# Geometric Decomposition of Serendipity Finite Element Spaces

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Joint Mathematics Meetings 2013 AMS Special Session Finite Element Exterior Calculus and Applications

#### What is a serendipity finite element method?

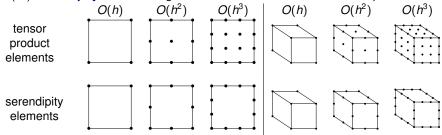
**Goal:** Efficient, accurate approximation of the solution to a PDE over  $\Omega \subset \mathbb{R}^n$ .

Standard  $O(h^r)$  tensor product finite element method in  $\mathbb{R}^n$ :

- $\rightarrow$  Mesh  $\Omega$  by *n*-dimensional cubes of side length *h*.
- $\rightarrow$  Set up a linear system involving  $(r+1)^n$  degrees of freedom (DoFs) per cube.
- $\rightarrow$  For unknown continuous solution u and computed discrete approximation  $u_h$ :

$$\underbrace{||u-u_h||_{H^1(\Omega)}}_{\text{approximation error}} \leq \underbrace{C \, h^r \, |u|_{H^{r+1}(\Omega)}}_{\text{optimal error bound}}, \quad \forall u \in H^{r+1}(\Omega).$$

A  $O(h^r)$  serendipity FEM converges at the same rate with fewer DoFs per element:

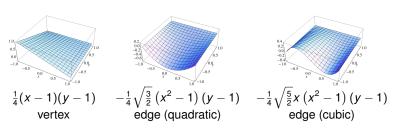


**Example:** For  $O(h^3)$ , d=3, 50% fewer DoFs  $\rightarrow \infty$  smaller linear system

#### What is a geometric decomposition?

A **geometric decomposition** for a finite element space is an explicit correspondence:

- → Previously known basis functions employ Legendre polynomials
- → These functions bear no symmetrical correspondence to the domain points

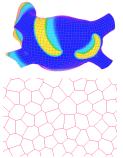


SZABÓ AND I. BABUŠKA Finite element analysis, Wiley Interscience, 1991.

#### Motivations and Related Topics

**Goal:** Construct geometric decompositions of serendipity spaces using linear combinations of standard tensor product functions. **Focus:** Cubic Hermites.





- Modern mathematics: Finite Element Exterior Calculus, Discrete Exterior Calculus, Virtual Element Methods...
   ARNOLD, AWANOU The serendipity family of finite elements, Found. Comp. Math, 2011.
- Isogeometric analysis: Finding basis functions suitable for both domain description and PDE approximation avoids the expensive computational bottleneck of re-meshing.
  - COTTRELL, HUGHES, BAZILEVS Isogeometric Analysis: Toward Integration of CAD and FEA, Wiley, 2009.
- Computational Efficiency: Modern applications such as patient-specific cardiac electrophysiology need efficient, stable, error-bounded 'real-time' methods
- Flexible Domain Meshing: Serendipity type elements for Voronoi meshes provides various computational benefits.
   RAND, GILLETTE, BAJAJ Quadratic Serendipity Finite Elements on Polygons Using Generalized Barycentric

Coordinates arXiv:1109.3259, 2011.

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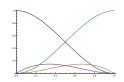
1 The Cubic Case: Hermite Functions, Serendipity Spaces

Geometric Decompsitions of Cubic Serendipity Spaces

Applications and Future Directions

# Cubic Hermite Geometric Decomposition: 1D

Cubic Hermite Basis on [0, 1] 
$$\begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \end{bmatrix} := \begin{bmatrix} 1 - 3x^2 + 2x^3 \\ x - 2x^2 + x^3 \\ x^2 - x^3 \\ 3x^2 - 2x^3 \end{bmatrix}$$



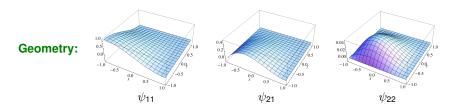
**Approximation:** 
$$x^r = \sum_{i=1}^4 \varepsilon_{r,i} \psi_i$$
, for  $r = 0, 1, 2, 3$ , where  $[\varepsilon_{r,i}] = \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & -2 & 1 \\ 0 & 0 & -3 & 1 \end{pmatrix}$ 

Geometry: 
$$u=u(0)\psi_1+u'(0)\psi_2-u'(1)\psi_3+u(1)\psi_4, \qquad \forall u\in\underbrace{\mathcal{P}_3([0,1])}_{\text{cubic polynomials}}$$

## Cubic Hermite Geometric Decomposition: 2D

$$\underbrace{ \left\{ \begin{array}{c} x^r y^s \\ 0 \leq r, s \leq 3 \end{array} \right\} }_{\mathcal{Q}_3([0,1]^2)} \qquad \longleftrightarrow \qquad \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_j(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_i(y) \\ 1 \leq i, j \leq 4 \end{array} \right\} }_{\text{uniform or monomials}} \qquad \longleftrightarrow \qquad \underbrace{ \left\{ \begin{array}{c} \psi_i(x) \psi_i(y) \\ 1 \leq i,$$

**Approximation:** 
$$x^r y^s = \sum_{i=1}^4 \sum_{j=1}^4 \varepsilon_{r,i} \varepsilon_{s,j} \psi_{ij}$$
, for  $0 \le r, s \le 3$ ,  $\varepsilon_{r,i}$  as in 1D.

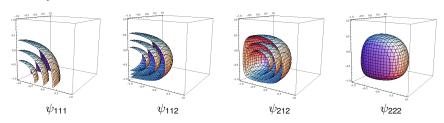


# Cubic Hermite Geometric Decomposition: 3D

$$\underbrace{\left\{\begin{array}{c} x^r y^s z^t \\ 0 \leq r, s, t \leq 3 \end{array}\right\}}_{\mathcal{Q}_3([0,1]^3)} \qquad \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 1 \leq i,j,k \leq 4 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_j(y)\psi_k(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomials}} \longleftrightarrow \qquad \underbrace{\begin{array}{c} \psi_i(x)\psi_i(y)\psi_i(z) \\ 0 \leq r,s,t \leq 3 \end{array}}_{\text{monomial$$

**Approximation:** 
$$x^r y^s z^t = \sum_{i=1}^4 \sum_{j=1}^4 \sum_{k=1}^4 \varepsilon_{r,i} \varepsilon_{s,j} \varepsilon_{t,k} \psi_{ijk}$$
, for  $0 \le r, s, t \le 3$ ,  $\varepsilon_{r,i}$  as in 1D.

Geometry: Contours of level sets of the basis functions:



#### Two families of finite elements on cubical meshes

- $\mathcal{Q}_r \Lambda^k([0,1]^n) \longrightarrow \text{standard tensor product spaces} \qquad (\leq \text{degree } r \text{ in each variable})$  early work: RAVIART, THOMAS 1976, NEDELEC 1980 more recently: ARNOLD, BOFFI, BONIZZONI arXiv:1212.6559, 2012
- $S_r\Lambda^k([0,1]^n) \longrightarrow \text{serendipity finite element spaces} \qquad \text{(superlinear degree } r\text{)}$  early work: STRANG, FIX An analysis of the finite element method 1973 more recently: ARNOLD, AWANOU FoCM 11:3, 2011, and arXiv:1204.2595, 2012.

The **superlinear** degree of a polynomial ignores linearly-appearing variables.

$$n = 2: \quad \underbrace{\{\underbrace{1, x, y, x^2, y^2, xy, x^3, y^3, x^2y, xy^2, x^3y, xy^3, x^2y^2, x^3y^2, x^2y^3, x^3y^3\}}_{\mathcal{S}_3 \wedge^0([0,1]^2) \text{ (dim=12)}}$$

$$n = 3: \quad \underbrace{\{\underbrace{1, x, y, x^2, y^2, xy, x^3, y^3, x^2y, xy^2, x^3y, xy^3, x^2y^2, x^3y^2, x^2y^3, x^3y^3\}}_{\mathcal{S}_3 \wedge^0([0,1]^3) \text{ (dim=64)}}$$

$$n = 3: \quad \underbrace{\{\underbrace{1, \dots, xyz, \ x^3y, x^3z, y^3z, \dots, x^3yz, xyz^3, xyz^3, x^3y^2, \dots, x^3y^3z^3\}}_{\mathcal{S}_3 \wedge^0([0,1]^3) \text{ (dim=32)}}$$

 $Q_r\Lambda^k$  and  $S_r\Lambda^k$  and have the **same** key mathematical properties needed for FEEC (degree, inclusion, trace, subcomplex, unisolvence, commuting projections) but for fixed  $k \geq 0$ ,  $r, n \geq 2$  the serendipity spaces have **fewer** degrees of freedom

#### Outline

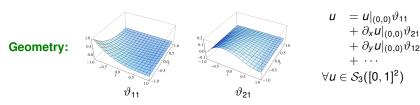
The Cubic Case: Hermite Functions, Serendipity Spaces

Geometric Decompsitions of Cubic Serendipity Spaces

Applications and Future Directions

**Theorem [G, 2012]:** A Hermite-like geometric decomposition of  $S_3([0,1]^2)$  exists.

**Approximation:** 
$$x^r y^s = \sum_{\ell m} \varepsilon_{r,i} \varepsilon_{s,j} \vartheta_{\ell m}$$
, for superlinear degree  $(x^r y^s) \leq 3$ 



#### **Proof Overview:**

1 Fix index sets and basis orderings based on domain points:

$$V = \text{vertices (11, 14, ...)}$$
 $E = \text{edges (12, 13, ...)}$ 
 $D = \text{interior (22, 23, ...)}$ 
 $V = \text{vertices (11, 14, ...)}$ 
 $V = \text{vertices (11, 14, ...)}$ 
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- **2** Define a 12 × 16 matrix  $\mathbb{H}$  with entries  $h_{ij}^{\ell m}$  so that  $\ell m \in V \cup E$ ,  $ij \in V \cup E \cup D$ .
- **3** Define the serendipity basis functions  $\theta_{\ell m}$  via

$$\boxed{[\vartheta_{\ell m}] := \mathbb{H}[\psi_{ij}]}$$

and show that the approximation and geometry properties hold.

#### **Proof Details:**

**2** Define a 12 × 16 matrix  $\mathbb{H}$  with entries  $h_{ij}^{\ell m}$  so that  $\ell m \in V \cup E$ ,  $ij \in V \cup E \cup D$ .

**3** Define  $[\vartheta_{\ell m}] := \mathbb{H}[\psi_{ij}]$ . The geometry property holds since for  $\ell m \in V \cup E$ ,

$$\vartheta_{\ell m} = \underbrace{\psi_{\ell m}}_{\text{bicubic Hermite}} + \underbrace{\sum_{ij \in D} h_{ij}^{\ell m} \psi_{ij}}_{\text{zero on boundary}} \implies \vartheta_{\ell m} \equiv \psi_{\ell m} \text{ on edges}$$

#### **Proof Details:**

To prove that the approximation property holds, observe:

$$[\vartheta_{\ell m}] := \mathbb{H}[\psi_{ij}] \quad ext{implies} \quad \sum_{ij} \mathit{h}^{\ell m}_{ij} \psi_{ij} = \vartheta_{\ell m}$$

For all (r, s) pairs such that  $sldeg(x^ry^s) \le 3$ , the matrix entries in column ij satisfy

$$\varepsilon_{r,i}\varepsilon_{s,j} = \sum_{\ell m \in V \cup E} \varepsilon_{r,\ell}\varepsilon_{s,m} h_{ij}^{\ell m}$$

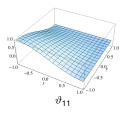
Substitute these into the Hermite 2D approximation property:

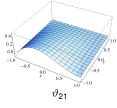
$$x^{r}y^{s} = \sum_{ij \in V \cup E \cup D} \varepsilon_{r,i}\varepsilon_{s,i}\psi_{ij} = \sum_{ij} \sum_{\ell m} \varepsilon_{r,\ell}\varepsilon_{s,m}h_{ij}^{\ell m}\psi_{ij}$$
$$= \sum_{\ell m} \varepsilon_{r,\ell}\varepsilon_{s,m}\sum_{ij} h_{ij}^{\ell m}\psi_{ij} = \sum_{\ell m} \varepsilon_{r,\ell}\varepsilon_{s,m}\vartheta_{\ell m}$$

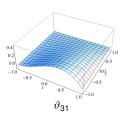
Hence  $[\vartheta_{\ell m}]$  is a basis for  $S_2([0,1]^2)$ , completing the geometric decomposition.

#### Hermite Style Serendipity Functions (2D)

$$[\vartheta_{\ell m}] = \begin{bmatrix} \vartheta_{11} \\ \vartheta_{14} \\ \vartheta_{41} \\ \vartheta_{44} \\ \vartheta_{12} \\ \vartheta_{43} \\ \vartheta_{21} \\ \vartheta_{21} \\ \vartheta_{34} \end{bmatrix} = \begin{bmatrix} -(x-1)(y-1)(-2+x+x^2+y+y^2) \\ (x-1)(y+1)(-2-x+x^2-y+y^2) \\ (x+1)(y-1)(-2-x+x^2+y+y^2) \\ -(x+1)(y-1)(y-1)^2(y+1) \\ (x-1)(y-1)^2(y+1) \\ (x+1)(y-1)^2(y+1) \\ -(x+1)(y-1)(y+1)^2 \\ (x+1)(y-1)^2(y+1) \\ (x-1)(x+1)(y-1) \\ (x-1)^2(x+1)(y-1) \\ (x-1)^2(x+1)(y+1) \\ -(x-1)(x+1)^2(y+1) \end{bmatrix} \cdot \frac{1}{8}$$





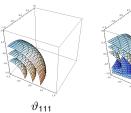


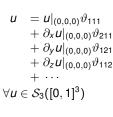
**Theorem [G, 2012]:** A Hermite-like geometric decomposition of  $S_3([0,1]^3)$  exists.

$$\underbrace{\left\{\begin{array}{c} x^ry^sz^t\\ \text{sldeg} \leq 3 \end{array}\right\}}_{\mathcal{S}_3([0,\,1]^3)} \quad \longleftrightarrow \quad \left\{\begin{array}{c} \vartheta_{\ell mn}\\ \text{(to be defined)} \end{array}\right\} \quad \longleftrightarrow \quad \underbrace{\left\{\begin{array}{c} \vartheta_{\ell mn}\\ \text{in} \\ \text{in} \end{array}\right\}}_{\text{in}} \underbrace{\left\{\begin{array}{c} 21 \\ \text{in} \end{array}\right\}}_{\text{in}} \underbrace{\left\{\begin{array}{c}$$

**Approximation:**  $x^r y^s z^t = \sum_{\ell mn} \varepsilon_{r,i} \varepsilon_{s,j} \varepsilon_{t,k} \vartheta_{\ell mn}$ , for superlinear degree  $(x^r y^s z^t) \le 3$ 







 $\vartheta_{112}$ 

#### **Proof Overview:**

1 Fix index sets and basis orderings based on domain points:

$$\begin{split} & \left[\vartheta_{\ell mn}\right] := \left[\begin{array}{ccc} \vartheta_{111}, \dots, \vartheta_{444}, & \vartheta_{112}, \dots, \vartheta_{443} \end{array}\right], \\ & \left[\psi_{ijk}\right] := \left[\begin{array}{ccc} \underline{\psi_{111}}, \dots, \underline{\psi_{444}}, & \underline{\psi_{112}}, \dots, \underline{\psi_{443}}, & \underline{\psi_{122}}, \dots, \underline{\psi_{433}}, & \underline{\psi_{222}}, \dots, \underline{\psi_{333}} \end{array}\right] \\ & \text{indices in } V & \text{indices in } E & \text{indices in } F & \text{indices in } M \end{split}$$

- **2** Define a 32 × 64 matrix  $\mathbb{W}$  with entries  $h_{ijk}^{\ell mn}$  (where  $\ell mn \in V \cup E$ )
- **3** Define the serendipity basis functions  $\vartheta_{\ell mn}$  via

$$[\vartheta_{\ell mn}] := \mathbf{W}[\psi_{ijk}]$$

and show that the approximation and geometry properties hold.

#### **Proof Details:**

**2** Define a 32  $\times$  64 matrix  $\overline{\mathbb{W}}$  with entries  $h_{ijk}^{\ell mn}$  so that  $\ell mn \in V \cup E$ 

$$\mathbb{W} := \left[ \begin{array}{c|c} \mathbb{I} & \text{specific full rank} \\ (32x32 \text{ identity matrix}) & 32x32 \text{ matrix} \\ \text{with entries -1, 0, or 1} \end{array} \right]$$

- **3** Define  $[\vartheta_{\ell mn}] := \mathbb{W}[\psi_{ijk}]$ .
  - $\longrightarrow$  Confirm directly that  $[\vartheta_{\ell mn}]$  restricts to  $[\vartheta_{\ell m}]$  on faces.
  - ---- Similar proof technique confirms geometry and approximation properties.

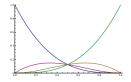
$$[\vartheta_{\ell mn}] = \begin{bmatrix} \vartheta_{111} \\ \vartheta_{114} \\ \vdots \\ \vartheta_{442} \\ \vartheta_{443} \end{bmatrix} = \begin{bmatrix} (x-1)(y-1)(z-1)(-2+x+x^2+y+y^2+z+z^2) \\ -(x-1)(y-1)(z+1)(-2+x+x^2+y+y^2-z+z^2) \\ \vdots \\ (x+1)(y+1)(z-1)^2(z+1) \\ -(x+1)(y+1)(z-1)(z+1)^2 \end{bmatrix} \cdot \frac{1}{16}$$

Complete list and more details in my paper:

GILLETTE Hermite and Bernstein Style Basis Functions for Cubic Serendipity Spaces on Squares and Cubes, arXiv:1208.5973, 2012

# Cubic Bernstein Serendipity Geom. Decomp: 2D, 3D

Cubic Bernstein Basis on [0, 1] 
$$\begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \end{bmatrix} := \begin{bmatrix} (1-x)^3 \\ (1-x)^2 x \\ (1-x)x^2 \\ x^3 \end{bmatrix}$$

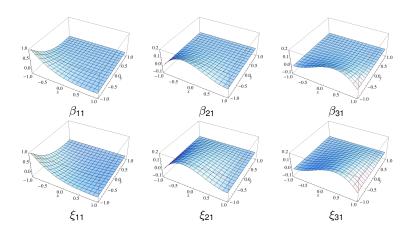


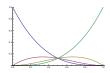
**Theorem [G, 2012]:** Bernstein-like geometric decompositions of  $S_3([0,1]^2)$  and  $S_3([0,1]^3)$  exist.

$$[\xi_{\ell m}] = \begin{bmatrix} \xi_{11} \\ \xi_{14} \\ \vdots \\ \xi_{24} \\ \xi_{34} \end{bmatrix} = \begin{bmatrix} (x-1)(y-1)(-2-2x+x^2-2y+y^2) \\ -(x-1)(y+1)(-2-2x+x^2+2y+y^2) \\ \vdots \\ (x-1)^2(x+1)(y+1) \\ -(x-1)(x+1)^2(y+1) \end{bmatrix} \cdot \frac{1}{16}$$

$$[\xi_{\ell mn}] = \begin{bmatrix} \xi_{111} \\ \xi_{114} \\ \vdots \\ \xi_{442} \\ \xi_{443} \end{bmatrix} = \begin{bmatrix} -(x-1)(y-1)(z-1)(-5-2x+x^2-2y+y^2-2z+z^2) \\ (x-1)(y-1)(z+1)(-5-2x+x^2-2y+y^2+2z+z^2) \\ \vdots \\ (x+1)(y+1)(z-1)^2(z+1) \\ -(x+1)(y+1)(z-1)(z+1)^2 \end{bmatrix} \cdot \frac{1}{32}$$

## Bernstein Style Serendipity Functions (2D)





Bicubic Bernstein functions (top) and Bernstein-style serendipity functions (bottom).

→ Note boundary agreement with Bernstein functions.

#### Outline

The Cubic Case: Hermite Functions, Serendipity Spaces

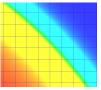
Geometric Decompsitions of Cubic Serendipity Spaces

Applications and Future Directions

## Application: Cardiac Electrophysiology

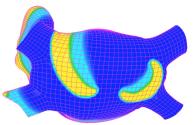
- → Initial implementation of cubic Hermite serendipity functions as part of the Continuity software package for cardiac electrophysiology models.
- → Used to solve the monodomain equations, similar to the heat equation:

$$\partial_t u - \Delta u = f$$



- → Cubic Hermite serendipity functions resolve the voltage wavefront equivalently (in eveball norm) to regular Hermite bicubics.

→ Initial results show a 4x computational speedup in 3D.



→ Current research into isogeometric analysis with applications to clinical modeling.

## Future Directions and Open Questions

- $\rightarrow$  Construction of Bernstein-like bases for  $S_r \Lambda^k([0,1]^n)$  for
  - $\rightarrow$  Higher order scalar cases (k = 0, r > 3, n = 2, 3)
  - $\rightarrow$  Higher form order cases (k > 0)
- → Generalized Bernstein-like construction process:

$$[\vartheta_\ell] := \mathbf{M} \ [\beta_i]$$
 [ basis for  $\mathcal{S}_r \Lambda^k([0,1]^n)$  ] :=  $\mathbf{M}$  [ Bernstein basis for  $\mathcal{Q}_r \Lambda^k([0,1]^n)$  ]

Define entries of M so that the appropriate approximation and geometry properties hold for  $[\vartheta_s]$ .

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Slides and pre-prints: http://ccom.ucsd.edu/~agillette

