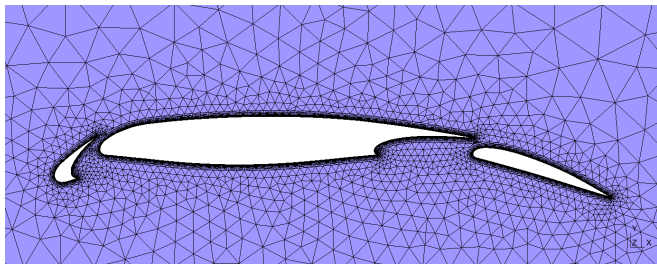


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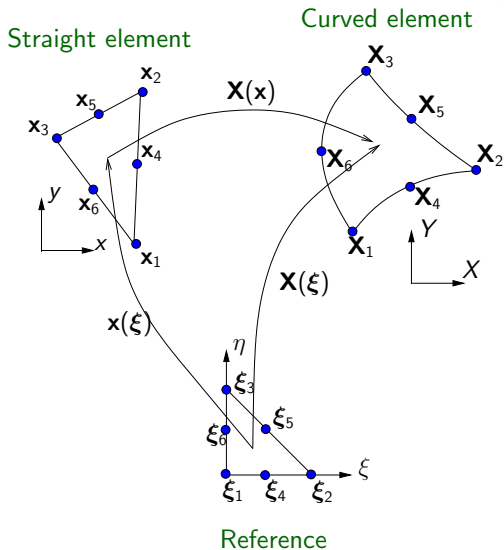
Optimization

**Computationally
expensive !**

Example of Optimization: Wing (3-order Triangles)



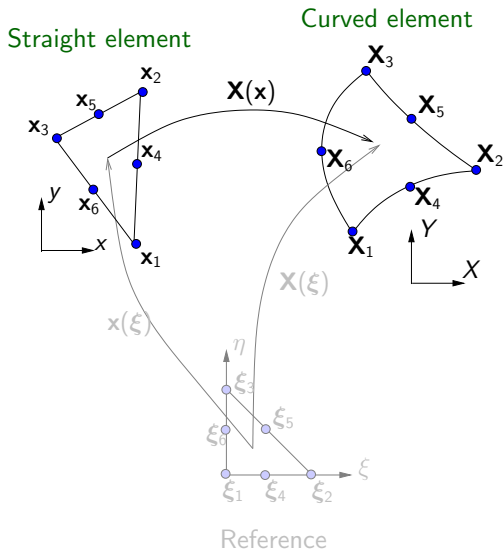
Quality Deterioration Control



Quality Deterioration Control

Straight element
→ good quality

Mapping $\mathbf{X}(\mathbf{x})$
⇒ distortion that should be
controlled

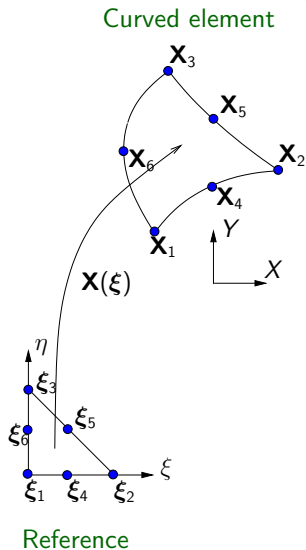


- ▶ Introduction
- ▶ **Bounds for Geometrical Validity**
- ▶ Bounds on Geometrical Quality

Element Validity

An element is valid. . .

if its mapping $\mathbf{X}(\xi)$ is bijective



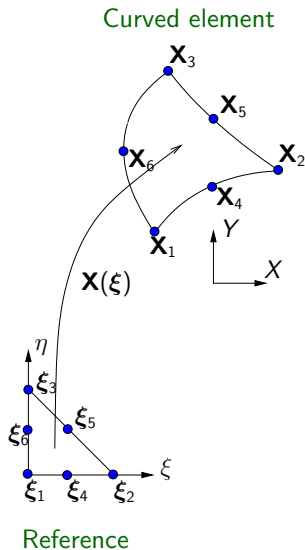
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i.e. if $J(\xi) := \det \mathbf{X}_{,\xi} > 0$

Jacobian determinant $J(\xi)$



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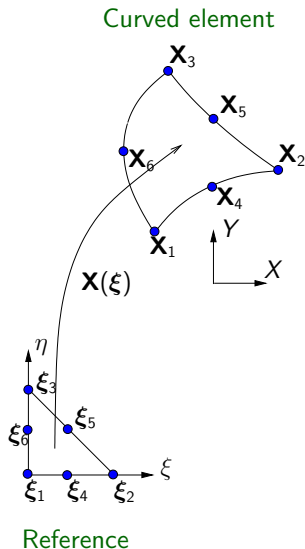
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Necessary and sufficient condition :

$$J_{\min} = \min_{\xi} J(\xi) > 0$$



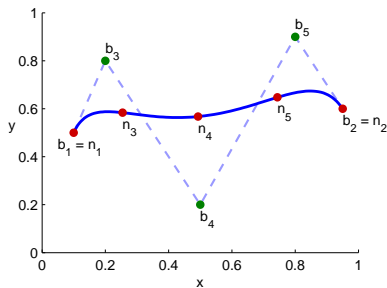
Use of Bézier for Computing Bounds on J_{\min} - Arbitrary Element Case

$$J(\xi) = \sum_{i=1}^N b_i B_i(\xi)$$

Bézier functions $B_i(\xi)$

Bézier coefficients b_i

Property: $J(\xi) \in \text{conv}(b_1, b_2, \dots, b_N)$

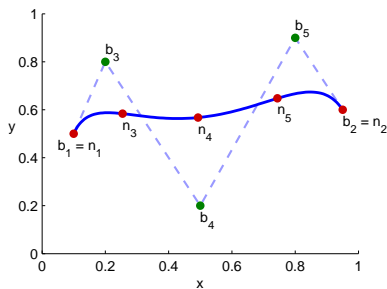


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Lower bound :

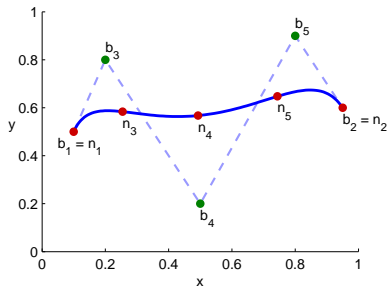
$$b_{\min} = \min(b_1, b_2, \dots, b_N) \leq J_{\min}$$

Use of Bézier for Computing Bounds on J_{\min} - Arbitrary Element Case

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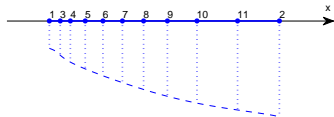
Property: $J(\xi) \in \text{conv}(b_1, b_2, \dots, b_N)$

Also: at the corners, $b_k = J(\xi_k)$ ($k = 1, \dots, N_{\text{corner}}$)

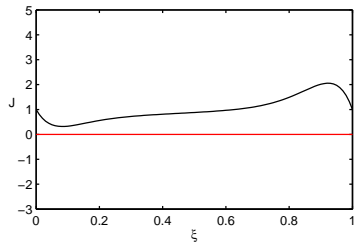
Lower and upper bound :

$$b_{\min} = \min(b_1, b_2, \dots, b_N) \leq J_{\min} \leq \min(b_1, b_2, \dots, b_{N_{\text{corner}}}) = c_{\min}$$

Accurate Bounds

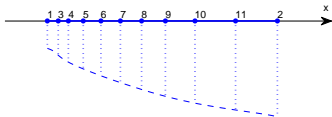


Element Geometry

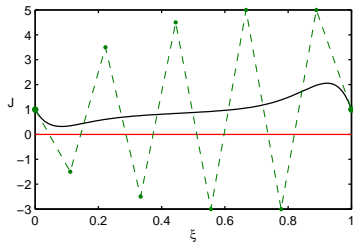


Jacobian $J(\xi)$

Accurate Bounds



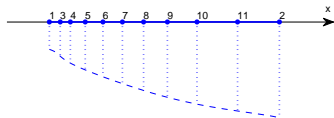
Element Geometry



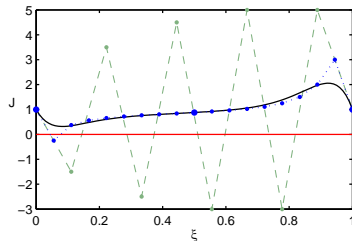
Jacobian $J(\xi)$

$$J_{\min} \in \quad 1. \quad [-3, 1]$$

Accurate Bounds

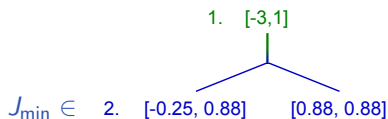


Element Geometry

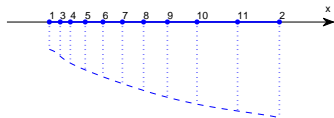


Jacobian $J(\xi)$

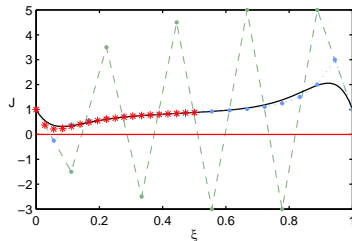
Subdivision = Restriction to Q subdomains



Accurate Bounds



Element Geometry

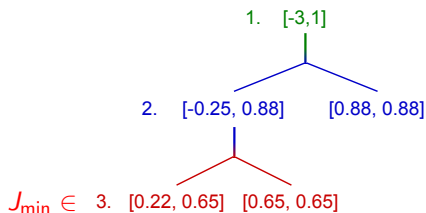


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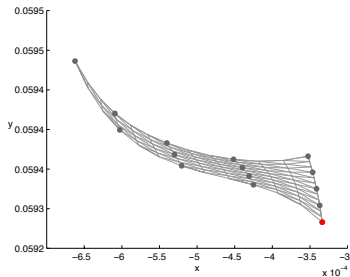
Each subdivision provides better bounds :

$$\begin{cases} b_{\min}^{\text{sub}} \geq b_{\min} \\ c_{\min}^{\text{sub}} \leq c_{\min} \end{cases}$$

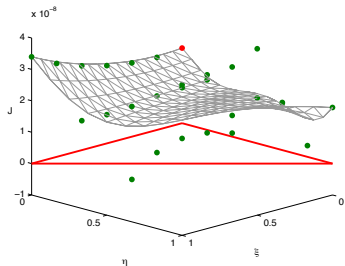


Application : 4-order Triangle

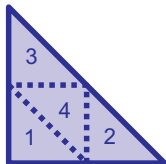
Element Geometry



Jacobian

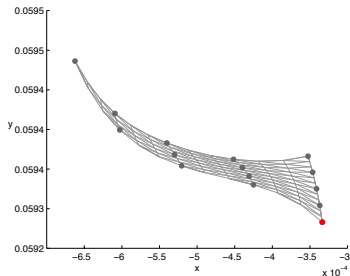


Subdomains :

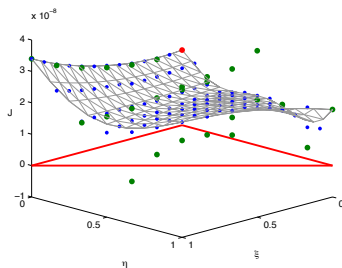


Application : 4-order Triangle

Element Geometry



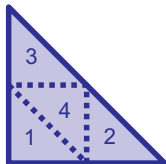
Jacobian



Subdivision

- ▶ adaptative scheme
- ▶ quadratic convergence

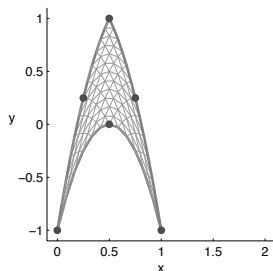
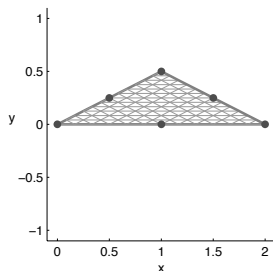
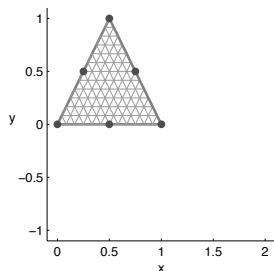
Subdomains :



- ▶ Introduction
- ▶ Bounds for Geometrical Validity
- ▶ **Bounds on Geometrical Quality**

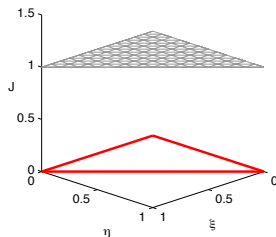
Jacobian Determinant is Not Enough for Optimization

Jacobian Determinant captures only **area changes** !



The three triangles have

- ▶ same total area
- ▶ same constant Jacobian

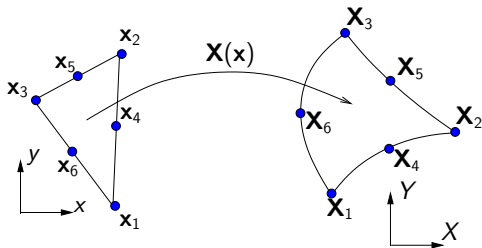


Measure of Variations of Lengths

Metric tensor of the mapping:

$$\mathbf{M} := \mathbf{J}^T \mathbf{J}$$

Jacobian matrix \mathbf{J}



Ratio between deformed and undeformed lengths:

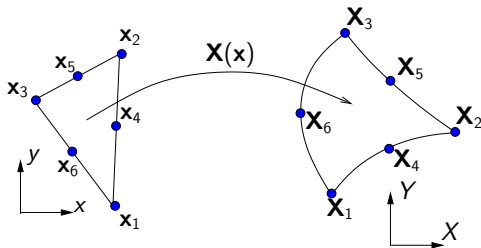
$$r = \frac{\|\mathbf{J}\mathbf{v}\|}{\|\mathbf{v}\|} = \sqrt{\frac{\mathbf{v}^T \mathbf{M} \mathbf{v}}{\mathbf{v}^T \mathbf{v}}}$$

Measure of Variations of Lengths

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\Rightarrow

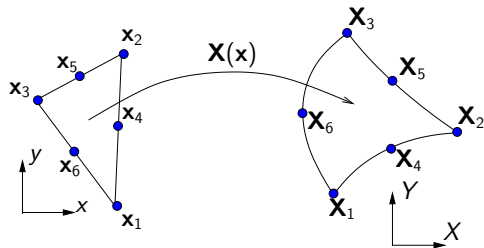
$$r_{\min} = \sqrt{\lambda_{\min}(\mathbf{M})}$$

eigenvalue λ

and resp. for the **max**

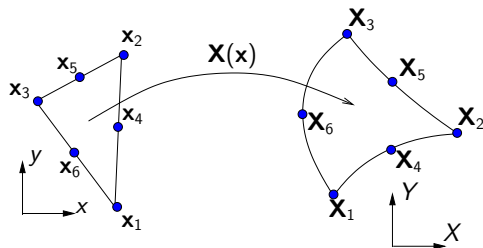
Eigenvalues of Metric Tensor - in 2D

$$\mathbf{M} = \begin{pmatrix} \|\mathbf{X}_{,x}\|^2 & \mathbf{X}_{,x} \cdot \mathbf{X}_{,y} \\ \mathbf{X}_{,x} \cdot \mathbf{X}_{,y} & \|\mathbf{X}_{,y}\|^2 \end{pmatrix}$$



Eigenvalues of Metric Tensor - in 2D

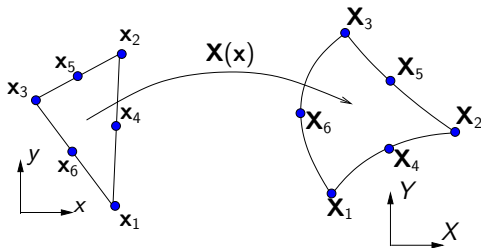
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$$\lambda(\mathbf{M}) = \frac{1}{2} \left[\|\mathbf{X}_{,x}\|^2 + \|\mathbf{X}_{,y}\|^2 \pm \sqrt{(\|\mathbf{X}_{,x}\|^2 - \|\mathbf{X}_{,y}\|^2)^2 + (2 \mathbf{X}_{,x} \cdot \mathbf{X}_{,y})^2} \right]$$

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$$\Rightarrow \boxed{\lambda(\mathbf{M}) = \frac{1}{2} \left[q \pm \sqrt{s^2 + t^2} \right]}$$

$$\left. \begin{array}{l} q(\xi) \\ s(\xi) \\ t(\xi) \end{array} \right] \begin{array}{l} \blacktriangleright \text{polynomial functions} \\ \blacktriangleright \text{same order as } J(\xi) \end{array}$$

Operator definition

- For
- ▶ polynomial function f ,
 - ▶ exact Bézier expansion after k subdivisions,
 - ▶ subdomain $d \in D(k)$:

$\min_{[d]}^b f(\cdot)$ & $\max_{[d]}^b f(\cdot)$:= min/max of Bézier coefficients

$\min_{[d]}^l f(\cdot)$ & $\max_{[d]}^l f(\cdot)$:= min/max of Bézier coefficients at corners

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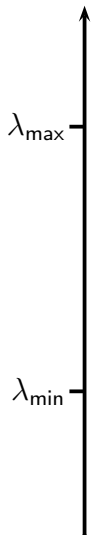
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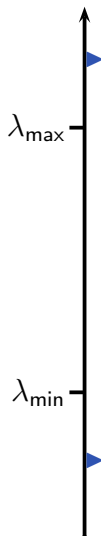
Properties:

- ▶ $\min_{d \in D(k)} \min_{[d]} b f \leq \min f \leq \min_{d \in D(k)} \min_{[d]} l f$
- ▶ $\mu \left(\left[\min_{d \in D(k)} \min_{[d]} b f, \min_{d \in D(k)} \min_{[d]} l f \right] \right) = \mathcal{O} \left(\frac{1}{k^2} \right)$

Bounds - in 2D



Bounds - in 2D


$$\lambda_{\max}^{UB}(\mathbf{M}) = \max_{d \in D(k)} \frac{1}{2} \left[\max_{[d]} q + \sqrt{\left(\max_{[d]} s \right)^2 + \left(\max_{[d]} t \right)^2} \right]$$
$$\lambda_{\min}^{LB}(\mathbf{M}) = \min_{d \in D(k)} \frac{1}{2} \left[\min_{[d]} q - \sqrt{\left(\max_{[d]} s \right)^2 + \left(\max_{[d]} t \right)^2} \right]$$

Bounds - in 2D

λ_{\max}

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Eigenvalues in 3D ?

Eigenvalues in 3D ? Yes

$$\lambda_i = \frac{\|\mathbf{x}_{,x}\|^2 + \|\mathbf{x}_{,y}\|^2 + \|\mathbf{x}_{,z}\|^2}{3} + \frac{2}{\sqrt{6}} \rho(\cdot) \cos\left(\phi(\cdot) + i\frac{2\pi}{3}\right)$$

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$$\rho(\cdot) = \sqrt{\begin{aligned} &\left(\frac{2\|\mathbf{x}_x\|^2}{3} - \frac{\|\mathbf{x}_y\|^2 + \|\mathbf{x}_z\|^2}{3}\right)^2 + \left(\frac{2\|\mathbf{x}_y\|^2}{3} - \frac{\|\mathbf{x}_x\|^2 + \|\mathbf{x}_z\|^2}{3}\right)^2 \\ &+ \left(\frac{2\|\mathbf{x}_z\|^2}{3} - \frac{\|\mathbf{x}_x\|^2 + \|\mathbf{x}_y\|^2}{3}\right)^2 \\ &+ 2[(\mathbf{x}_x \cdot \mathbf{x}_y)^2 + (\mathbf{x}_x \cdot \mathbf{x}_z)^2 + (\mathbf{x}_y \cdot \mathbf{x}_z)^2] \end{aligned}}$$

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$\phi(\cdot)$ is awful

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$\phi(\cdot)$ is awful

$$\left. \begin{array}{l} \lambda_{\max} \\ \lambda_{\min} \end{array} \right\} \simeq \frac{\|\mathbf{x}_x\|^2 + \|\mathbf{x}_y\|^2 + \|\mathbf{x}_z\|^2}{3} \pm \frac{2}{\sqrt{6}} \rho(\cdot)$$