

# Geometric Modeling

Summer Semester 2010

## Spline Surfaces

Tensor Product Surfaces • Total Degree Surfaces

# Overview...

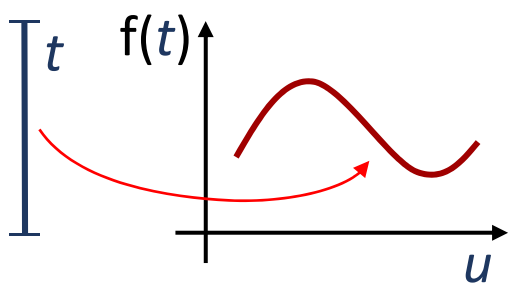
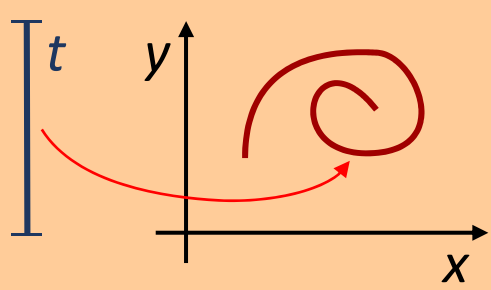
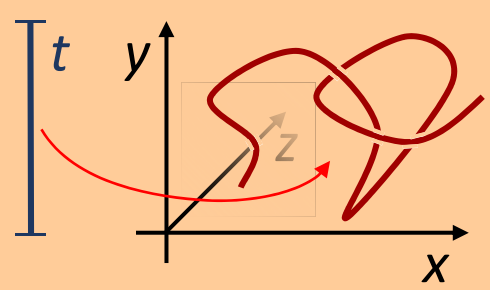
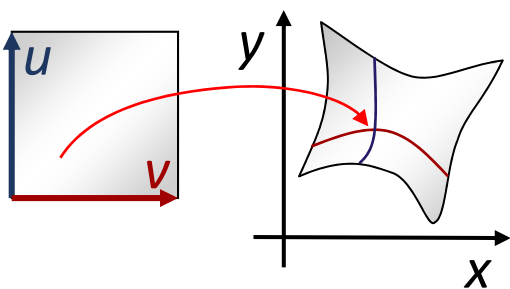
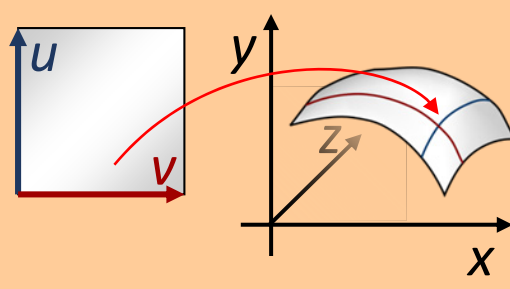
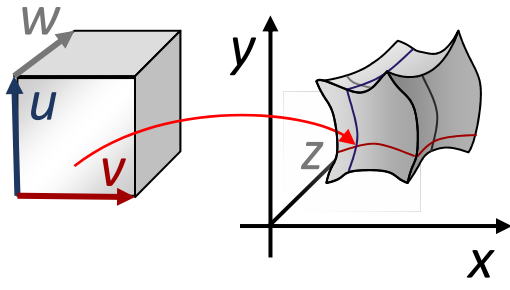
## Topics:

- Polynomial Spline Curves
- Blossoming and Polars
- Rational Spline Curves
- Spline Surfaces
  - Introduction
  - Tensor Product Surfaces
  - Total Degree Surfaces



# **Introduction:**

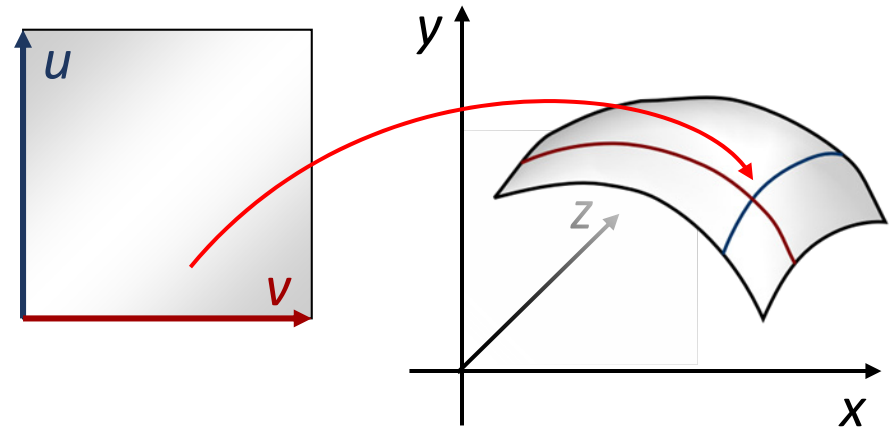
## Spline Surfaces

	output: 1D	output: 2D	output: 3D
input: 1D	 <p>function graph</p>	 <p>plane curve</p>	 <p>space curve</p>
input: 2D		 <p>plane warp</p>	 <p>surface</p>
input: 3D			 <p>space warp</p>

# Spline Surfaces

## Parametric spline surfaces:

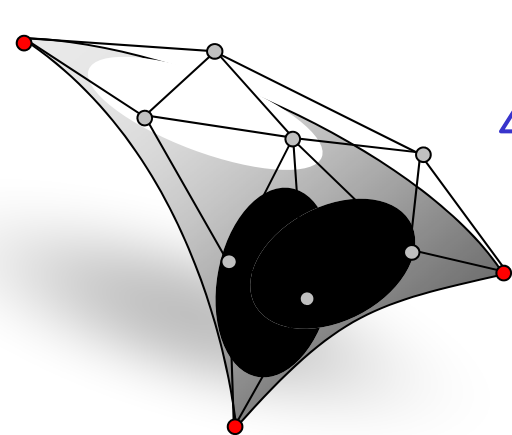
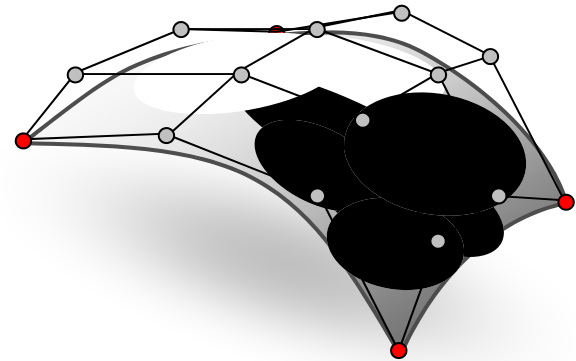
- Two parameter coordinates  $(u,v)$
- Piecewise bivariate polynomials (rational surfaces → homogeneous coords)
- Assemble multiple pieces to form a surface with continuity
- Each piece is called *spline patch*



# Spline Surfaces

## Two different approaches

- Tensor product surfaces
  - Simple construction
  - Everything carries over from curve case
  - Quad patches
  - Degree anisotropy
- Total degree surfaces
  - Not as straightforward (blossoming will help)
  - Isotropic degree
  - Triangle patches



# Tensor Product Surfaces

# Tensor Product Surfaces

## Simple Idea:

- Given a basis for a one dimensional function space on the interval  $t \in [t_0, t_1] \rightarrow \mathbb{R}^d$ :

$$\mathbf{B}^{(\text{curv})} := \{b_1(t), \dots, b_n(t)\}$$

- Build a new basis with two parameters by taking all possible products:

$$\mathbf{B}^{(\text{surf})} := \{b_1(u)b_1(v), b_1(u)b_2(v), \dots, b_n(u)b_n(v)\}$$



# Tensor Product Surfaces

## Tensor product basis

	$b_1(u)$	$b_2(u)$	$b_3(u)$	$b_4(u)$
$b_1(v)$	$b_1(v)b_1(u)$	$b_1(v)b_2(u)$	$b_1(v)b_3(u)$	$b_1(v)b_4(u)$
$b_2(v)$	$b_2(v)b_1(u)$	$b_2(v)b_2(u)$	$b_2(v)b_3(u)$	$b_2(v)b_4(u)$
$b_3(v)$	$b_3(v)b_1(u)$	$b_3(v)b_2(u)$	$b_3(v)b_3(u)$	$b_3(v)b_4(u)$
$b_4(v)$	$b_4(v)b_1(u)$	$b_4(v)b_2(u)$	$b_4(v)b_3(u)$	$b_4(v)b_4(u)$

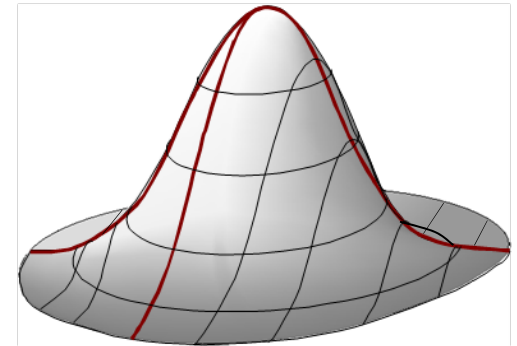
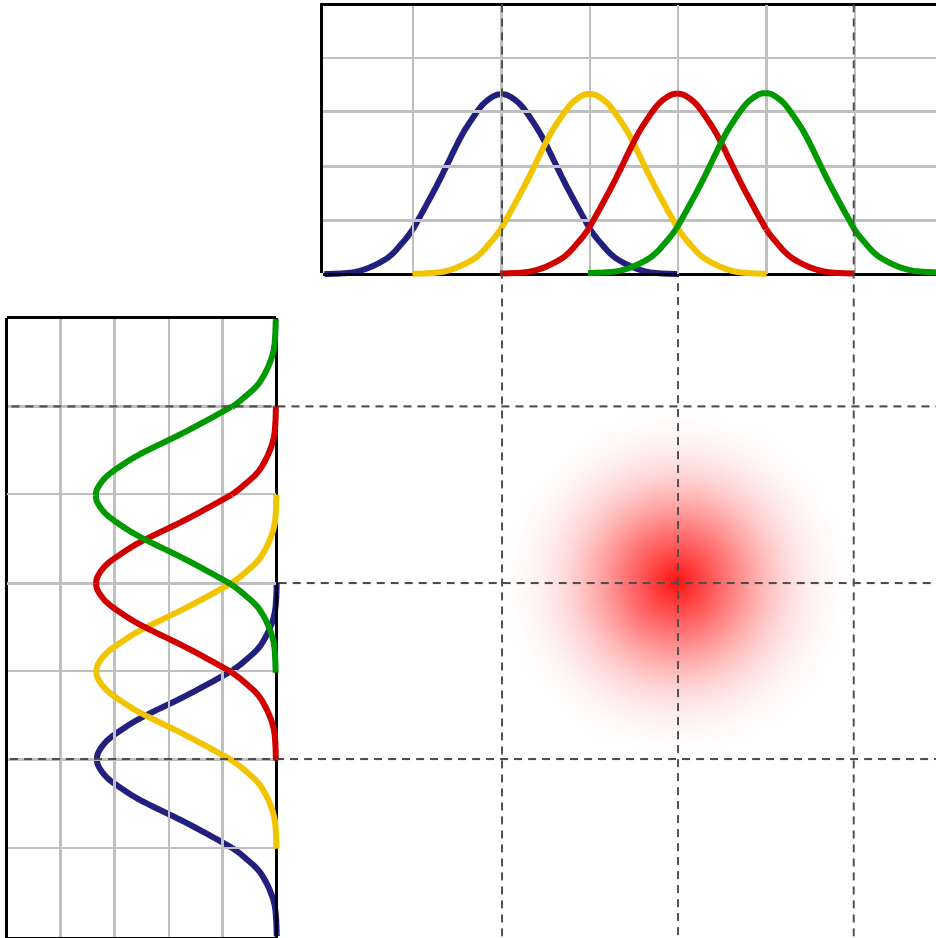
# Monomial Example

## Tensor product basis of cubic monomials

	1	$u$	$u^2$	$u^3$
1	1	$u$	$u^2$	$u^3$
$v$	$v$	$uv$	$u^2v$	$u^3v$
$v^2$	$v^2$	$uv^2$	$u^2v^2$	$u^3v^3$
$v^3$	$v^3$	$uv^3$	$u^2v^3$	$u^3v^3$

**Degree Anisotropy:**  $b_{33}(t, t)$  is of degree 6 in  $t$

# Example

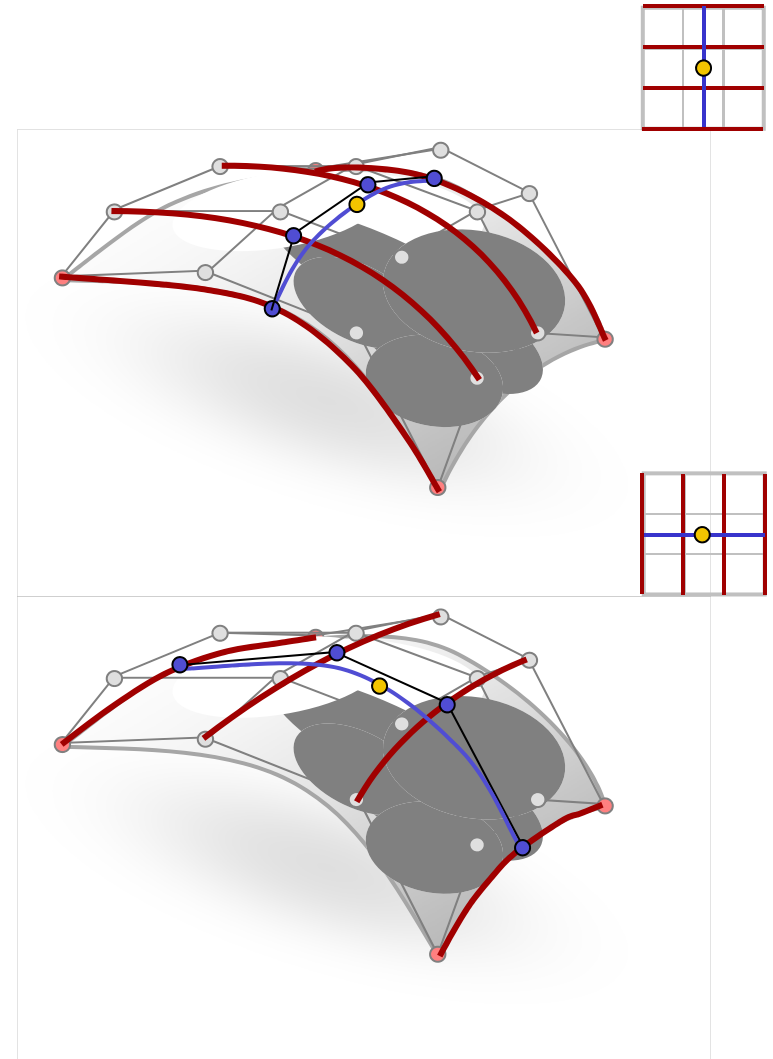


# Tensor Product Surfaces

## Tensor Product Surfaces:

$$\begin{aligned}\mathbf{f}(u,v) &= \sum_{i=1}^n \sum_{j=1}^n b_i(u)b_j(v)\mathbf{p}_{i,j} \\ &= \sum_{i=1}^n b_i(u) \sum_{j=1}^n b_j(v)\mathbf{p}_{i,j} \\ &= \sum_{j=1}^n b_j(u) \sum_{i=1}^n b_i(v)\mathbf{p}_{i,j}\end{aligned}$$

- “Curves of Curves”
- Order does not matter



# Properties

## Properties of tensor product surfaces:

- Linear invariance: Obvious
- Affine invariance?
  - Needs partition of unity property
  - Assume basis  $\mathbf{B}^{(\text{curv})} := \{b_1(t), \dots, b_n(t)\}$  forms a partition of unity, i.e.:  $\sum_{i=1}^n b_i(v) = 1$

- Then we get:

$$\sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) = \sum_{i=1}^n b_i(u) \sum_{j=1}^n b_j(v) = \sum_{j=1}^n b_j(u) \cdot 1 = 1$$

- Affine invariance for tensor product surfaces is induced by the corresponding property of the employed curve basis

# Properties

## Properties of tensor product surfaces:

- Convex Hull?
  - Assume basis  $\mathbf{B}^{(\text{curv})} := \{b_1(t), \dots, b_n(t)\}$  forms a partition of unity and it is positive ( $\geq 0$ ) on  $t \in [t_0, t_1]$

- Obviously, we then have:

$$\sum_{i=1}^n \sum_{j=1}^n \underbrace{b_i(u)}_{\geq 0} \underbrace{b_j(v)}_{\geq 0} \geq 0$$

- So we have the convex hull property on  $[t_0, t_1]^2$
- The convex hull property for tensor product surfaces is induced by the property of the employed curve basis.

# Partial Derivatives

## Computing partial derivatives:

- First derivatives:

$$\frac{\partial}{\partial u} \sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) \mathbf{p}_{i,j} = \sum_{j=1}^n b_j(v) \sum_{i=1}^n \left( \frac{d}{du} b_i \right) (u) \mathbf{p}_{i,j}$$

$$\frac{\partial}{\partial v} \sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) \mathbf{p}_{i,j} = \sum_{i=1}^n b_i(u) \sum_{j=1}^n \left( \frac{d}{dv} b_j \right) (v) \mathbf{p}_{i,j}$$

- Just spline-curve combinations of curve derivatives

# Partial Derivatives

## Computing partial derivatives:

- Second derivatives:

$$\frac{\partial^2}{\partial u^2} \sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) \mathbf{p}_{i,j} = \sum_{j=1}^n b_j(v) \sum_{i=1}^n \left( \frac{d}{du^2} b_i \right)(u) \mathbf{p}_{i,j}$$

$$\begin{aligned} \frac{\partial^2}{\partial u \partial v} \sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) \mathbf{p}_{i,j} &= \frac{\partial}{\partial v} \sum_{j=1}^n b_j(v) \sum_{i=1}^n \left( \frac{d}{du} b_i \right)(u) \mathbf{p}_{i,j} \\ &= \sum_{j=1}^n \left( \frac{d}{dv} b_j \right)(v) \sum_{i=1}^n \left( \frac{d}{du} b_i \right)(u) \mathbf{p}_{i,j} \end{aligned}$$



# Partial Derivatives

## Computing partial derivatives:

- General derivatives:

$$\begin{aligned}\frac{\partial^{r+s}}{\partial u^r \partial v^s} \sum_{i=1}^n \sum_{j=1}^n b_i(u) b_j(v) \mathbf{p}_{i,j} &= \sum_{j=1}^n \left( \frac{d^s}{dv^s} b_j \right)(v) \sum_{i=1}^n \left( \frac{d^r}{du^r} b_i \right)(u) \mathbf{p}_{i,j} \\ &= \sum_{i=1}^n \left( \frac{d^r}{du^r} b_i \right)(u) \sum_{j=1}^n \left( \frac{d^s}{dv^s} b_j \right)(v) \mathbf{p}_{i,j}\end{aligned}$$

# Normal Vectors

We can compute normal vectors from partial derivatives:

$$\bullet \mathbf{n}(u,v) = \frac{\left( \sum_{j=1}^n b_j(v) \sum_{i=1}^n \frac{d}{du} b_i(u) \mathbf{p}_{i,j} \right) \times \left( \sum_{j=1}^n \frac{d}{dv} b_j(v) \sum_{i=1}^n b_i(u) \mathbf{p}_{i,j} \right)}{\left\| \left( \sum_{j=1}^n b_j(v) \sum_{i=1}^n \frac{d}{du} b_i(u) \mathbf{p}_{i,j} \right) \times \left( \sum_{j=1}^n \frac{d}{dv} b_j(v) \sum_{i=1}^n b_i(u) \mathbf{p}_{i,j} \right) \right\|}$$

- Problem: degenerate cases
  - Colinear tangents
  - Irregular parametrization
- Need extra code to handle special cases

# Bezier Patches

## Bezier Patches:

- Use tensor product Bernstein basis

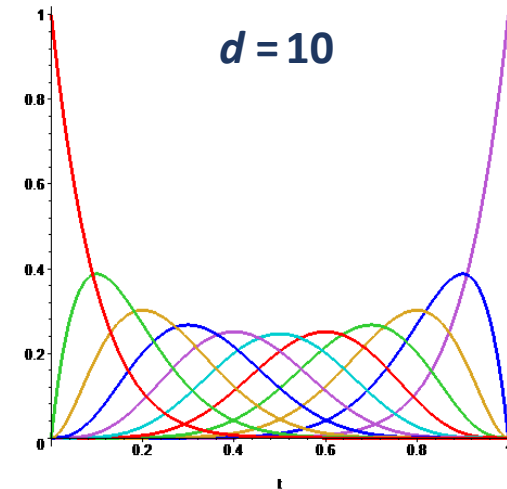
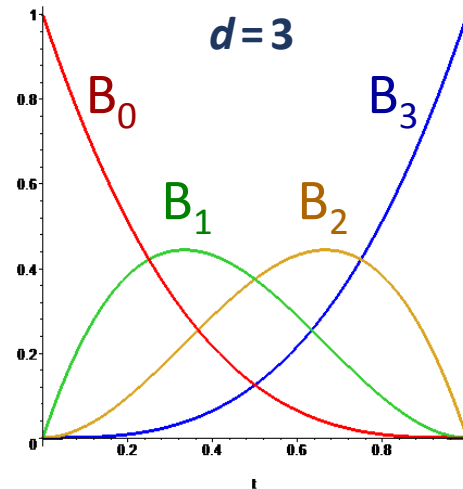
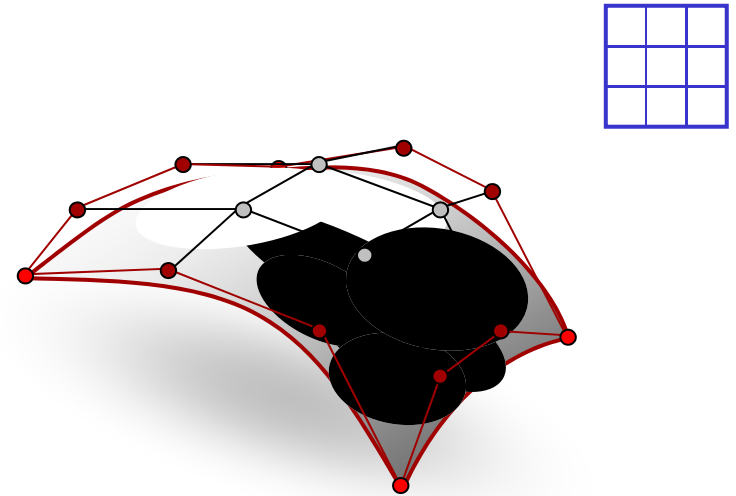
$$\mathbf{f}(u,v) = \sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \mathbf{p}_{i,j}$$

- We get automatically:
  - Affine invariance
  - Convex hull property

# Bezier Patches

## Bezier Patches:

- Interpolation:
  - Boundary curves are Bezier curves of the boundary control points



# Bezier Patches

## Bezier Patches

- Tangent vectors:
  - First derivatives at boundary points are proportional to differences of control points:

$$\begin{aligned}\left. \frac{\partial}{\partial u} \mathbf{f}(u, v) \right|_{u=0} &= \sum_{i=0}^d \sum_{j=0}^d B_j^{(d)}(v) B_i'^{(d)}(0) \mathbf{p}_{i,j} \\ &= d \sum_{j=0}^d B_j^{(d)}(v) (\mathbf{p}_{1,j} - \mathbf{p}_{0,j})\end{aligned}$$

$$\left. \frac{\partial}{\partial u} \mathbf{f}(u, v) \right|_{u=1} = d \sum_{j=0}^d B_j^{(d)}(v) (\mathbf{p}_{d,j} - \mathbf{p}_{d-1,j})$$

# Continuity Conditions

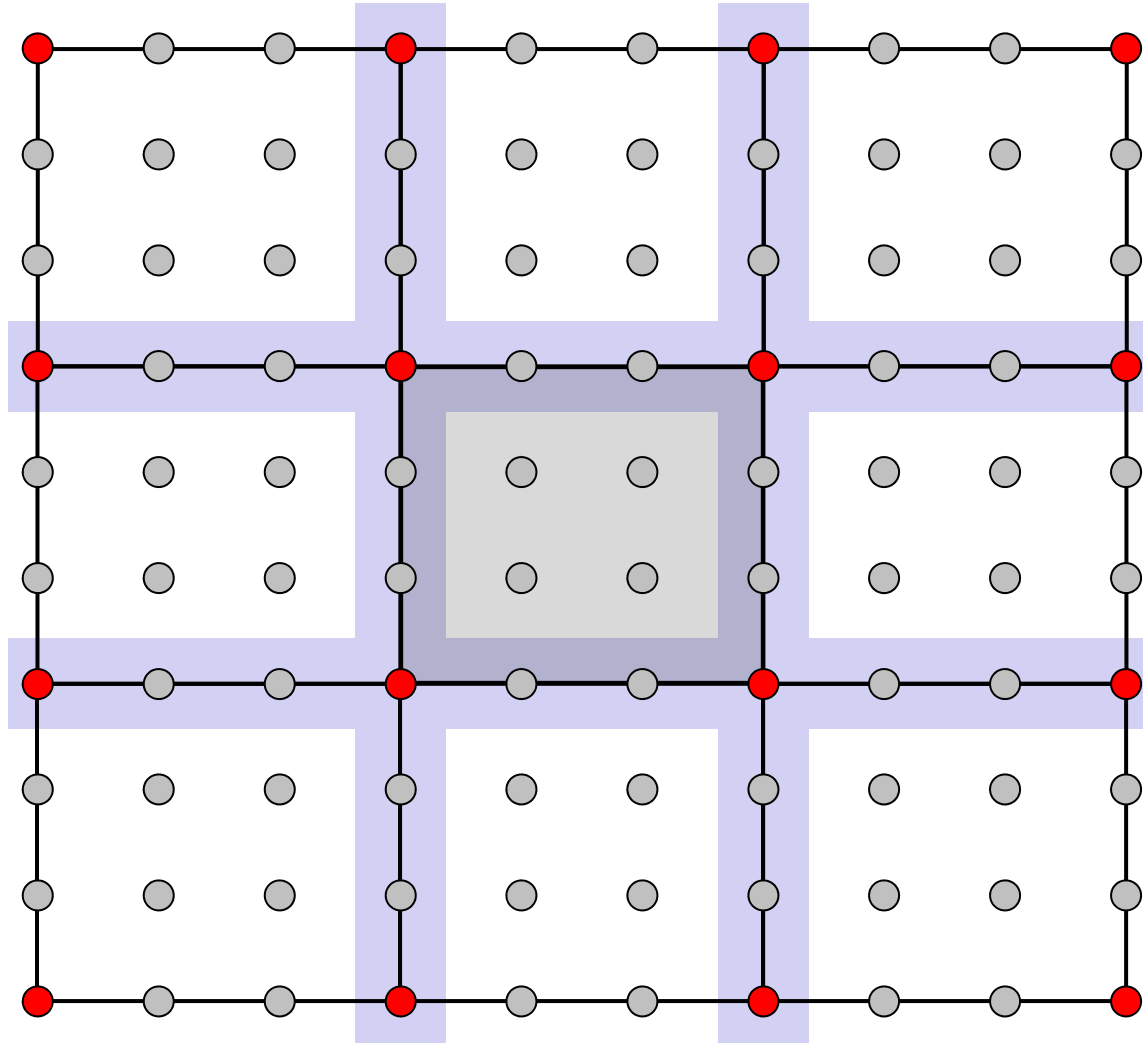
## For $C^0$ continuity:

- Boundary control points must match

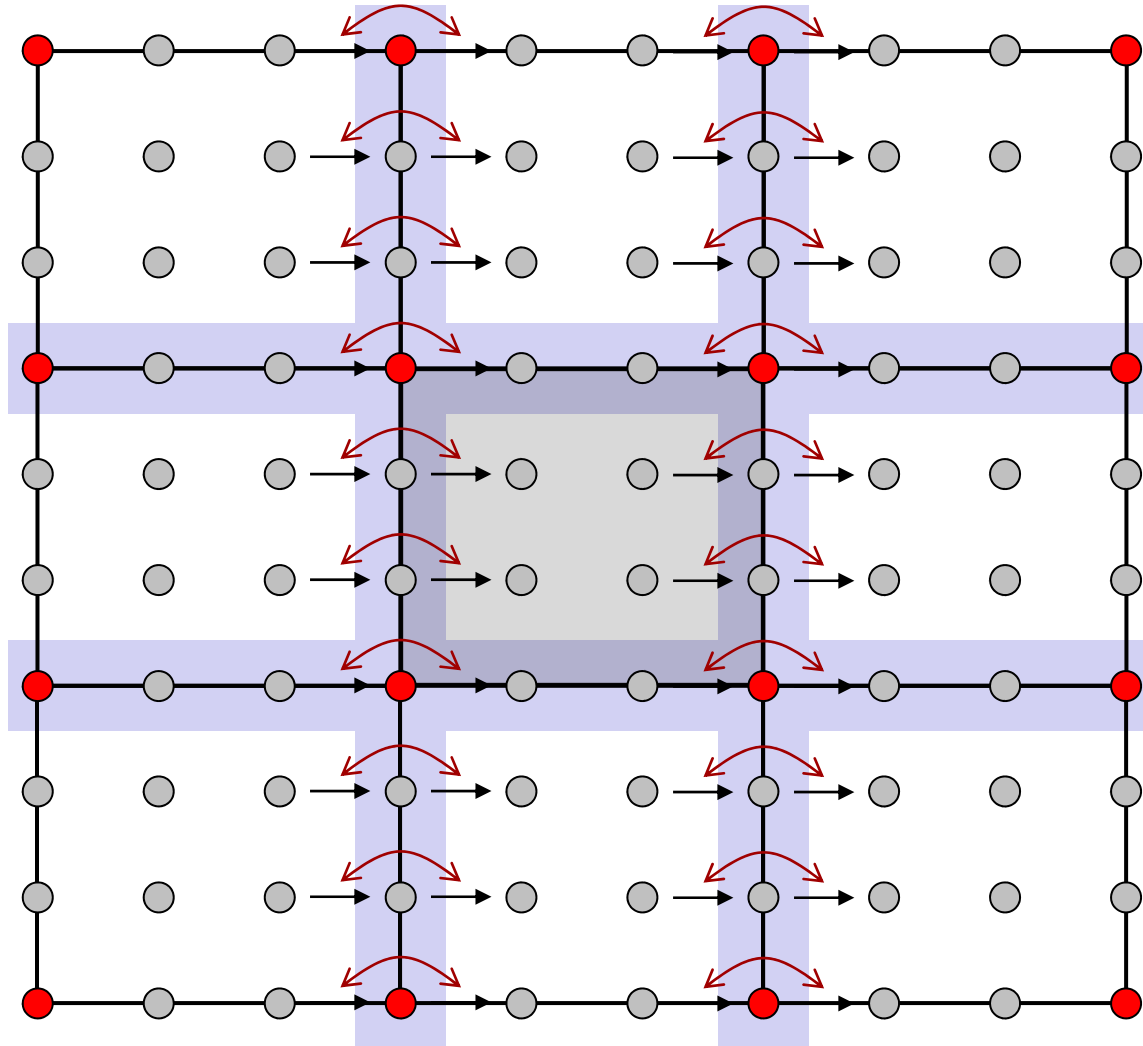
## For $C^1$ continuity:

- Difference vectors must match at the boundary

# $C^0$ Continuity

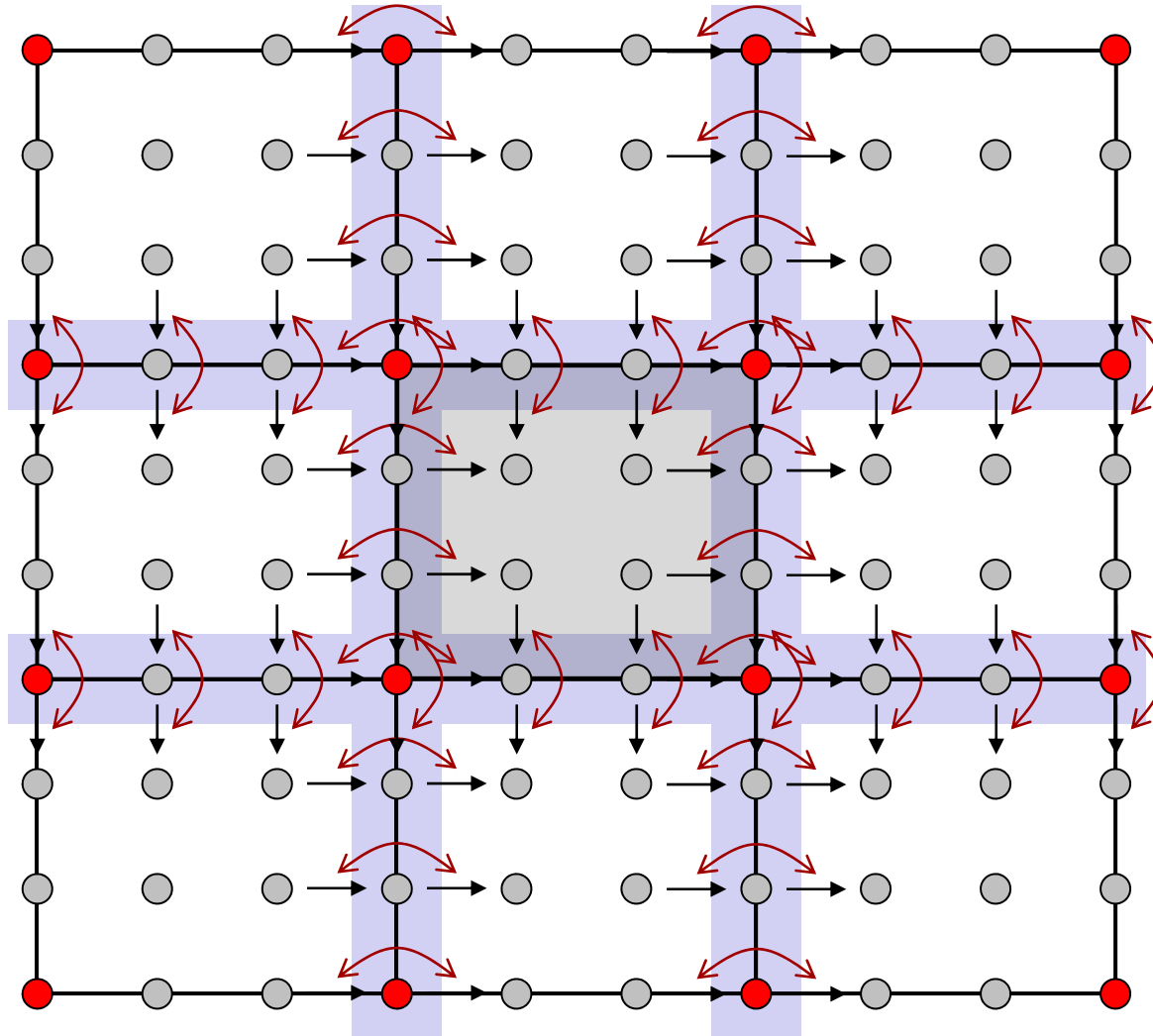


# C<sup>1</sup> Continuity





# C<sup>1</sup> Continuity



# Polars & Blossoms

## Blossoms for tensor product surfaces:

- Polar form of a polynomial tensor product surface of degree  $d$ :

$$\mathbf{F}: \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}^n \quad \mathbf{F}(u, v)$$

$$\mathbf{f}: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}^n \quad \mathbf{f}(u_1, \dots, u_d; v_1, \dots, v_d)$$

- Required Properties:

- Diagonality:  $\mathbf{f}(u, \dots, u; v, \dots, v) = \mathbf{F}(u, v)$

- Symmetry:  $\mathbf{f}(u_1, \dots, u_d; v_1, \dots, v_d) = \mathbf{f}(u_{\pi(1)}, \dots, u_{\pi(d)}; v_{\mu(1)}, \dots, v_{\mu(d)})$   
for all permutations of indices  $\pi, \mu$ .

- Multi-affine:  $\sum \alpha_k = 1$

$$\begin{aligned} \Rightarrow \mathbf{f}(u_1, \dots, \sum \alpha_k u_i^{(k)}, \dots, u_d; v_1, \dots, v_d) \\ = \alpha_1 \mathbf{f}(u_1, \dots, u_i^{(1)}, \dots, u_d; v_1, \dots, v_d) + \dots + \alpha_n \mathbf{f}(u_1, \dots, u_i^{(n)}, \dots, u_d; v_1, \dots, v_d) \end{aligned}$$

$$\begin{aligned} \text{and } \mathbf{f}(u_1, \dots, u_d; v_1, \dots, \sum \alpha_k v_i^{(k)}, \dots, v_d) \\ = \alpha_1 \mathbf{f}(u_1, \dots, u_d; v_1, \dots, v_i^{(1)}, \dots, v_d) + \dots + \alpha_n \mathbf{f}(u_1, \dots, u_d; v_1, \dots, v_i^{(n)}, \dots, v_d) \end{aligned}$$

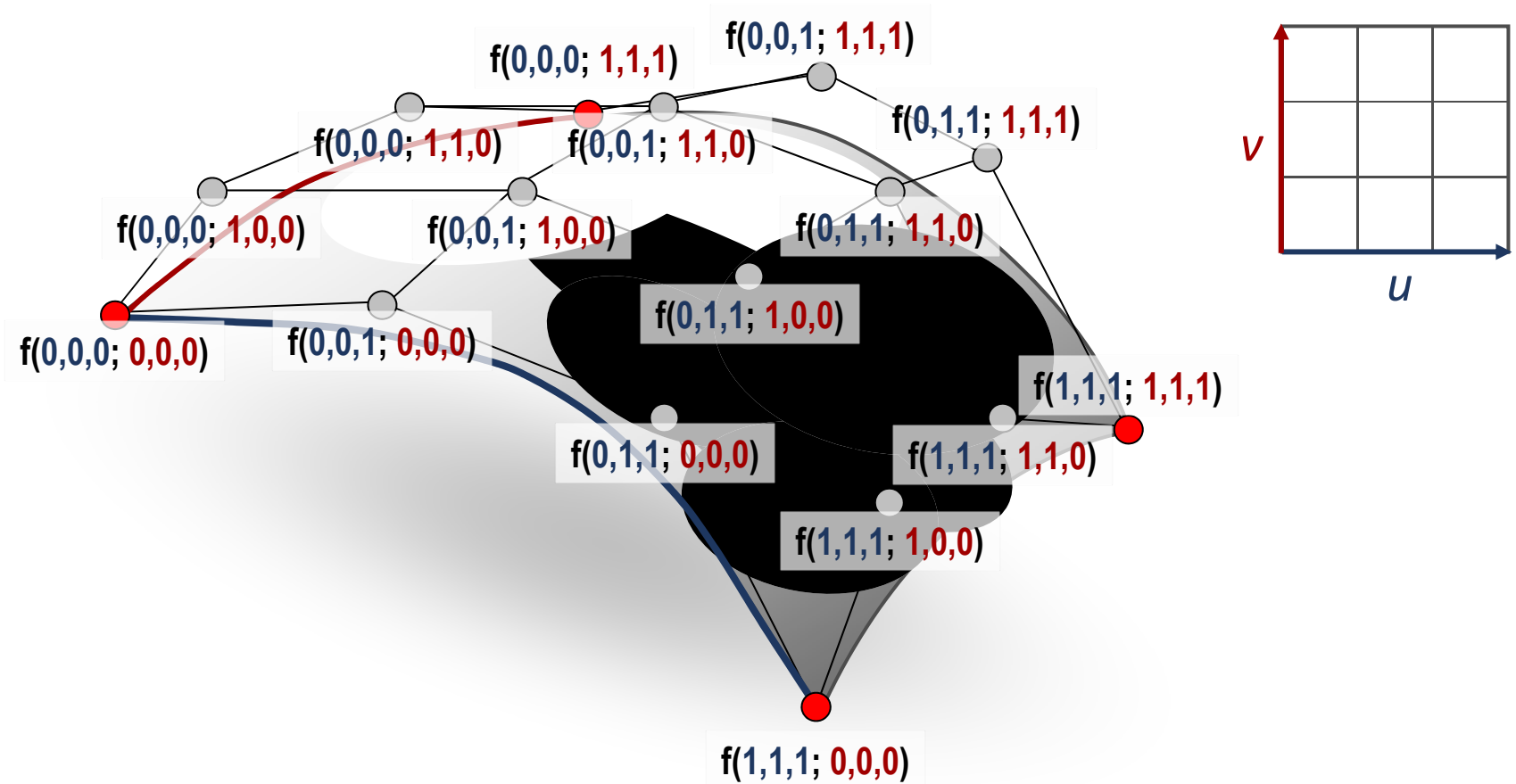
# Short Summary

## Polar forms for tensor product surfaces:

- Polarize separately in  $u$  and  $v$ .
- Notation:  $\mathbf{f}(\underbrace{u_1, \dots, u_d}_{u\text{-parameters}}; \underbrace{v_1, \dots, v_d}_{v\text{-parameters}})$
- Can be used to derive properties/algorithms similar to the curve case
- More interesting: Polar forms for total degree surfaces (we will see this later)

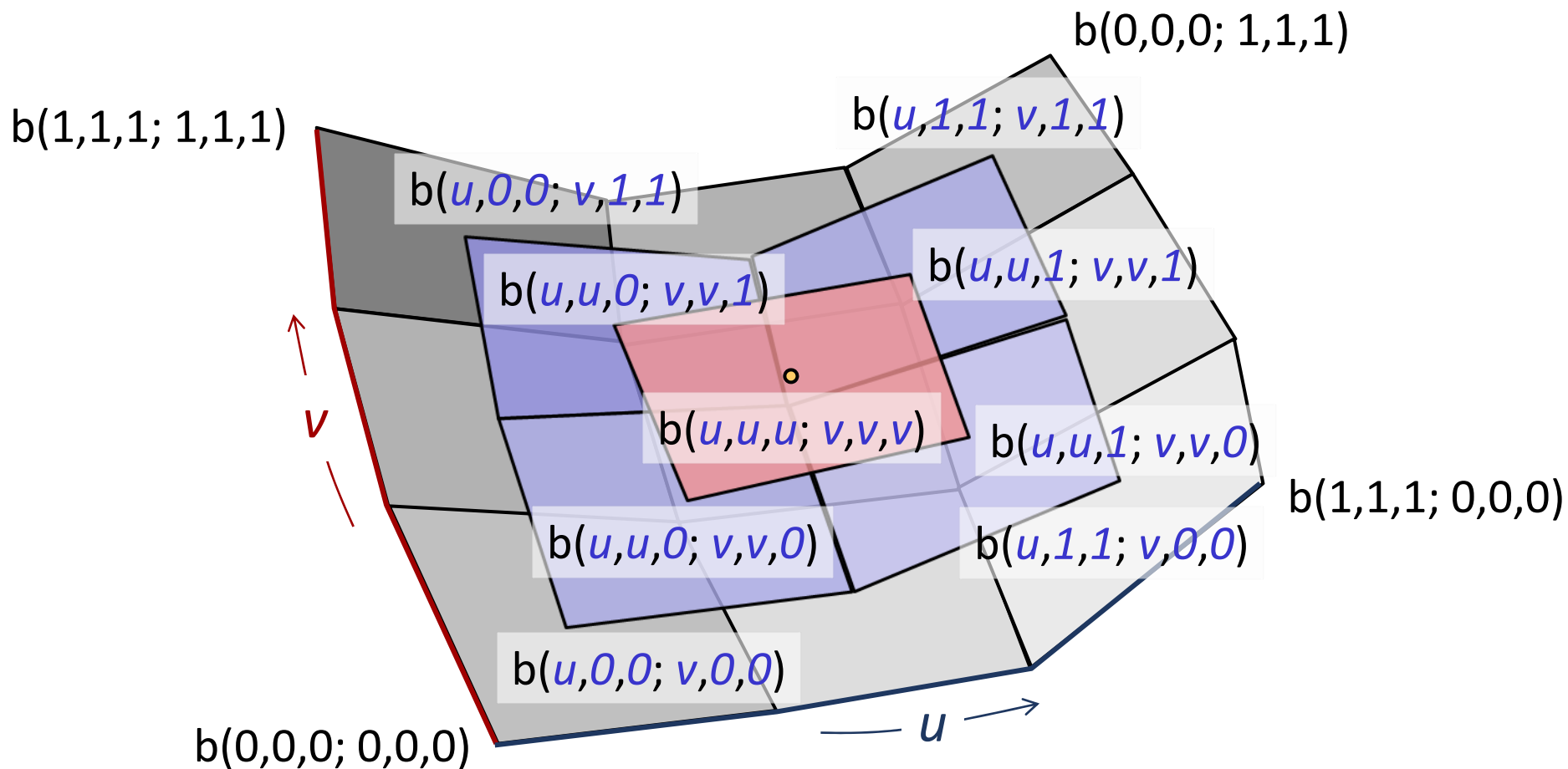
# Bezier Control Points

Bezier control points in blossom notation:



# De Casteljau Algorithm

## De Casteljau algorithm for tensor product surfaces:



# B-Spline Patches

## B-Spline Patches

- More general than Bezier patches  
(we get Bezier patches as a special case)
- First, we fix a degree  $d$ .
- Then, we need knot sequences in  $u$  and  $v$  direction:  
 $(u_1, \dots, u_n), (v_1, \dots, v_m)$
- And a corresponding array of control points:

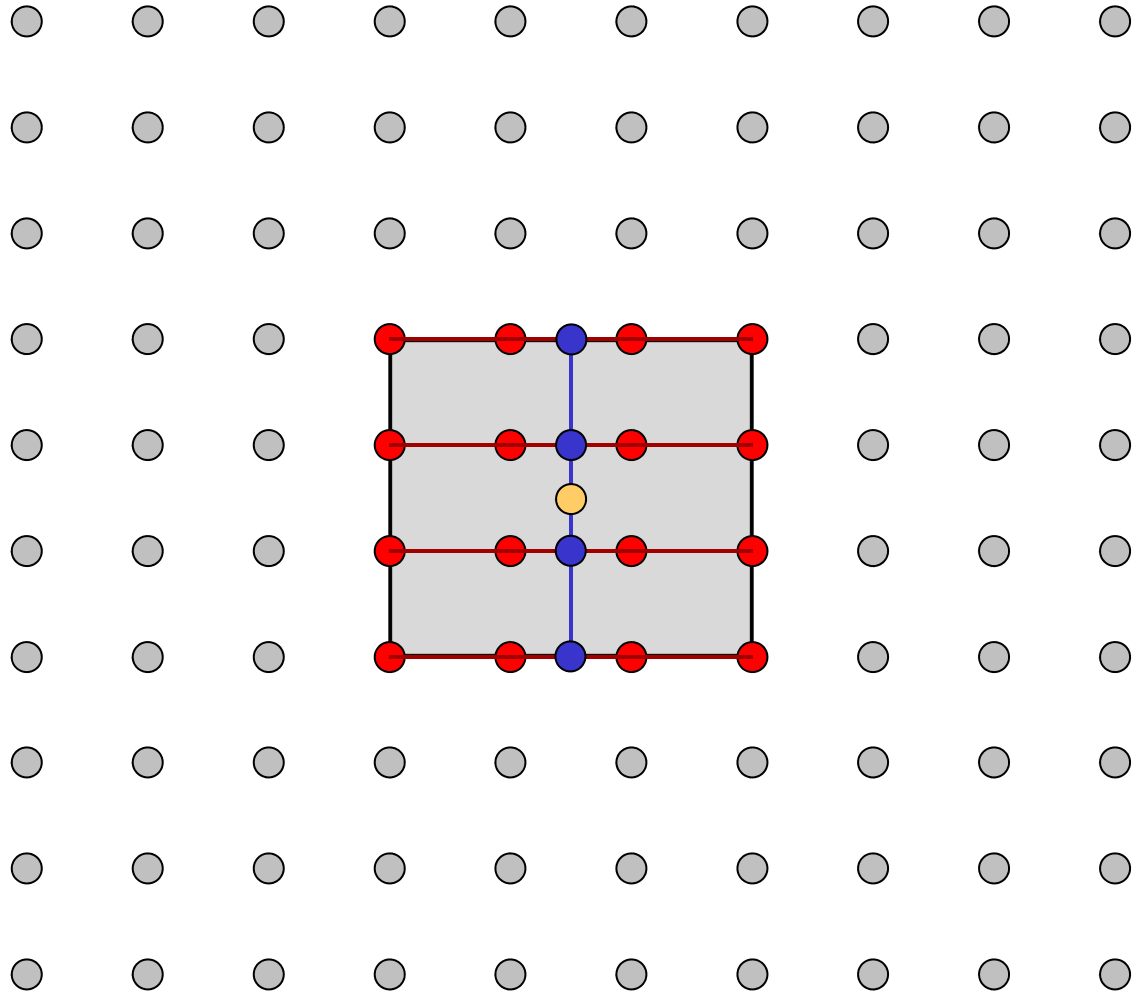
$$\begin{array}{ccc} \mathbf{d}_{0,0} & \dots & \mathbf{d}_{n-d+1,0} \\ \vdots & & \vdots \\ \mathbf{d}_{0,m-d+1} & \dots & \mathbf{d}_{n-d+1,m-d+1} \end{array}$$

# B-Spline Patches

Then, obtain a parametric B-spline patch as:

- $$\mathbf{f}(u,v) = \sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \mathbf{p}_{i,j}$$
- We can evaluate the patches using the de Boor Algorithm:
  - “Curves of curves” idea
  - Determine the knots/control points influencing  $(u,v)$ .  
These will be no more than  $(d+1) \times (d+1)$  points.
  - Compute  $(d+1)$   $v$ -direction control points along  $u$  direction, performing  $(d+1)$  curve evaluations.
  - Then evaluate the curve in  $v$ -direction.
  - (or the other way round, interchanging  $u,v$ -directions)

# Illustration





# B-Spline Patches

## Alternative:

- 2D de Boor algorithm
- Works similar to the 2D de Casteljau algorithm but with different weights  
(we can use tensor-product blossoming to derive the weights)

# Rational Patches

## Rational Patches:

- We can use rational Bezier/B-splines to create the patches (“rational Bezier patches” / “NURBS-patches”)
- Idea:
  - Form a parametric surface in 4D, homogenous space
  - Then project to  $\omega = 1$  to obtain the surface in Euclidian 3D space
- In short: Just use homogeneous coordinates everywhere

# Rational Patch

## Rational Bezier Patch:

$$\mathbf{f}^{(\text{hom})}(u, v) = \sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \begin{pmatrix} \omega_{i,j} \mathbf{p}_{i,j} \\ \omega_{i,j} \end{pmatrix}$$

$$\mathbf{f}^{(\text{Eucl})}(u, v) = \frac{\sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \mathbf{p}_{i,j}}{\sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \omega_{i,j}}$$

# Rational Patch

## Rational B-Spline Patch:

$$\mathbf{f}^{(\text{hom})}(u,v) = \sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \begin{pmatrix} \omega_{i,j} \mathbf{p}_{i,j} \\ \omega_{i,j} \end{pmatrix}$$

$$\mathbf{f}^{(\text{Eucl})}(u,v) = \frac{\sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \mathbf{p}_{i,j}}{\sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \omega_{i,j}}$$

# Remark: Rational Patches

## Interesting Observation:

- The Euclidian surface is not a tensor product surface (denominator of  $B_i(u)$ ,  $B_j(v)$  depends on both  $u$  and  $v$ )
- In the homogeneous space, the 4D surface is of course still a tensor product surface.

$$\mathbf{f}^{(Eucl)}(u, v) = \frac{\sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \mathbf{p}_{i,j}}{\sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \omega_{i,j}}$$

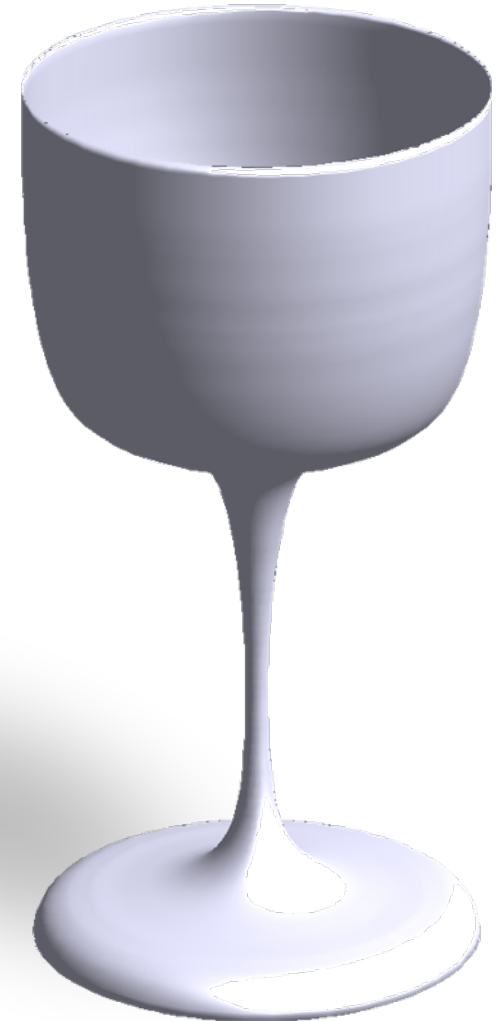
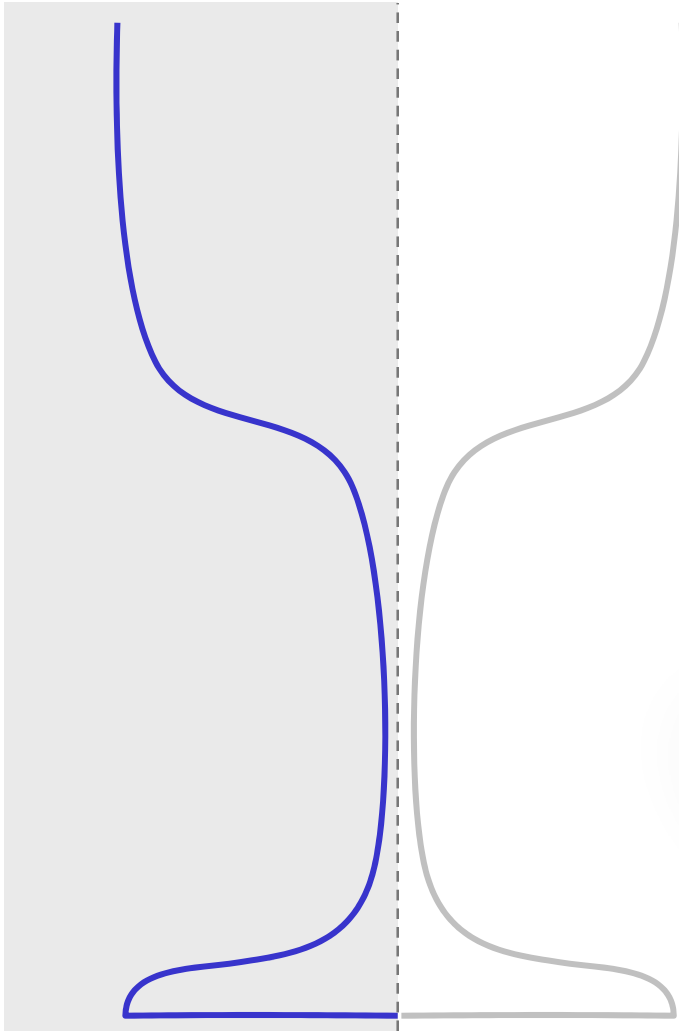
$$\mathbf{f}^{(Eucl)}(u, v) = \frac{\sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \mathbf{p}_{i,j}}{\sum_{i=0}^n \sum_{j=0}^m N_i^{(d)}(u) N_j^{(d)}(v) \omega_{i,j}}$$

# Surfaces of Revolution

## Advantages of rational patches:

- Rational patches can represent surfaces of revolution exactly.
- Examples:
  - Cylinders
  - Cones
  - Spheres
  - Ellipsoids
  - Tori
- Question: given a cross section curve, how do we get the control points for the 3D surface?

# Surfaces of Revolution



# Surfaces of Revolution

## Given:

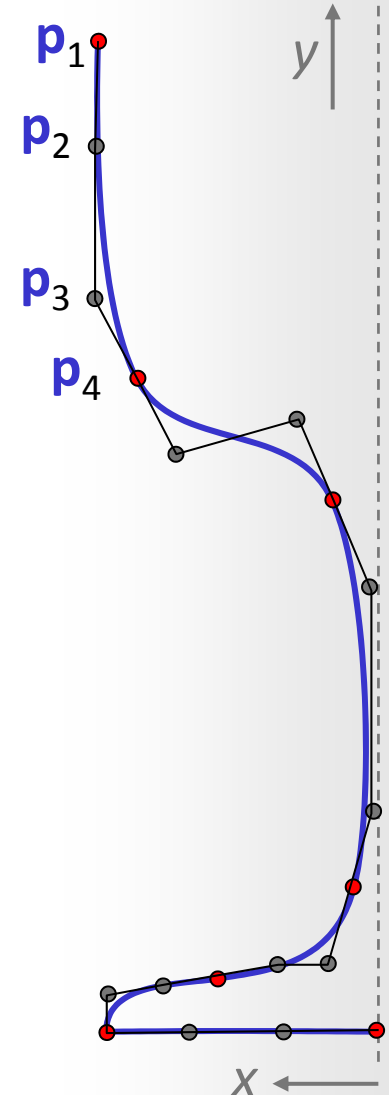
- Control points  $\mathbf{p}_1, \dots, \mathbf{p}_n$  of curve (“generatrix”)

## We want to compute:

- Control points  $\mathbf{p}_{i,j}$  of a rational surface

## Such that:

- The surface describes the surface of revolution that we obtain by rotating the curve around the  $y$  axis (w.l.o.g.)

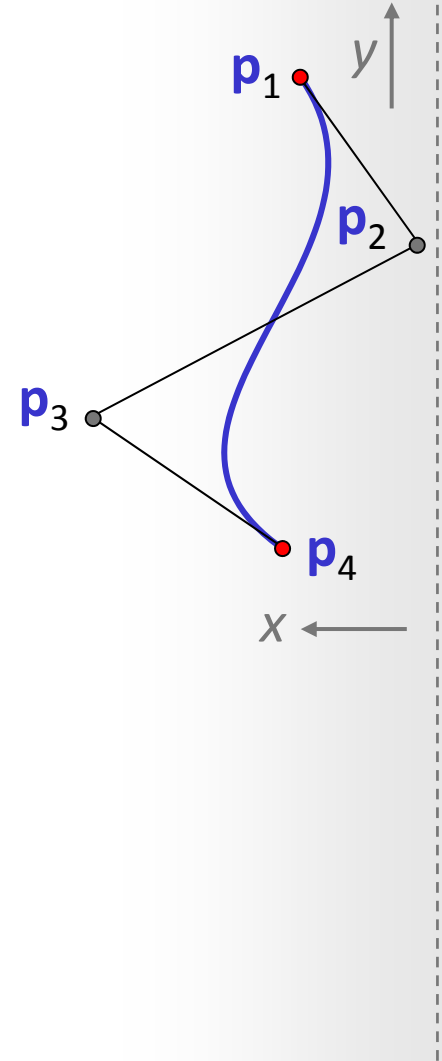




# Surfaces of Revolution

## Simplification:

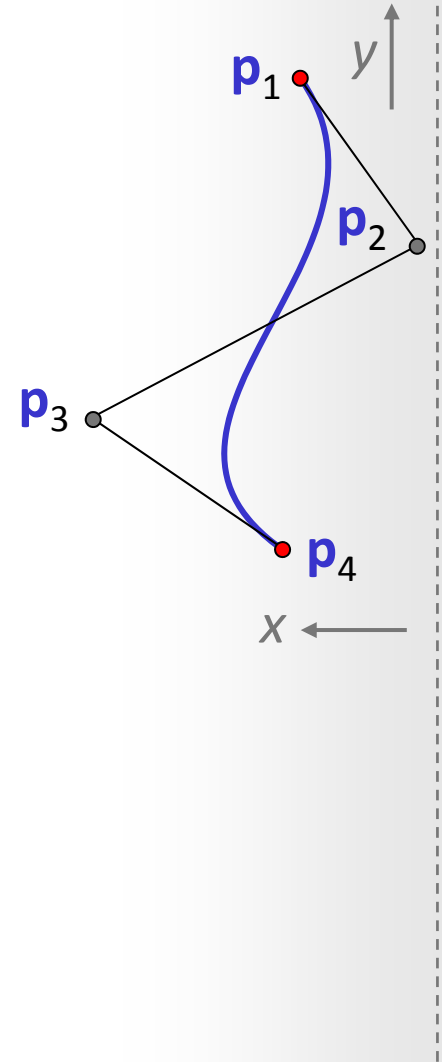
- We look only at a single rational Bezier segment.
- Applying the scheme to multiple segments together is straightforward.
- The same idea also works for B-splines.



# Surfaces of Revolution

## Construction:

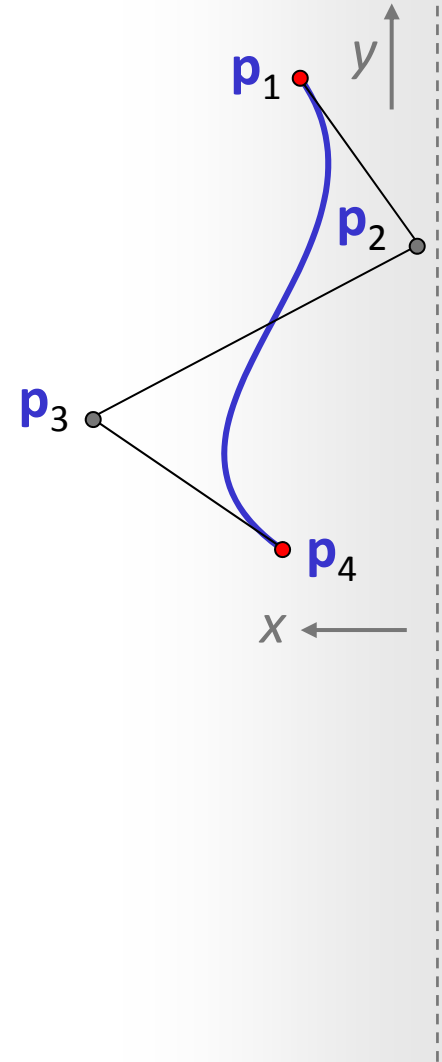
- We are given control points  $\mathbf{p}_1, \dots, \mathbf{p}_{d+1}$  ( $d$  is the degree in  $u$  direction)
- We introduce a new parameter  $v$ .
- In  $v$  direction, we use quadratic Bezier curves (2nd degree basis in  $v$ -direction)



# Surfaces of Revolution

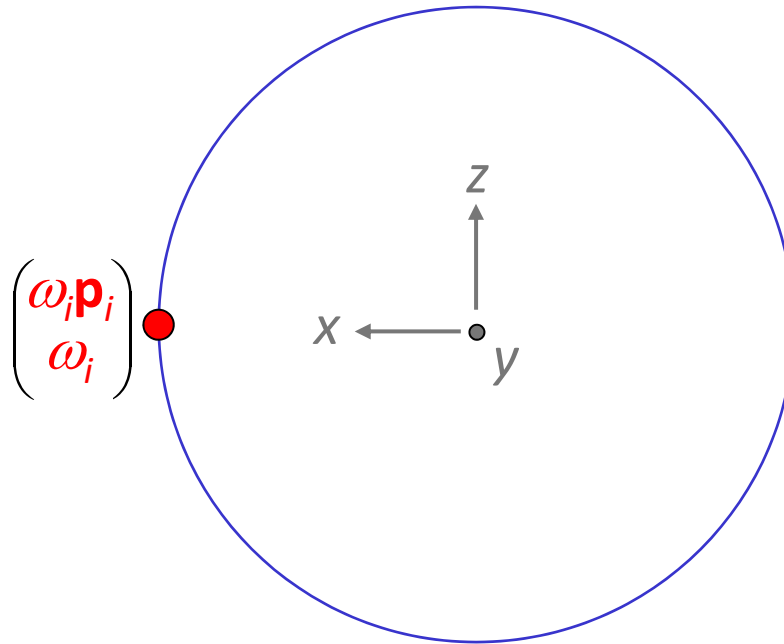
## Key Idea:

- For  $u$ -direction curves: Control points (and thus the curves) must move on circles around the  $y$ -axis.
- Circles must have the same parametrization (this is easy)
- This means, the control points rotate around the  $y$ -axis.
- Affine invariance will make the whole curve rotate, we get the desired surface of revolution.



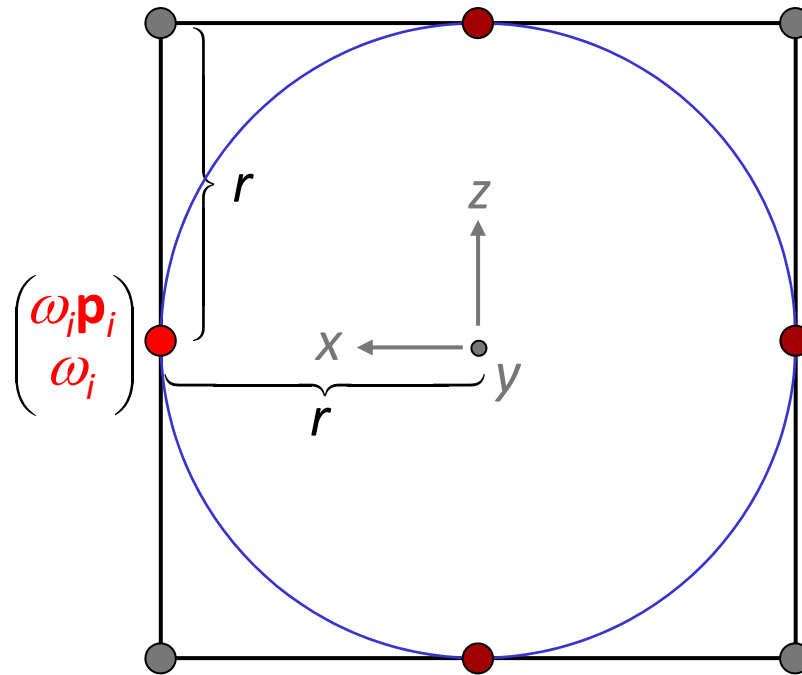
# Surfaces of Revolution

Making one point rotate around the y-axis:



# Surfaces of Revolution

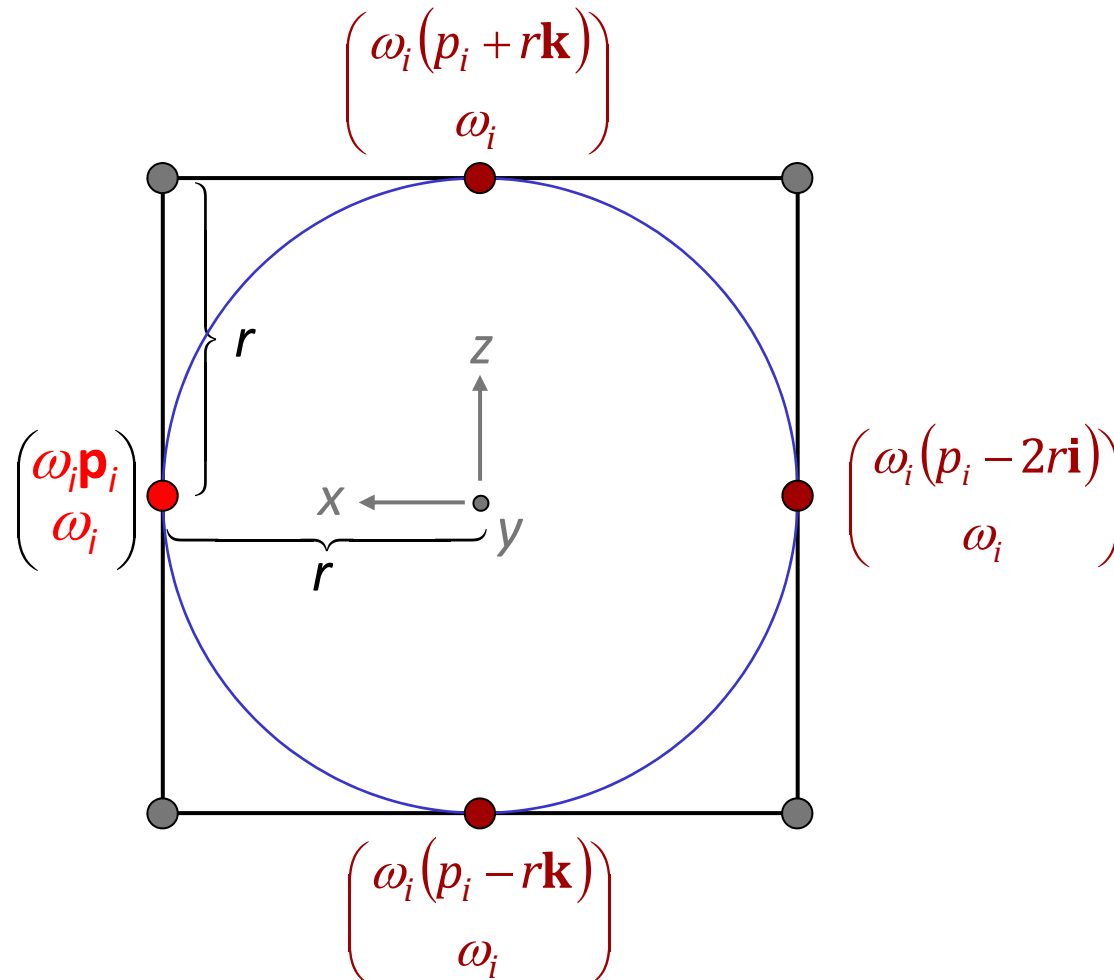
Making one point rotate around the y-axis:



# Surfaces of Revolution

Making one point rotate around the y-axis:

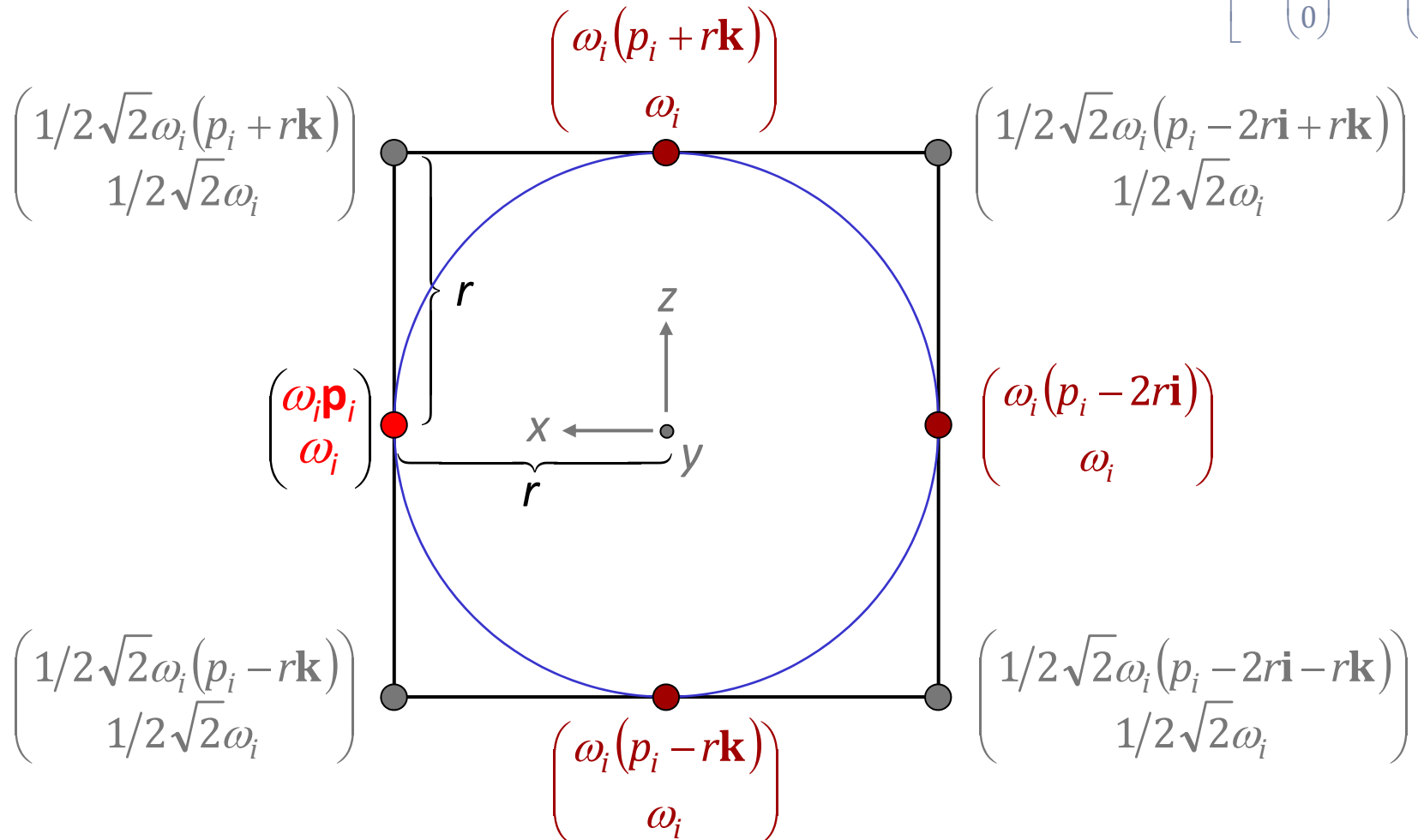
$$\mathbf{i} := \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \mathbf{k} := \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$



# Surfaces of Revolution

Making one point rotate around the y-axis:

$$\mathbf{i} := \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \mathbf{k} := \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$



# Remark

## What we get:

- We obtain 4 segments, i.e. 4 patches for each Bezier segment
- A similar construction with 3 segments exists as well

## Does the scheme yield a circle for weights $\neq 1$ in the generatrix curve?

- Common factors in weights cancel out
- Therefore, we still obtain a circle at these points
- Parametrization does not change either



# Benefit

---

**With this construction, it is straightforward to create:**

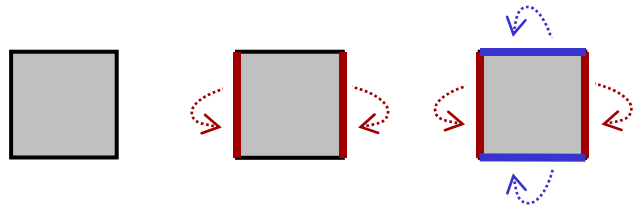
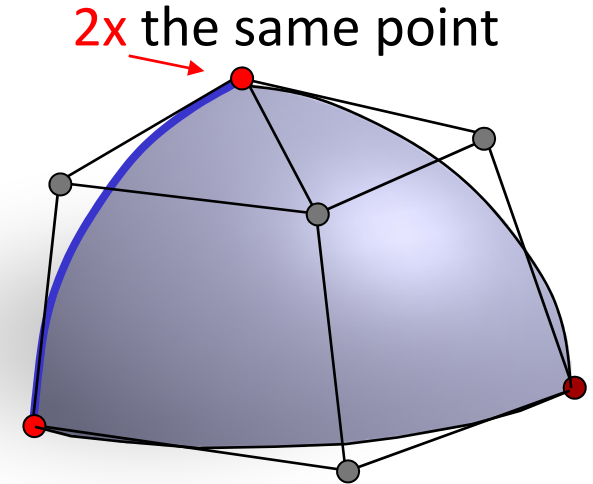
- Spheres
- Tori
- Cylinders
- Cones

**And affine transformations of these (e.g. ellipsoids)**

# Parametrization Restrictions

## Remaining problem:

- The sphere and the cone are not regularly parametrized (double control points)
- Might cause trouble (normals computation, tessellation)
- In general: no spheres, or  $n$ -tori ( $n > 1$ ) can be parametrized without degeneracies
- What works: open surfaces, cylinders, tori



# Curves on Surfaces, trimmed NURBS

## Quad patch problem:

- All of our shapes are parameterized over rectangular regions
- General boundary curves are hard to create
- Topology fixed to a disc (or cylinder, torus)
- No holes in the middle
- Assembling complicated shapes is painful
  - Lots of pieces
  - Continuity conditions for assembling pieces become complicated
  - Cannot use  $C^2$  B-Splines continuity along boundaries when using multiple pieces

# Curves on Surfaces, trimmed NURBS

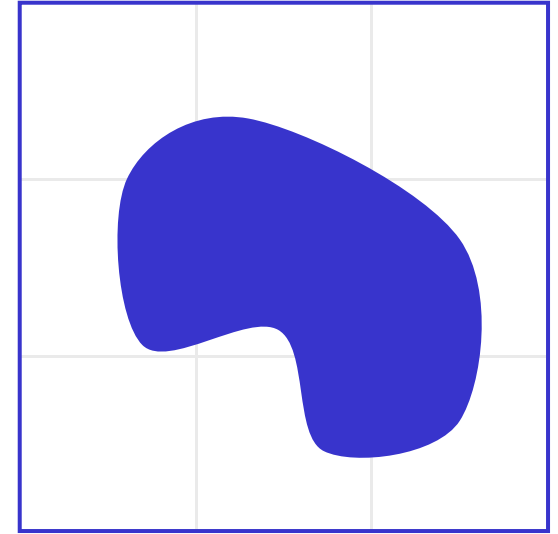
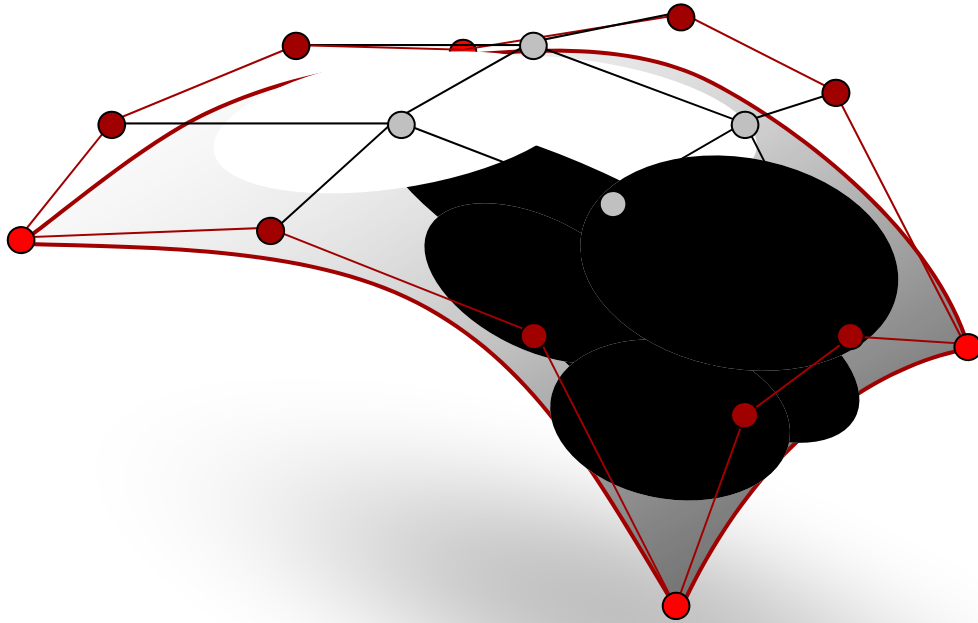
## Consequence:

- We need more control over the parameter domain
- One solution is *trimming* using *curves on surfaces (CONS)*
- Standard tool in CAD: *trimmed NURBS*

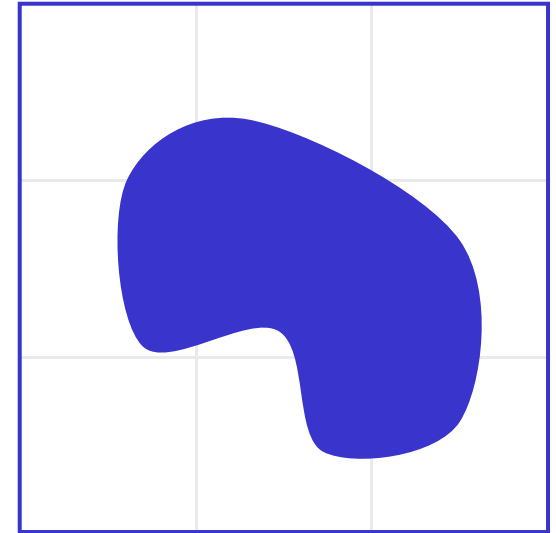
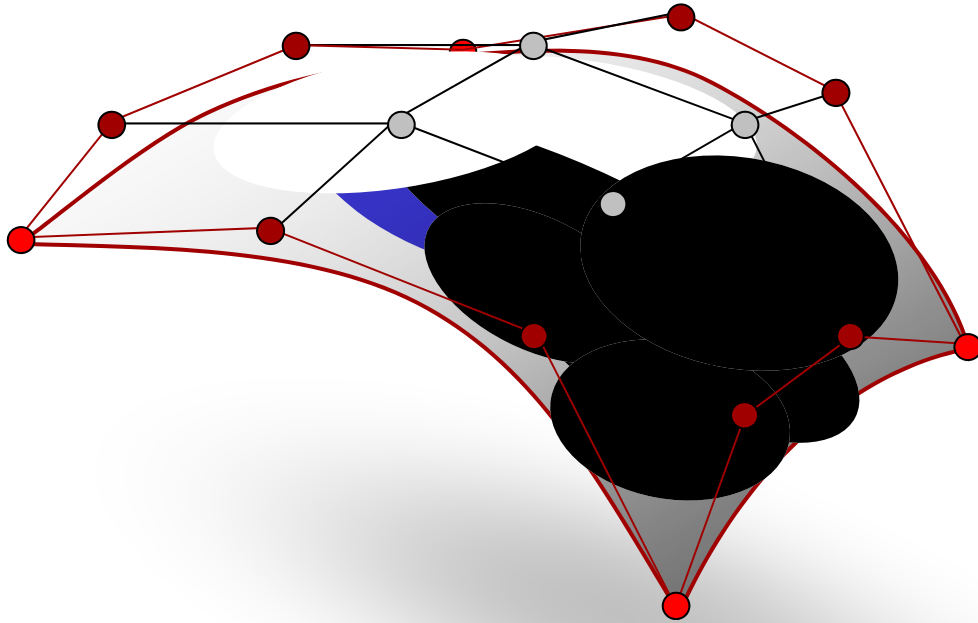
## Basic idea:

- Specify a curve in the parameter domain that encapsulates one (or more) pieces of area
- Tessellate the parameter domain accordingly to cut out the trimmed piece (rendering)

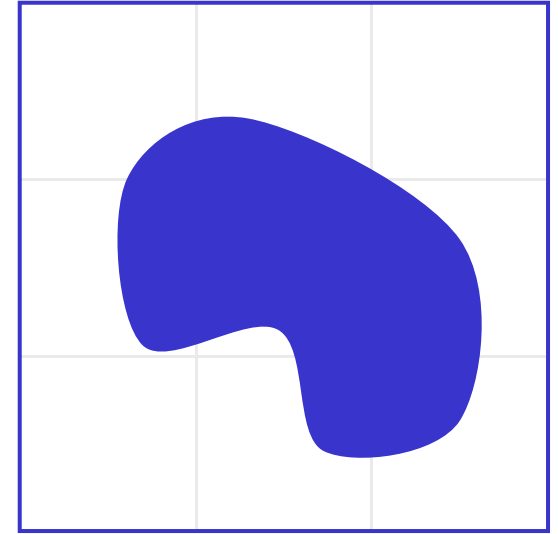
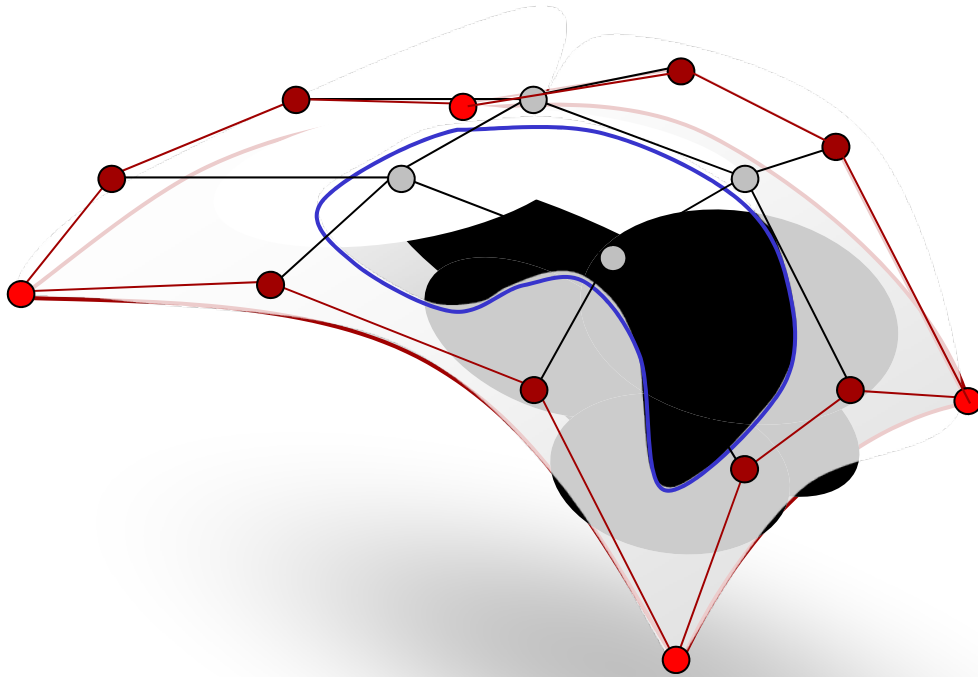
# Curves-on-Surfaces (CONS)



# Curves-on-Surfaces (CONS)



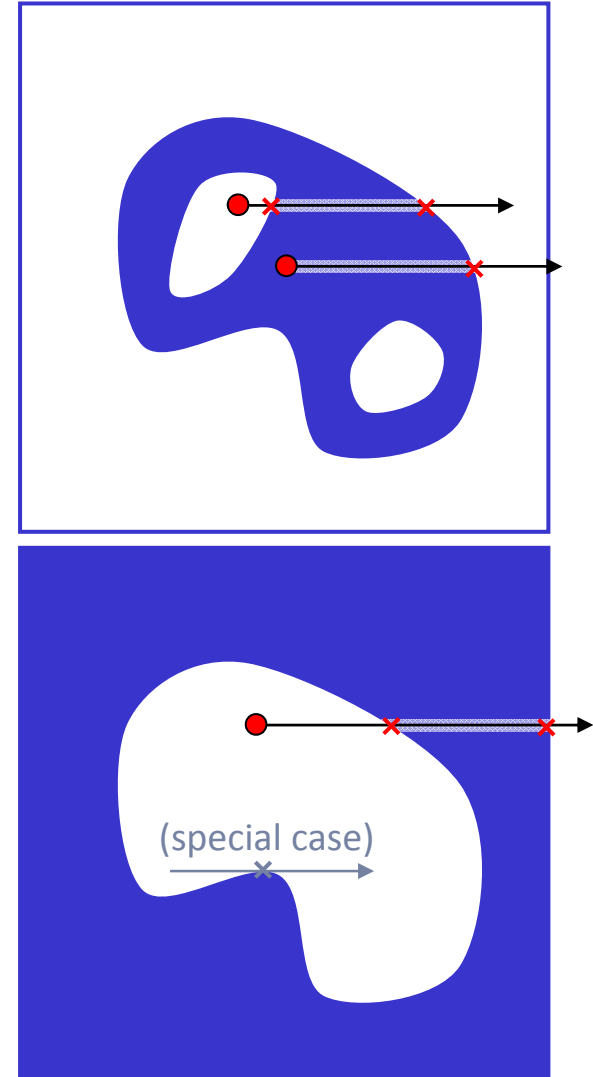
# Curves-on-Surfaces (CONS)



# General Shapes

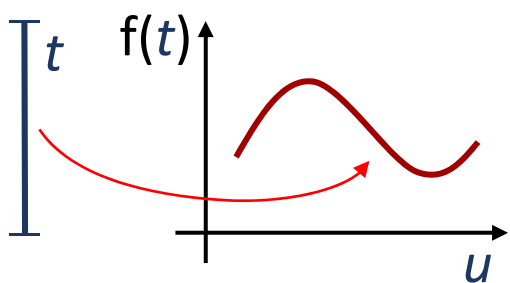
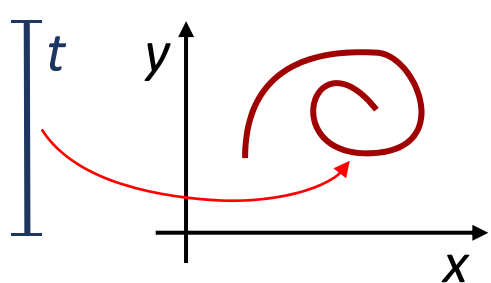
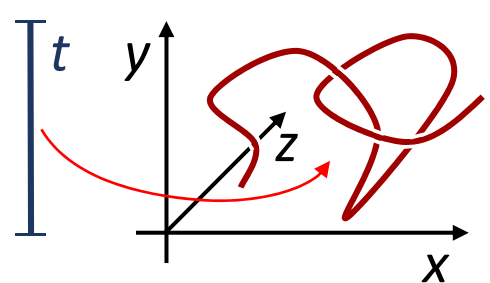
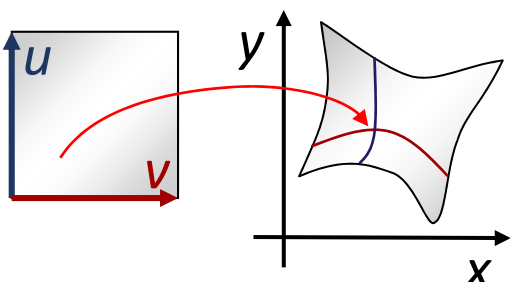
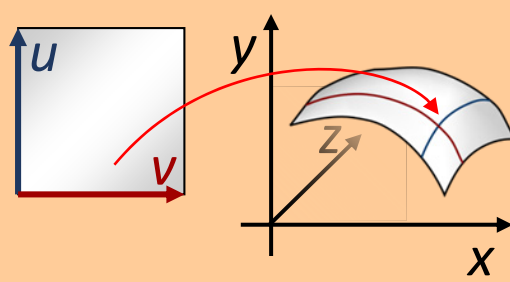
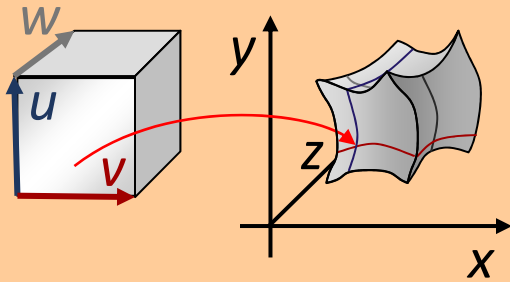
## General shapes with holes:

- Draw multiple curves
- Inside / outside test:
  - If any ray in the parameter domain intersects the boundary curves an odd number of times, the point is inside
  - Outside otherwise
  - Implementation needs to take care of special cases (critical points with respect to normal of the ray)
  - Nasty, but doable





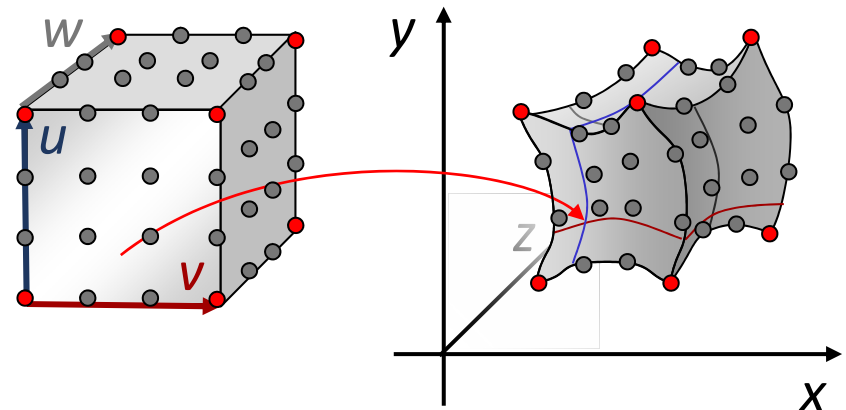
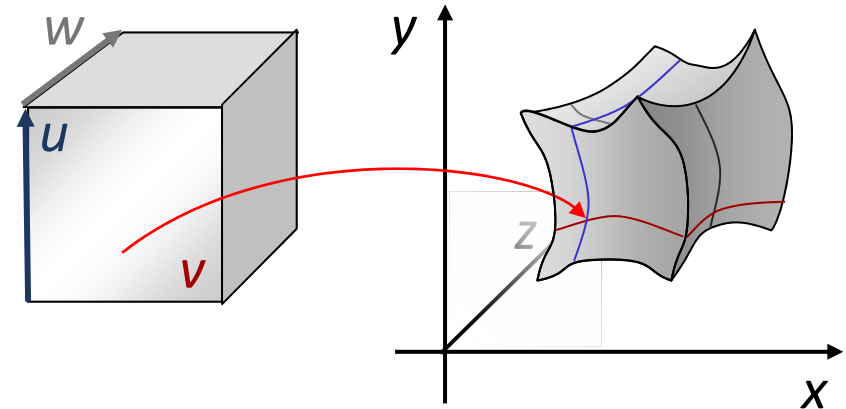
# Free Form Deformation

	output: 1D	output: 2D	output: 3D
input: 1D	 <p>function graph</p>	 <p>plane curve</p>	 <p>space curve</p>
input: 2D		 <p>plane warp</p>	 <p>surface</p>
input: 3D			 <p>space warp</p>

# FFD

## Free Form Deformations

- Use a 3D tensor-product B-Spline (or Bezier spline)
- Defines a bend mapping  $\mathbb{R}^3 \rightarrow \mathbb{R}^3$
- Can be used to change the shape of objects globally
- We will see other shape deformation techniques later in the lecture (time permitting)

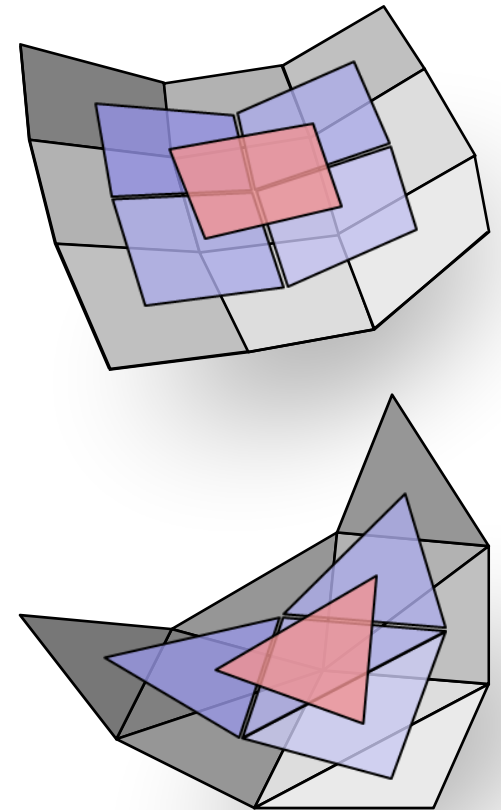


# **Total Degree Surfaces**

# Bezier Triangles

## Alternative surface definition: Bezier triangles

- Constructed according to given total degree
  - Completely symmetric:  
No degree anisotropy
- Can be derived using a triangular de Casteljau algorithm
  - Blossoming formalism is very helpful for defining Bezier Triangles
  - Barycentric interpolation of blossom values



# Blossoms for Total Degree Surfaces

## Blossoms with points as arguments:

- Polar form degree  $d$  with points as input and output:

$$\begin{aligned} \mathbf{F}: \mathbb{R}^n &\rightarrow \mathbb{R}^m \\ \mathbf{f}: \mathbb{R}^{d \times n} &\rightarrow \mathbb{R}^m \end{aligned}$$

points as arguments

- Required Properties:

- Diagonality:  $\mathbf{f}(\mathbf{t}, \mathbf{t}, \dots, \mathbf{t}) = \mathbf{F}(\mathbf{t})$
- Symmetry:  $\mathbf{f}(\mathbf{t}_1, \mathbf{t}_2, \dots, \mathbf{t}_d) = \mathbf{f}(\mathbf{t}_{\pi(1)}, \mathbf{t}_{\pi(2)}, \dots, \mathbf{t}_{\pi(d)})$   
for all permutations of indices  $\pi$ .
- Multi-affine:  $\sum \alpha_k = 1$   
 $\Rightarrow \mathbf{f}(\mathbf{t}_1, \mathbf{t}_2, \dots, \sum \alpha_k \mathbf{t}_i^{(k)}, \dots, \mathbf{t}_d)$   
 $= \alpha_1 \mathbf{f}(\mathbf{t}_1, \mathbf{t}_2, \dots, \mathbf{t}_i^{(1)}, \dots, \mathbf{t}_d) + \dots + \alpha_n \mathbf{f}(\mathbf{t}_1, \mathbf{t}_2, \dots, \mathbf{t}_i^{(n)}, \dots, \mathbf{t}_d)$

# Example

## Example: bivariate monomial basis

- In powers of  $(u, v)$ :

$$B = \{1, u, v, u^2, uv, v^2\}$$

- Blossom form: multilinear in  $(u_1, u_2, v_1, v_2)$

$$B = \left\{ 1, \right. \\ \left. \frac{1}{2}(u_1 + u_2), \quad \frac{1}{2}(v_1 + v_2), \right. \\ \left. u_1 u_2, \quad \frac{1}{4}(u_1 v_1 + u_1 v_2 + u_2 v_1 + v_2 u_2), \quad v_1 v_2 \right\}$$

# Barycentric Coordinates

## Barycentric Coordinates:

- Planar case:

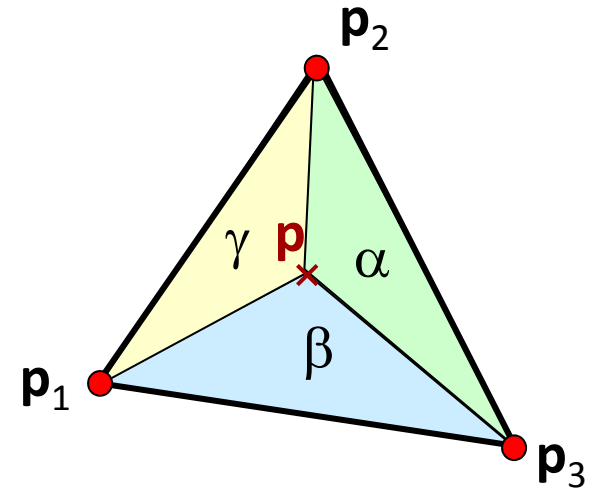
Barycentric combinations of 3 points

$$\mathbf{p} = \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \gamma \mathbf{p}_3, \text{ with } : \alpha + \beta + \gamma = 1$$

$$\gamma = 1 - \alpha - \beta$$

- Area formulation:

$$\alpha = \frac{\text{area}(\Delta(\mathbf{p}_2, \mathbf{p}_3, \mathbf{p}))}{\text{area}(\Delta(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3))}, \beta = \frac{\text{area}(\Delta(\mathbf{p}_1, \mathbf{p}_3, \mathbf{p}))}{\text{area}(\Delta(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3))}, \gamma = \frac{\text{area}(\Delta(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}))}{\text{area}(\Delta(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3))}$$



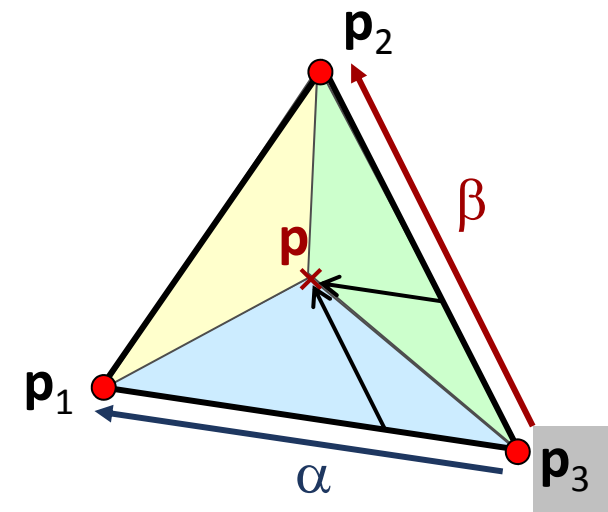
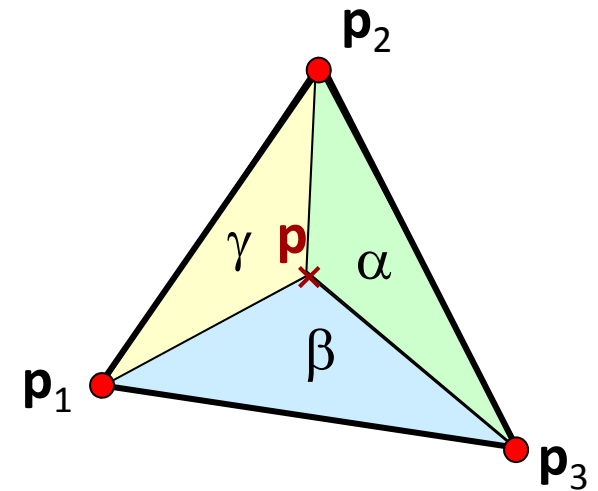


# Barycentric Coordinates

## Barycentric Coordinates:

- Linear formulation:

$$\begin{aligned}\mathbf{p} &= \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \gamma \mathbf{p}_3 \\ &= \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + (1 - \alpha - \beta) \mathbf{p}_3 \\ &= \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \mathbf{p}_3 - \alpha \mathbf{p}_3 - \beta \mathbf{p}_3 \\ &= \mathbf{p}_3 + \alpha(\mathbf{p}_1 - \mathbf{p}_3) + \beta(\mathbf{p}_2 - \mathbf{p}_3)\end{aligned}$$



# Bezier Triangles: Overview

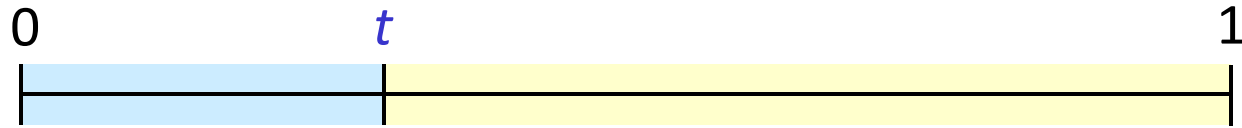
## Bezier Triangles: Main Ideas

- Use 3D points as inputs to the blossoms
- These are Barycentric coordinates of a parameter triangle  $\{\mathbf{a}, \mathbf{b}, \mathbf{c}\}$
- Use 3D points as outputs
- Form control points by multiplying parameter points, just as in the curve case:  $\mathbf{p}(\underbrace{\mathbf{a}, \dots, \mathbf{a}}_i, \underbrace{\mathbf{b}, \dots, \mathbf{b}}_j, \underbrace{\mathbf{c}, \dots, \mathbf{c}}_k)$
- De Casteljau Algorithm: Compute polynomial values  $\mathbf{p}(\mathbf{x}, \dots, \mathbf{x})$  by barycentric interpolation

# Plugging in the Barycentric Coord's

**Analogy:** 2D Curves in barycentric coordinates

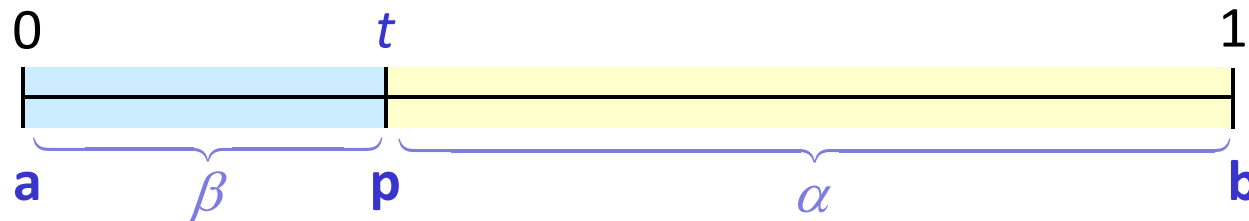
- Barycentric coordinates for 2D curves:



# Plugging in the Barycentric Coord's

## Analogy: 2D Curves in barycentric coordinates

- Barycentric coordinates for 2D curves:



- $\mathbf{p} = \alpha \mathbf{a} + \beta \mathbf{b}, \quad \alpha + \beta = 1$

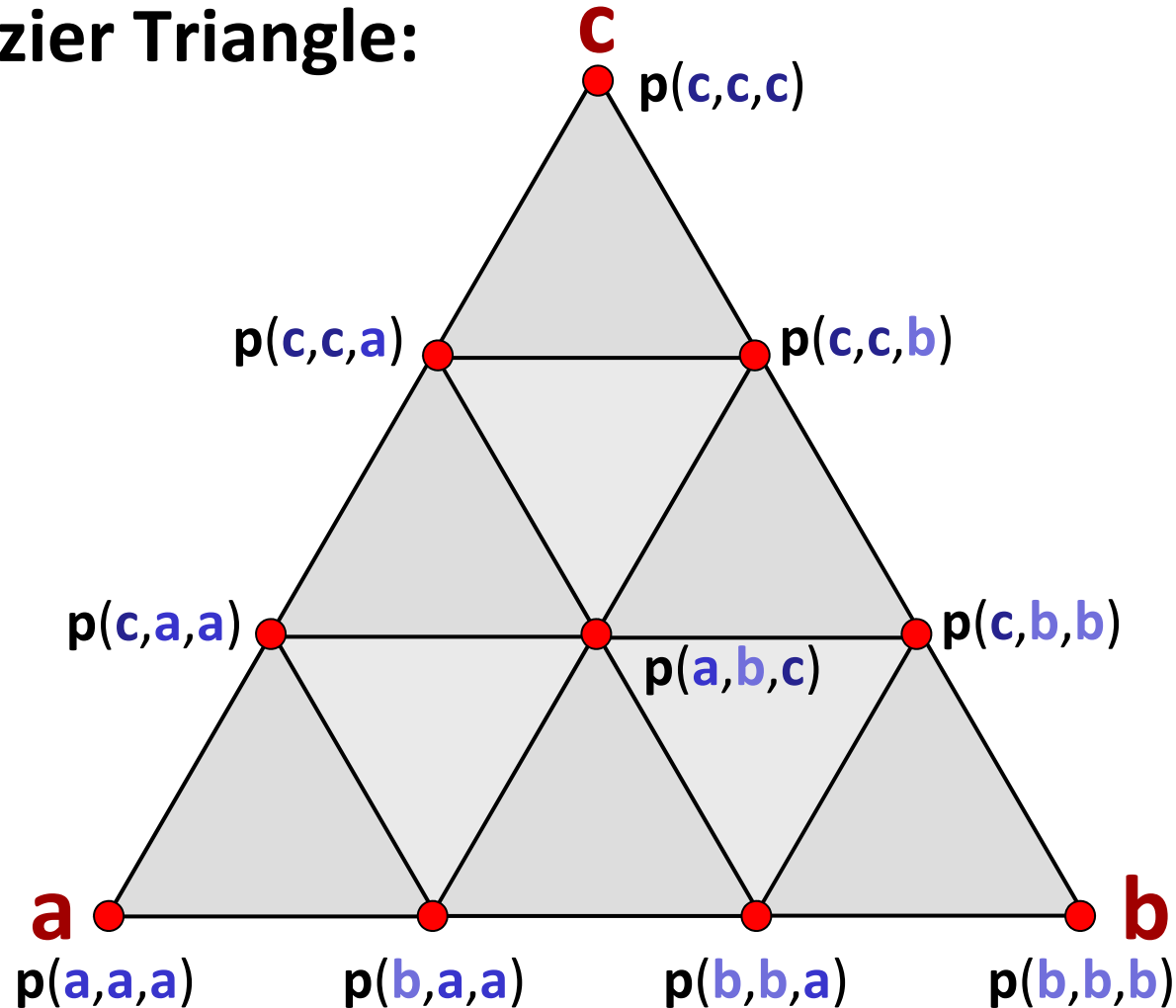
- Bezier splines:

$$\mathbf{F}(t) = \sum_{i=0}^d \binom{d}{i} (1-t)^i t^{d-i} \mathbf{f}(\underbrace{\mathbf{a}, \dots, \mathbf{a}}_i, \underbrace{\mathbf{b}, \dots, \mathbf{b}}_{d-i}) \quad (\text{standard form})$$

$$\mathbf{F}(\mathbf{p}) = \sum_{\substack{i+j=d \\ i \geq 0, j \geq 0}} \frac{d!}{i! j!} \alpha^i \beta^j \mathbf{f}(\underbrace{\mathbf{a}, \dots, \mathbf{a}}_i, \underbrace{\mathbf{b}, \dots, \mathbf{b}}_j) \quad (\text{barycentric form})$$

# Example

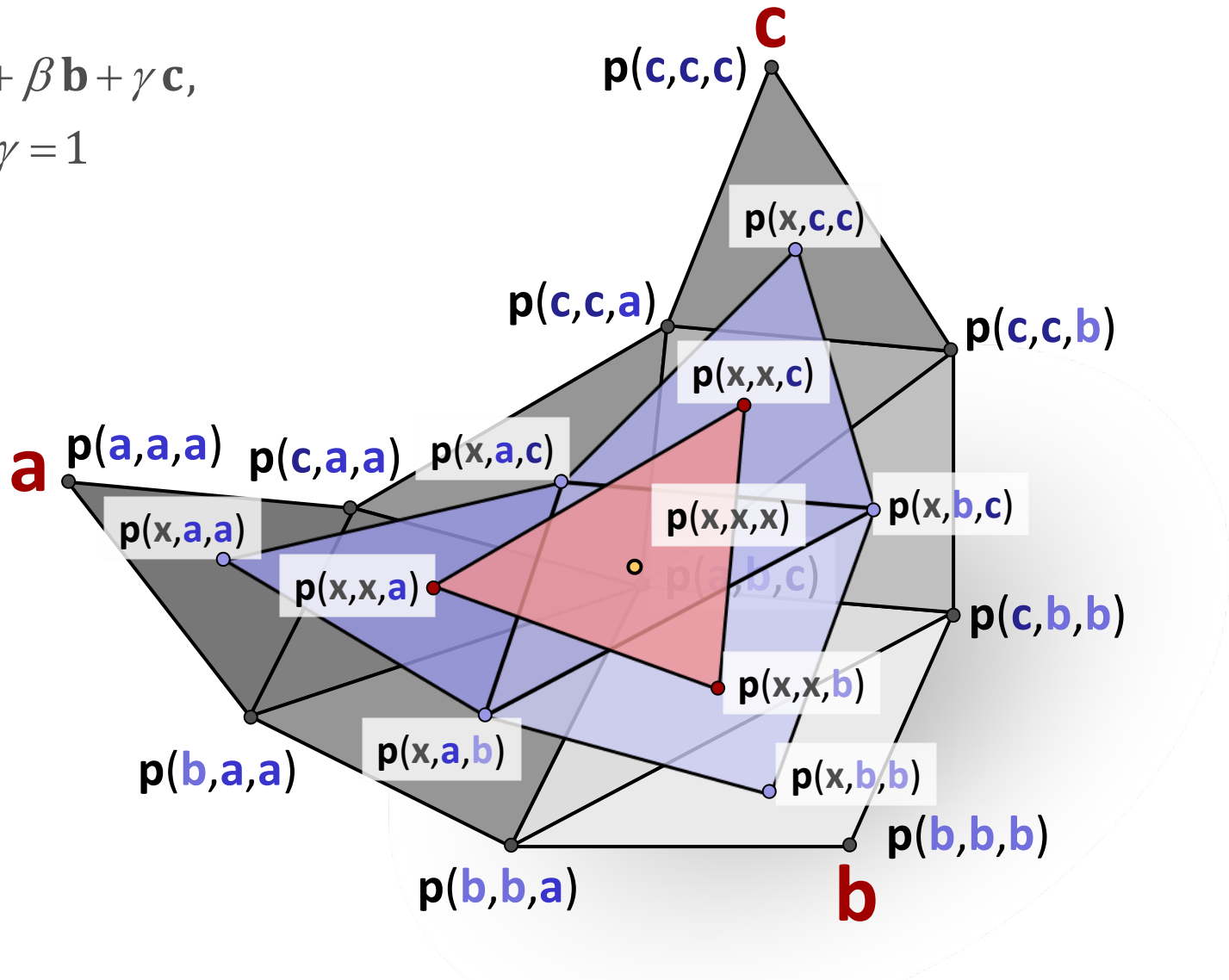
Cubic Bezier Triangle:



# De Casteljau Algorithm

$$\mathbf{x} = \alpha \mathbf{a} + \beta \mathbf{b} + \gamma \mathbf{c},$$

$$\alpha + \beta + \gamma = 1$$



# Bernstein Form

Writing this recursion out, we obtain:

- $$F(\mathbf{x}) = \sum_{\substack{i+j+k=d \\ i,j,k \geq 0}} \frac{d!}{i!j!k!} \alpha^i \beta^j \gamma^k \mathbf{f}(\underbrace{a, \dots, a}_i, \underbrace{b, \dots, b}_j, \underbrace{c, \dots, c}_k)$$

$$\mathbf{x} = \alpha \mathbf{a} + \beta \mathbf{b} + \gamma \mathbf{c},$$

$$\alpha + \beta + \gamma = 1$$

- This is the *Bernstein form* of a Bezier triangle surface
- (Proof by induction)

# Continuity

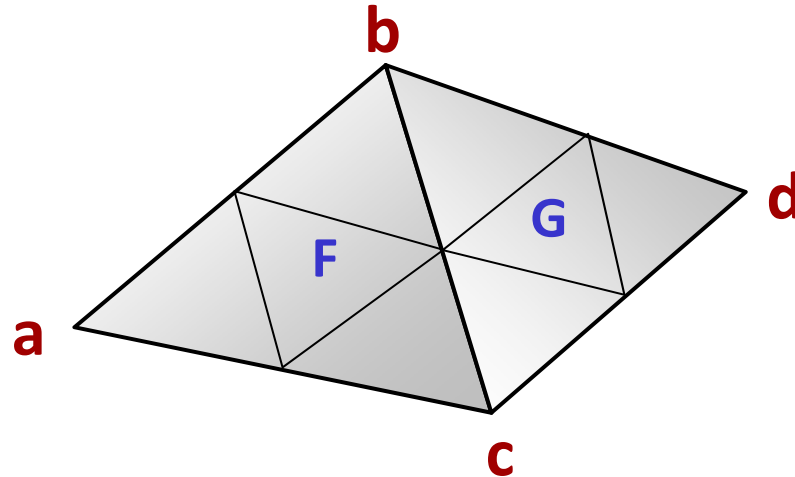
## **We need to assemble Bezier triangles continuously:**

- What are the conditions for  $C^0$ ,  $C^1$  continuity?
- As an example, we will look at the quadratic case...
- (Try the cubic case as an exercise)



# Continuity

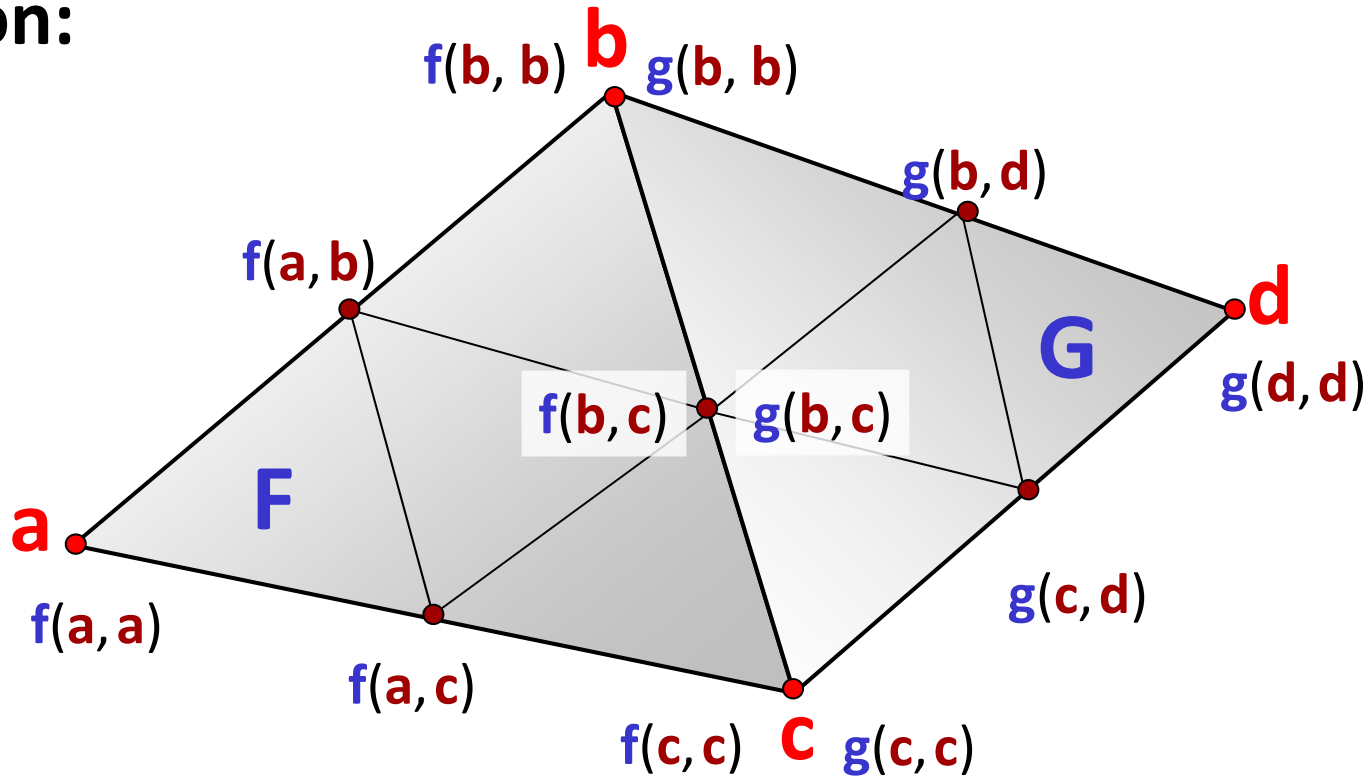
Situation:



- Two Bezier triangles meet along a common edge.
  - Parametrization:  $T_1 = \{a, b, c\}$ ,  $T_2 = \{c, b, d\}$
  - Polynomial surfaces  $F(T_1)$ ,  $G(T_2)$
  - Control points:
    - $F(T_1)$ :  $f(a, a)$ ,  $f(a, b)$ ,  $f(b, b)$ ,  $f(a, c)$ ,  $f(c, c)$ ,  $f(b, c)$
    - $G(T_2)$ :  $g(d, d)$ ,  $g(d, b)$ ,  $g(b, b)$ ,  $g(d, c)$ ,  $g(c, c)$ ,  $g(b, c)$

# Continuity

Situation:



# Continuity

## $C^0$ Continuity:

- The points on the boundary have to agree:

$$f(\mathbf{b}, \mathbf{b}) = g(\mathbf{b}, \mathbf{b})$$

$$f(\mathbf{b}, \mathbf{c}) = g(\mathbf{b}, \mathbf{c})$$

$$f(\mathbf{c}, \mathbf{c}) = g(\mathbf{c}, \mathbf{c})$$

- Proof: Let  $\mathbf{x} := \beta \mathbf{b} + \gamma \mathbf{c}$ ,  $\beta + \gamma = 1$

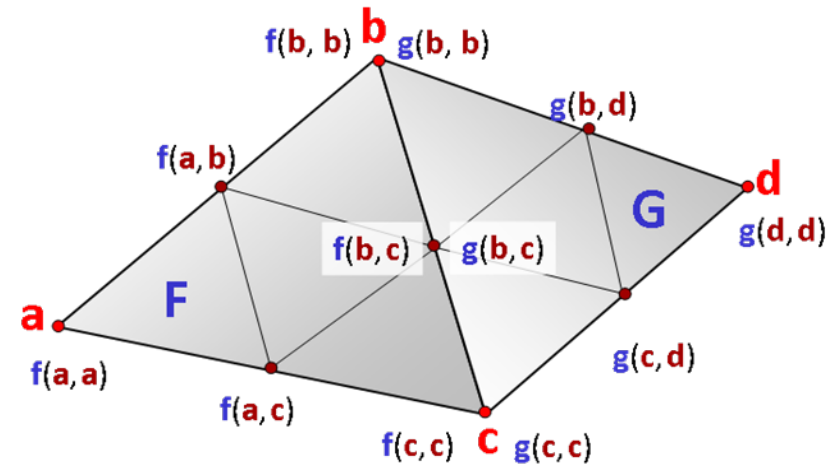
$$f(\mathbf{x}, \mathbf{x}) = \beta f(\mathbf{b}, \mathbf{x}) + \gamma f(\mathbf{c}, \mathbf{x})$$

$$= \beta^2 f(\mathbf{b}, \mathbf{b}) + 2\beta\gamma f(\mathbf{b}, \mathbf{c}) + \gamma^2 f(\mathbf{c}, \mathbf{c})$$

$$\begin{array}{ccc} \parallel & \parallel & \parallel \\ g(\mathbf{b}, \mathbf{b}) & g(\mathbf{b}, \mathbf{c}) & g(\mathbf{c}, \mathbf{c}) \end{array}$$

$$= \beta^2 g(\mathbf{b}, \mathbf{b}) + 2\beta\gamma g(\mathbf{b}, \mathbf{c}) + \gamma^2 g(\mathbf{c}, \mathbf{c})$$

$$= \beta g(\mathbf{b}, \mathbf{x}) + \gamma g(\mathbf{c}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$$



# Continuity

## $C^1$ Continuity:

- We need  $C^0$  continuity.
- In addition:

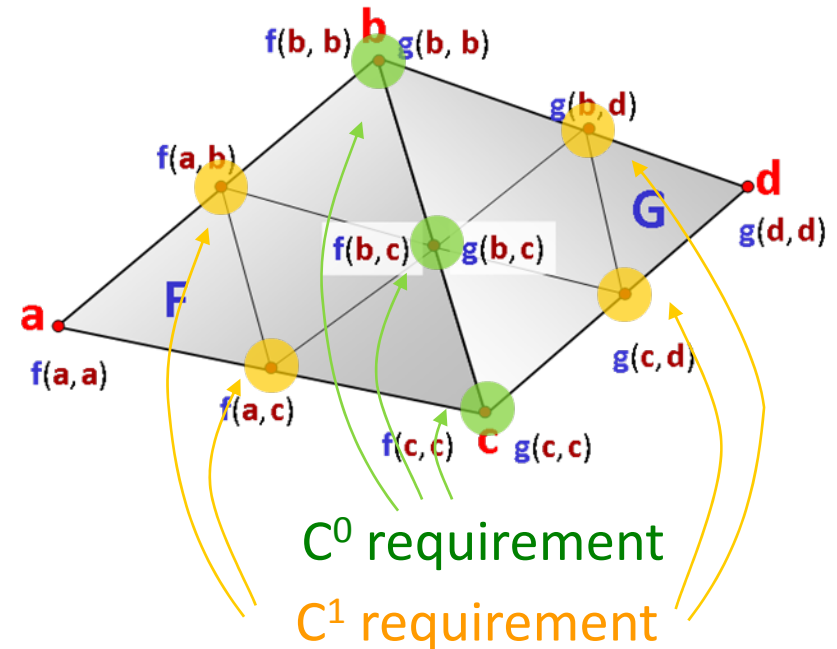
The Blossoms have  
have to agree partially:

$$f(\mathbf{a}, \mathbf{b}) = g(\mathbf{a}, \mathbf{b})$$

$$f(\mathbf{b}, \mathbf{d}) = g(\mathbf{b}, \mathbf{d})$$

$$f(\mathbf{a}, \mathbf{c}) = g(\mathbf{a}, \mathbf{c})$$

$$f(\mathbf{c}, \mathbf{d}) = g(\mathbf{c}, \mathbf{d})$$



# Continuity

## C<sup>1</sup> Continuity: Proof

- Derivatives:

$$\frac{\partial}{\partial \hat{\mathbf{d}}} \mathbf{F}(\mathbf{x}) \Big|_{\mathbf{x}=\mathbf{p}} = \mathbf{f}(\mathbf{p}, \hat{\mathbf{d}})$$

(similar to the curve case)

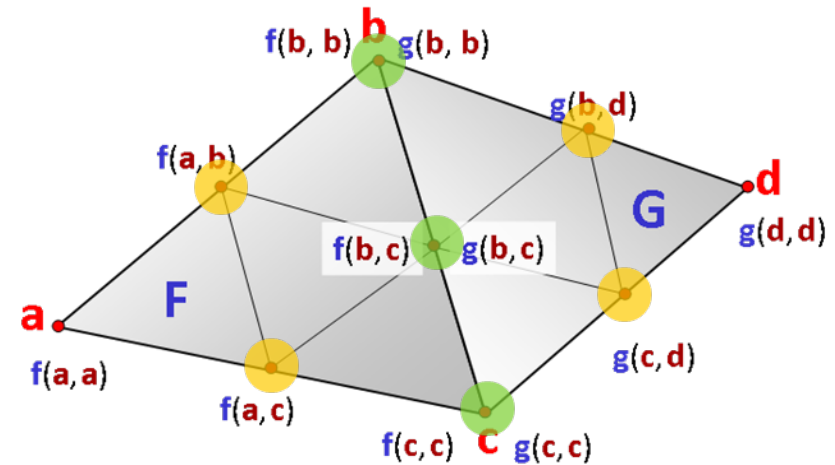
- C<sup>1</sup>-Continuity:

$$\forall \mathbf{x} \in \mathbb{R}^3 : \mathbf{f}(\mathbf{p}, \mathbf{x}) = \mathbf{g}(\mathbf{p}, \mathbf{x})$$

- We have to show:

$$\forall \mathbf{x} \in \mathbb{R}^3 : \begin{cases} \mathbf{f}(\mathbf{b}, \mathbf{x}) = \mathbf{g}(\mathbf{b}, \mathbf{x}) \\ \mathbf{f}(\mathbf{c}, \mathbf{x}) = \mathbf{g}(\mathbf{c}, \mathbf{x}) \end{cases}$$

$\Rightarrow$  C<sup>1</sup> continuity follows for all boundary points (by interp.)



# Continuity

## $C^1$ Continuity: Proof

- So we have to show:

$$\forall \mathbf{x} \in \mathbb{R}^3 : \begin{cases} \mathbf{f}(\mathbf{b}, \mathbf{x}) = \mathbf{g}(\mathbf{b}, \mathbf{x}) \\ \mathbf{f}(\mathbf{c}, \mathbf{x}) = \mathbf{g}(\mathbf{c}, \mathbf{x}) \end{cases}$$

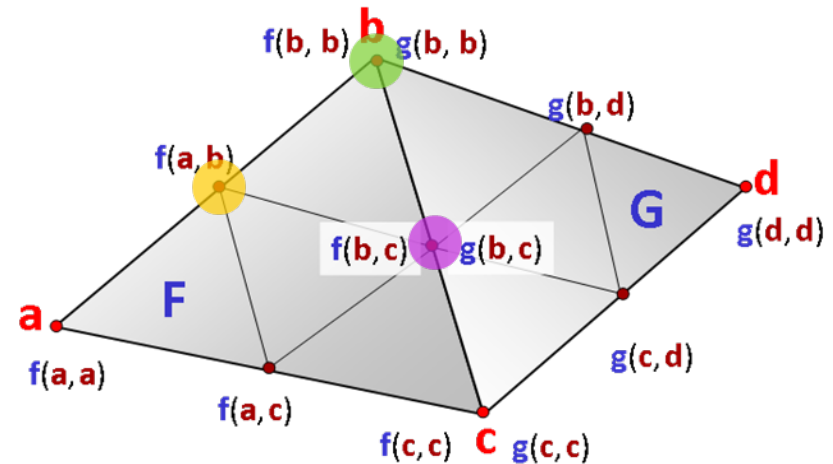
- Proof:

Write  $\mathbf{x} = \alpha \mathbf{f}(\mathbf{a}, \mathbf{b}) + \beta \mathbf{f}(\mathbf{b}, \mathbf{b}) + \gamma \mathbf{f}(\mathbf{b}, \mathbf{c})$  (coordinate system)

$$\mathbf{f}(\mathbf{b}, \mathbf{x}) = \alpha \mathbf{f}(\mathbf{a}, \mathbf{b}) + \beta \mathbf{f}(\mathbf{b}, \mathbf{b}) + \gamma \mathbf{f}(\mathbf{b}, \mathbf{c})$$

$$\mathbf{g}(\mathbf{b}, \mathbf{x}) = \alpha \mathbf{g}(\mathbf{a}, \mathbf{b}) + \beta \mathbf{g}(\mathbf{b}, \mathbf{b}) + \gamma \mathbf{g}(\mathbf{b}, \mathbf{c}) \quad C^0$$

$$\begin{aligned} \mathbf{f}(\mathbf{b}, \mathbf{x}) = \mathbf{g}(\mathbf{b}, \mathbf{x}) &\Leftrightarrow \alpha \mathbf{f}(\mathbf{a}, \mathbf{b}) + \beta \mathbf{f}(\mathbf{b}, \mathbf{b}) + \gamma \mathbf{f}(\mathbf{b}, \mathbf{c}) \\ &= \alpha \mathbf{g}(\mathbf{a}, \mathbf{b}) + \beta \mathbf{g}(\mathbf{b}, \mathbf{b}) + \gamma \mathbf{g}(\mathbf{b}, \mathbf{c}) \quad C^0 \\ &\Leftrightarrow \mathbf{f}(\mathbf{a}, \mathbf{b}) = \mathbf{g}(\mathbf{a}, \mathbf{b}) \end{aligned}$$



(the same for the other three conditions)

# Continuity

## So what does this mean?

- The Blossoms agree partially:

$$f(a, b) = g(a, b)$$

$$f(b, d) = g(b, d)$$

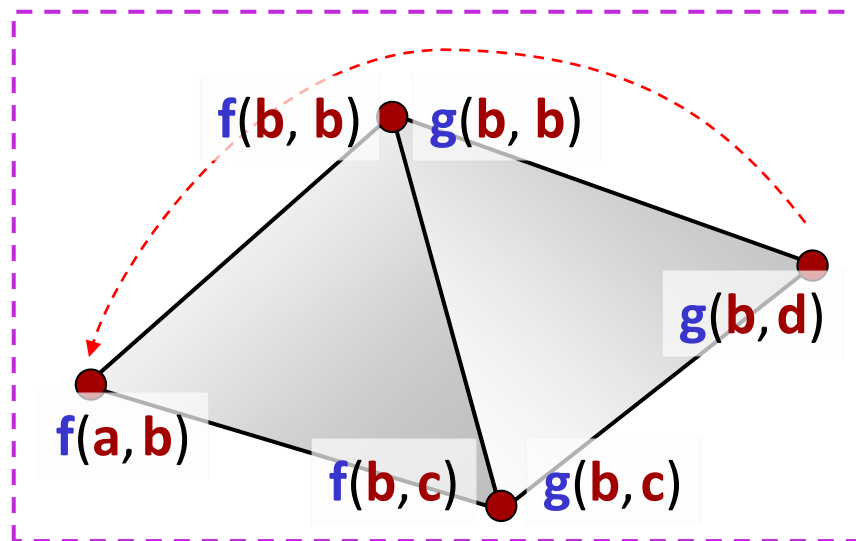
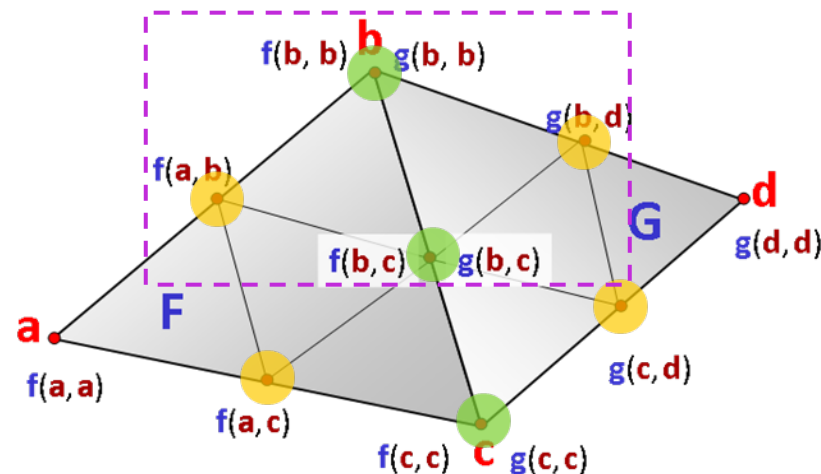
$$f(a, c) = g(a, c)$$

$$f(c, d) = g(c, d)$$

- The points must be coplanar (with edge points):

$$f(a, b), g(b, d), g(b, b), g(b, c)$$

- The points in **F** must be affine images of the points in **G**



# Rendering



# Rendering trimmed NURBS

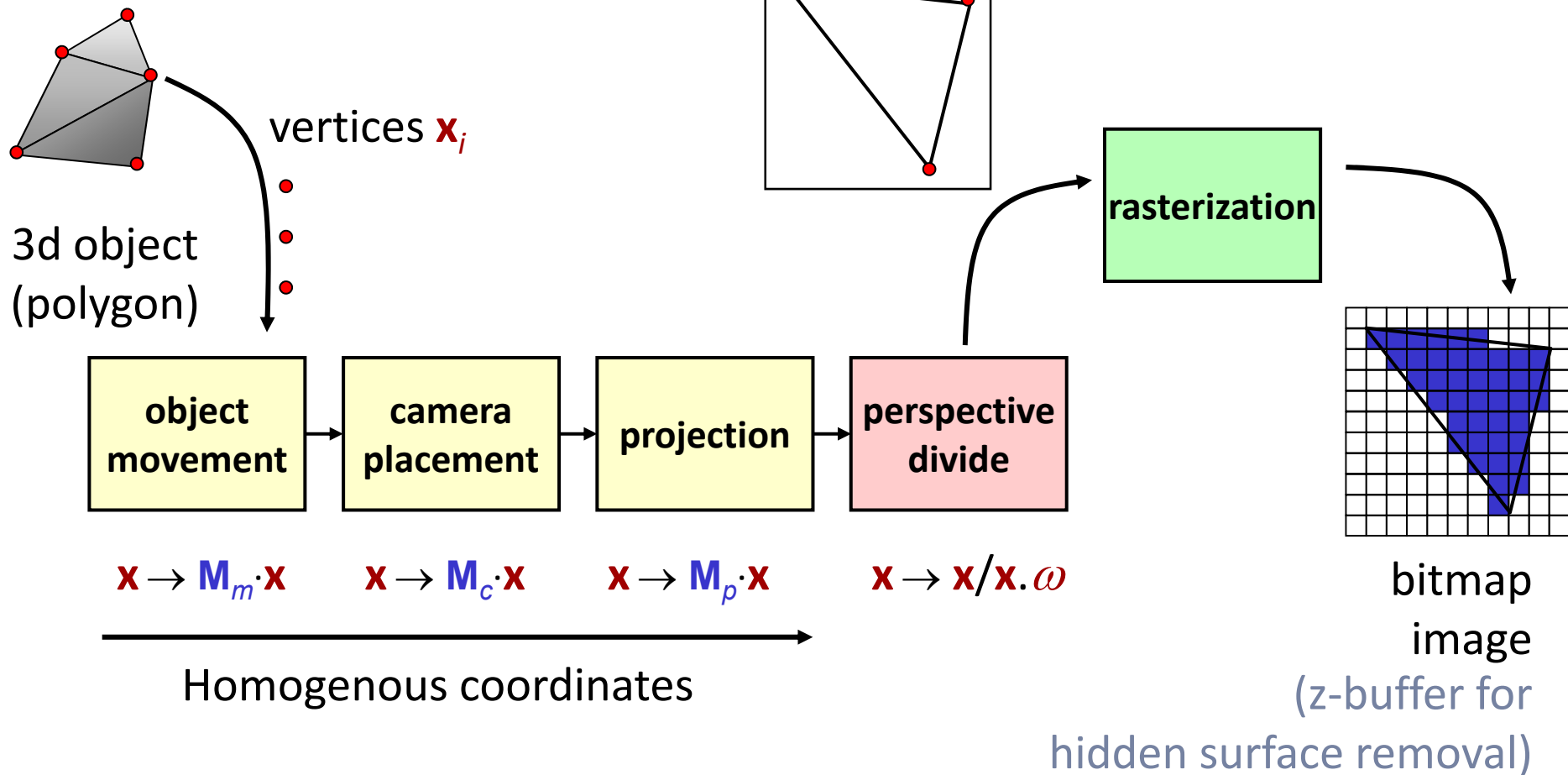
**How can we render trimmed NURBS?**

**We will look at three variants:**

- Rasterization
- Raytracing
- Hardware-friendly rasterization algorithm

# Rasterization

## Basic pipeline:



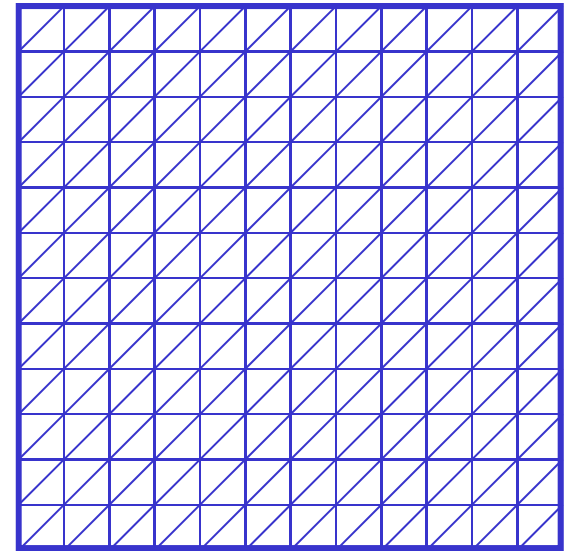
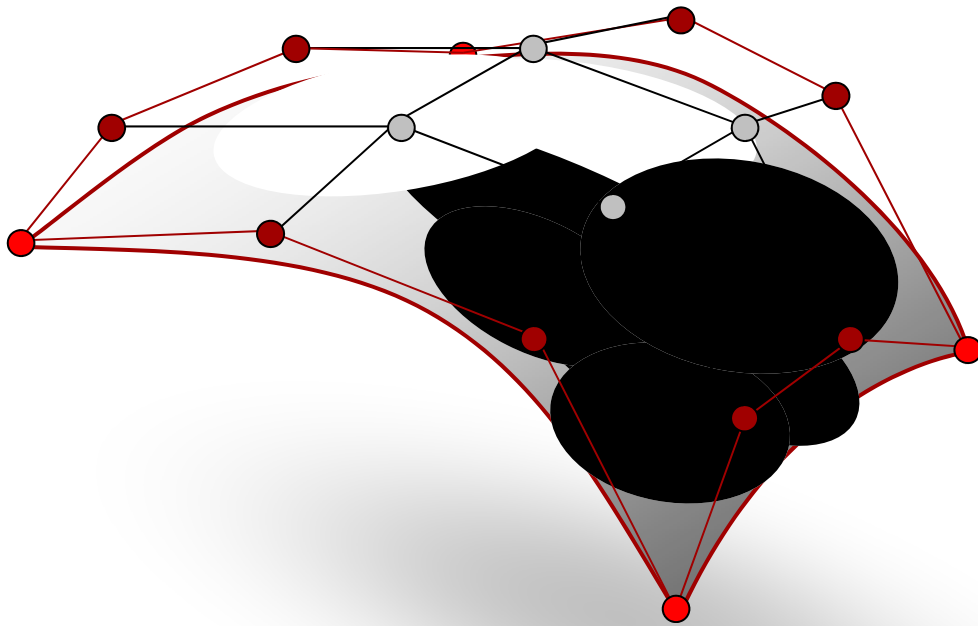
# Rasterization Pipeline

## Basically:

- We can draw triangles
- Very efficient due to hardware support  
(standard GPU: 100 M triangles/sec, 1000 M pixels/sec)
- We need to convert our surfaces into triangles  
("tessellation")
- Nowadays: We can afford high resolution tessellations

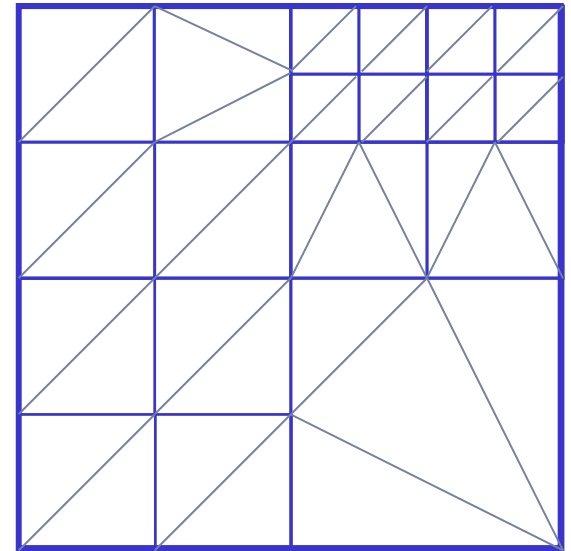
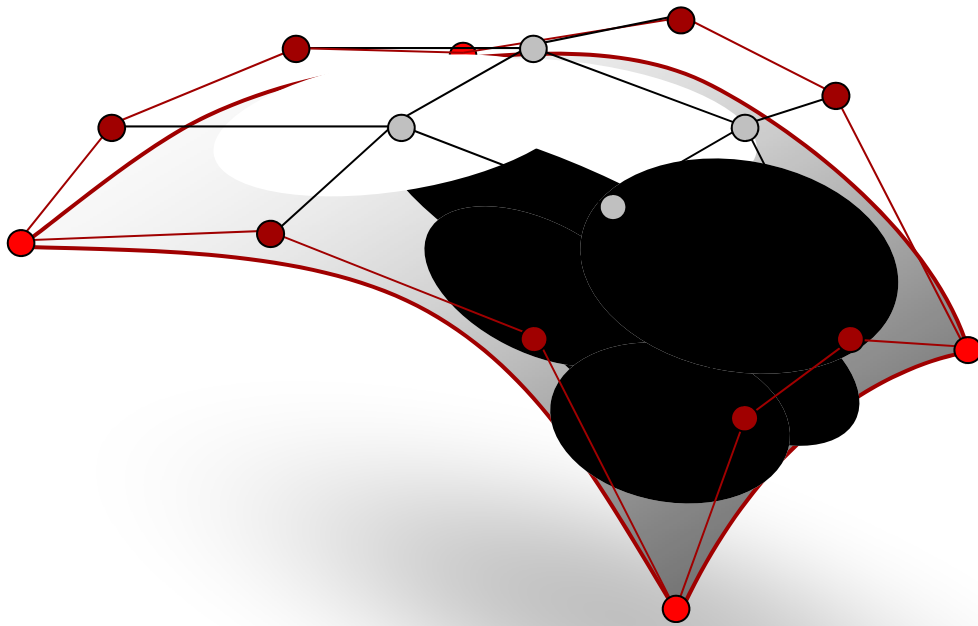
# Simple Idea

**Simplest solution:** Uniform tessellation



# Fancier Idea

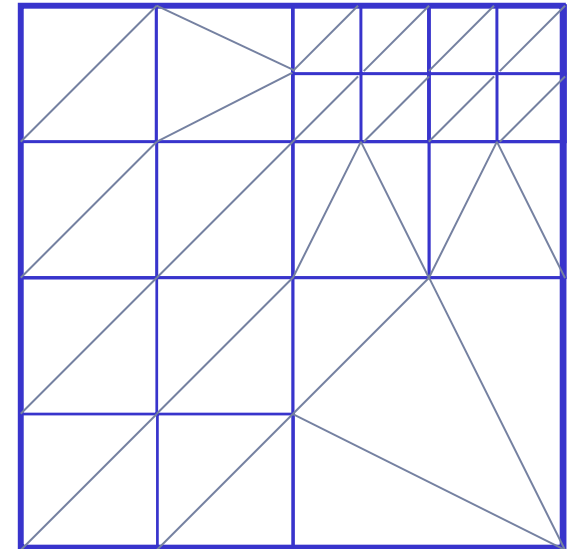
**Better solution:** Adaptive tessellation



# Adaptive Tesselation

## Adaptive Tesselation:

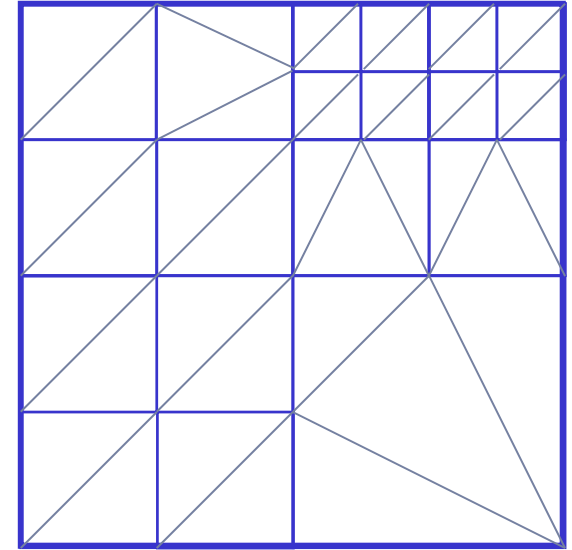
- Subdivide parameter domain recursively
- Divide rectangle into four smaller parts (“Quadtree”)
- Possible stopping criterion:
  - Distance between planar faces and surface
  - Approximately: planarity of control points



# Adaptive Tesselation

## Adaptive Tesselation:

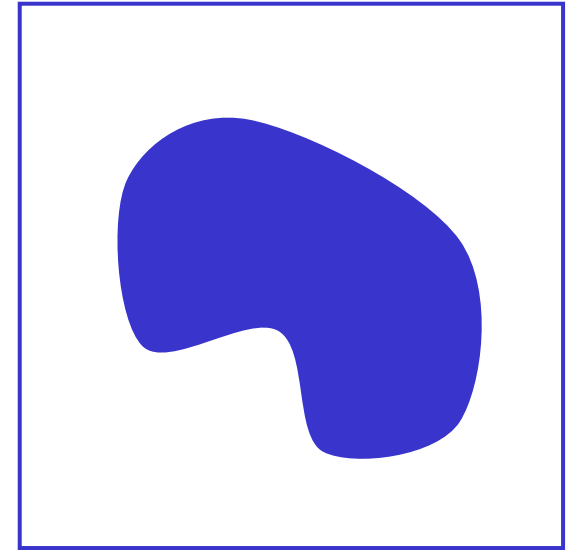
- Balanced Quadtree:
  - Make sure that the subdivision level of adjacent cells does not differ by more than one level
- Divide cells into triangles
- Look at direct neighbors to create a closed mesh
- Only  $2^4 = 16$  cases



# So what about the curves?

## Remaining problem:

- Need to render trimmed patches
- Super-simple solution (“cheating”):
  - add a texture map, remove “white” pixels with (do not draw empty space)
  - Supported in hardware (“alpha test”)
  - But this looks ugly
  - And does not help in geometric computation (if we need a triangulation of the trimmed object for further processing)

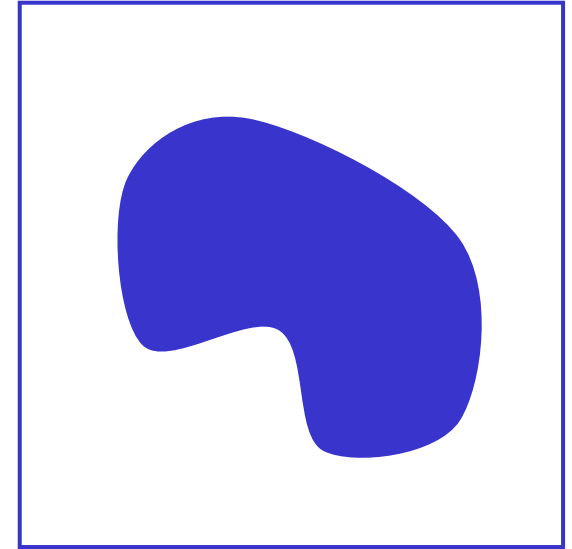




# So what about the curves?

## Second try:

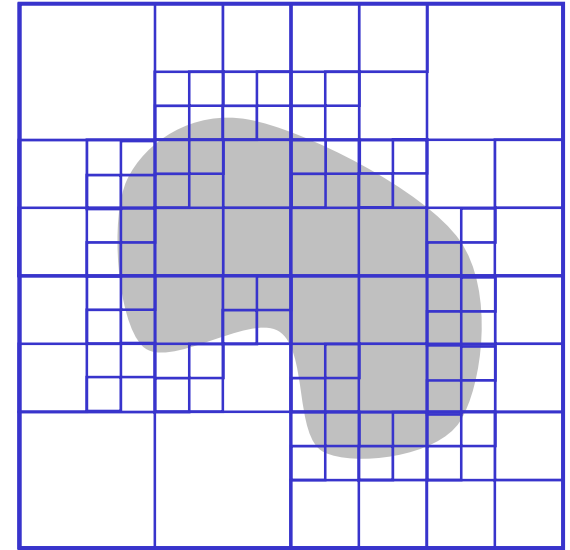
- We have to tessellate the trimming area in the domain
- Need to place triangles in the domain that approximate the shape
- Curve tessellation problem
  - Classic computational geometry problem
  - Several solutions
  - E.g. constrained Delaunay triangulation
- Easy to implement: Quadtree triangulation method



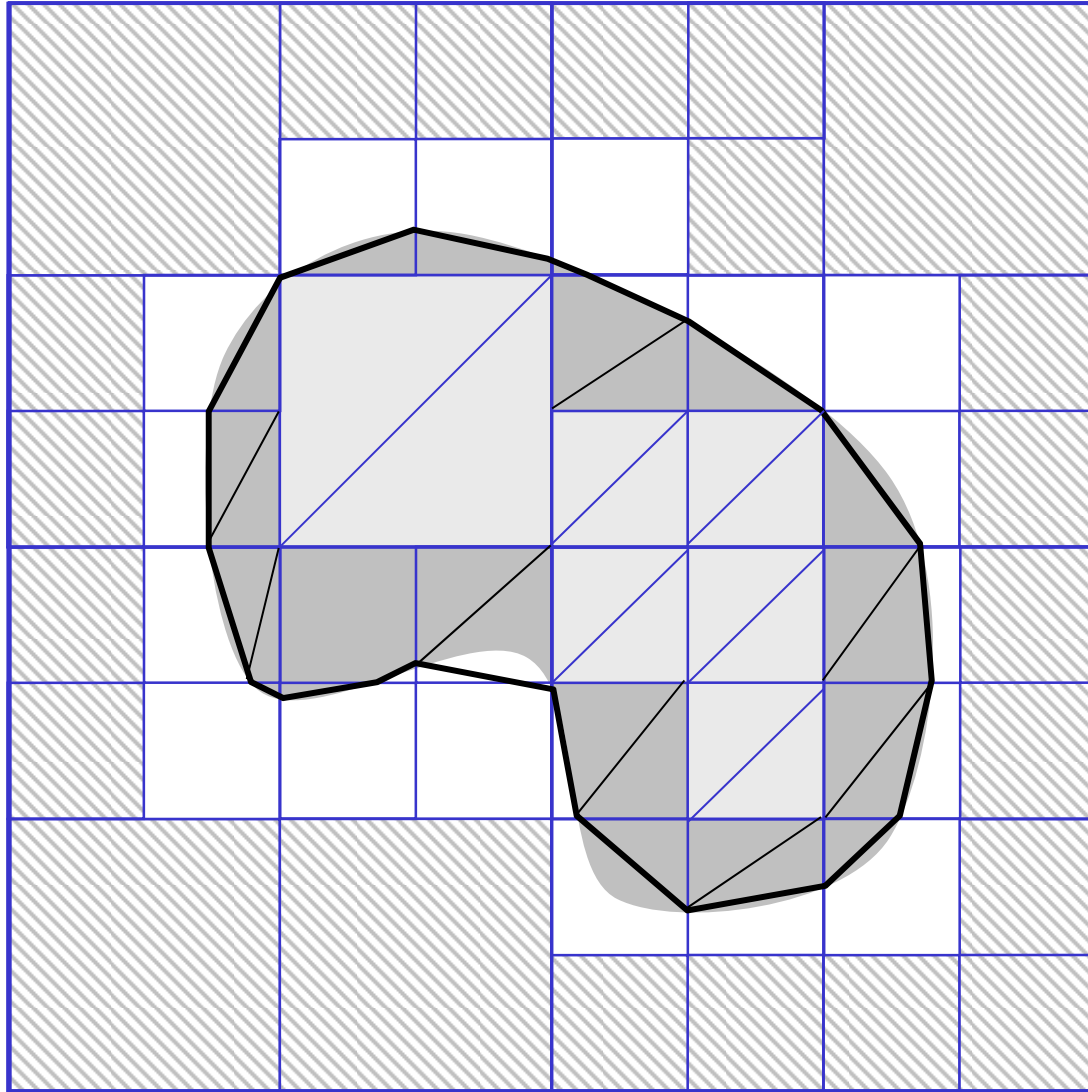
# Quadtree triangulation

## Quadtree triangulation:

- Subdivide recursively as before
- New stopping criterion
  - If the bounding box intersects the area:
    - Do not stop until surface is well approximated
    - **And:** No boundary curve inside, or the boundary curves intersects exactly twice
    - Limit recursion depth to avoid trouble at degeneracies
  - If the bounding box covers empty space:
    - Stop immediately



# Quadtree triangulation



# Quadtree triangulation

## Tessellation Algorithm:

- Compute balanced quadtree
- Stop when accuracy is met and only two curve intersections are in each box
- Tessellate interior the same way as before
- Tessellate intersections with fixed scheme (at most two triangles)
- Drop exterior boxes

## Interior holes:

- Use ray-based inside/outside test

# Hardware friendly version

## Problem:

- The adaptive tessellation is computationally costly
- Algorithm with complex data structures and pointers, not easy to implement on special purpose hardware
- Even a standard CPU needs its time

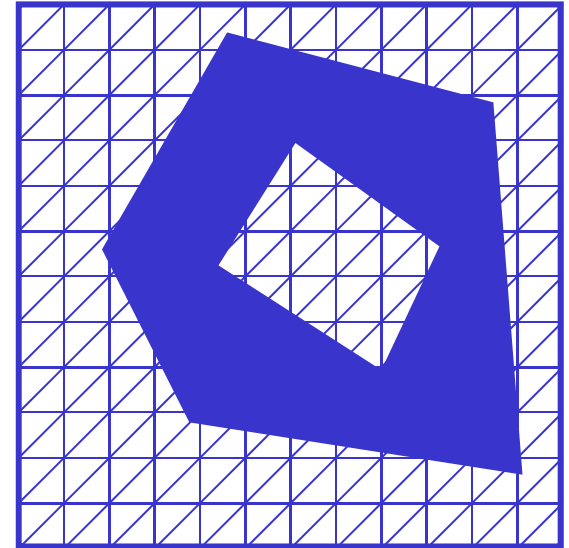
## Hardware friendly algorithm: [Guthe et al. 2005]

- Basic idea: graphics hardware is so fast, we can waste a few triangles
- Runs completely on programmable graphics hardware
- We will discuss a simplified version (no gory GPU details)

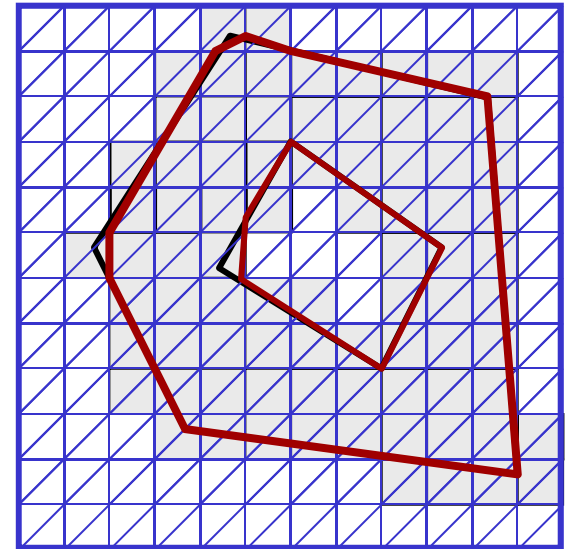
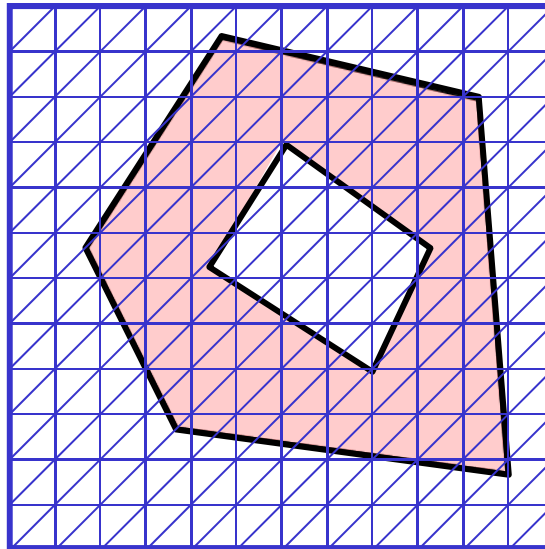
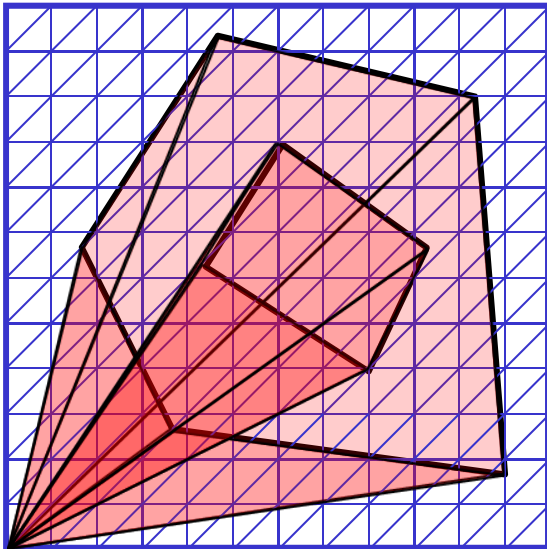
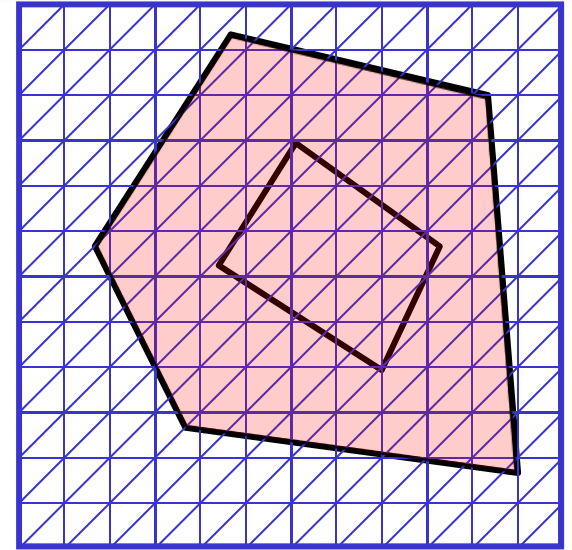
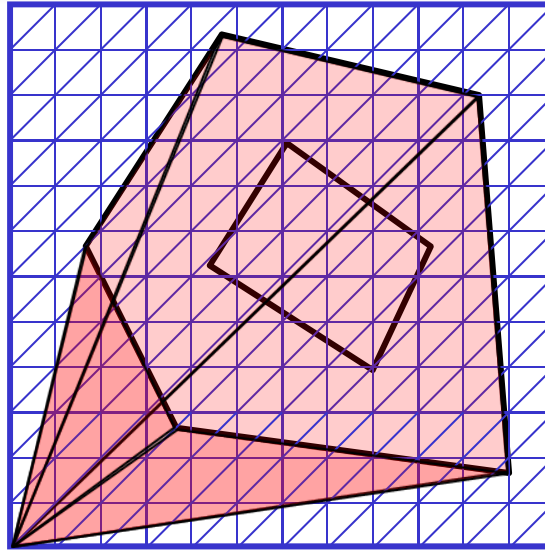
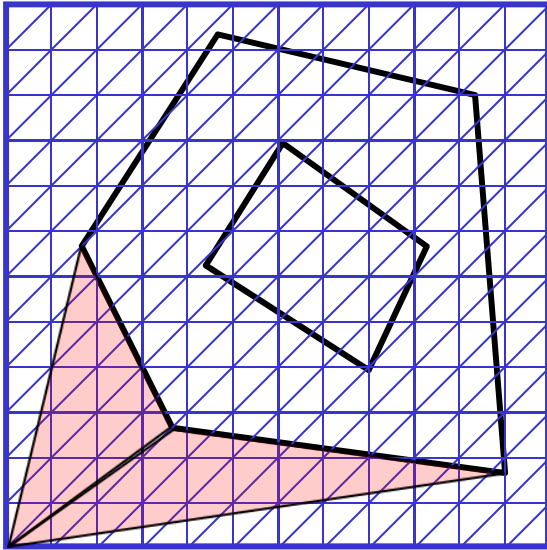
# Guthe's Algorithm

## Basic Idea:

- Use a uniform grid
- Represent each quad as a pixel
- Now render sequence of triangles along the curve, connected with one corner, in XOR mode



# Guthe's Algorithm



# Hardware friendly algorithm

## After XOR-polygon drawing:

- Knowing the pixels that cover the domain, each one can be easily tessellated
- The spline surface is evaluated on the graphics hardware (programmable shaders)
- This algorithm is much faster than standard techniques
- In case the accuracy is not sufficient, a hierarchical refinement “on demand” is implemented
- Increases the resolution in surface parts close to the viewer



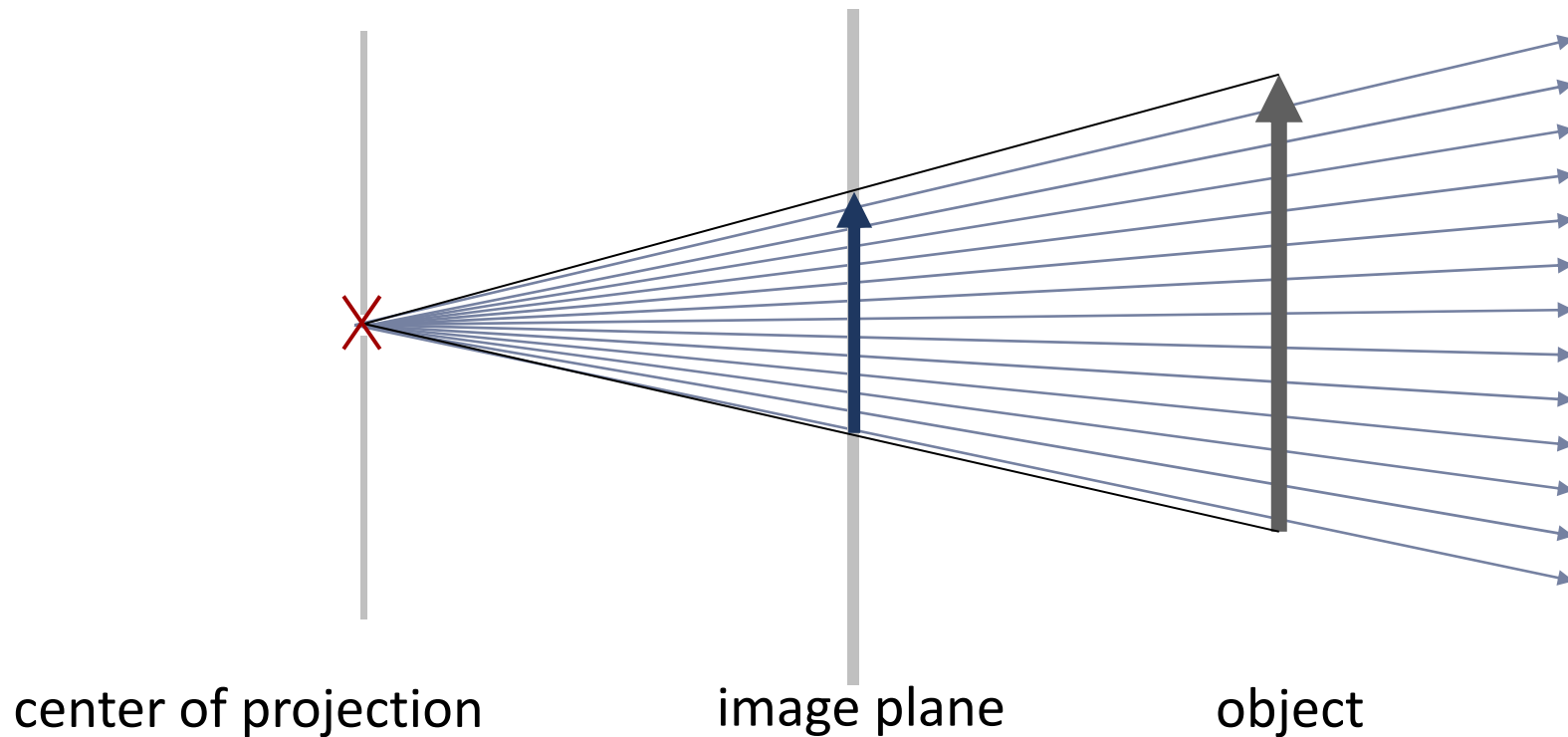
# Raytracing

## How can we raytrace NURBS patches?

### Raytracing algorithm:

- Shoot a ray through each pixel of the image
- Test objects in the scene for intersection
- Display closest object
- For shading the object, further rays can be sent recursively
  - Shadow rays to the light source(s) – if blocked, object is in shadow
  - Reflected / refracted rays for mirroring / refractions

# Raytracing



# Intersection Problem

## Intersection Problem

- Rendering with raytracing reduces to determine whether a ray intersects a spline patch
- Non-linear system of equations:

$$\begin{aligned} \mathbf{f}(u,v) &= \sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \mathbf{p}_{i,j} \\ \mathbf{r}(t) &= t\mathbf{a} + \mathbf{b} \end{aligned} \quad \left\{ \begin{aligned} &\underbrace{\sum_{i=0}^d \sum_{j=0}^d B_i^{(d)}(u) B_j^{(d)}(v) \mathbf{p}_{i,j} - t\mathbf{a} + \mathbf{b}}_{\mathbf{F}(u,v,t)} = 0 \end{aligned} \right.$$

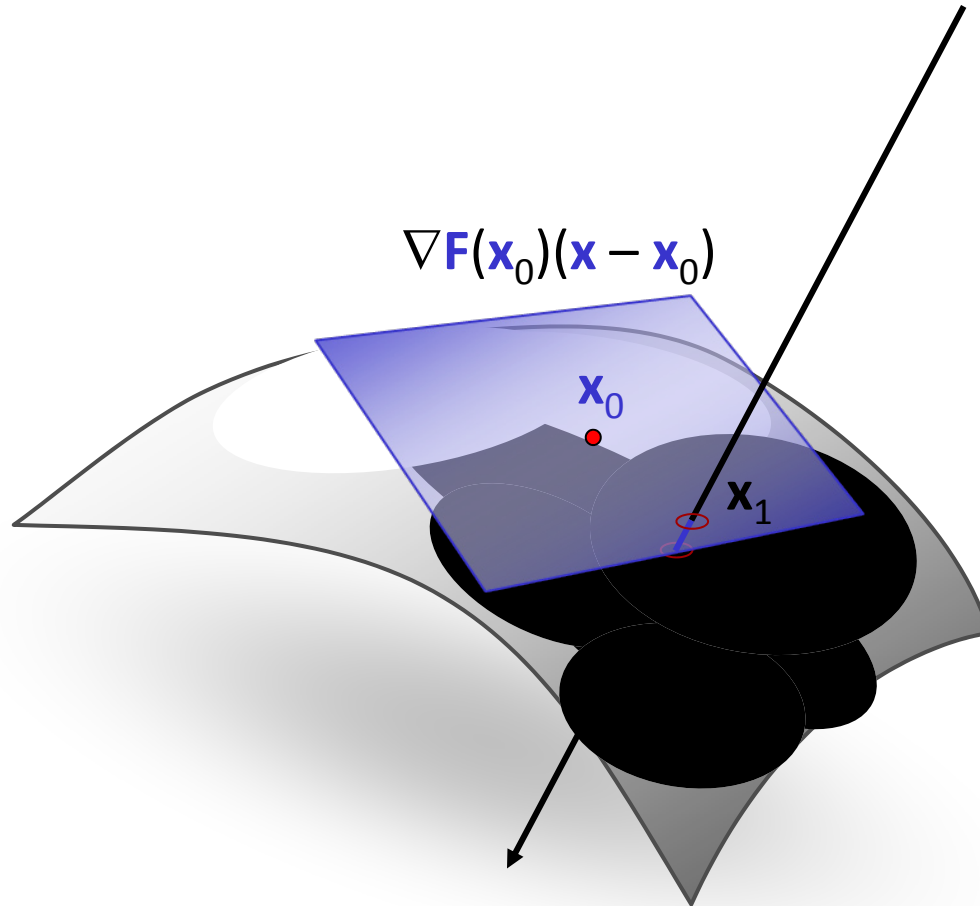
$\mathbf{F}(u,v,t) = 0$   
solve for  $u, v, t$

# Solution Strategies

## Numerical optimization

- No closed form solution
- Therefore: Numerical approach
  - Need a starting value  $\mathbf{x}_0$  (e.g.  $\mathbf{x}_0 = (u, v, t) = (0, 0, 0)$ )
  - Then iteratively improve solution
- Numerical techniques
  - (Gradient decent on squared residue)
  - Newton's method: Linearize problem
    - Compute Jacobian
    - Solve linear system  $\nabla \mathbf{F}(\mathbf{x}_0)(\mathbf{x} - \mathbf{x}_0) + \mathbf{F}(\mathbf{x}_0) = 0$
    - Iterate
  - Newton-like geometric technique

# Newton-like technique



# Problem

## Properties of Newton-based algorithm

- Quite efficient – typically needs only a few iterations
- However: No convergence guarantees
  - In general: does not always converge to the correct solution
- Need good initialization

## Brute-Force approach:

- Restart iteration from a number of starting points on the surface
- But that takes forever to compute

# Alternative

## **Alternative:** Hierarchical subdivision algorithm

- Compute bounding volume of control points (convex hull property)
  - We can use the convex hull
  - Simpler to implement: bounding sphere
- Test for intersection
  - No intersection found → return false, we are done
  - Otherwise continue recursively
- Recursion: subdivide patch into four parts (de Casteljau)
- Call recursive test for all patches
- Always terminate, if precision is sufficient

# Alternative

## **Alternative:** Hierarchical subdivision algorithm

- Guaranteed to converge
- But slower
  - Linear convergence, i.e. number of correct digits in solution increases proportional to #iterations (asymptotical)
  - Newton method typically converges quadratically (number of correct digits increases quadratically)

## **“Best of both worlds”**

- Start with a few iterations of hierarchical subdivision
  - Stopping criterion: Test for “flatness of control points”
- Then use Newton iteration to boost accuracy rapidly