

Integral Equations and Fast Algorithms

Lecture 20: High-order numerics, IE discretizations

CS 598AK · October 31, 2013

Outline

Intro to Numerics

IE discretization: Nyström

Generalized Vandermonde matrices

$$\begin{pmatrix} \varphi_0(x_0) & \varphi_1(x_0) & \cdots & \varphi_n(x_0) \\ \varphi_0(x_1) & \varphi_1(x_1) & \cdots & \varphi_n(x_1) \\ \vdots & \vdots & \ddots & \vdots \\ \varphi_0(x_n) & \varphi_1(x_n) & \cdots & \varphi_n(x_n) \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = ?$$

Generalized Vandermonde matrices

$$\begin{pmatrix} \varphi_0(x_0) & \varphi_1(x_0) & \cdots & \varphi_n(x_0) \\ \varphi_0(x_1) & \varphi_1(x_1) & \cdots & \varphi_n(x_1) \\ \vdots & \vdots & \ddots & \vdots \\ \varphi_0(x_n) & \varphi_1(x_n) & \cdots & \varphi_n(x_n) \end{pmatrix} \begin{pmatrix} \text{MODAL COEFFS} \end{pmatrix} = \begin{pmatrix} \text{NODAL COEFFS} \end{pmatrix}$$

Instantly generalizes to multiple dimensions

Generalized Vandermonde matrices

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Instantly generalizes to multiple dimensions

Invertible?

Generalized Vandermonde matrices

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Instantly generalizes to multiple dimensions

Invertible?

Conditioning?

Interpolation

Interpolation node placement demo

Vandermonde conditioning

Vandermonde conditioning demo

2D basis demo

2D basis demo

Common operations

(Generalized) Vandermonde matrices simplify common operations:

- Modal \leftrightarrow Nodal (“Global interpolation”)

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- Differentiation

Common operations

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- Modal \leftrightarrow Nodal (“Global interpolation”)
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- Differentiation
- Indefinite Integration

Common operations

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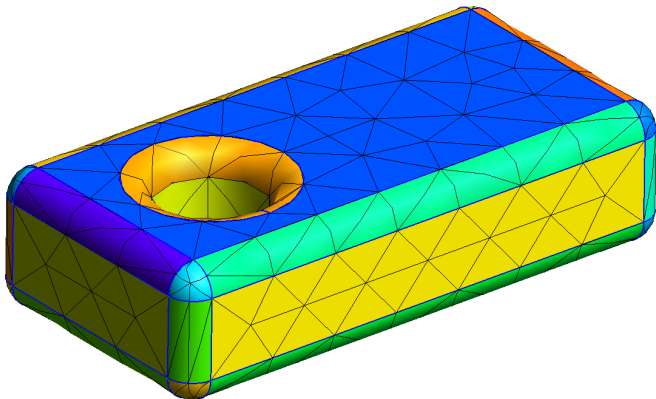
- Modal \leftrightarrow Nodal (“Global interpolation”)
 - Filtering
 - Up-/Oversampling
- Point interpolation (*Hint: solve using V^T*)
- Differentiation
- Indefinite Integration
- Inner product

Common operations

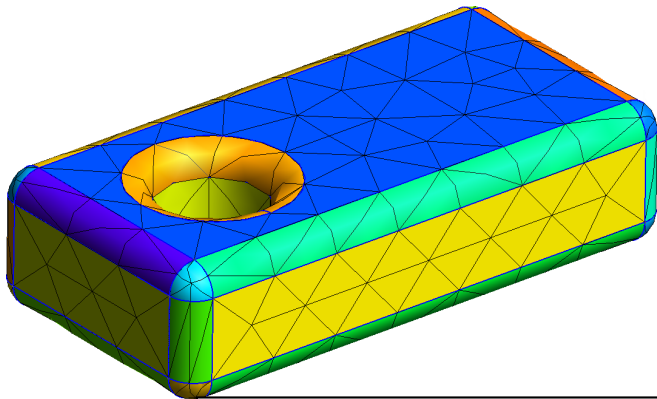
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- Inner product
- Definite integration

Unstructured mesh

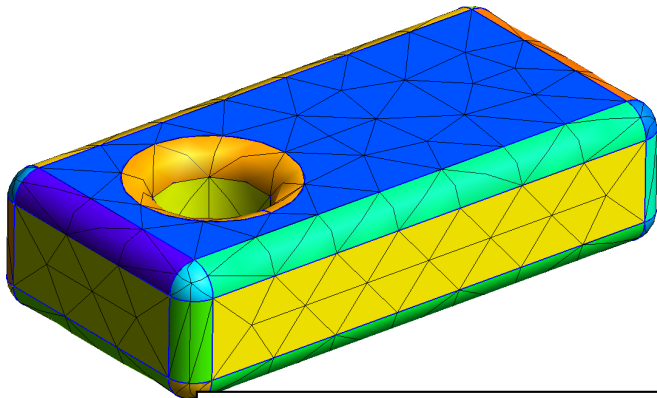


Unstructured mesh



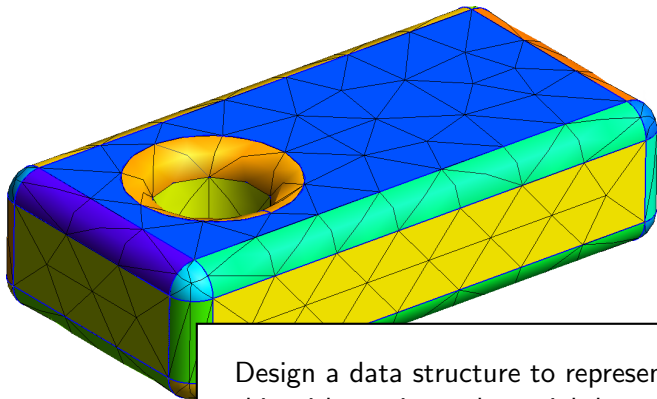
Design a data structure to represent this

Unstructured mesh



Design a data structure to represent this with varying polynomial degrees

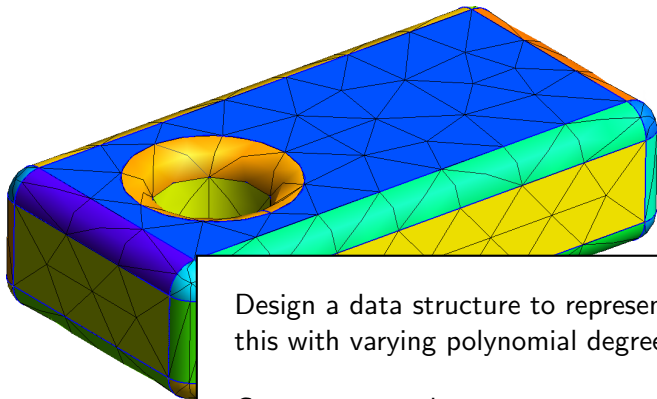
Unstructured mesh



Design a data structure to represent this with varying polynomial degrees

Compute normal vectors

Unstructured mesh

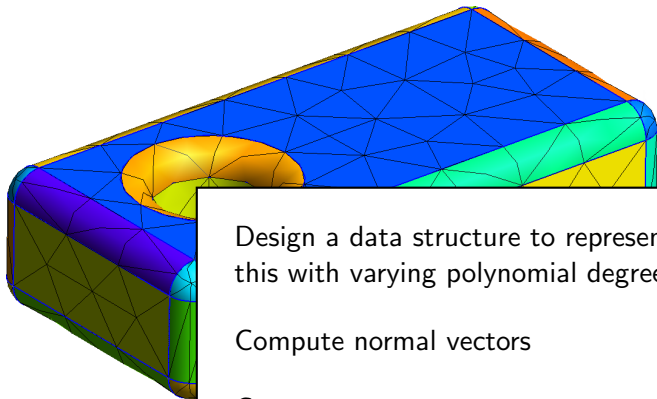


Design a data structure to represent this with varying polynomial degrees

Compute normal vectors

Compute area

Unstructured mesh



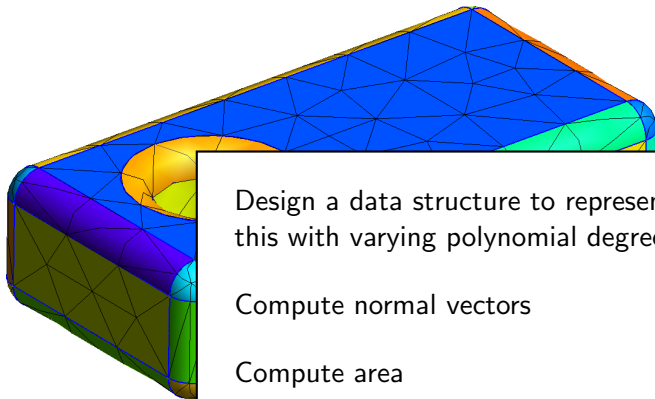
Design a data structure to represent this with varying polynomial degrees

Compute normal vectors

Compute area

Compute integral of a function

Unstructured mesh



Design a data structure to represent this with varying polynomial degrees

Compute normal vectors

Compute area

Compute integral of a function

How is the function represented?

Outline

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IE discretization: Nyström

IE discretization: Overview

$$\varphi(x) - \int_{\Gamma} K(x,y)\varphi(y)dy = f(x)$$

Nyström

- Approximate integral by quadrature:

$$\int_{\Gamma} f(y)dy \rightarrow \sum_{k=1}^n \omega_k f(y_k)$$

- Evaluate quadrature'd IE at quadrature nodes, solve

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Projection

- Consider residual: $R := \varphi - A\varphi - f$
- Pick projection P_n onto finite-dimensional subspace
 $P_n\varphi := \sum_{k=1}^n \langle \varphi, v_k \rangle w_k \rightarrow$ DOFs $\langle \varphi, v_k \rangle$
- Solve $P_n R = 0$

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Important in projection methods:
subspace (e.g. of $C(\Gamma)$)

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Possible bases for projection?

- Galerkin: $v_k = w_k$
- Collocation:
 $v_k = \delta(x_k), w_k(x_j) = \delta_{jk}$
- Petrov-Galerkin: $v_k \neq w_k$

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Equivalent to projection:

Test IE with test functions

IE discretization: Overview

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Nyström

- Approximate integral: $\int_{\Gamma} f(y)dy \rightarrow \sum_{k=1}^n w_k f(x_k)$
- Evaluate quadrature weights

Projection

- Consider residual: $R = \varphi - P_n \varphi$
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Equivalent to projection:

Test IE with test functions

Collocation and Nyström: the same?

Questions?

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