Computer Graphics

Curves and Surfaces

Hermite/Bezier Curves, (B-)Splines, and

NURBS

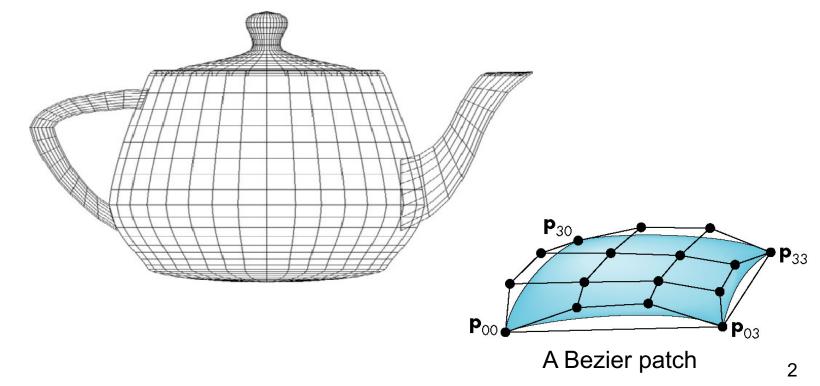
Ulf Assarsson

By

Most of the material is originally made by Edward Angel and is adapted to this course by Ulf Assarsson. Some material is made by Magnus Bondesson

Utah Teapot

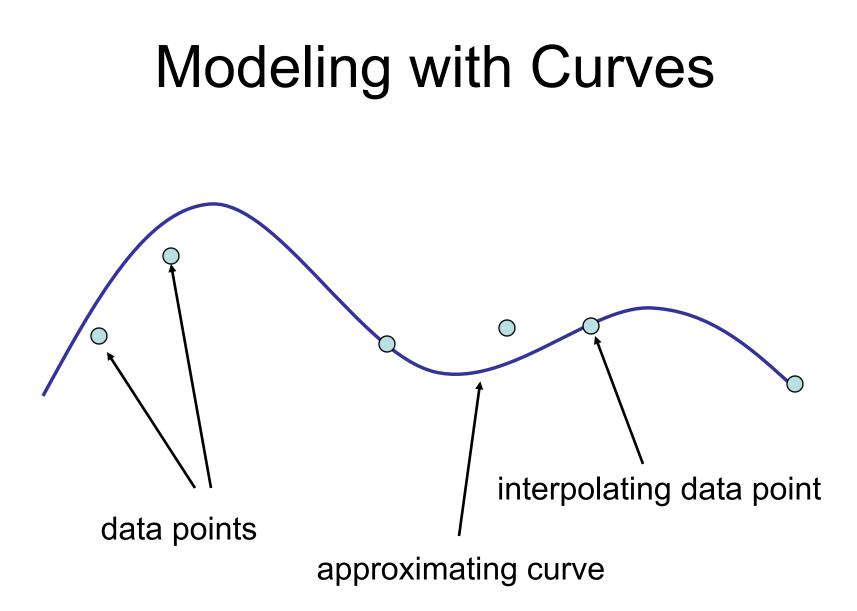
- Most famous data set in computer graphics
- Widely available as a list of 306 3D vertices and the indices that define 32 Bezier patches



Outline

Goal is to explain NURBS curves/surfaces...

- Introduce types of curves and surfaces
 - Explicit not general, easy to compute.
 - Implicit general, non-easy to compute.
 - Parametric general + simple to compute. We choose this.
- A complete curve is split into curve segments, each defined by a cubical polynomial.
 - Introducing Interpolating/Hermite/Bezier curves.
- Adjacent segments should have C² continuity.
 - Leads to B-Splines with a blending function (a spline) per control point
 - Each spline consists of 4 cubical polynomials, forming a bell shape translated along *u*.
 - (Also, four bells will overlap at each point on the complete curve.)
- NURBS a generalization of B-Splines:
 - Control points at non-uniform locations along parameter *u*.
 - Individual weights (i.e., importance) per control point



What Makes a Good Representation?

- There are many ways to represent curves and surfaces
- Want a representation that is
 - -Stable
 - -Smooth
 - -Easy to evaluate
 - –Must we interpolate or can we just come close to data?
 - -Do we need derivatives?

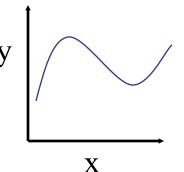
Explicit Representation

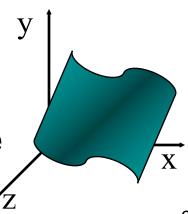
- Most familiar form of curve in 2D
- Cannot represent all curves
 –Vertical lines
 –Circles

y=f(x)

- Extension to 3D
 - -y=f(x), z=g(x) gives a curve

-The form y = f(x,z) defines a surface





Implicit Representation

Two dimensional curve(s)

g(x,y)=0

Much more robust

-Circles $x^2+y^2-r^2=0$

• Three dimensions g(x,y,z)=0 defines a surface

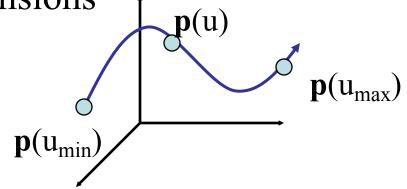
-(we could intersect two surfaces to get a curve)

Parametric Curves

Separate equation for each spatial variable

x=x(u) y=y(u) $p(u)=[x(u), y(u), z(u)]^T$ z=z(u)

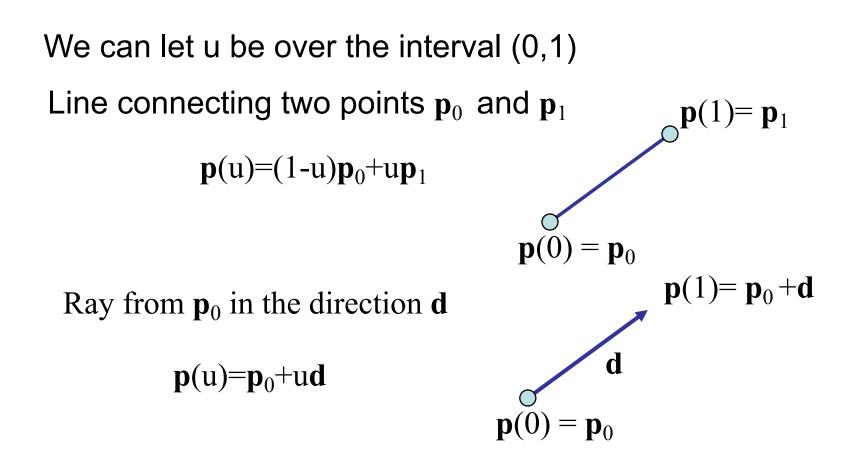
• For $u_{max} \ge u \ge u_{min}$ we trace out a curve in two or three dimensions





- Usually we can select "good" functions
 - not unique for a given spatial curve
 - Approximate or interpolate known data
 - -Want functions which are easy to evaluate
 - Want functions which are easy to differentiate
 - Computation of normals
 - Connecting pieces (segments)
 - -Want functions which are smooth

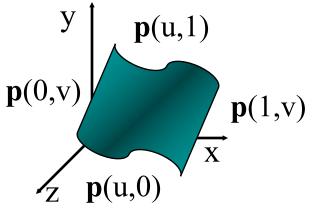
Parametric Lines



Parametric Surfaces

Surfaces require 2 parameters

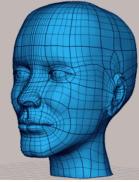
x=x(u,v)y=y(u,v)z=z(u,v)



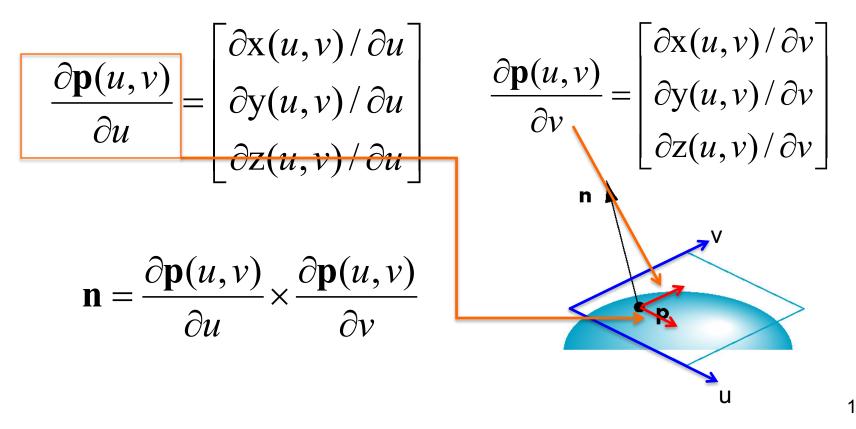
 $\mathbf{p}(u,v) = [x(u,v), y(u,v), z(u,v)]^{T}$

- Want same properties as curves:
 - -Smoothness
 - -Differentiability
 - -Ease of evaluation

Normals

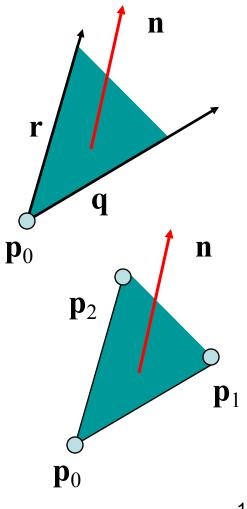


We can differentiate with respect to \mathbf{u} and \mathbf{v} to obtain the normal at any point p



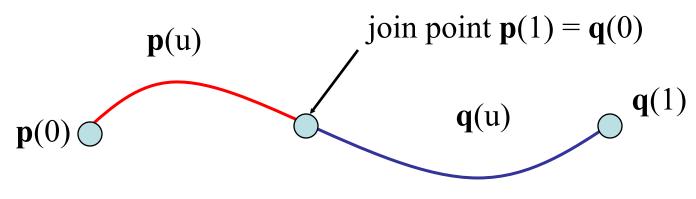
Parametric Planes

point-vector form $\mathbf{p}(\mathbf{u},\mathbf{v})=\mathbf{p}_0+\mathbf{u}\mathbf{q}+\mathbf{v}\mathbf{r}$ $\frac{\partial \mathbf{p}(u,v)}{\partial \mathbf{p}(u,v)} \times \frac{\partial \mathbf{p}(u,v)}{\partial \mathbf{p}(u,v)}$ ∂v ди $\mathbf{n} = \mathbf{q} \mathbf{x} \mathbf{r}$ (three-point form $\mathbf{q} = \mathbf{p}_1 - \mathbf{p}_0$ $\mathbf{r} = \mathbf{p}_2 - \mathbf{p}_0$)



Curve <u>Segments</u>

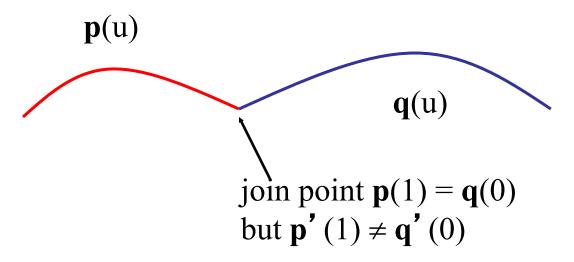
- After normalizing u, each curve is written $\mathbf{p}(u)=[x(u), y(u), z(u)]^{T}, 1 \ge u \ge 0$
- In classical numerical methods, we design a single global curve
- In computer graphics and CAD, it is better to design small connected curve *segments*



How should we describe curve segments?

We choose Polynomials

- Easy to evaluate
- Continuous and differentiable everywhere
 - -Must worry about continuity at join points including continuity of derivatives



Let's worry about that later. First let's scrutinize the polynomials!

Parametric Polynomial Curves

$$x(u) = \sum_{i=0}^{N} c_{xi} u^{i} \quad y(u) = \sum_{j=0}^{M} c_{yj} u^{j} \quad z(u) = \sum_{k=0}^{L} c_{zk} u^{k}$$

•Cubic polynomials gives N=M=L=3

Remember:

$$x=x(u)$$

 $y=y(u)$
 $z=z(u)$
 $p(u)$
 $p(u)$
 $p(u_{max})$

•Noting that the curves for x, y and z are independent, we can define each independently in an identical manner

•We will use the form
$$p(u) = \sum_{k=0}^{L} c_k u^k$$

where p can be any of x, y, z

Let's assume cubic polynomials!

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Cubic Parametric Polynomials

 Cubic polynomials give balance between ease of evaluation and flexibility in design

$$\mathbf{p}(u) = \sum_{k=0}^{3} c_k u^k$$

- Four coefficients to determine for each of x, y and z
- Seek four independent conditions for various values of u resulting in 4 equations in 4 unknowns for each of x, y and z

-Conditions are a mixture of continuity requirements at the join points and conditions for fitting the data

Some Types of Curves \mathbf{p}_1 р_з \mathbf{p}_2 \mathbf{p}_0 Introduce the types of curves – Interpolating Blending polynomials for interpolation of 4 control points (fit curve to 4 control points) p'(1) p'(0)– Hermite – fit curve to 2 control points + 2 derivatives (tangents) -Bezier p(0)p(1) 2 interpolating control points + 2 intermediate points to define the tangents - B-spline - use points of adjacent curve segments To get C¹ and C² continuity -NURBS p3 \mathbf{p}_0 Different weights of the control points **p**₁ (The control points can be at non-uniform intervalls • **p**₂ Analyze them **p**(0) $\mathbf{p}(1)$ \mathbf{p}_0 18

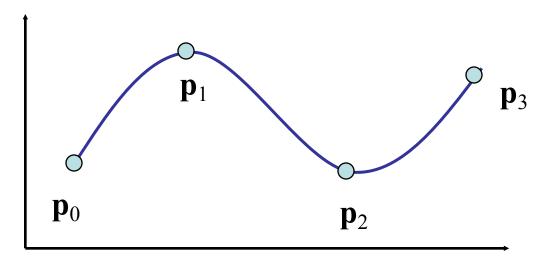
Matrix-Vector Form

$$\mathbf{p}(u) = \sum_{k=0}^{3} c_{k} u^{k}$$

define
$$\mathbf{c} = \begin{bmatrix} c_{0} \\ c_{1} \\ c_{2} \\ c_{3} \end{bmatrix} \quad \mathbf{u} = \begin{bmatrix} 1 \\ u \\ u^{2} \\ u^{3} \end{bmatrix}$$

then $\mathbf{p}(u) = \mathbf{u}^T \mathbf{c} = \mathbf{c}^T \mathbf{u}$

Interpolating Curve

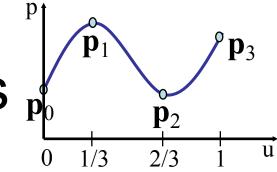


Given four data (control) points \mathbf{p}_0 , \mathbf{p}_1 , \mathbf{p}_2 , \mathbf{p}_3 determine cubic $\mathbf{p}(\mathbf{u})$ which passes through them

Must find \mathbf{c}_0 , \mathbf{c}_1 , \mathbf{c}_2 , \mathbf{c}_3

Let's create an equation system! 20

Interpolation Equations



 $p(u) = c_0 + c_1 u + c_2 u^2 + c_3 u^3$

apply the interpolating conditions at u=0, 1/3, 2/3, 1

$$\begin{pmatrix}
p_0 = p(0) = c_0 \\
p_1 = p(1/3) = c_0 + (1/3)c_1 + (1/3)^2c_2 + (1/3)^3c_3 \\
p_2 = p(2/3) = c_0 + (2/3)c_1 + (2/3)^2c_2 + (2/3)^3c_3 \\
p_3 = p(1) = c_0 + c_1 + c_2 + c_3
\end{cases}$$
or in matrix form with $\mathbf{p} = [\mathbf{p}_0 \ \mathbf{p}_1 \ \mathbf{p}_2 \ \mathbf{p}_3]^T$

$$\begin{bmatrix}
1 & 0 & 0 & 0 \\
1 & (1) & (1)^2 & (1)^3
\end{bmatrix}
\begin{bmatrix}
c_0
\end{bmatrix}$$

$$\mathbf{\hat{p}} = \mathbf{A}\mathbf{c} \qquad \mathbf{p} = \begin{bmatrix} p_1 \\ p_2 \\ p_3 \end{bmatrix} = \mathbf{A}\mathbf{c} = \begin{bmatrix} 1 & (\overline{3}) & (\overline{3}) & (\overline{3}) \\ 1 & (\overline{2}) & (\overline{3})^2 & (\overline{3})^3 \\ 1 & (\overline{3}) & (\overline{2})^2 & (\overline{2})^3 \\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

Interpolation Matrix

Solving for $\ensuremath{\mathbf{c}}$ we find the interpolation matrix

$$\mathbf{M}_{I} = \mathbf{A}^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -5.5 & 9 & -4.5 & 1 \\ 9 & -22.5 & 18 & -4.5 \\ -4.5 & 13.5 & -13.5 & 4.5 \end{bmatrix}$$

 $c = M_I p$

Note that \mathbf{M}_I does not depend on input data and can be used for each segment in x, y, and z p_1

$$p(u) = c_0 + c_1 u + c_2 u^2 + c_3 u^3$$

p₃

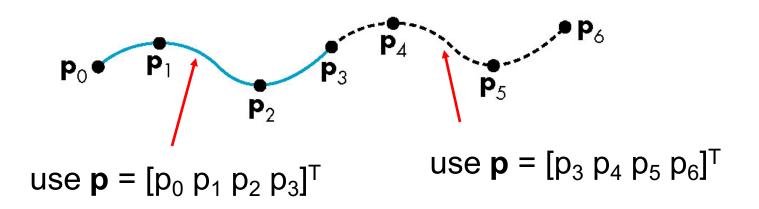
Interpolation Matrix

 $p(u) = c_0 + c_1 u + c_2 u^2 + c_3 u^3$ means:

$$\begin{array}{c} x = x(u) = c_{x0} + c_{x1}u + c_{x2}u^2 + c_{x3}u^3 \\ y = y(u) = c_{y0} + c_{y1}u + c_{y2}u^2 + c_{y3}u^3 \\ z = z(u) = c_{z0} + c_{z1}u + c_{z2}u^2 + c_{z3}u^3 \\ \end{array}$$
where
$$\begin{array}{c} \mathbf{p}_{1} \\ \mathbf{p}_{2} \\ \mathbf{p}_{2} \end{array}$$

 \mathbf{p}_x are the x coordinates of $p_0 \dots p_3$ \mathbf{p}_y are the y coordinates of $p_0 \dots p_3$ \mathbf{p}_z are the z coordinates of $p_0 \dots p_3$

Interpolating Multiple Segments



Get continuity at join points but not continuity of derivatives

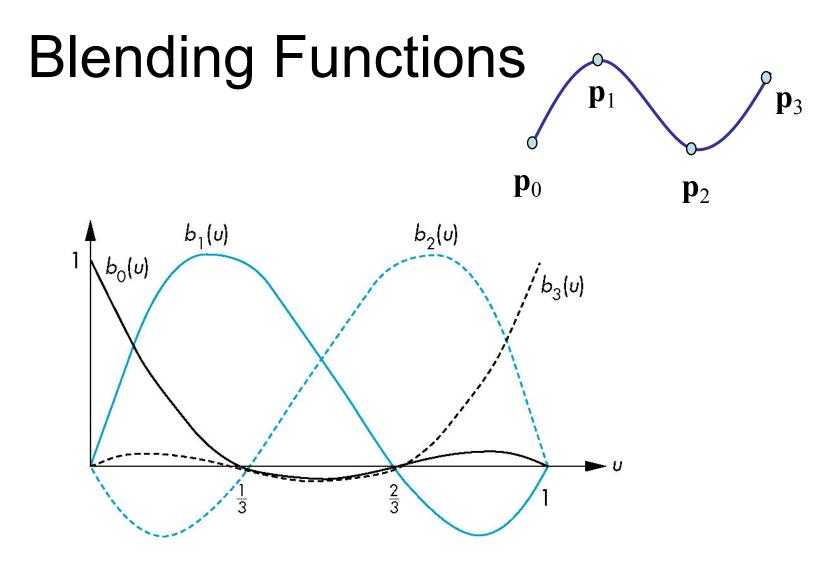
Blending Functions $p_1 p_3$

Rewriting the equation for p(u)

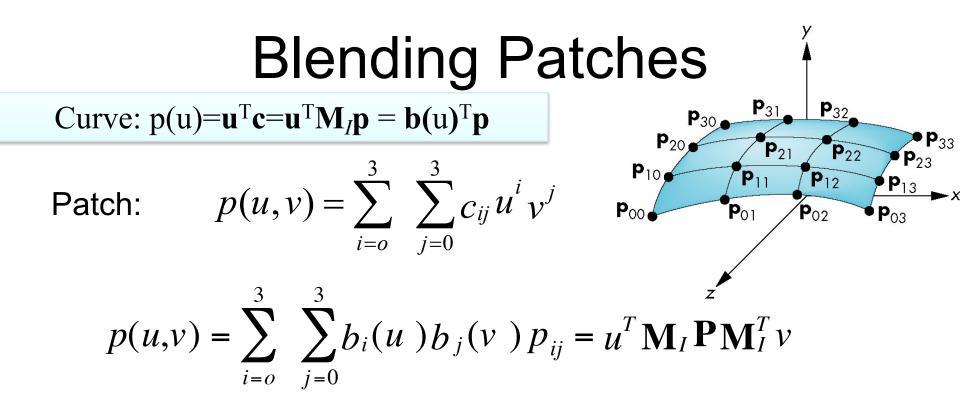
 $p(u)=u^{T}c=u^{T}M_{I}p = b(u)^{T}p$

where $b(u) = [b_0(u) b_1(u) b_2(u) b_3(u)]^T$ is an array of *blending polynomials* such that $p(u) = b_0(u)p_0 + b_1(u)p_1 + b_2(u)p_2 + b_3(u)p_3$

> $b_0(u) = -4.5(u-1/3)(u-2/3)(u-1)$ $b_1(u) = 13.5u (u-2/3)(u-1)$ $b_2(u) = -13.5u (u-1/3)(u-1)$ $b_3(u) = 4.5u (u-1/3)(u-2/3)$



 $p(u) = b_0(u)p_0 + b_1(u)p_1 + b_2(u)p_2 + b_3(u)p_3$



Each $b_i(u)b_j(v)$ is a blending patch

Shows that we can build and analyze surfaces from our knowledge of curves

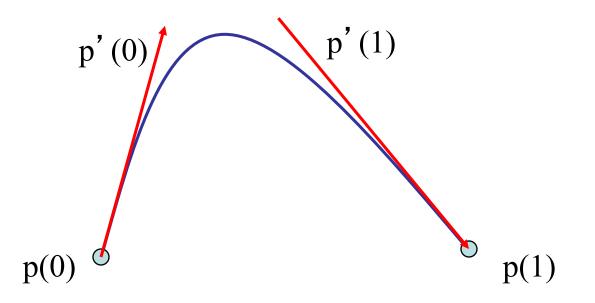
Hermite Curves and Surfaces

• How can we get around the limitations of the interpolating form –Lack of smoothness $P_0 \bullet P_1 \bullet P_2 \bullet P_3 \bullet P_4 \bullet P_5$

-Discontinuous derivatives at join points

- We have four conditions (for cubics) that we can apply to each segment
 - –Use them other than for interpolation
 - -Need only come close to the data

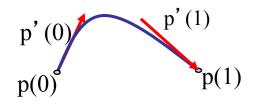
Hermite Form



Use two interpolating conditions and two derivative conditions per segment

Ensures continuity and first derivative continuity between segments

Equations



 $p(u) = c_0 + uc_1 + u^2 c_2 + u^3 c_3$

Interpolating conditions are the same at ends

$$p(0) = p_0 = c_0$$

$$p(1) = p_1 = c_0 + c_1 + c_2 + c_3$$

Differentiating we find p' (u) = $c_1 + 2uc_2 + 3u^2c_3$

Evaluating at end points $p'(0) = p'_{0} = c_{1}$ $p'(1) = p'_{1} = c_{1} + 2c_{2} + 3c_{3}$ $\mathbf{q} = \begin{bmatrix} p_{0} \\ p_{1} \\ p'_{0} \\ p'_{1} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2 & 3 \end{bmatrix} \mathbf{c}$

p'(0) p(0) p(1)

Matrix Form

$$\mathbf{q} = \begin{bmatrix} p_0 \\ p_1 \\ p'_0 \\ p'_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2 & 3 \end{bmatrix} \mathbf{c}$$

Solving, we find $c=M_H q$ where M_H is the Hermite matrix

$$\mathbf{M}_{H} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -3 & 3 & -2 & -1 \\ 2 & -2 & 1 & 1 \end{bmatrix}$$

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$$p(\mathbf{u}) = \mathbf{u}^{\mathrm{T}} \mathbf{c} \Longrightarrow$$
$$p(\mathbf{u}) = \mathbf{u}^{\mathrm{T}} \mathbf{M}_{\mathrm{H}} \mathbf{q}$$

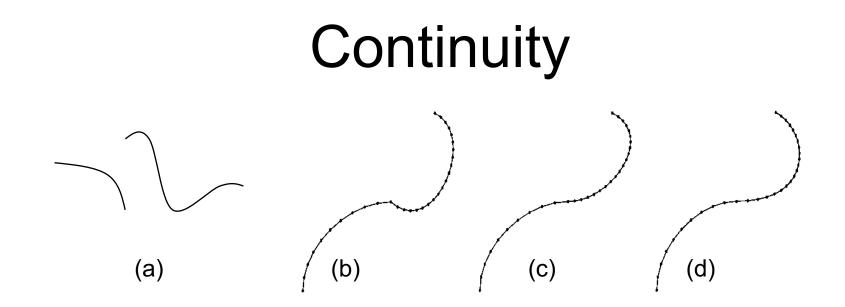
Blending Polynomials

 $p(u) = u^T M_H q \Longrightarrow p(u) = b(u)^T q$

$$\mathbf{b}(u) = \begin{bmatrix} 2u^3 - 3u^2 + 1 \\ -2u^3 + 3u^2 \\ u^3 - 2u^2 + u \\ u^3 - u^2 \end{bmatrix} \qquad \mathbf{M}_H = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -3 & 3 & -2 & -1 \\ 2 & -2 & 1 & 1 \end{bmatrix}$$

Although these functions are smooth, the Hermite form is not used directly in Computer Graphics and CAD because we usually have control points but not derivatives

However, the Hermite form is the basis of the Bezier form

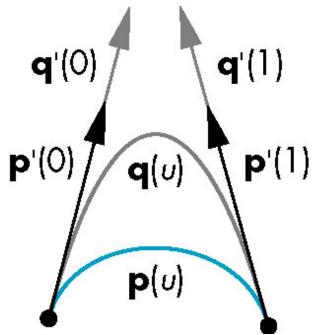


- A) Non-continuous
- B) C⁰-continuous
- C) G¹-continuous
- D) C¹-continuous
- (C²-continuous)

See page 726-727 in Real-time Rendering, 4th ed.

G¹-continuity Example

- Here the p and q have the same tangents at the ends of the segment but different derivatives
- Generate different
 Hermite curves
- This techniques is used in drawing applications



Reflections should be at least C¹



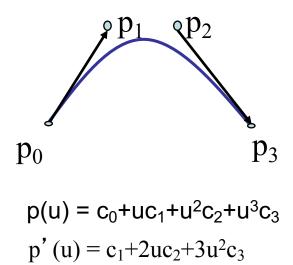


Bezier Curves

- In graphics and CAD, we do not usually have derivative data
- Bezier suggested using the same 4 data points as with the cubic interpolating curve to approximate the derivatives in the Hermite form

Approximating Derivatives \mathbf{p}_2 p_1 located at u=1/3 p_2 located at u=2/3 $\frac{dp(u=0)}{du} = p'(0) \approx \frac{p_1 - p_0}{1/3}$ $p'(1) \approx \frac{p_3 - p_2}{1/3}$ slope p'(1) slope p' (0) p_3 p_0

Equations



Interpolating conditions are the same

$$p(0) = p_0 = c_0$$

$$p(1) = p_3 = c_0 + c_1 + c_2 + c_3$$

Approximating derivative conditions

$$p'(0) \approx \frac{p_1 - p_0}{1/3}$$
 $p'(0) = 3(p_1 - p_0) = c_1$
 $p'(1) \approx \frac{p_3 - p_2}{1/3}$
 $p'(1) = 3(p_3 - p_2) = c_1 + 2c_2 + 3c_3$

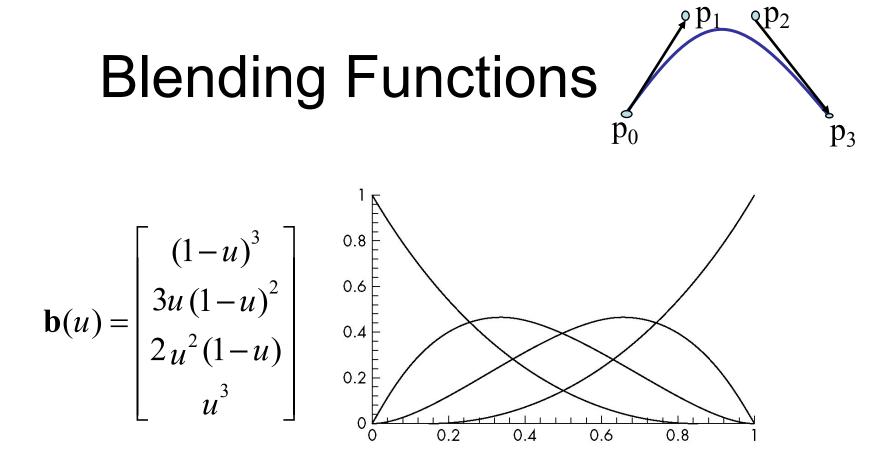
$$\Rightarrow \mathbf{Bp} = \mathbf{Ac}$$
$$\Rightarrow \mathbf{c} = \mathbf{A}^{-1}\mathbf{Bp}$$

Solve four linear equations for $c=M_B p$

Bezier Matrix

$$\mathbf{M}_{B} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -3 & 3 & 0 & 0 \\ 3 & -6 & 3 & 0 \\ -1 & 3 & -3 & 1 \end{bmatrix}$$

 $p(\mathbf{u}) = \mathbf{u}^{\mathrm{T}} \mathbf{M}_{B} \mathbf{p} = \mathbf{b}(\mathbf{u})^{\mathrm{T}} \mathbf{p}$ blending functions

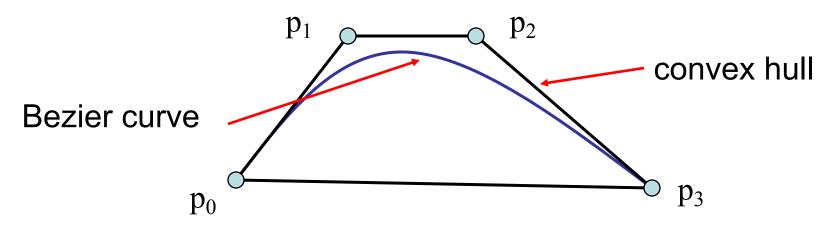


Note that all zeros are at 0 and 1 which forces the functions to be smoother over (0,1)

Smoother because the curve stays inside the convex hull, and therefore does not have room to fluctuate so much.

Convex Hull Property

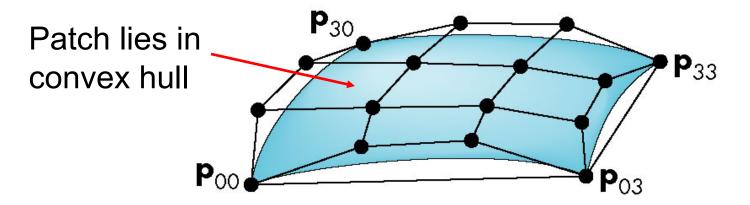
- All weights within [0,1] and sum of all weights = 1 (at given u) ensures that all Bezier curves lie in the convex hull of their control points
- Hence, even though we do not interpolate all the data, we cannot be too far away



Bezier Patches
$$p(u,v) = \sum_{i=0}^{3} \sum_{j=0}^{3} C_{ij} u^{i} v^{j}$$

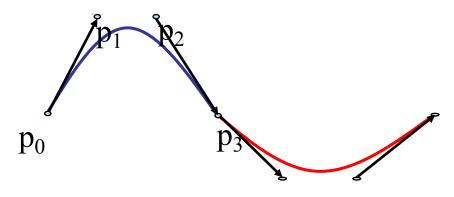
Using same data array $\mathbf{P}=[p_{ij}]$ as with interpolating form

$$p(u,v) = \sum_{i=0}^{3} \sum_{j=0}^{3} b_i(u) b_j(v) p_{ij} = u^T \mathbf{M}_B \mathbf{P} \mathbf{M}_B^T v$$



Analysis

 Although the Bezier form is much better than the interpolating form, the derivatives are not continuous at join points



•What shall we do to solve this?

B-Splines

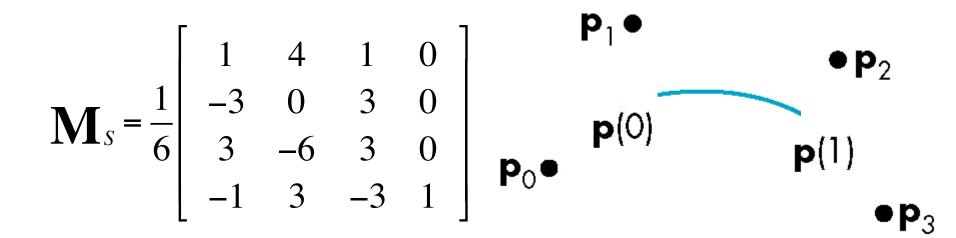
• Basis splines: use the data at

 $\pmb{p}{=}[p_{i\text{-}2} \; p_{i\text{-}1} \; p_i \; p_{i\text{+}1}]^T$ to define curve only between $p_{i\text{-}1}$ and p_i

- Allows us to apply more continuity conditions to each segment
- For cubics, we can have continuity of the function and first and second derivatives at the join points

Cubic B-spline

$$\mathbf{p}(\mathbf{u}) = \mathbf{u}^{\mathrm{T}} \mathbf{M}_{S} \mathbf{p} = \mathbf{b}(\mathbf{u})^{\mathrm{T}} \mathbf{p}$$

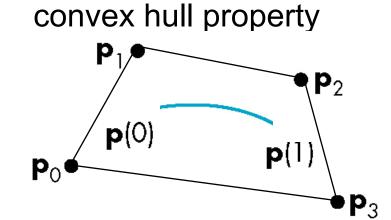


Blending Functions

A

 $p(u) = u^{T}M_{S}p = b(u)^{T}p =>$ $p(u) = b_{0}(u)p_{0}+b_{1}(u)p_{1}+b_{2}(u)p_{2}+b_{3}(u)p_{3}$

$$\mathbf{u}^{T}\mathbf{M}_{s} = \begin{bmatrix} 1 & u & u^{2} & u^{3} \end{bmatrix} \frac{1}{6} \begin{bmatrix} 1 & 4 & 1 & 0 \\ -3 & 0 & 3 & 0 \\ 3 & -6 & 3 & 0 \\ -1 & 3 & -3 & 1 \end{bmatrix}$$



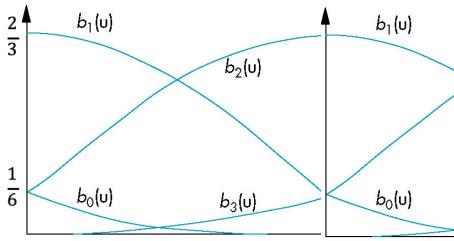
Blending Functions

$\mathbf{p}(\mathbf{u}) = \mathbf{c}^{\mathrm{T}}\mathbf{u} = \mathbf{u}^{\mathrm{T}}\mathbf{M}_{S}\mathbf{p} = \mathbf{b}(\mathbf{u})^{\mathrm{T}}\mathbf{p}$

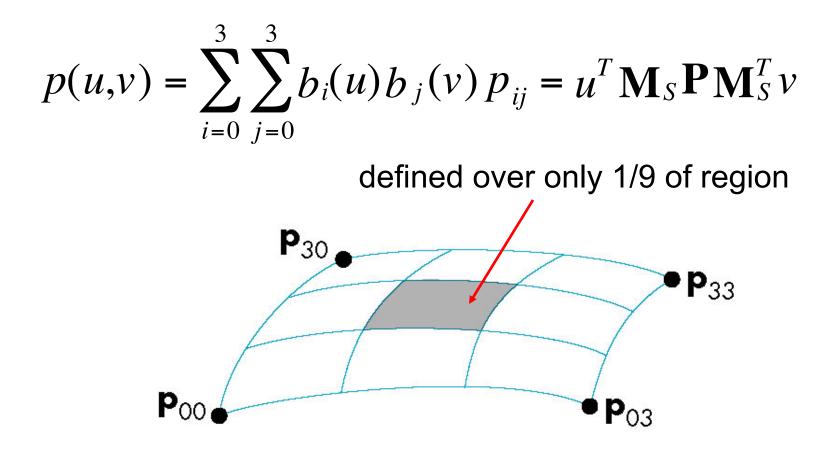
16 unknowns in M_s . We need 16 equations:

- 5 for endpoint values: $b_0(0)=b_1(1)$. $b_0(1)=0$. $b_1(0)=b_2(1)$. $b_2(0)=b_3(1)$. $b_3(0)=0$.
- Same 5 for endpoint 1^{st} derivatives: $b_0'(0)=b_1'(1)$. $b_0'(1)=0$. $b_1'(0)=b_2'(1)$. $b_2'(0)=b_3'(1)$. $b_3'(0)=0$.
- Same 5 for endpoint 2nd derivatives: b₀"(0)=b₁"(1). b₀"(1)=0. b₁"(0)=b₂"(1). b₂"(0)=b₃"(1). b₃"(0)=0.
- Sum = 1, everywhere: $b_0(u)+b_1(u)+b_2(u)+b_3(u)=1$, for $u \in [0,1]$. E.g., for u=0.

$$\mathbf{u}^{T}\mathbf{M}_{s} = \begin{bmatrix} 1 & u & u^{2} & u^{3} \end{bmatrix} \frac{1}{6} \begin{bmatrix} 1 & 4 & 1 & 0 \\ -3 & 0 & 3 & 0 \\ 3 & -6 & 3 & 0 \\ -1 & 3 & -3 & 1 \end{bmatrix}$$



B-Spline Patches

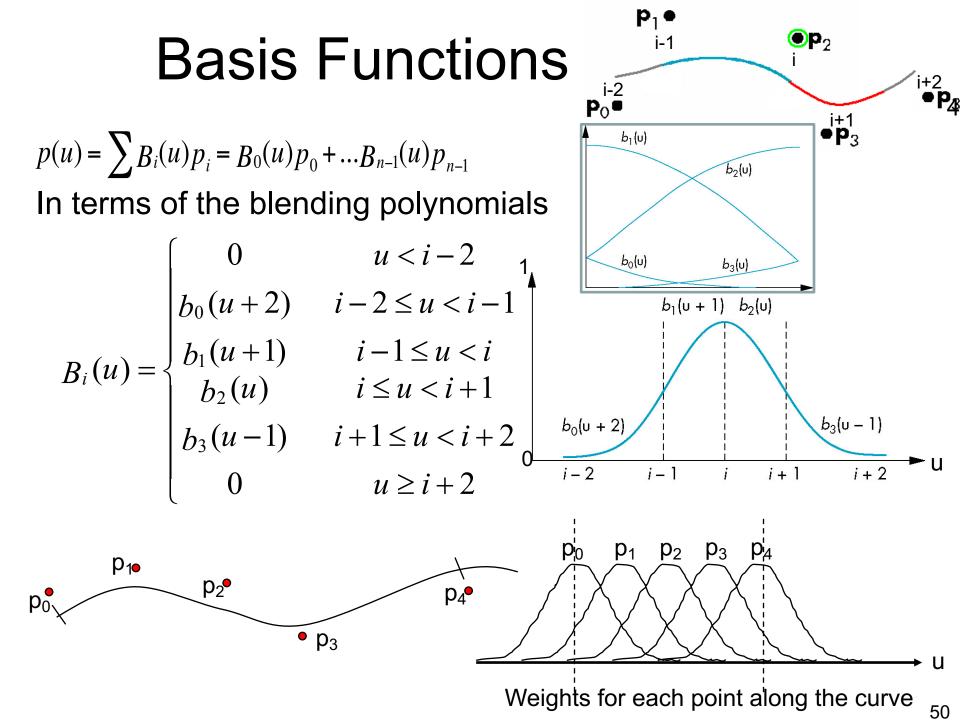


Splines and Basis

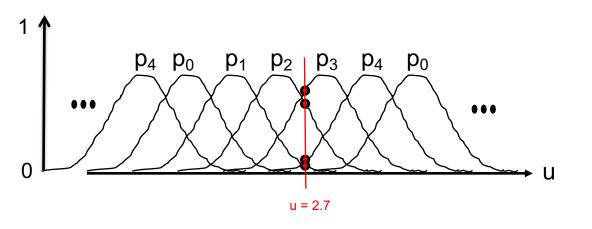
- If we examine the cubic B-spline from the perspective of each control (data) point, each interior point contributes (through the blending functions) to four segments
- •We can rewrite p(u) in terms of *all* the data points along the curve as

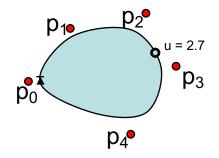
$$p(u) = \sum B_i(u) p_i$$

defining the basis functions $\{B_i(u)\}$

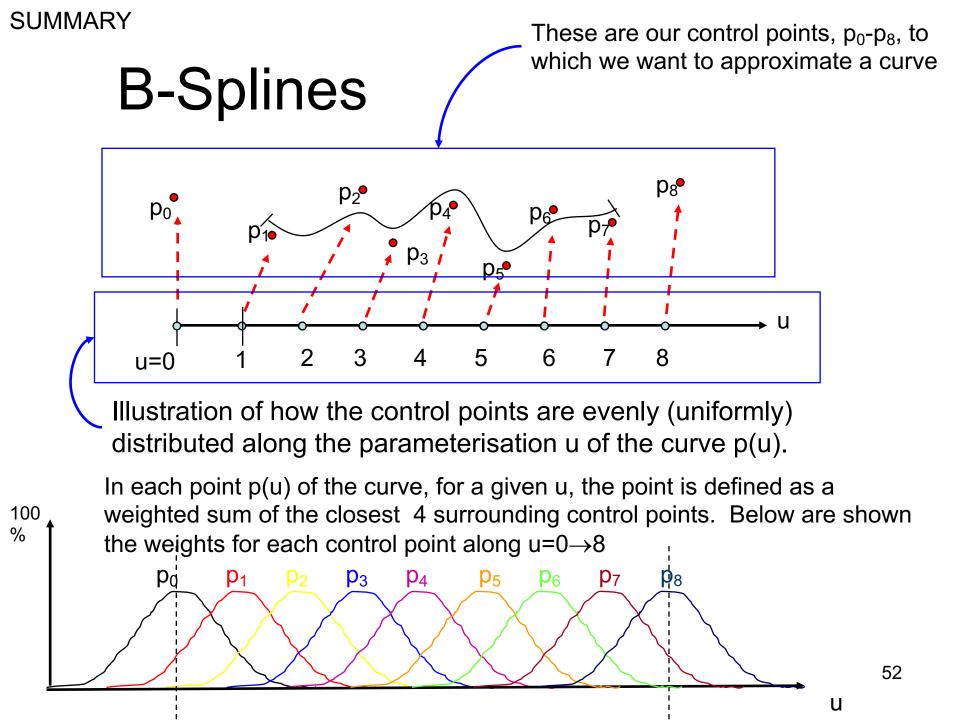


One more example





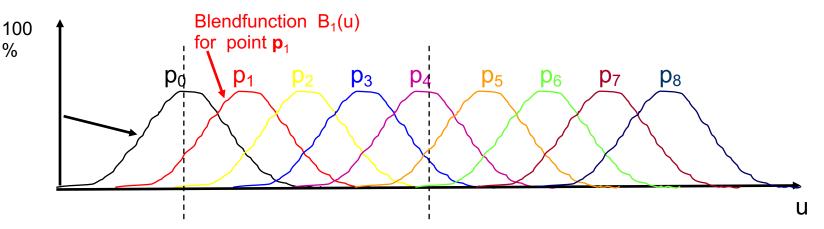
 $p(u) = B_0(u)p_0 + B_1(u)p_1 + B_2(u)p_2 + B_3(u)p_3 + B_4(u)p_4$ I.e.,: $p(u) = \sum B_i(u) p_i$



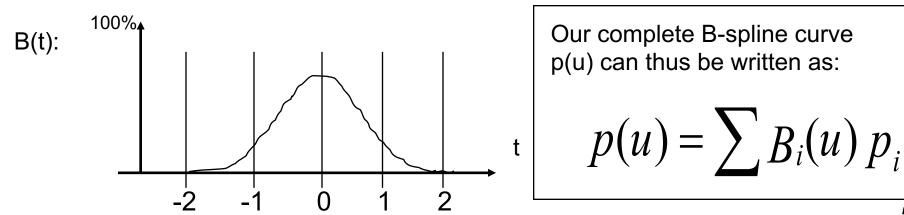
SUMMARY

B-Splines

In each point p(u) of the curve, for a given u, the point is defined as a weighted sum of the closest 4 surrounding points. Below are shown the weights for each point along $u=0\rightarrow 8$



The weight function (blend function) B_i (u) for a point p_i can thus be written as a translation of a basis function B(t). $B_i(u) = B_t(u-i)$



Generalizing Splines

- •We can extend to splines of any degree
- Data and conditions do not have to be given at equally spaced values (the knots)
 - –Nonuniform and uniform splines
 - -Can have repeated knots
 - Easiest implemented by just repeating a ctrl point – Can force spline to interpolate points
- (Cox-deBoor recursion gives method of evaluation (also known as de Casteljau-recursion, see page 721, RTR 4:th edition for details))

NURBS = <u>N</u>on-<u>U</u>niform <u>R</u>ational <u>B</u>-<u>S</u>plines

NURBS is similar to B-Splines except that:

- 1. The control points can have different weights, w_i, (heigher weight makes the curve go closer to that control point)
- The control points do not have to be at uniform distances (u=0,1,2,3...) along the parameterisa-tion u. E.g.: u=0, 0.5, 0.9, 4, 14,...

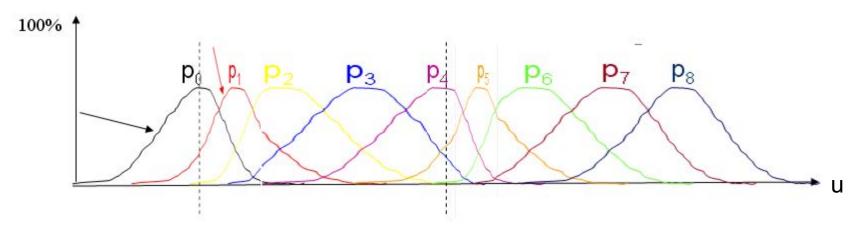
The NURBS-curve is thus defined as

 $\mathbf{p}(u) = B_{i}(u)W_{i}$ 55

Division with the sum of the weights, to make the combined weights sum up to 1, at each position along the curve. Otherwise, a translation of the curve is introduced (which is not desirable)

 Allowing control points at non-uniform distances means that the basis functions B_{pi}() are being streched and non-uniformly located.

• E.g.:



Each curve $B_{pi}()$ should of course look smooth and C^2 –continuous. But it is not so easy to draw smoothly by hand...

(The sum of the weights are still =1 due to the division in previous slide)

- <u>Nonuniform Rational B-Spline curves and</u> surfaces add a fourth variable w to x,y,z
 - -Can interpret as weight to give more importance to some control data
 - -Can also interpret as moving to homogeneous coordinate
- (Requires a perspective division –NURBS act correctly for perspective viewing
- Quadrics are a special case of NURBS)

- Concider a control point in 3 dimensions: $\mathbf{p}_i = [x_i, y_i, z_i]$
- The weighted homogeneous-coordinate is:

$$\mathbf{q}_{i} = w_{i} \begin{bmatrix} x_{i} \\ y_{i} \\ z_{i} \\ 1 \end{bmatrix}$$

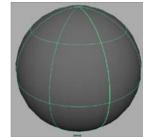
• The idea is to use the weights w_i to increase or decrease the importance of a particular control point

- The w-component may not be equal to 1.
- Thus we must do a perspective division to get the three-dimensional points:

$$\mathbf{p}(u) = \frac{1}{w(u)} \mathbf{q}(u) = \frac{\sum_{i=0}^{n} B_{i,d} w_i \mathbf{p}(i)}{\sum_{i=0}^{n} B_{i,d} w_i}$$

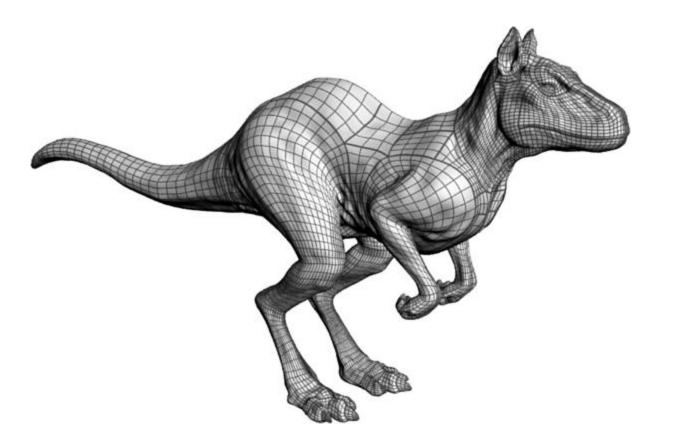
 Each component of p(u) is now a rational function in u, and because we have not restricted the knots (the knots does not have to be uniformly distributed), we have derived a nonuniform rational B-spline (NURBS) curve

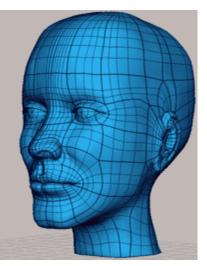
- If we apply an affine transformation to a B-spline curve or surface, we get the same function as the B-spline derived from the transformed control points.
- Because perspective transformations are not affine, most splines will not be handled correctly in perspective viewing.
- However, the perspective division embedded in the NURBS ensures that NURBS curves are handled correctly in perspective views.



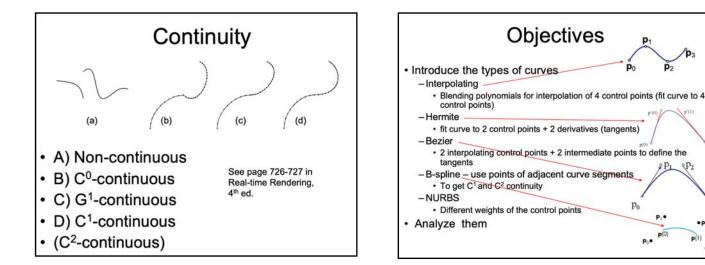
Quadrics can be shown to be a special case of quadratic NURBS curve; thus, we can use a single modeling method, NURBS curves, for the most widely used curves and surfaces
 THE FNIC

NURBS Surfaces - examples





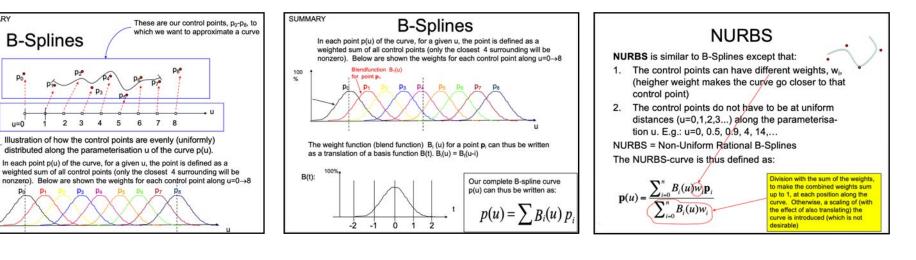
What you need to know:



SUMMARY

100

u=0



P₀

 p_3

• P3

·P.

Bonus slides

- Every polynomial curve can be exactly described by a bezier curve (by properly adjusting the control points).
- Rasterization of Bezier curves can be implemented highly efficiently using *de Casteljau recursion*.
- Thus, NURBS curves are often first converted to Bezier curves, to be efficiently rasterized.
- See following bonus slides for explanations...

Every Polynomial Curve is a Bezier Curve

- We can render a given polynomial using the recursive method if we find control points for its representation as a Bezier curve
- Suppose that p(u) is given as an interpolating curve with control points q p(u)=u^TM_Iq
- There exist Bezier control points \boldsymbol{p} such that $p(\boldsymbol{u}) = \boldsymbol{u}^T \boldsymbol{M}_{\mathcal{B}} \boldsymbol{p}$
- Equating and solving, we find $\mathbf{p}=\mathbf{M}_{B}^{-1}\mathbf{M}_{I}$

Bonus

deCasteljau¹ Recursion

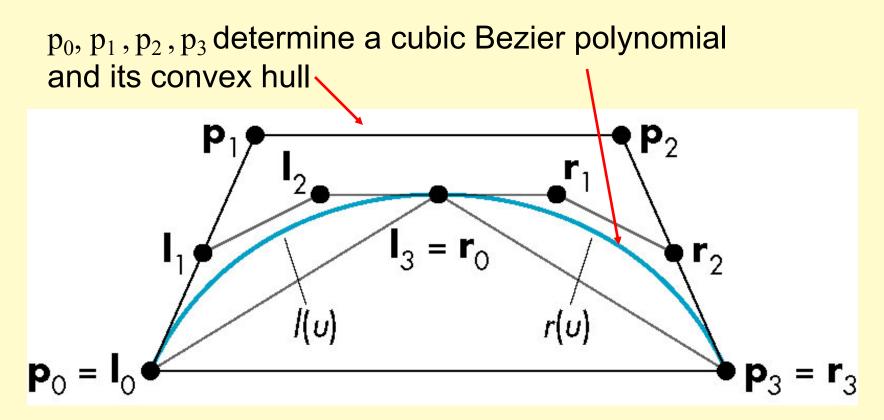
 We can use the convex hull property of Bezier curves to obtain an efficient recursive method that does not require any function evaluations

-Uses only the values at the control points

 Based on the idea that "any polynomial and any part of a polynomial is a Bezier polynomial for properly chosen control data"

¹ Paul de Casteljau and Pierre Bezier where engineers in the car industry. De Casteljau at Peugot at Bezier at Renault. Both developed Bezier-surfaces, unaware of each other.

Splitting a Cubic Bezier

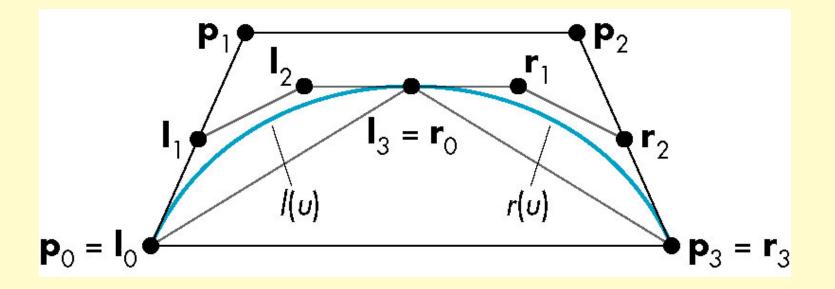


Consider left half l(u) and right half r(u)

Bonus

l(u) and r(u)

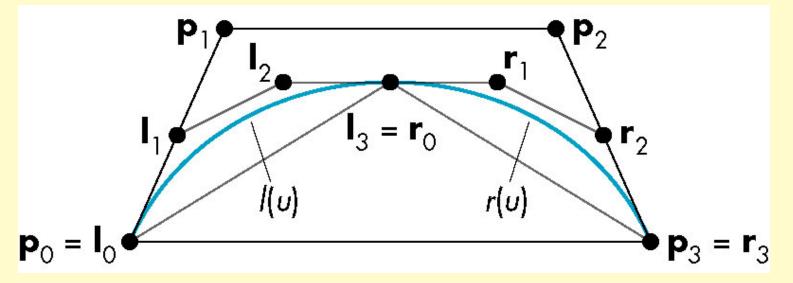
Since l(u) and r(u) are Bezier curves, we should be able to find two sets of control points $\{l_0, l_1, l_2, l_3\}$ and $\{r_0, r_1, r_2, r_3\}$ that determine them



Convex Hulls

 $\{l_0, l_1, l_2, l_3\}$ and $\{r_0, r_1, r_2, r_3\}$ each have a convex hull that that is closer to p(u) than the convex hull of $\{p_0, p_1, p_2, p_3\}$ This is known as the *variation diminishing property*.

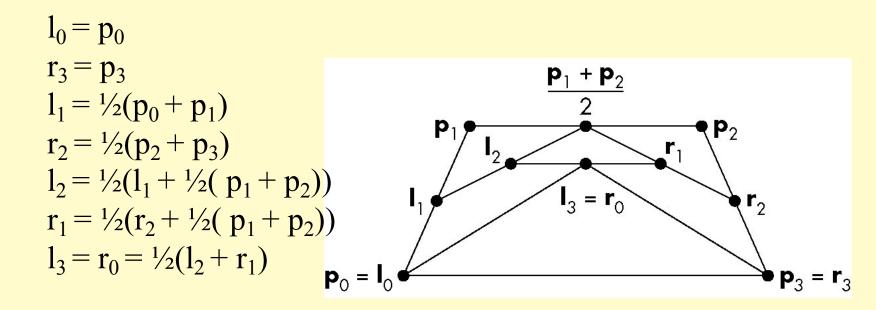
The polyline from l_0 to l_3 (= r_0) to r_3 is an approximation to p(u). Repeating recursively we get better approximations.



deCasteljau¹ Recursion:

Bonus

Efficient Form



Requires only shifts and adds!

Then, recursively continue for the two new bezier curves $\{I_0, I_1, I_2, I_3\}$ and $\{r_0, r_1, r_2, r_3\}$ until desired precission is reached.