
Geometric errors in surface finite element methods

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parts joint work with

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Outline

1. Canonical surface FEM

- Implicit surface representations
- FEM
- Consistency errors and a priori estimates

2. Eigenvalue problems

- Definitions
- Results of canonical arguments
- Quadrature-based superconvergence argument

3. A posteriori estimates

- Estimates assuming canonical implicit representation
- Parametric representation of surfaces
- New estimates

4. Conclusion

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Prologue: Basic FEM terminology

Poisson model problem: Solve

$$-\Delta u = f \text{ in } \Omega \subset \mathbb{R}^2, \quad u = 0 \text{ on } \partial\Omega.$$

Sobolev space, forms, and norm: Let

$$a(u, v) := \int_{\Omega} \nabla u \nabla v, \quad m(f, v) := \int_{\Omega} f v.$$

$$\|u\|_{H_0^1(\Omega)} = \|\nabla u\|_{L_2(\Omega)} = a(u, u)^{1/2},$$

$$H_0^1(\Omega) = \{u \text{ s.t. } \|u\|_{H_0^1(\Omega)} < \infty, \quad u = 0 \text{ on } \partial\Omega.\}$$

Infinite dimensional vector space.

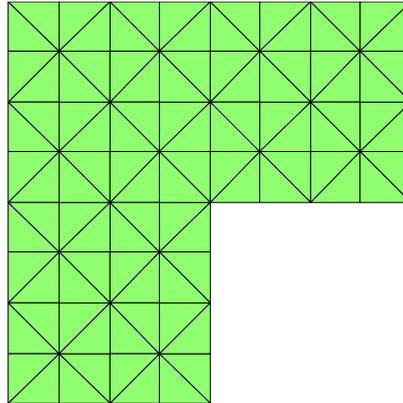
Weak form: Find $u \in H_0^1(\Omega)$ s.t.

$$a(u, v) = m(f, v), \quad \text{all } v \in H_0^1(\Omega).$$

Find basis for $H_0^1(\Omega) \rightsquigarrow$ infinite-dimensional set of linear equations.

Euclidean FEM

Mesh: \mathcal{T}_h is a decomposition of Ω into triangles of diameter h .



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Finite element subspace: Elements of $S_h \subset H_0^1(\Omega)$ are

- Continuous
- 0 on $\partial\Omega$
- **polynomials of degree r** over each $T \in \mathcal{T}_h$.

S_h is a (*finite dimensional vector space*).

Galerkin's method: Find $u_h \in S_h$ s.t.

$$a(u_h, v_h) = m(f, v_h), \quad v_h \in S_h.$$

Basis for $S_h \rightsquigarrow$ finite dimensional set of linear equations.

Projection property: u_h is the orthogonal projection of u onto S_h w.r.t. a :

$$\|u - u_h\|_{H_0^1(\Omega)} = \inf_{\chi \in S_h} \|u - \chi\|_{H_0^1(\Omega)}.$$

Basic error estimates

If u is smooth enough,

$$\|u - u_h\|_{H_0^1(\Omega)} \leq \inf_{\chi \in S_h} \|u - \chi\|_{H_0^1(\Omega)} \leq Ch^r,$$

$$\|u - u_h\|_{L_2(\Omega)} \leq Ch^{r+1}.$$

Relates error to cost and properties of method: h , r .

Variational crimes/Consistency errors

Variational crimes/Consistency errors: Sometimes we define u_h via forms “close to” a and m :

$$A(u_h, v_h) = M(u_h, v_h), \quad \text{all } v_h \in S_h,$$

with $A \approx a$ and $M \approx m$.

Must account for loss of projection property in error analysis:

$$\|u - u_h\|_{H^1(\Omega)} \leq \inf_{\chi \in S_h} \|u - \chi\|_{H^1(\Omega)} + \|A - a\|_* + \|M - m\|_*.$$

(Effects on convergence rate depend on the situation...).

1. Laplace-Beltrami problem

Definitions:

- γ is a compact, 2-dimensional C^2 surface without boundary in \mathbb{R}^3 .
- f is (given) data satisfying $\int_{\gamma} f \, d\sigma = 0$.
- ∇_{γ} , Δ_{γ} are the tangential gradient and Laplace-Beltrami operator.

Model problem (strong form):

$$-\Delta_{\gamma}u = f \text{ on } \gamma.$$

Dirichlet form and L_2 inner product:

$$a(u, v) := \int_{\gamma} \nabla_{\gamma}u \nabla_{\gamma}v \, d\sigma, \quad m(f, v) := \int_{\gamma} f v \, d\sigma.$$

Weak form of the Laplace-Beltrami problem: Find $u \in H^1(\gamma)$ s.t.

$$a(u, v) = m(f, v) \text{ for all } v \in H^1(\gamma).$$

We require $\int_{\gamma} u \, d\sigma = 0$ to ensure uniqueness.

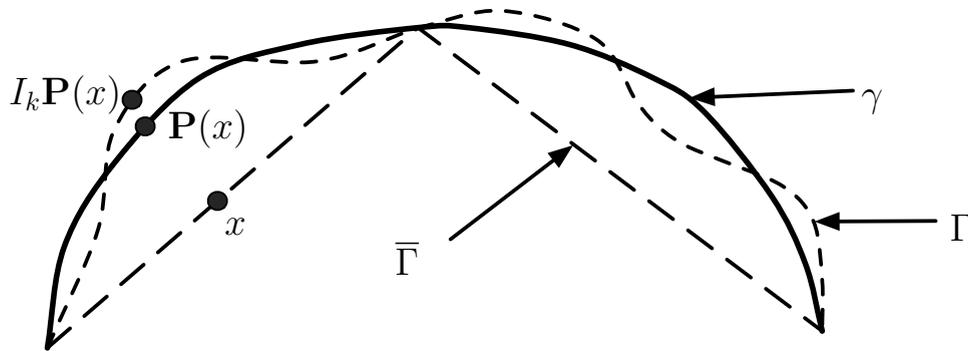
Applications of surface PDE

Why solve the Laplace-Beltrami problem?

1. Geometry: Mean curvature flow, etc.
2. Image and surface processing
3. Physical modeling: Surface tension in two-phase flow; biomembranes
4. Shape registration: Spectrum can serve as a “shape DNA”

Surface FEM

- *Base discrete surface:* $\bar{\Gamma}$ is a polyhedron with triangular or quad faces.
- **Basic mapping assumption:** There is a “reasonably nice” map $\mathbf{P} : \bar{\Gamma} \rightarrow \gamma$.
- *Polynomial surface approximation:* $\Gamma = I_k \mathbf{P}(\bar{\Gamma})$ with I_k a **degree- k** Lagrange interpolant.



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- *Polynomial surface approximation:* $\Gamma = I_k \mathbf{P}(\bar{\Gamma})$ with I_k a **degree- k** Lagrange interpolant.
- *Meshes:* $\bar{\mathcal{T}}$ is the set of faces of $\bar{\Gamma}$, \mathcal{T} is the faces of Γ .
- *Finite element space:* $S_{\mathcal{T}}$ is the piecewise **degree- r** polynomials over Γ .
- *Data:* f is defined on γ , so have to define data F on Γ .
- *Forms on Γ :*

$$A(U, V) := \int_{\Gamma} \nabla_{\Gamma} U \nabla_{\Gamma} V \, d\sigma_{\mathcal{T}}, \quad M(F, V) := \int_{\Gamma} FV \, d\sigma_{\mathcal{T}}.$$

- *Finite element method:* Find $U \in S_{\mathcal{T}}$ such that $\int_{\Gamma} U \, d\sigma_{\mathcal{T}} = 0$ and

$$A(U, V) = M(F, V), \quad V \in S_{\mathcal{T}}.$$

Choosing P

Canonical choice historically for C^2 surfaces: Implicit representation.

Viewpoint: $\gamma = \{x : d(x) = 0\}$ with d the signed distance function.

Then: For x lying in a sufficiently small tubular neighborhood U of γ ,

- *Orthogonal closest-point projection onto γ :*

$$\mathbf{P}_d(x) := x - d(x)\vec{\nu}(x)$$

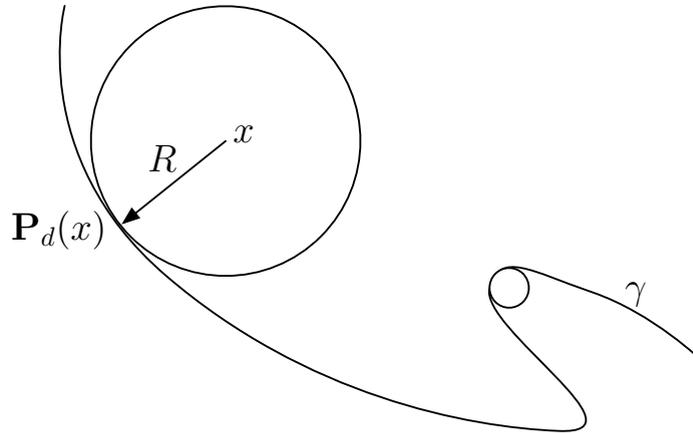
with $\nu = \nabla d$ the unit normal on γ .

Ups and downs:

- + Correct theoretical properties in FEM.
- Often difficult to access in codes (explicit formulas only for sphere, torus).
- Surface regularity less than C^2 ?

We'll look at other options later...

Curvature and the closest point projection



Notes:

- *Curvature*: $\kappa(\mathbf{P}_d(x)) = 1/R$ with R the maximum radius of open balls tangent to but not intersecting γ .
- *Closest point projection*: Uniquely defined on a tubular neighborhood of γ having width $\inf_{x \in \gamma} \frac{1}{|\kappa(x)|}$.

Surface regularity and the distance function

Two distinct regimes of surface regularity:

1. γ is $C^{1,1}$ or smoother (can be locally described via a $C^{1,1}$ diffeomorphism):
 - The distance function, closest point projection behave as described above.
 - Distance function inherits surface regularity: γ is $C^k \Rightarrow d$ is also C^k .
2. For any γ not $C^{1,1}$ (say, $C^{1,\alpha}$ with $\alpha < 1$):
 - d does NOT inherit surface regularity: d is only Lipschitz.
 - \mathbf{P}_d is not uniquely defined on ANY open neighborhood of γ .
 - Established in [Lucas, 1957] and [Federer, 1959].

Geometric consistency error

Dirichlet consistency matrix: With $\mathbf{E}_{\mathbf{P}_d}$ a matrix determined by using change of variables formulas for the mapping \mathbf{P}_d and $U^\ell = U \circ \mathbf{P}_d^{-1}$,

$$A(U, V) - a(U^\ell, V^\ell) = \int_{\gamma} \mathbf{E}_{\mathbf{P}_d} \nabla_{\gamma} U^\ell \nabla_{\gamma} V^\ell d\sigma.$$

- Computing $\mathbf{E}_{\mathbf{P}_d}$ requires computing distance function d and derivatives.
- *Order of consistency error:* On a triangle T of size h ,

$$\|\mathbf{E}_{\mathbf{P}_d}\|_{L^\infty(T)} \lesssim \|d\|_{L^\infty(T)} + \|\vec{\nu} - \vec{\nu}_h\|_{L^\infty(T)}^2 \lesssim h^{k+1} + h^{2k} \lesssim h^{k+1},$$

where $\vec{\nu}$ and $\vec{\nu}_h$ are normals to γ and Γ .

- An $O(h^{k+1})$ consistency error is observed *essentially independently of the method used to construct Γ* (interpolation of \mathbf{P}_d isn't necessary in practice!).

A priori estimates for surface FEM

Theorem 1 (Dz88, De09). *For discrete data F consistently chosen,*

$$\begin{aligned}\|\nabla_\gamma(u - U^\ell)\|_{L_2(\gamma)} &\lesssim h^r \|u\|_{H^{r+1}(\gamma)} + \|\mathbf{E}_{\mathbf{P}_d}\|_{L_\infty(\gamma)} \|\nabla_\gamma u\|_{L_2(\gamma)} \\ &\lesssim h^r \|u\|_{H^{r+1}(\gamma)} + h^{k+1} \|\nabla_\gamma u\|_{L_2(\gamma)}, \\ \|u - U_T^\ell - \frac{1}{|\gamma|} \int_\gamma (u - U_T^\ell)\|_{L_2(\gamma)} &\lesssim h^{r+1} \|u\|_{H^{r+1}(\gamma)} + \|\mathbf{E}_{\mathbf{P}_d}\|_{L_\infty(\gamma)} \|\nabla_\gamma u\|_{L_2(\gamma)} \\ &\lesssim h^{r+1} \|u\|_{H^{r+1}(\gamma)} + h^{k+1} \|\nabla_\gamma u\|_{L_2(\gamma)}.\end{aligned}$$

Notes:

- Error consists of a **Galerkin error** and a **geometric consistency error**.
- Geometric error is the same for energy and L_2 norms.
- $r = k = 1$: Previous estimates require C^3 regularity. This is too much since $C^2 \Rightarrow u \in H^2(\gamma)$. Requirement is reduced to C^2 in recent joint work w/Bonito and Nochetto.

A (too?) general statement

Metatheorem: Geometric consistency errors are of order $k + 1$ for any quantity of interest (various norms, point values...) and any standard surface FEM (mixed, DG, HDG, cut/trace, FEEC, parabolic problems...) for elliptic problems on surfaces.

Proof: See lots of FEM literature starting with [Dziuk 88] (also BEM literature starting with [Nedelec '78], [Bendali '84]...)

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Theme of this talk: Things aren't always that simple!

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Laplace-Beltrami Eigenvalue Problem

- **Strong form:** Find (u, λ) such that:

$$-\Delta_\gamma u = \lambda u,$$

- **Weak eigenvalue problem:** Find $(u, \lambda) \in H^1(\gamma)/\mathbb{R} \times \mathbb{R}^+$ such that

$$a(u, v) = \lambda m(u, v) \quad \forall v \in H^1(\gamma).$$

- **Finite element approximation:** Find $(U, \Lambda) \in S_{\mathcal{T}}/\mathbb{R} \times \mathbb{R}^+$ such that

$$A(U, V) = \Lambda M(U, V), \quad V \in S_{\mathcal{T}}.$$

- **Eigenfunction bound:** Let \mathbf{P}_λ be the $L_2(\gamma)$ projection onto eigenspace associated with λ . For an SFEM eigenpair (U, Λ) associated to an eigenvalue λ of $-\Delta_\gamma$, we have

$$\begin{aligned} \|U - \mathbf{P}_\lambda U\|_{H^1(\gamma)} &\lesssim h^r + h^{k+1}, \\ \|U - \mathbf{P}_\lambda U\|_{L_2(\gamma)} &\lesssim h^{r+1} + h^{k+1}. \end{aligned}$$

Eigenvalue Errors

Theorem 2 (Eigenvalue Bound). *Let λ be an eigenvalue of the surface eigenvalue problem and let (U, Λ) be a surface FEM eigenpair associated with λ . Then*

$$|\lambda - \Lambda| \leq \underbrace{\|U - \mathbf{P}_\lambda U\|_{H^1(\gamma)}^2}_{O(h^{2r}) + O(h^{2k+2})} + \lambda \underbrace{\|U - \mathbf{P}_\lambda U\|_{L_2(\gamma)}^2}_{O(h^{2r+2}) + O(h^{2k+2})} \\ + \Lambda \underbrace{|m(U, U) - M(U, U)|}_{\text{Geometric}} + \underbrace{|a(U, U) - A(U, U)|}_{\text{Geometric}}.$$

Obvious eigenvalue error bound:

$$|\lambda - \Lambda| \lesssim h^{2r} + \|\mathbf{E}_{\mathbf{P}_d}\|_{L_\infty} \lesssim h^{2r} + h^{k+1}.$$

Some Test Shapes

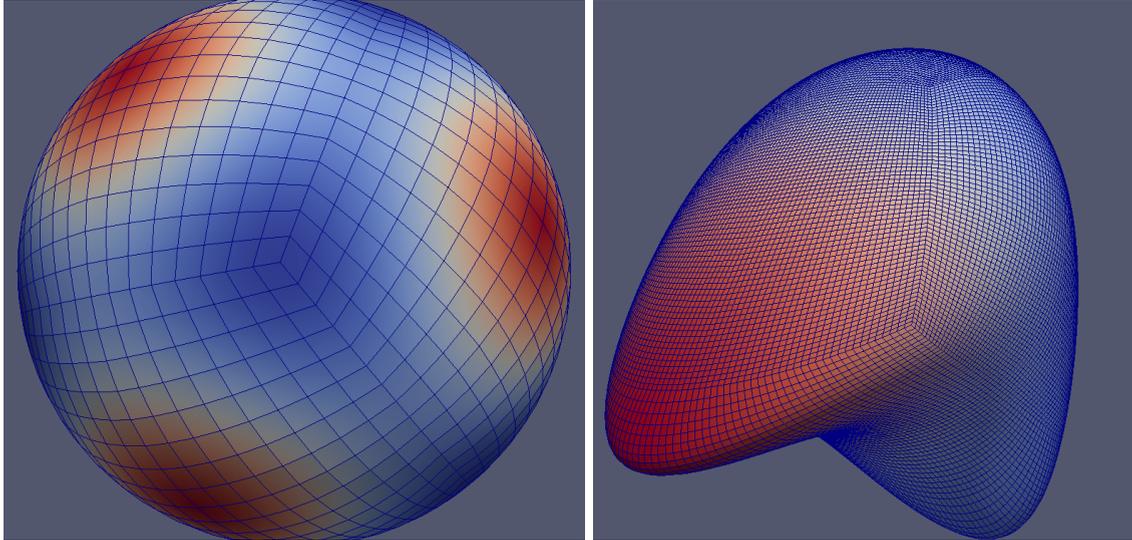
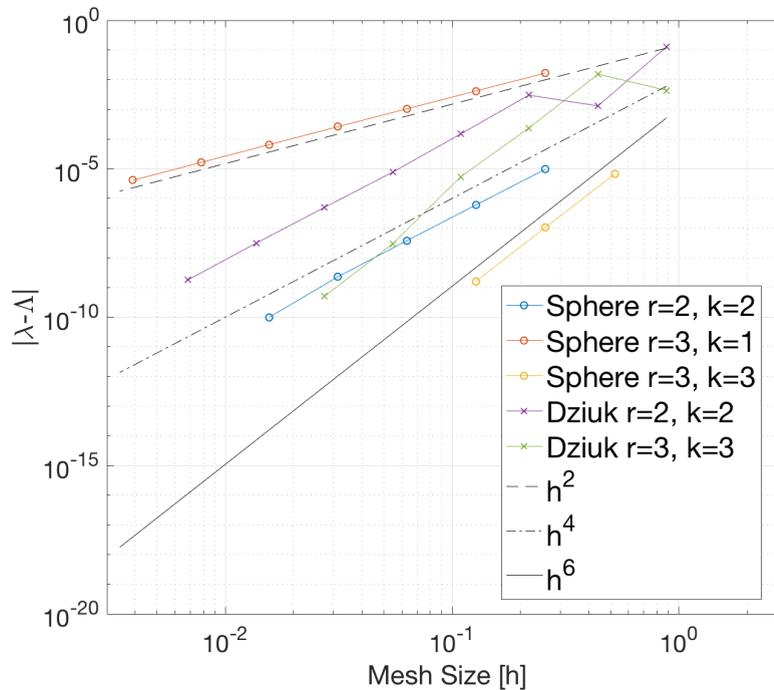


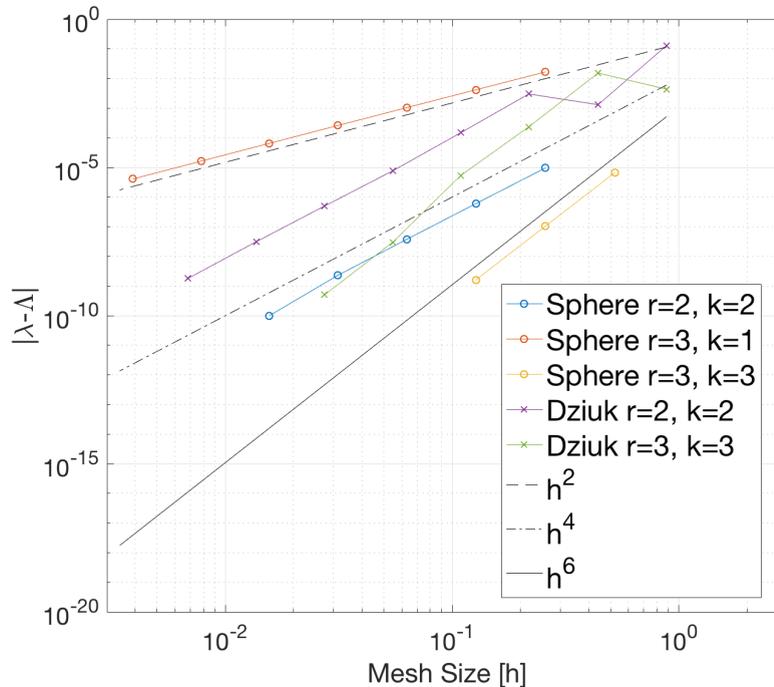
Figure 1: Sphere and Dziuk surface used in deal.ii computations of eigenvalues.

Numerical Experiments: Quadrilateral Elements



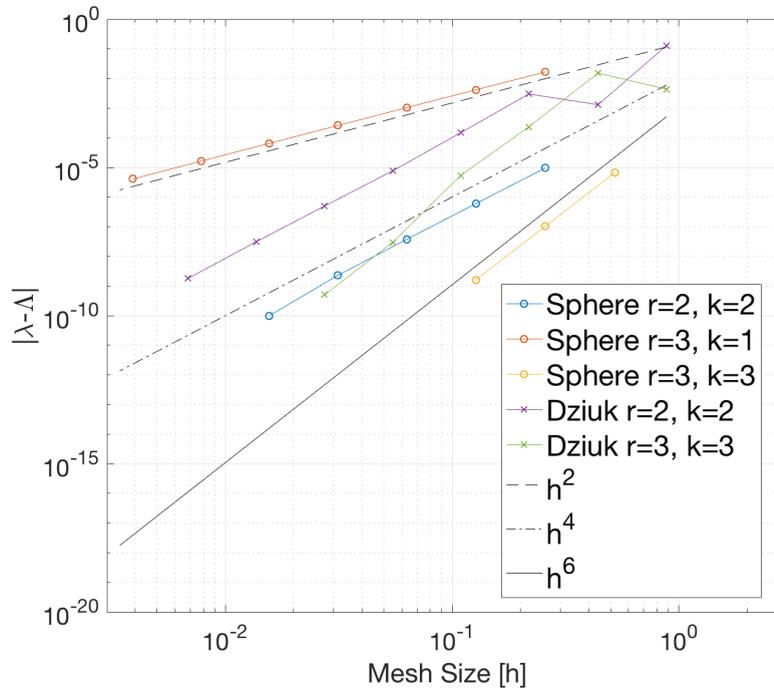
Looking for : $O(h^{k+1})$.

Numerical Experiments: Quadrilateral Elements



Strange Behavior: Geometric error is $O(h^{2k})$ rather than the expected $O(h^{k+1})$.

Numerical Experiments: Quadrilateral Elements



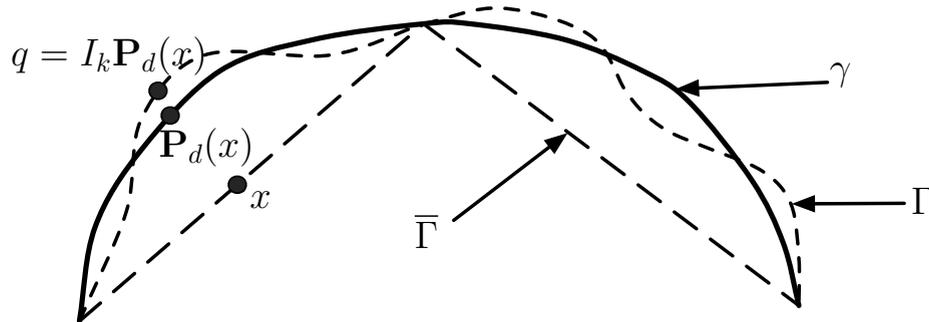
Important observation: deal.ii constructs Γ using interpolation of \mathbf{P}_d at Gauss-Lobatto interpolation points, NOT canonical Lagrange points.

An Explanation of Superconvergence

Lemma 1. Up to terms of order h^{2k} ,

$$|m(V, V) - M(V, V)| \leq \left| \int_{\Gamma} V(q)^2 d(q) \sum_{i=1}^n \frac{\kappa_i(\mathbf{P}_d(q))}{1 + d(q)\kappa_i(\mathbf{P}_d(q))} d\Sigma \right|,$$

where $\{\kappa_i\}_{i=1}^n$ are the principal curvatures of the surface.



Geometric Error Acts Like Quadrature Error

- **Exploit distance function:** The zeros of $d(q)$, $\{q_j\}_{j=1}^N$, on each face of Γ are the interpolation points used to create Γ .
- **Create quadrature rule:** Use the zeros of $d(q)$ to create a quadrature rule:

$$QUAD := \sum_{T \subset \Gamma} \sum_{j=1}^N W_j V(q_j)^2 \cancel{d(q_j)} \overset{0}{\sum_{i=1}^n} \frac{\kappa_i(\mathbf{P}(q_j))}{1 + d(q_j) \kappa_i(\mathbf{P}(q_j))} = 0$$

Theorem 3 (Quadrature Error). *Up to terms of order h^{2k} ,*

$$\begin{aligned} |m(V, V) - M(V, V)| &\leq \left| \int_{\Gamma} V(q)^2 d(q) \sum_{i=1}^n \frac{\kappa_i(\mathbf{P}(q))}{1 + d(q) \kappa_i(\mathbf{P}(q))} d\Sigma \right| \\ &= \left| \int_{\Gamma} V(q)^2 d(q) \sum_{i=1}^n \frac{\kappa_i(\mathbf{P}(q))}{1 + d(q) \kappa_i(\mathbf{P}(q))} d\Sigma - QUAD \right|. \end{aligned}$$

Conclusion for quad meshes

Corollary 4 (Superconvergence in deal.ii Computations). *If degree k interpolation points based on Gauss-Lobatto quadrature are used in the construction of Γ , U is the SFEM eigenfunction of Λ , and $\mathbf{P}_\lambda U$ has enough regularity, then*

$$|m(U, U) - M(U, U)| \lesssim h^{2k},$$

$$|a(U, U) - A(U, U)| \lesssim h^{2k},$$

and

$$|\lambda - \Lambda| \lesssim h^{2r} + h^{2k}.$$

Note: Tensor product of $k + 1$ points used in the 1D Gauss-Lobatto quadrature rule yields a quadrature rule exact for degree $2k - 1$.

Triangular meshes

Notes:

- Interpolation points are standard Lagrange points.
- Elementwise quadrature error for associated quadrature rule is $O(h^{k+1})$.
- Computational observation: Expected order h^{k+1} for odd k , superconvergent order h^{k+2} for even k .
- Observed orders were robust: Only exception was nodes perturbed off of surface with bias in one direction (e.g., outside of surface).
- Could be explained within our framework by known superconvergence phenomena for semi-structured meshes such as ones in which adjacent triangles form near-parallelograms.
- We didn't seem to have such structured meshes, but did not explore further down the superconvergence rabbit hole.

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Setting

Back to the Laplace-Beltrami source problem:

$$-\Delta_\gamma u = f \text{ on } \gamma.$$

FEM: Find $U_\mathcal{T} \in S_\mathcal{T}$ s.t.

$$A(U_\mathcal{T}, V) = (F, V), \quad V \in S_\mathcal{T}.$$

Goal: A posteriori (computable) estimates that bound the error:

$$\|\nabla_\gamma(u - U_\mathcal{T})\|_{L_2(\gamma)} \leq \mathcal{F}(U_\mathcal{T}, F) + \mathcal{G}(U_\mathcal{T}, F),$$

where \mathcal{F} is a computable term controlling the Galerkin error and \mathcal{G} is a computable term controlling the geometric error.

Note: Computing \mathcal{G} will require computing the map \mathbf{P} between Γ and γ .
How does our choice of \mathbf{P} affect the estimates?

A posteriori estimates on implicit surfaces

Fundamental assumption: γ is represented *in implementation* using the closest point projection $\mathbf{P}_d(x) = x - d(x)\vec{\nu}(x)$.

Residual indicator: For $T \in \mathcal{T}$,

$$\eta_T = h_T \|F + \Delta_\Gamma U_T\|_{L_2(T)} + h_T^{1/2} \|[\![\nabla_\Gamma U_T]\!] \|_{L_2(\partial T)}.$$

Theorem 5 (De-Dz '07). Assume $F(x) = J_{\mathbf{P}_d}(f \circ \mathbf{P}_d)$ with $J_{\mathbf{P}_d}$ the Jacobian of \mathbf{P}_d . Then

$$\|\nabla_\gamma(u - U_T^\ell)\|_{L_2(\gamma)}^2 \lesssim \sum_{T \in \mathcal{T}} \eta_T^2 + \|\mathbf{E}_{\mathbf{P}_d}\|_{L_\infty(\Gamma)}^2 \|\nabla_\Gamma U_T\|_{L_2(\Gamma)}^2.$$

Notes:

- Galerkin error + geometric consistency error
- Everything is computable IF we can compute d and its derivatives (needed to compute/estimate $\mathbf{E}_{\mathbf{P}_d}$).
- Can also work with a more general level set function, but still need to approximate d .

Summary: Estimates on implicit surfaces

Pluses:

- + Geometric error is of order h^{k+1} : “Superconvergent”.

Minuses:

- Analytical framework requires C^2 surface.
- **A posteriori estimates require evaluation of distance function:
Only explicitly available for sphere and torus!**

Second option for \mathbf{P}

Framework: There is an elementwise-smooth bi-Lipschitz map $\mathbf{P} : \bar{\Gamma} \rightarrow \gamma$ which we have access to in our code.

Simple example: γ is the graph of a function g over a Euclidean domain Ω ; $\mathbf{P} \neq \mathbf{P}_d$ is the “vertical” map induced by g .

Advantages:

1. More flexibility in representing smooth surfaces
2. Allows for less than C^2 surfaces.

Drawback: Theoretical properties aren't so nice!

Consistency errors

Assume \mathbf{P} is an arbitrary “reasonable” parametric map:

- Error representation: With $\mathbf{E}_{\mathbf{P}}$ a matrix derived from change of variables,

$$A(U, V) - a(U \circ \mathbf{P}^{-1}, V \circ \mathbf{P}^{-1}) = \int_{\gamma} \mathbf{E}_{\mathbf{P}} \nabla_{\gamma}(U \circ \mathbf{P}^{-1}) \nabla_{\gamma}(V \circ \mathbf{P}^{-1}) d\sigma.$$

- Computing $\mathbf{E}_{\mathbf{P}}$ only requires access to \mathbf{P} .
- Standard arguments for isoparametric FEM yield

$$\|\mathbf{E}_{\mathbf{P}}\|_{L^{\infty}(T)} \lesssim h^k.$$

The moral of the story: $O(h^{k+1})$ geometric errors are observed for smooth surfaces independent of \mathbf{P} used in implementation. *Thus we should use \mathbf{P}_d for theoretical purposes.*

A posteriori estimates: Parametric viewpoint

Theorem 6 (BCMMN, 2016). *Let $F = J_{\mathbf{P}}(f \circ \mathbf{P})$. Then under reasonable assumptions,*

$$\|\nabla_{\gamma}(u - U_{\mathcal{T}} \circ \mathbf{P})\|_{L_2(\gamma)}^2 \lesssim \sum_{T \in \mathcal{T}} \eta_T^2 + \|\nabla(\mathbf{P} - I_k \mathbf{P})\|_{L_{\infty}(\bar{\Gamma})}^2.$$

Properties:

- + Practical computation uses \mathbf{P} : Flexible!
- + Allows for less-than- C^2 surfaces.
- + AFEM convergence, optimality proved.
- Geometric consistency error $\|\nabla(\mathbf{P} - I_k \mathbf{P})\|_{L_{\infty}(\bar{\Gamma})}$ is only order h^k , *not* order h^{k+1} as in the implicit formulation.
- AFEM significantly overrefines to resolve geometric error.

A posteriori estimates: Merged perspective

Basic idea: Use generic \mathbf{P} for *implementation*, but use \mathbf{P}_d for *theory*.

The heart of our result:

$$\|\mathbf{E}_{\mathbf{P}_d}\|_{L^\infty(\gamma)} \lesssim \|\mathbf{P} - I_k \mathbf{P}\|_{L^\infty(\bar{\Gamma})} + \|\nabla(\mathbf{P} - I_k \mathbf{P})\|_{L^\infty(\bar{\Gamma})}^2 =: \epsilon_{\mathcal{T}}.$$

Theorem 7 (De.-Bonito). *Assume that γ is C^2 , and that a parametric FEM is used with $F = J_{\mathbf{P}}(f \circ \mathbf{P})$. Then under reasonable assumptions,*

$$\|\nabla_\gamma(u - U_{\mathcal{T}})\|_{L_2(\gamma)}^2 \lesssim \sum_{T \in \mathcal{T}} \eta_T^2 + \epsilon_{\mathcal{T}}^2.$$

Notes:

1. $\epsilon_{\mathcal{T}}$ is computable using only information from the parametric representation, but heuristically $\epsilon_{\mathcal{T}} \lesssim h^{k+1}$.
2. Central observation in proofs: \mathbf{P}_d is the closest point projection implies

$$|x - \mathbf{P}_d(x)| \leq |x - \mathbf{P}(x)|.$$

Numerical experiments

Computational setup for Experiment 1:

- **Smooth geometry:** γ is a half-sphere (smooth) \rightsquigarrow uniform geometric refinement.
- **Rough solution:** u is singular at the north pole \rightsquigarrow localized PDE refinement at pole.
- **Software:** Computations were performed using deal.ii.
- **Adaptive algorithm:** Selectively choose elements to subdivide based on elementwise quantities:

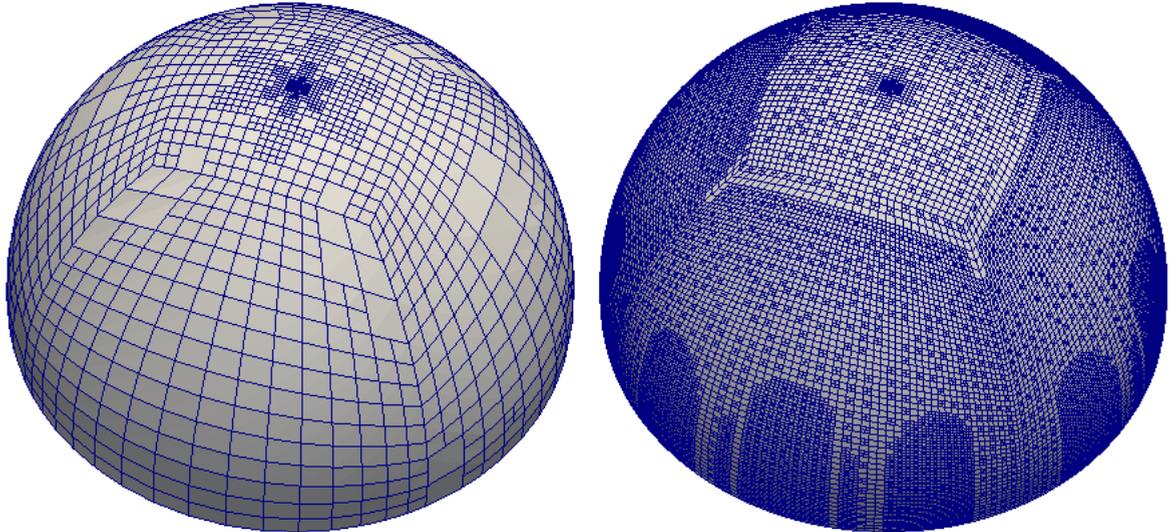
$$\eta_T \text{ (Galerkin error)}$$

and

$$\text{either } \epsilon_T \text{ or } \|\nabla(\mathbf{P} - I_k\mathbf{P})\|_{L^\infty(T)} \text{ (geometric error).}$$

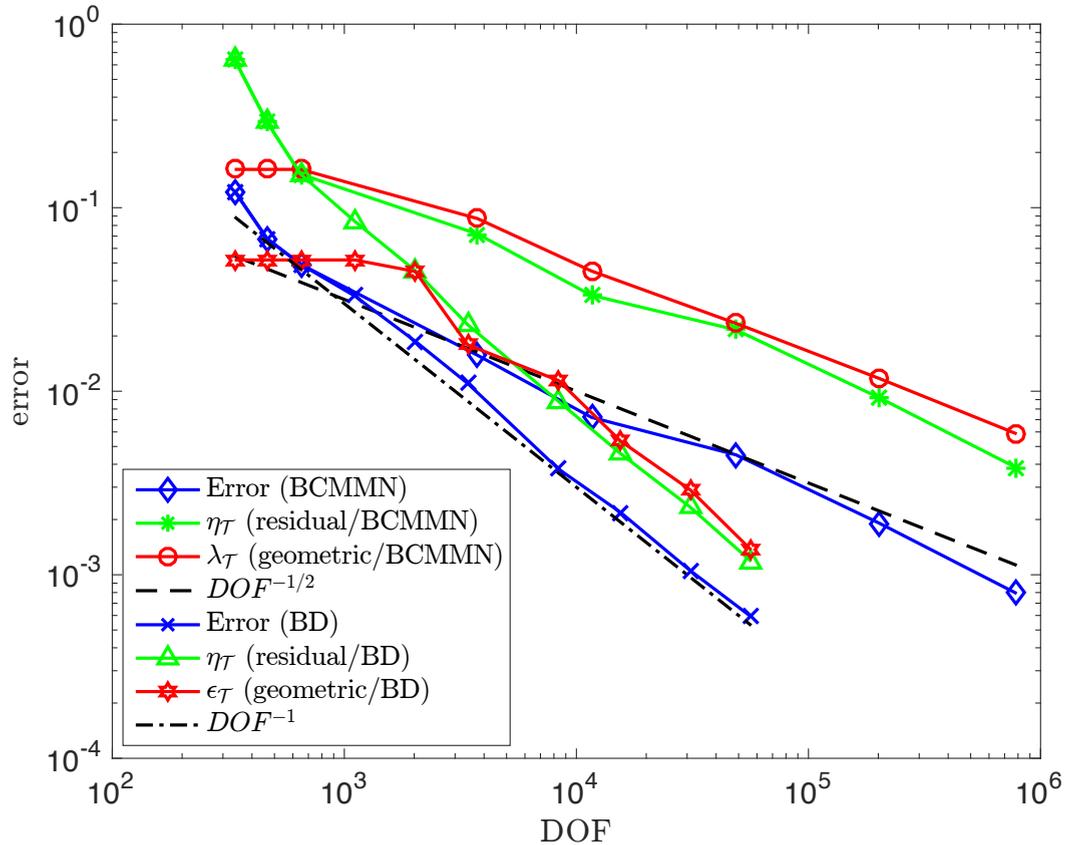
- **Polynomial degree:** We show results for $r = 2$, $k = 1$.
(Algorithms perform similarly for isoparametrics ($r = k = 1$)).

Meshes



Adaptive meshes after 10 AFEM iterations with $r = 2$, $k = 1$: BD refinement (left) and BCMMN refinement (right).

Error decrease

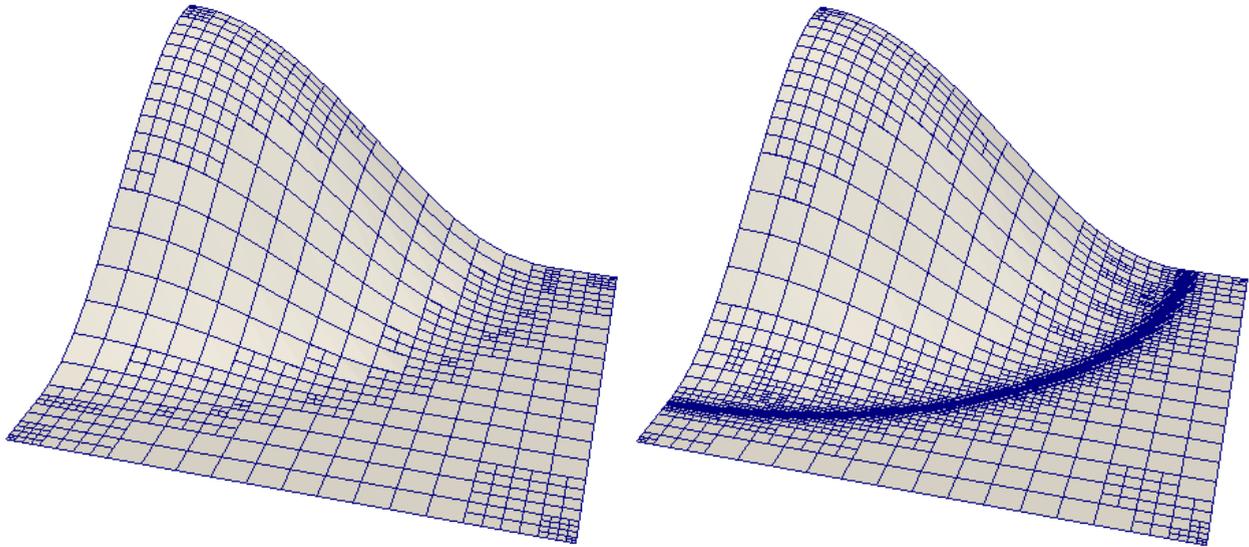


Experiment 2

Setup:

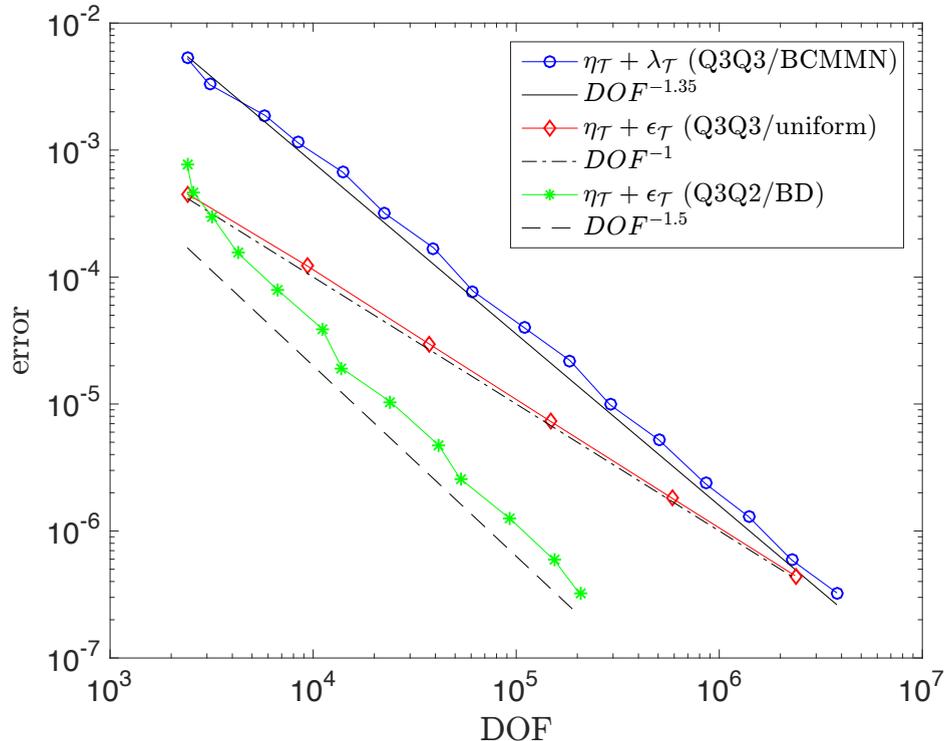
- γ is a $C^{2,\alpha}$ surface ($\alpha = 2/5$).
- $f \equiv 1$, u is unknown.
- Can show: $u \in H^{3-\epsilon}$, any $\epsilon > 0$.
- Three refinement routines:
 1. Uniform (measure geometric error w/BD estimator).
 2. AFEM: Q3/Q3 with BCMMN estimator, tolerance 5×10^{-7} .
 3. AFEM: Q3/Q2 with BD estimator, tolerance 5×10^{-7} .

Adaptive meshes



Adaptive meshes after 20 AFEM iterations: BD refinement with $r = 3, k = 2$ (left) and BCMMN refinement with $r = k = 3$ (right).

Error decrease



Note: Adaptive error decrease is suboptimal for Q3Q3/BCMMN, but NOT for Q3Q2/BD.

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Conclusions

Some future directions:

1. Nonsmooth (not C^2) surfaces
 - Closest point projection isn't immediately useful.
 - So far, only parametric viewpoint has been used in proofs: $O(h^\alpha)$ geometric error on a $C^{1,\alpha}$ surface.
 - Implies $O(h^\alpha)$ geometric error on $C^{1,\alpha}$ surfaces with $\alpha < 1$, but $O(h^2)$ on C^2 surfaces (!).
2. Vector Laplacians/Stokes on surfaces:
 - Several recent papers: Metatheorem doesn't always hold.
 - For some methods $O(h^{k+1})$ geometric error can be recovered if a better approximation to the normal is used in the definition of the discrete energy inner product.
 - For other methods the situation is less clear.