

# What are Multivariate Splines?

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Splines are smooth piecewise polynomial functions.

## Disclaimers, Sources, Acknowledgments

- I am going to ignore spline alternatives, such as tensor product schemes, radial basis functions, parametric interpolation and approximation schemes, and alternative spline definitions.
- For a lack of time, I am also going to skip over the rich history of the subject.
- A comprehensive and current source of information on splines is the new monograph: Ming-Jun Lai and Larry L. Schumaker, *Spline Functions on Triangulations*, Cambridge University Press, 2007, ISBN 9780521875929.
- The software demonstrated in this talk can be used and downloaded at  

[www.math.utah.edu/~pa](http://www.math.utah.edu/~pa)
- Much of my spline research since 1983 has been in collaboration with Larry Schumaker.
- Recently I have had the pleasure to work on several projects with Tanya Sorokina.
- *Two Tetrahedral  $C^1$  Cubic Macro Elements*, submitted

# Outline of Talk

- Splines
- Dimension Problem for Univariate Splines
- Bivariate Splines
- The Bernstein Bézier Form
- Smoothness Conditions
- Dimension Problem for Bivariate Splines
- $S_3^1$
- Trivariate Splines
- Macro Elements
- Open Problems

# Univariate Splines

$$a = x_0 < x_1 < \dots < x_N = b$$

- partitions  $[a, b]$  into  $N$  subintervals

$$I_i = [x_{i-1}, x_i], \quad i = 1, \dots, N$$

- spline space:

$$S_d^r = \{s \in C^r[a, b] : s|_{I_i} \in P_d, \quad i = 1, \dots, N\}$$

where  $P_d$  is the space of univariate polynomials of degree  $d$ .

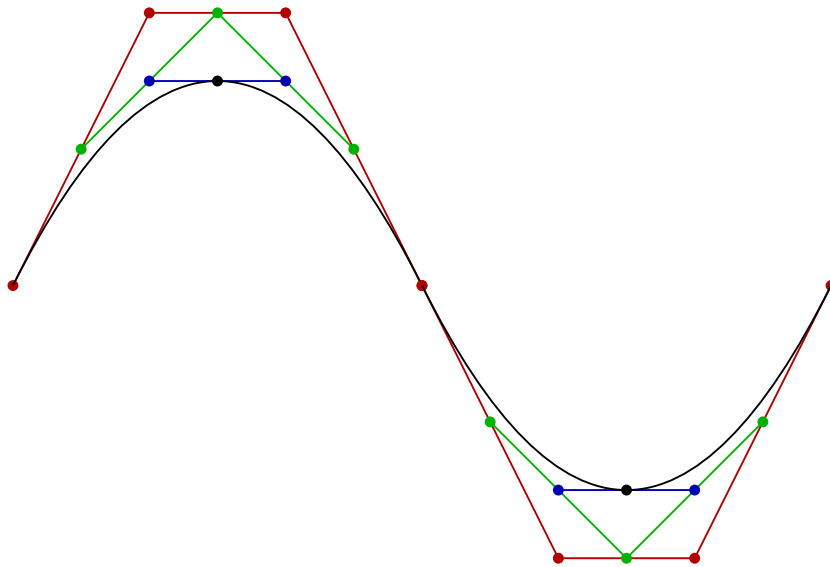
- Obviously:

$$\begin{aligned} \dim S_d^r &= (d + 1) + (N - 1)(d + 1 - (r + 1)) \\ &= N(d - r) + r + 1. \end{aligned}$$

What could be simpler?

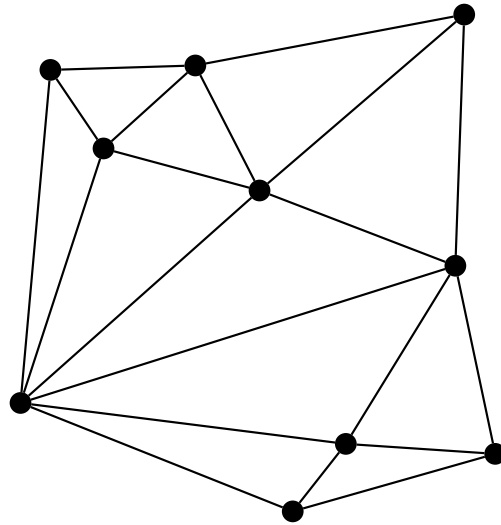
Examples:

- **piecewise linear:**  $\dim S_1^0 = N + 1$ ,
- **cubic splines:**  $\dim S_3^2 = N + 2$ ,
- **piecewise cubic Hermite:**  $\dim S_3^1 = 2(N + 1)$



**Figure 1.** A univariate spline in  $S_3^1$ .

- What about two variables?
- Need to partition a polygonal domain
- rectangular grid leads to tensor product schemes. They form a huge area. But they are essentially univariate splines done twice.
- If points are distributed irregularly then **triangles** work.



**Figure 2.** A triangulation.

Let  $\Omega$  be the union of the triangles  $T_i$ ,  $i = 1, \dots, N$ , and  $P^d$  the  $\binom{d+2}{2}$ -dimensional space of bivariate polynomials of degree  $d$ .

$$P_d = \left\{ p : p(x_1, x_2) = \sum_{i=0}^d \sum_{j=0}^{d-i} \alpha_{ij} x_1^i x_2^j \right\}.$$

$$S_d^r = \{ s \in C^r(\Omega) : s|_{T_i} \in P^d \}.$$

## The Bernstein-Bézier Form

Let  $\Delta$  be a triangle with vertices  $v_1$ ,  $v_2$ , and  $v_3$ .

For  $x \in \mathbb{R}^2$ , define its *barycentric coordinates*:

$$x = \sum_{i=1}^3 b_i v_i \quad \text{where} \quad \sum_{i=1}^3 b_i = 1.$$

Barycentric coordinates are linear functions of  $x$ .

Any polynomial  $p \in P_d$  can be written uniquely in its *Bernstein-Bézier form* as:

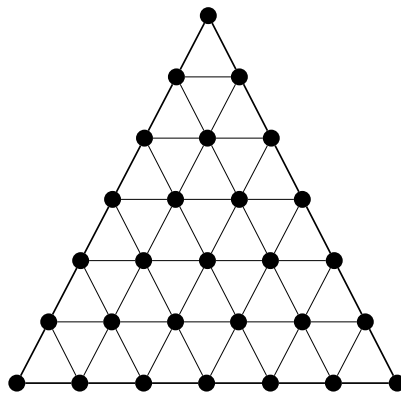
$$p(x) = \sum_{i+j+k=d} \frac{d!}{i!j!k!} c_{ijk} b_1^i b_2^j b_3^k.$$

where

$c_{ijk}$  : *Bézier ordinates*

$P_{ijk} = \frac{iv_1 + jv_2 + kv_3}{d} \in \mathbb{R}^2$  : *Domain Points*

$(P_{ijk}, c_{ijk}) \in \mathbb{R}^3$  : *Bézier control points*



**Figure 3.** Domain points,  $d = 6$ .

## Simple Geometry

- The control points at the vertices lie on the graph of the polynomial. This is because

$$p(v_1) = c_{d00}, \quad p(v_2) = c_{0d0}, \quad p(v_3) = c_{0d0}$$

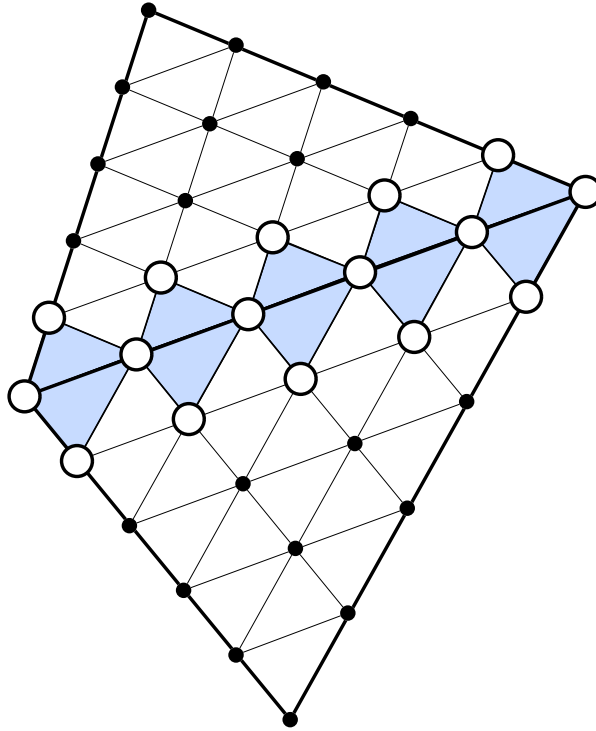
- The control points in the 1-disk, i.e.,

$$(P_{d00}, c_{d00}), \quad (P_{d-1,1,0}, c_{d-1,1,0}), \quad (P_{d-1,0,1}, c_{d-1,0,1})$$

lie in the tangent plane of  $p$  at  $v_1$ . Similarly for  $v_2$  and  $v_3$ .

- The control points along an edge determine the values of the polynomial along that edge.
- The control points along an edge, and in the first row parallel to the edge, determine the values of first derivatives of the polynomial along that edge.

## Smoothness Conditions



**Figure 4.** Smoothness conditions.

$C^0$ : Control points along edges coincide

$C^1$ : Quadrilaterals across edges are planar in  $\mathbb{R}^3$ .

$C^r$ : Evaluate subpolynomials.

## Idea of a particular proof:

Let  $D$  be a first order derivative operator.

$$p(x) = \sum_{i+j+k=d} \frac{d!}{i!j!k!} c_{ijk} b_1^i b_2^j b_3^k, \quad \text{as before}$$

On a neighboring triangle have

$$\tilde{p}(x) = \sum_{i+j+k=d} \frac{d!}{i!j!k!} \tilde{c}_{ijk} b_1^i b_2^j b_4^k.$$

For continuity,  $c_{ij0} = \tilde{c}_{ij0}$ ,  $i + j = d$ .

$$\begin{aligned} Dp(x) &= \sum_{i+j+k=d} \frac{d!}{i!j!k!} c_{ijk} \\ &\quad \left( i b_1^{i-1} D b_1 b_2^j b_3^k + b_1^i j b_2^{j-1} D b_2 b_3^k + b_1^i b_2^j k b_3^{k-1} D b_3 \right) \\ &= \sum_{i+j+k=d-1} \hat{c}_{ijk} b_1^i b_2^j b_3^k \end{aligned}$$

where

$$\hat{c}_{ijk} = d(c_{i+1,j,k}Db_1 + c_{i,j+1,k}Db_2 + c_{i,j,k+1}Db_3)$$

Differentiating on both triangles, restricting the derivative to the common edge, and equating coefficients gives the condition

$$d(a_{i+1j0}c_{i+1,j0} + a_{ij+10}c_{ij+10} + a_{ij11}c_{ij1} + \tilde{a}_{ij1}\tilde{c}_{ij1}) = 0 \quad (*)$$

for  $i + j = d - 1$ .

The coefficients in (\*) are independent of the particular quadrilateral and the polynomial degree.

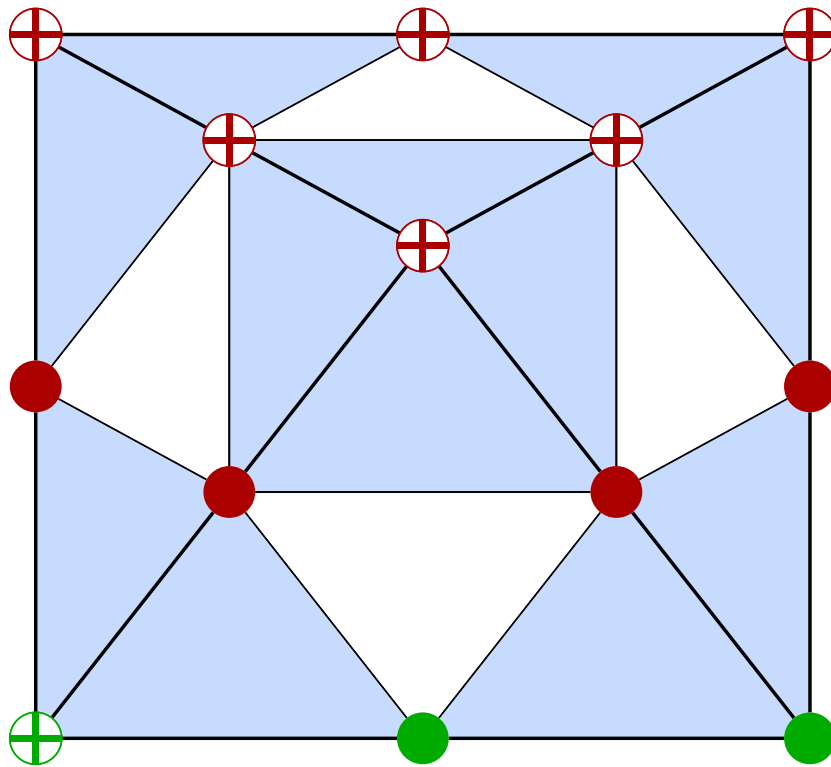
In particular, we obtain the same condition for the case  $d = 1$ . In that case, the condition means that a piecewise linear function is linear.

This means that in the large quadrilateral, formed by both triangles, for  $d = 1$ , the four control points lie in the same plane.

Since the small quadrilaterals are similar to the large one, the condition (\*) has the same geometric meaning.

Note the interplay of Geometry and Algebra!

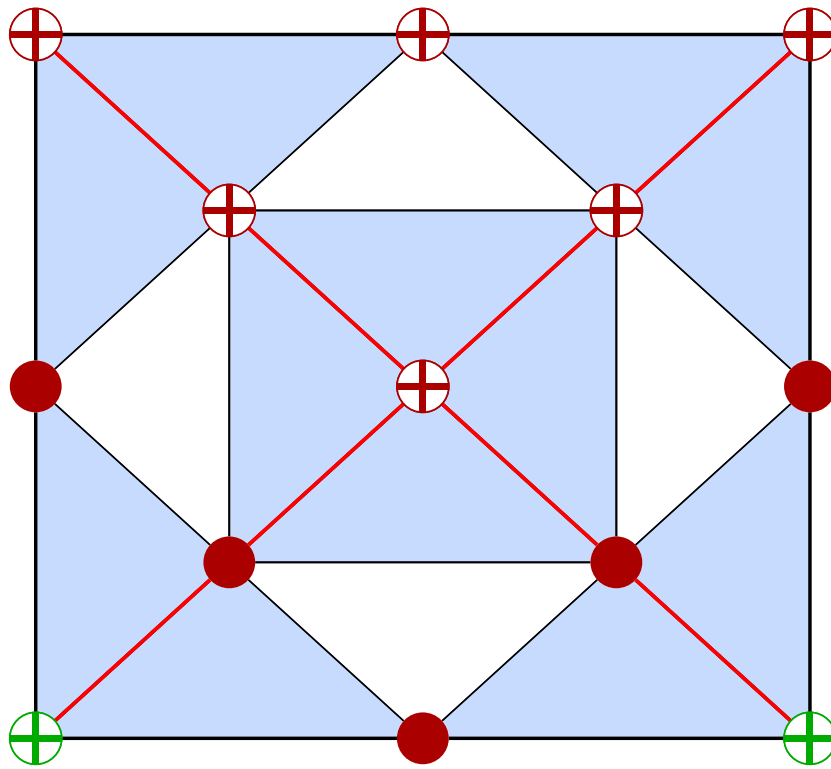
**Major Difficulty:** The dimension of  $S_d^r$  depends not just on the **combinatorics** of a triangulation, but also on its **geometry**.



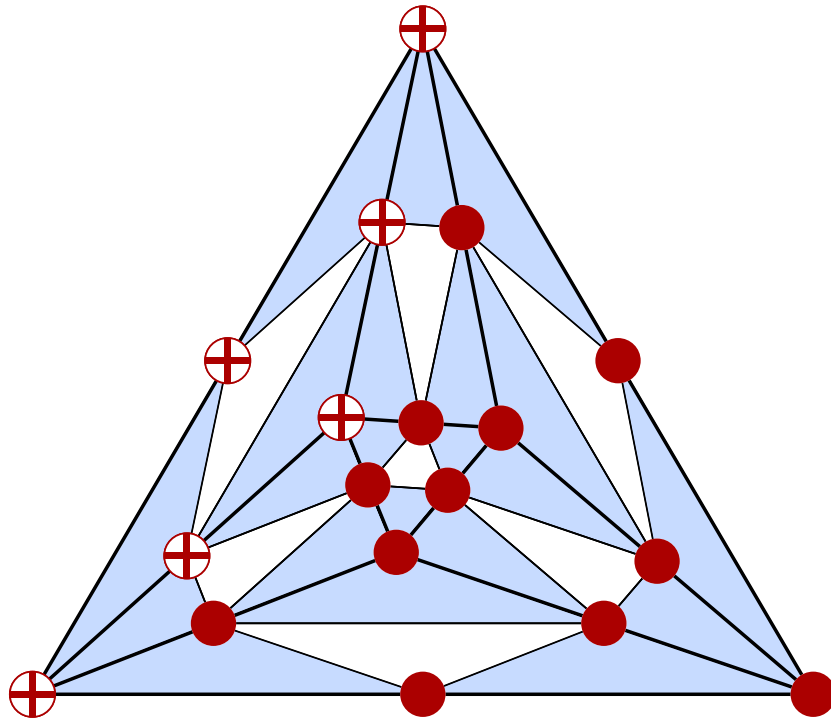
**Figure 5.**  $\dim S_2^1 = 7$  on a generic 4-star.

download software: [www.math.utah.edu/~pa/MDS](http://www.math.utah.edu/~pa/MDS)

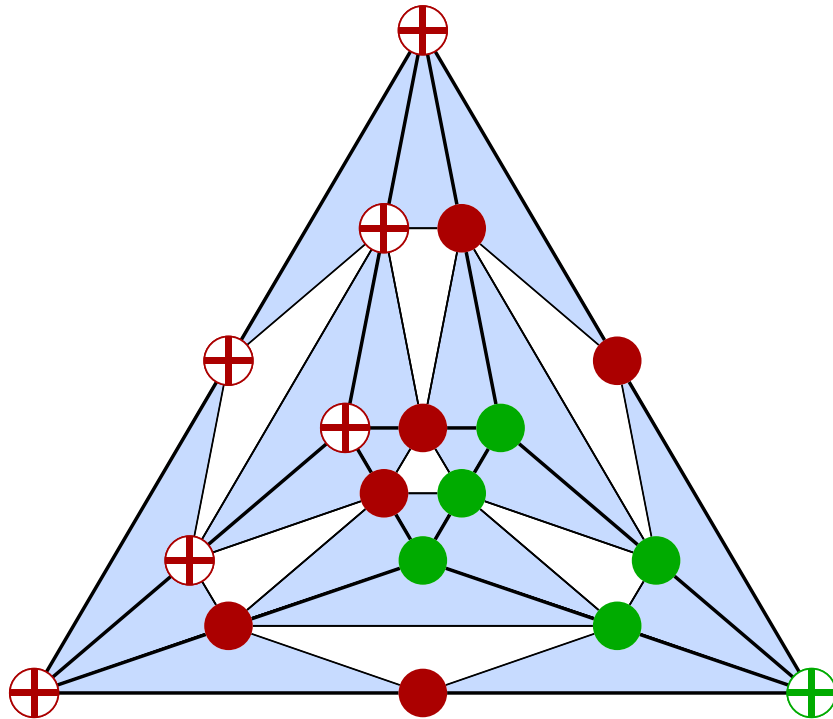
A **singular vertex** is an interior vertex of degree 4 whose edges form two intersecting lines.



**Figure 6.**  $\dim S_2^1 = 8$  on a singular vertex.

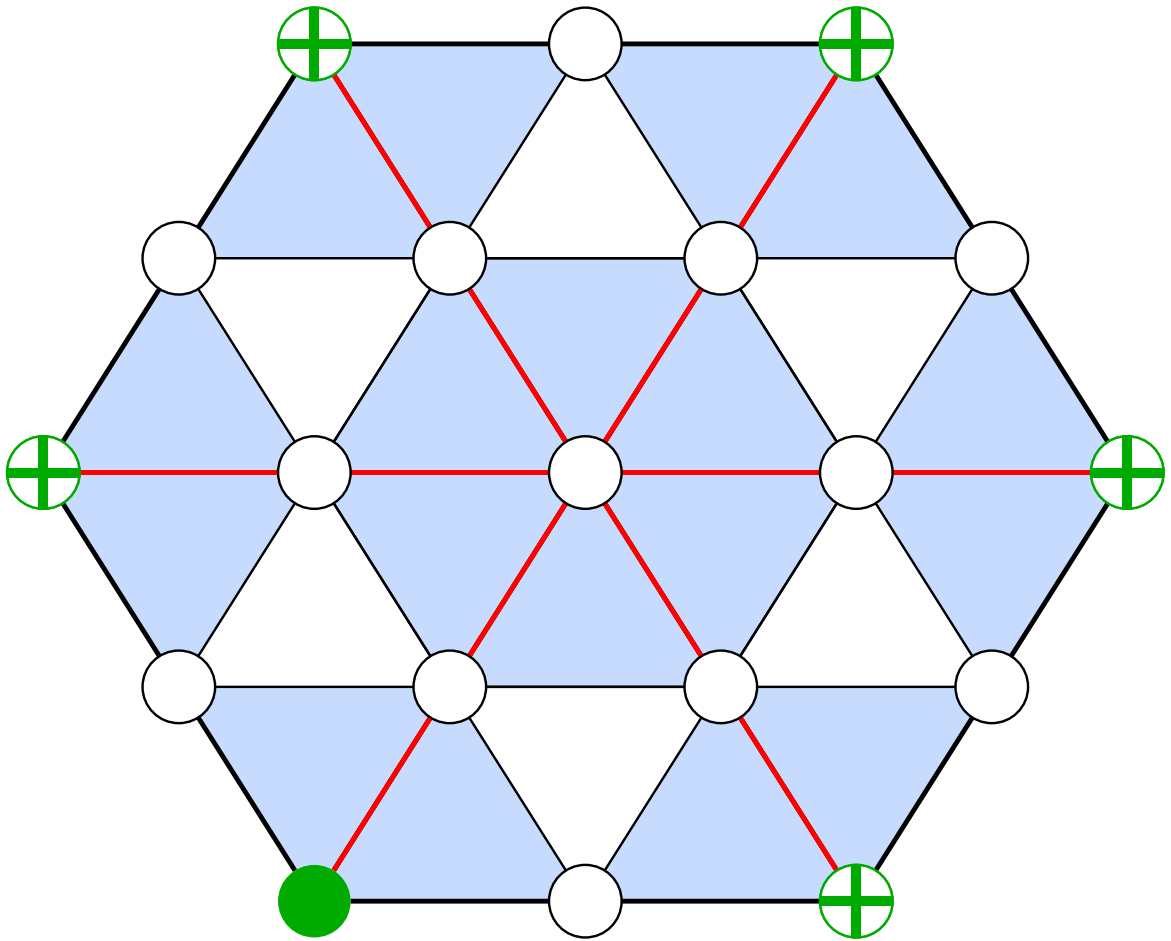


**Figure 7.**  $\dim S_2^1 = 6$  on the generic Morgan-Scott split.



**Figure 8.**  $\dim S_2^1 = 7$  on the symmetric Morgan-Scott split.

The geometry may also affect the Solvability of certain interpolation problems.



**Figure 9.**  $S_2^1$  interpolation on a regular hexagon.

What's happening?

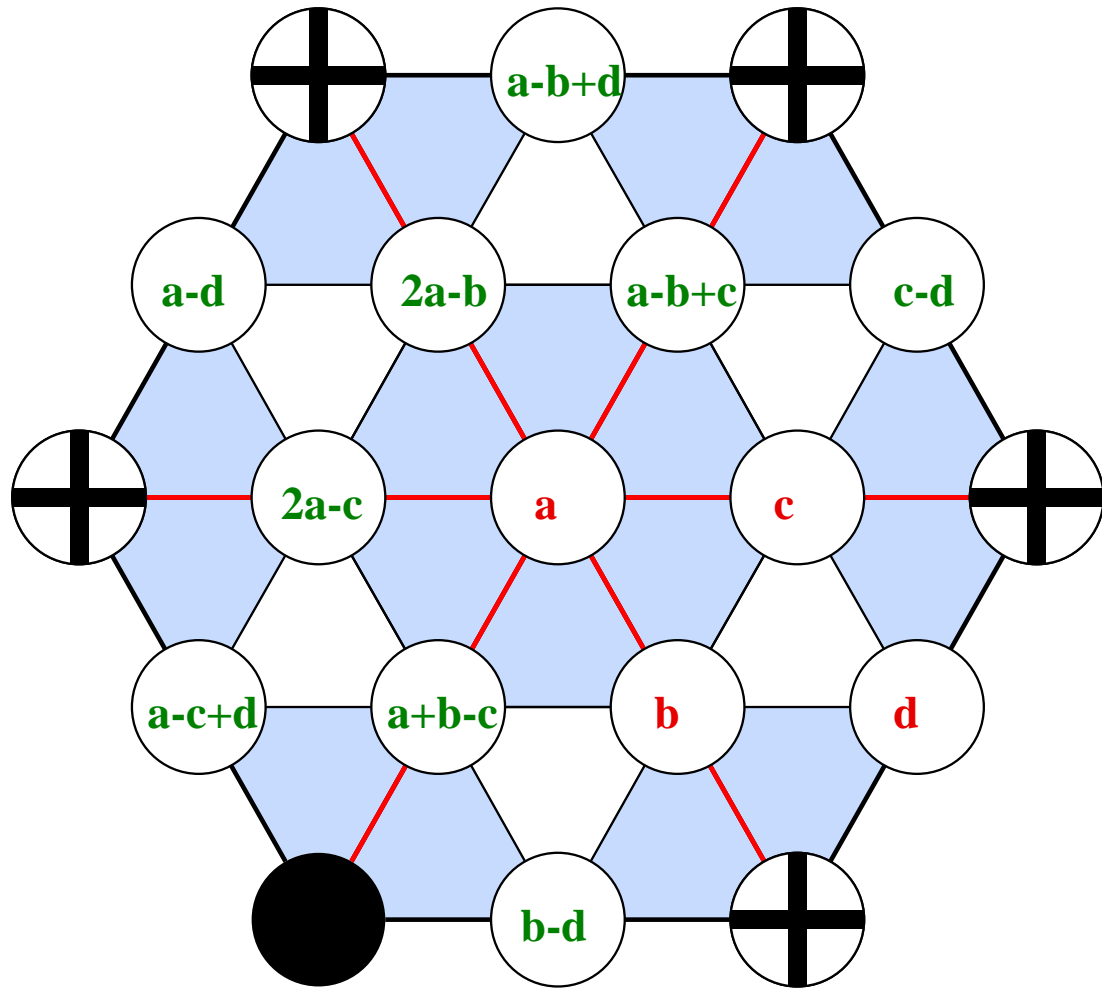
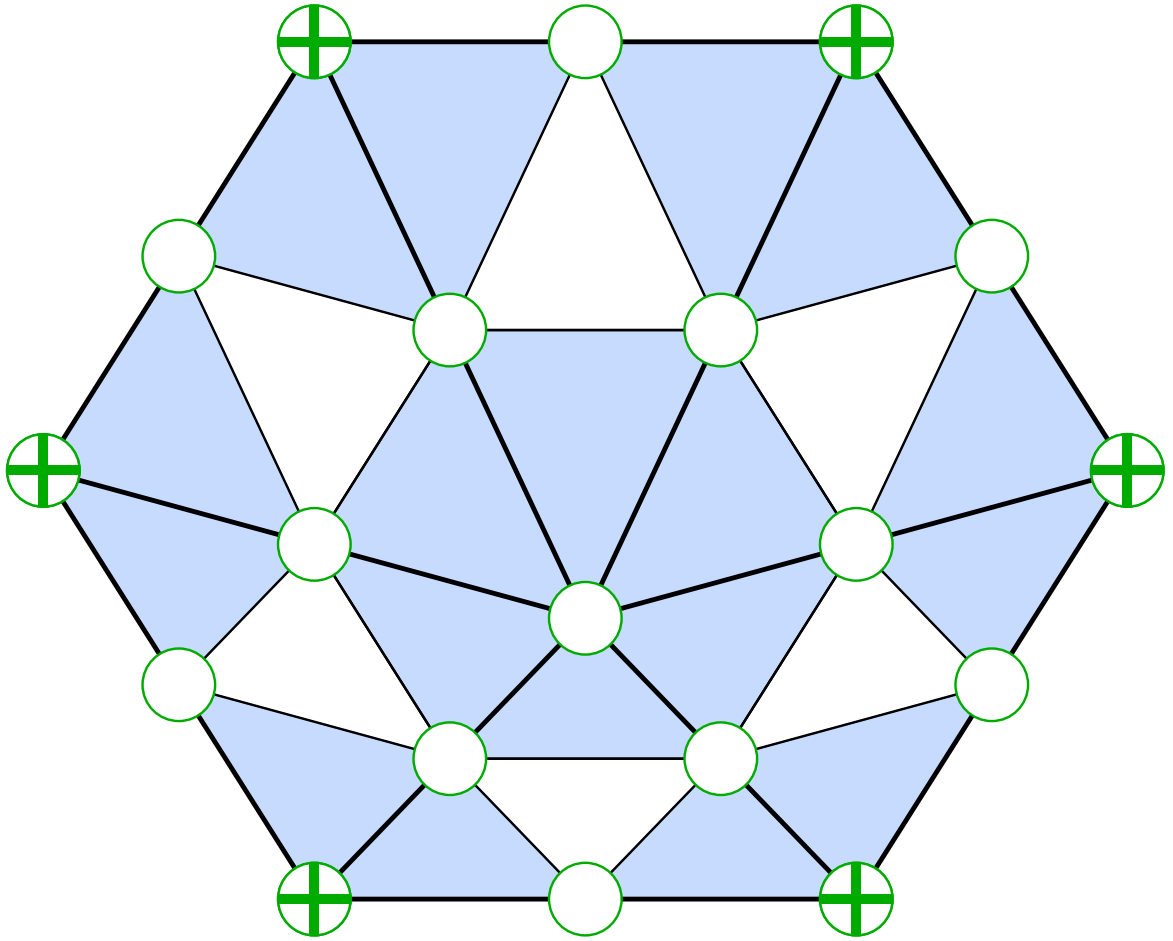
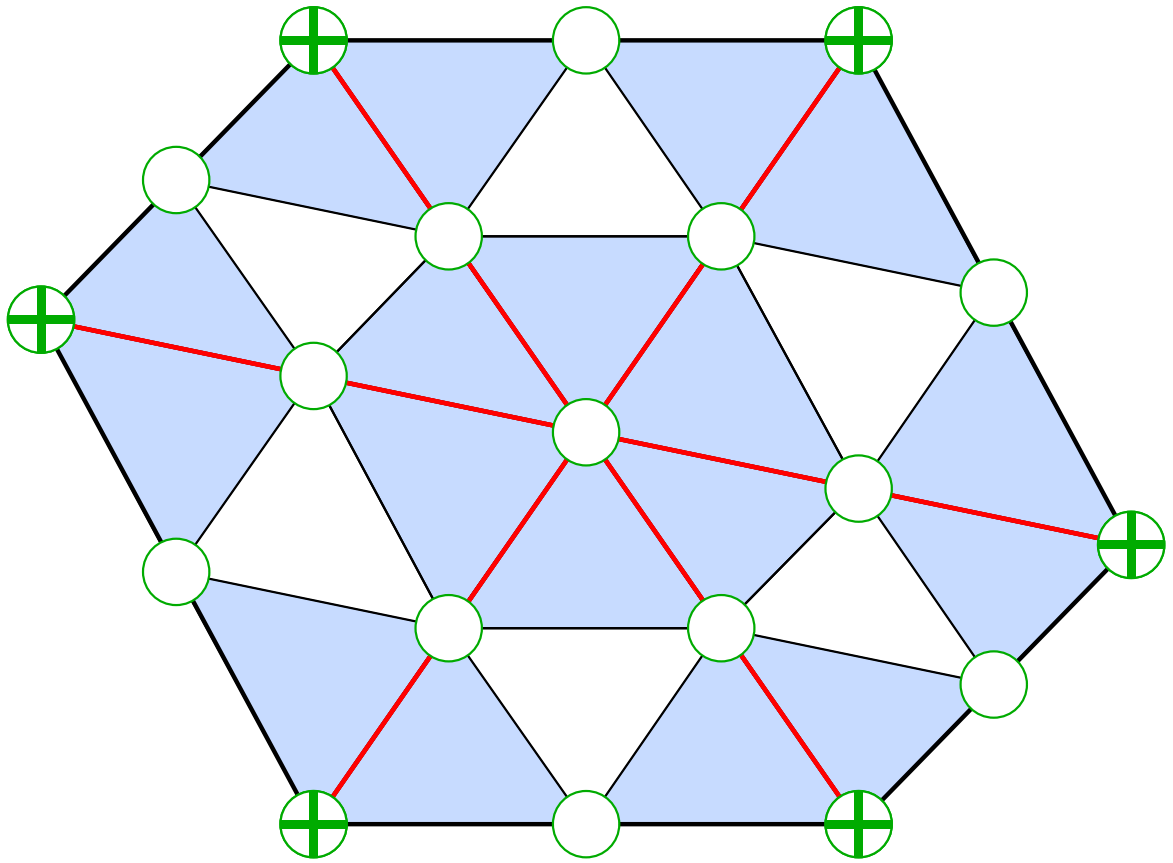


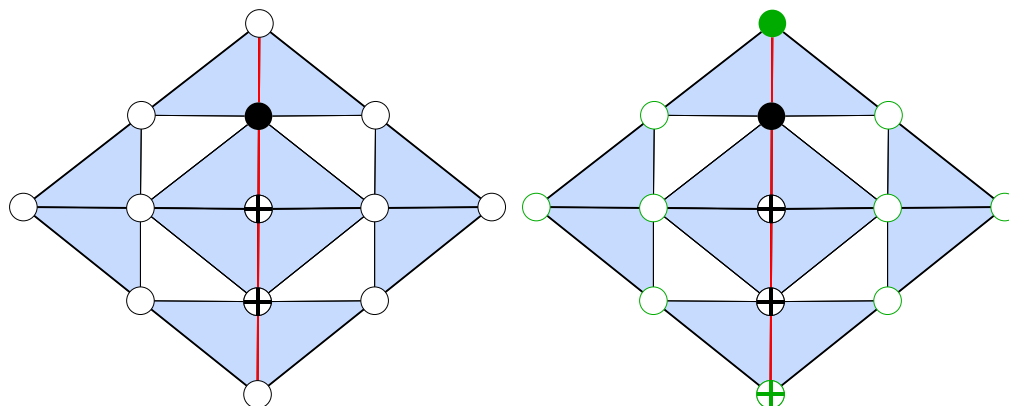
Figure 10. Reason for Failure.



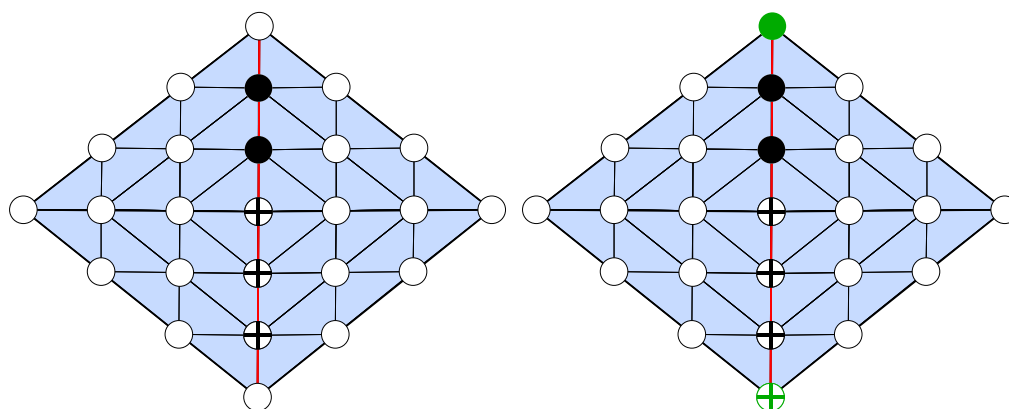
**Figure 11.**  $S_2^1$  interpolation on a generic hexagon.



**Figure 12.**  $S_2^1$  interpolation on a defective hexagon.



**Figure 13.**  $S_2^1$ : A peculiar effect.



**Figure 14.**  $S_3^2$ : The same effect.

Smoothness conditions can have peculiar long range effects. The above examples are due to Tanya Sorokina.

This does not work on singular or generic degree 4 vertices!

## Generic Dimension

Every spline space  $S$  has a generic dimension. If the dimension of  $S$  does not equal its generic value then there is an arbitrarily small perturbation of the location of the vertices such that the dimension of  $S$  does equal the generic value. Any other dimensions can only be larger than the generic dimension.

**Proof sketch:** Let  $S = S_d^r$  be the subspace of  $S_d^0$  with a coefficient vector  $c$  that satisfies the smoothness conditions

$$Ac = 0.$$

The entries of  $A$  are rational functions of the location of the vertices of the underlying triangulation.

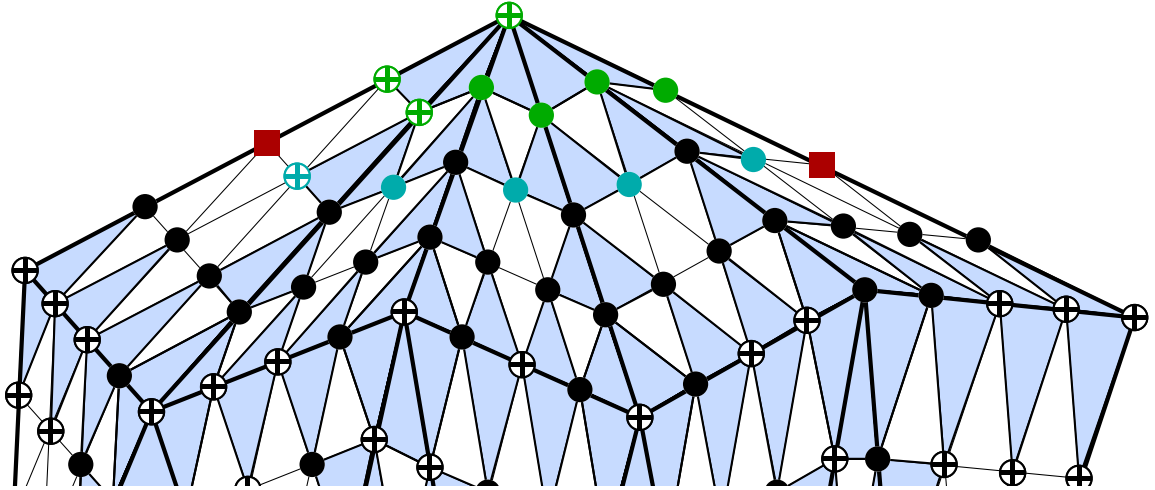
Let  $D$  be the minimum dimension of  $S$ . Then  $D = \dim S_d^0 - R$ , with  $R = \text{rank} A$ , and where (without loss of generality)

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix},$$

with  $A_{11}$  being a non-singular  $R \times R$  matrix.

The expression  $\det A_{11}$  is a non-zero rational function of the locations of the vertices, and can vanish only on a set of measure zero.

The generic dimension of  $S_4^1$  is  $6V - 3$ . Proof by Induction



**Figure 15.**  $\dim S_4^1 = 6V - 3$ .

Add the star of a boundary vertex to the growing triangulation  $T$ .

Black points are determined by the spline on  $T$ .

The green, red, and cyan points are newly imposed.

3 green, 2 red, and 1 cyan point are newly assigned.

Also shows that one can interpolate generically to function and gradient at vertices.

- Things get easier as the polynomial degree increases.
- Exact dimension known if  $r = 1$  and  $d \geq 4$ , or  $r > 1$  and  $d \geq 3r + 2$ .
- Generic dimension known for  $S_2^1$  and  $S_3^1$ , and for  $d = 3r + 1$  when  $r > 1$ .
- Dimensions and many other facts known on many types of special triangulations.
- Most famous outstanding problem:

$V_B$ : number of boundary vertices

$V_I$ : number of interior vertices

$\sigma$ : number of singular vertices

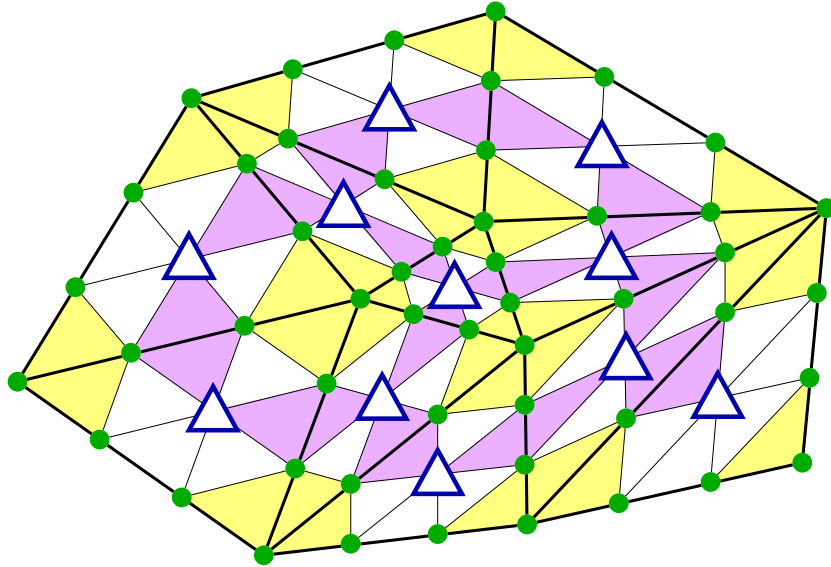
Problem first mentioned in a conjecture by Gil Strang in the early seventies. What's the dimension of  $S_3^1$ ?

$$\boxed{\dim S_3^1 \geq 3V_B + 2V_I + 1 + \sigma}$$

Does equality hold? Conjecture: yes.

If you solve this problem I want to know about it!

Why is this so hard? You can't localize things!



**Figure 16.**  $\dim S_3^1$ .

$$\dim S_3^1 \geq 3V + N - E_I + \sigma = 3V_B + 2V_I + 1 + \sigma.$$

What works for large  $d$ ? The Morgan Scott idea.

Use vertex globs (green), edge globs (blue), and inactive points (red). Requires  $d > 4r$ . Smoothness conditions decouple.

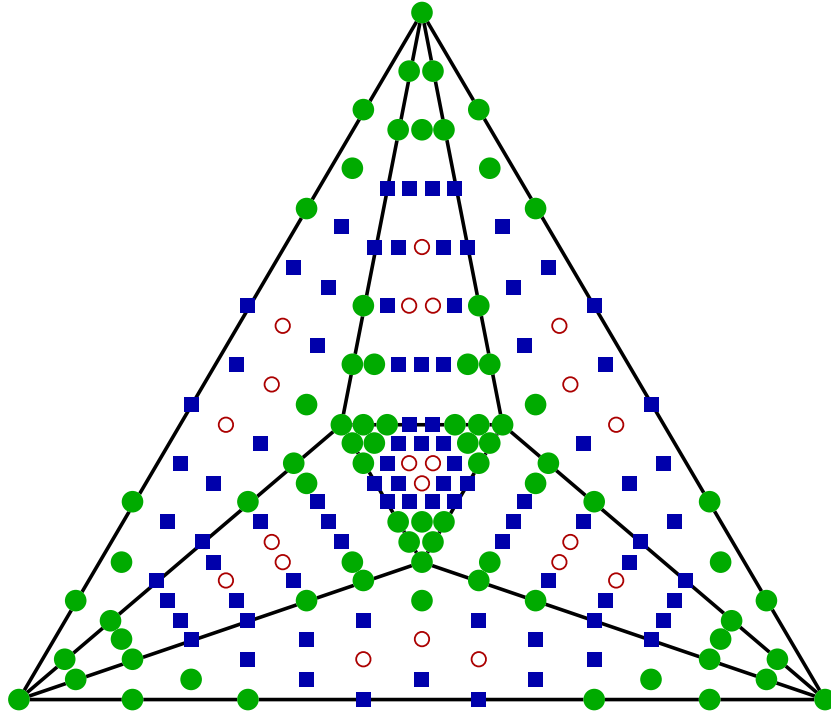
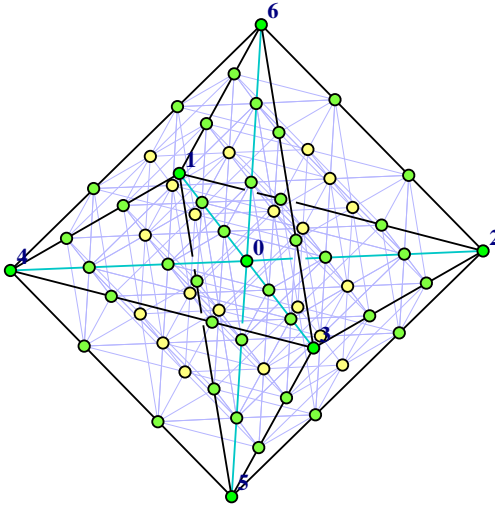


Figure 17. The Morgan-Scott idea,  $r = 1$ ,  $d = 6$ .

## Trivariate Splines

Consider a triangulation of a polyhedral domain by tetrahedra.  
Similar definition as before.

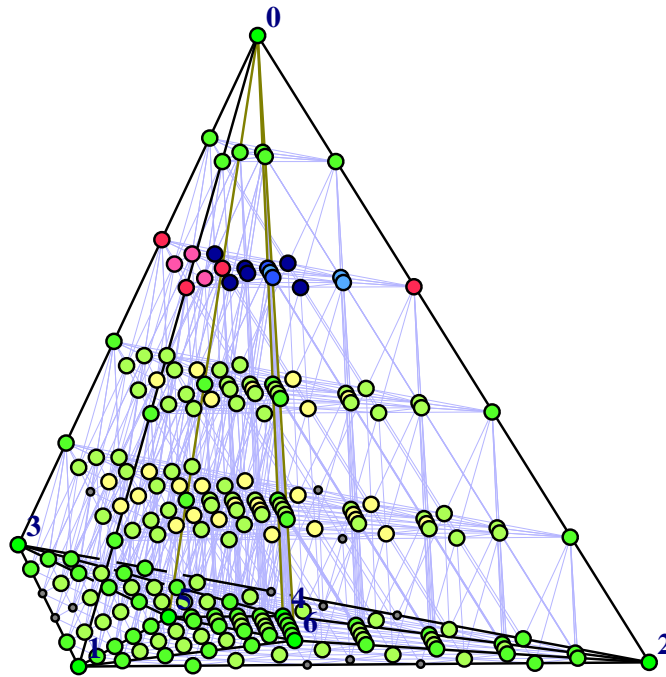


**Figure 18.**  $S_3^1$  on an octahedron.

download software: [www.math.utah.edu/~pa/3DMDS](http://www.math.utah.edu/~pa/3DMDS)

- Similar problems. Dimension depends on the geometry.
- Morgan-Scott idea requires  $d > 8r$ .
- Generic dimension is known for  $r = 1$  and  $d \geq 8$ .
- One new problem:
  - Knowing the dimension of the trivariate space  $S_d^r$  for sufficiently large  $d$  means we know the dimension of the bivariate space **for all**  $d$ .
  - To see this construct a three dimensional triangulation by starting with a planar triangulation  $T$  and then connecting every vertex of  $T$  to a new vertex in  $\mathbb{R}^3$  outside of the plane containing  $T$ .

# Coning



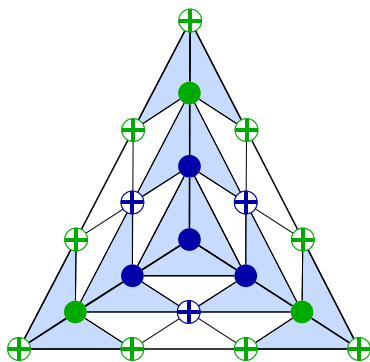
**Figure 19.** Coning.

Lift planar Triangulation to  $R^3$ . Smoothness conditions decouple.

## Macro Elements

Most applications of multivariate splines are based on so called “macro elements” (when approximating data) or “finite elements” (when solving differential equations).

- The interpolant is determined on each simplex by data on that simplex.
- Simplices may be subdivided.
- The overall spline space is a sub or superspace of the full space  $S_d^r$
- Many macro schemes are known in 2, 3, or  $n$  variables.



**Figure 20.** The planar Clough-Tocher element.

## Recent Tetrahedral Macro Elements

Schumaker, Sorokina, Worsey:  $r = 1$ ,  $d = 2$ , 504 microtetrahedra, no constraints, JAT, to appear.

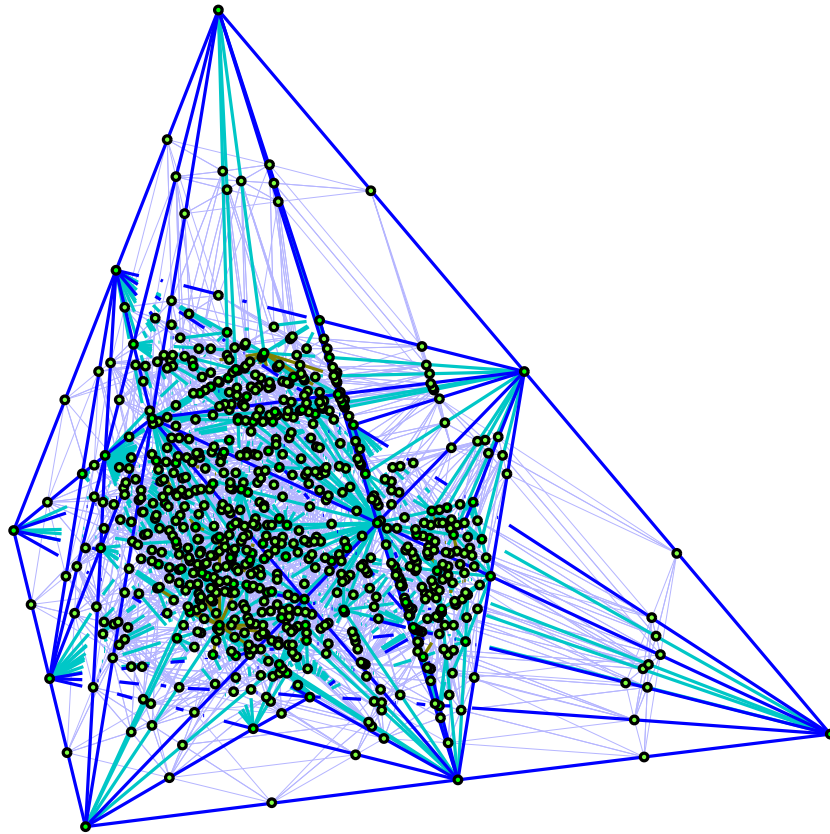
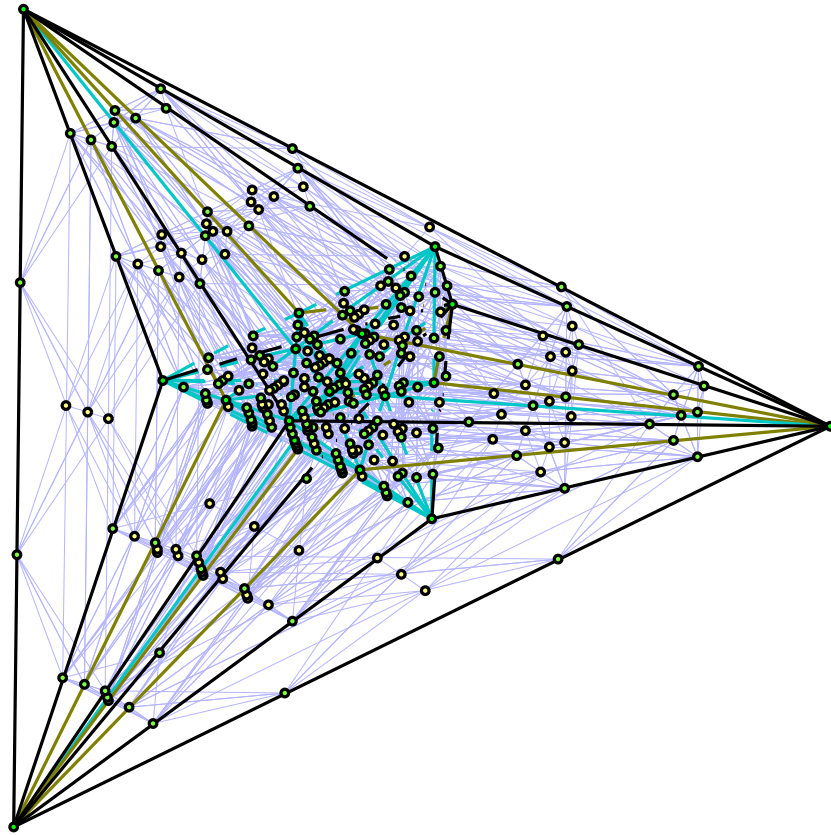


Figure 21. A quadratic macro element.

Alfeld and Sorokina.  $r = 1$ ,  $d = 3$ , 60 microtetrahedra, no constraints, submitted for publication.



**Figure 22.** A cubic macro element.

## Some Open Questions

### 2 Variables:

- What is the dimension of  $S_3^1$ ?
- Can one interpolate to vertex data with  $S_3^1$ ?
- What is the generic dimension of  $S_d^r$ ,  $r > 1$ ,  $r < d \leq 3r$ .
- What is the dimension of  $S_2^1$ ? (Much harder than  $S_3^1$ )
- Can one interpolate with  $S_4^1$  to function and gradient values at the vertices?

### $k > 2$ variables

- More or less tight bounds on the dimensions
- For given  $r$ , what are minimum values of  $d$ , as a function of  $k$ , for  $S_d^r$ 
  - to have a local basis
  - to be able to interpolate to function values at vertices
  - to be large enough to interpolate to function values at vertices.