

Topic 15:

Interpolating Curves

- Intro to curve interpolation & approximation
- Polynomial interpolation
- Bézier curves
- Cardinal splines

Interactive Design of 2D Curves

Goal: expand vocabulary of modeling primitives beyond global analytic shapes and 3D meshes

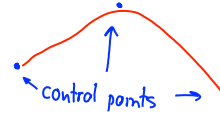
Criteria:

- Natural & intuitive interaction
- Controllable smoothness
- Adjustable resolution
- Analytic derivatives that are easy to compute
- Compact representation

Interactive Curve Design: Three Basic Tasks

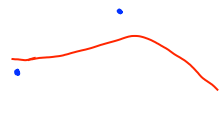
• Interpolation

Curve goes through control points



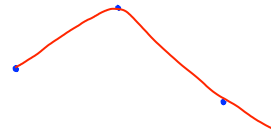
• Approximation

Curve approximates but does not go through the control points



• Extrapolation

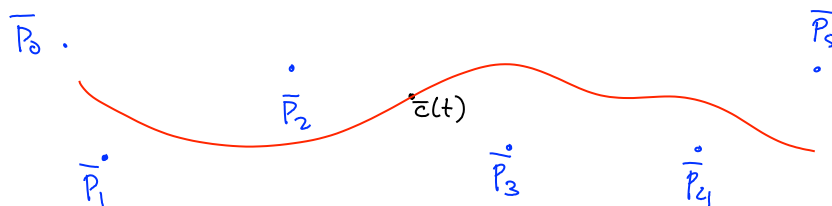
Extend curve beyond the domain of the control points



2D Curve Design: General Problem Statement

Given N control points $\bar{p}_i, i=0, \dots, N-1$

- ① define a curve $\bar{c}(t), t \in [0, N-1]$ that interpolates/approximates them



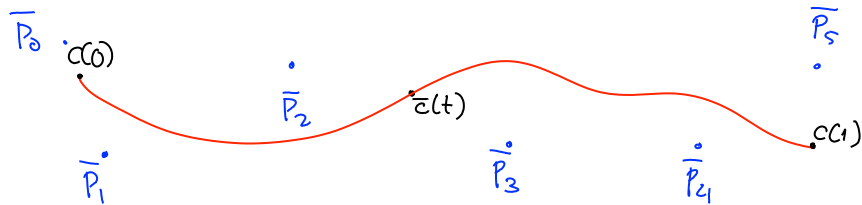
- ② compute its derivatives (and tangents, normals, etc)

2D Curve Design: General Problem Statement

Given N control points $\bar{P}_i, i=0, \dots, N-1$

- ① define a curve $\bar{c}(t), t \in [0,1]$ that interpolates/approximates them

convention: we assume that t ranges from 0 to 1

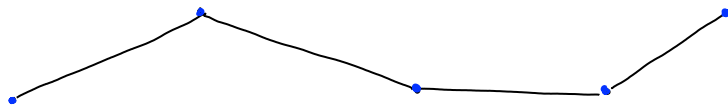


- ② compute its derivatives (and tangents, normals, etc)

Linear Interpolation

The simplest possible interpolation technique

Create a piecewise linear curve that connects the control points



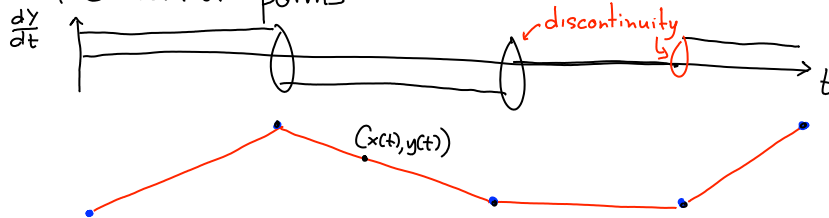
Q: What is the disadvantage of this technique?

Ans: The curve may be continuous, but its derivatives are not!

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Create a piecewise linear curve that connects the control points

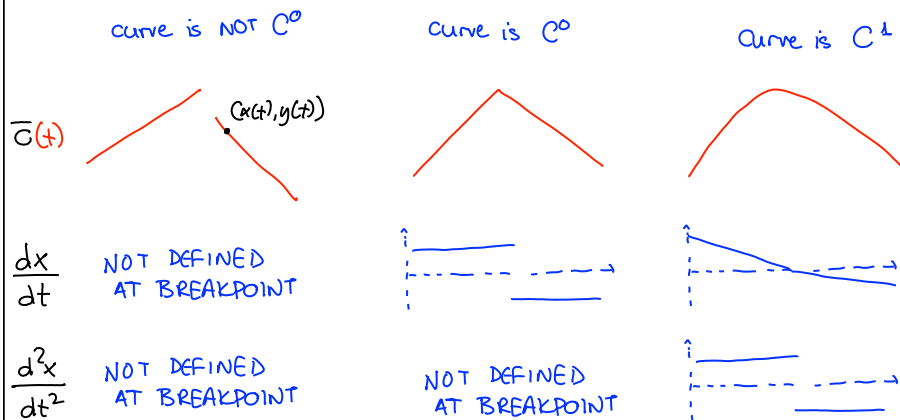


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C^n Continuity

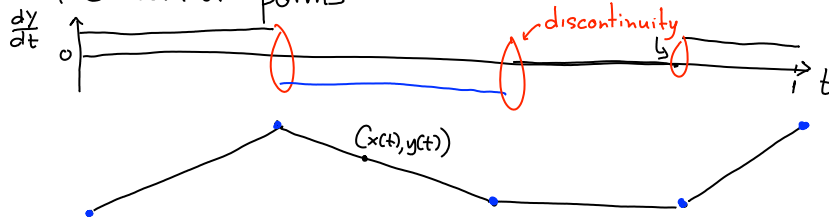
Definition A function is called C^n if its n -th order derivative is continuous everywhere



Linear Interpolation

The simplest possible interpolation technique

Create a piecewise linear curve that connects the control points



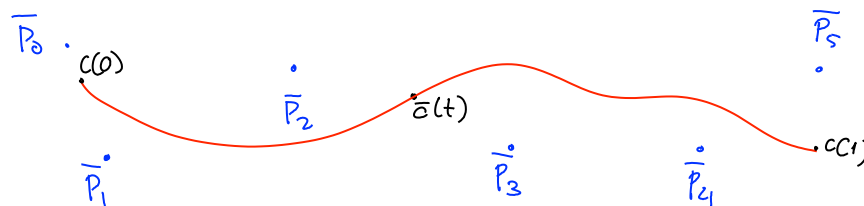
Q: What is the disadvantage of this technique?

Ans: Curve only has C^0 continuity

2D Curve Design: General Problem Statement

Given N control points $\bar{P}_i, i=0, \dots, N-1$

- ① define a curve $\bar{c}(t), t \in [0, 1]$ that interpolates/approximates them



- ② compute its derivatives (and tangents, normals, etc)

* We will seek functions that are at least C^1

Topic 15:

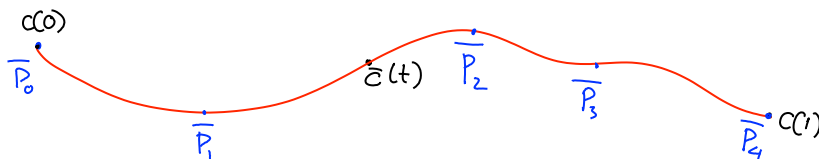
Interpolating Curves

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General Polynomial Interpolation

Given N control points $\bar{P}_i = (x_i, y_i)$ $i=0, \dots, N-1$

- ① define $(N-1)$ -order polynomials $x(t), y(t)$
such that $x(\frac{i}{N-1}) = x_i$, $y(\frac{i}{N-1}) = y_i$ for $i=0, \dots, N-1$

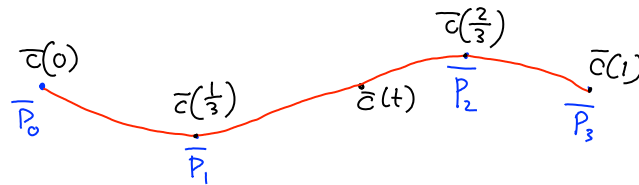


- ② compute its derivatives (and tangents, normals, etc)

Cubic Interpolation

Given 4 control points $\bar{P}_i = (x_i, y_i) \quad i=0, \dots, N-1$

- ① define 3rd-order polynomials $x(t), y(t)$
such that $x(\frac{t}{3}) = x_i, y(\frac{t}{3}) = y_i$ for $i=0, \dots, 3$



- ② compute its derivatives (and tangents, normals, etc)

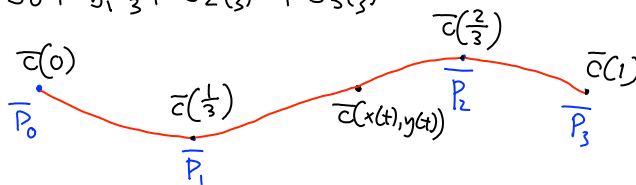
Cubic Interpolation: Basic Equations

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

Equations for one control point

$$x_1 = a_0 + a_1 \cdot \frac{1}{3} + a_2 \left(\frac{1}{3}\right)^2 + a_3 \left(\frac{1}{3}\right)^3$$

$$y_1 = b_0 + b_1 \cdot \frac{1}{3} + b_2 \left(\frac{1}{3}\right)^2 + b_3 \left(\frac{1}{3}\right)^3$$



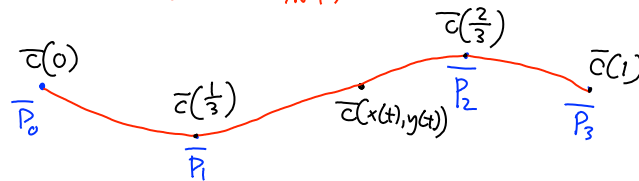
Equations in matrix form:

$$\begin{bmatrix} x_1 & y_1 \end{bmatrix} = \begin{bmatrix} 1 & \frac{1}{3} & \left(\frac{1}{3}\right)^2 & \left(\frac{1}{3}\right)^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

Cubic Interpolation: Computing the Coeffs

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{-given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

$$\underbrace{[x_i \ y_i]}_{\text{known}} = \underbrace{\begin{bmatrix} 1 & t_i & (t_i)^2 & (t_i)^3 \end{bmatrix}}_{\text{known } (t_i = \frac{i}{N-1})} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix} \leftarrow \text{unknown}$$



Equations in matrix form:

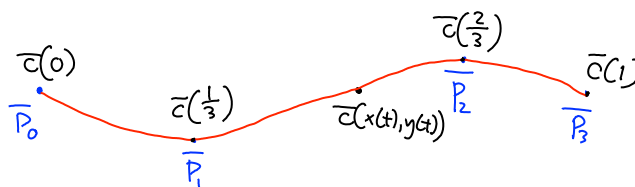
$$[x_1 \ y_1] = \begin{bmatrix} 1 & \frac{1}{3} & (\frac{1}{3})^2 & (\frac{1}{3})^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

Cubic Interpolation: Computing the Coeffs

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Eqs for 4 control points:

$$\underbrace{\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}}_{\text{known } \mathbf{C}} = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & \frac{1}{3} & (\frac{1}{3})^2 & (\frac{1}{3})^3 \\ 1 & \frac{2}{3} & (\frac{2}{3})^2 & (\frac{2}{3})^3 \\ 1 & 1 & 1 & 1 \end{bmatrix}}_{\text{known } \mathbf{A}} \underbrace{\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \\ a_4 & b_4 \end{bmatrix}}_{\text{unknown } \mathbf{X}} \Rightarrow \text{solve system in terms of unknown matrix } \mathbf{X} = \mathbf{A}^{-1} \mathbf{C}$$

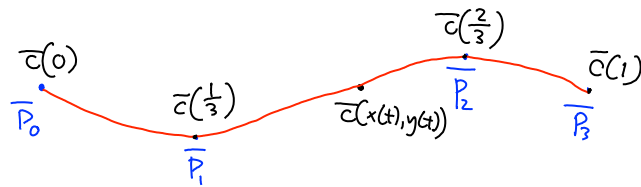


Cubic Interpolation: Computing the Coeffs

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{-given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

Coefficients of interpolating polynomial computed by

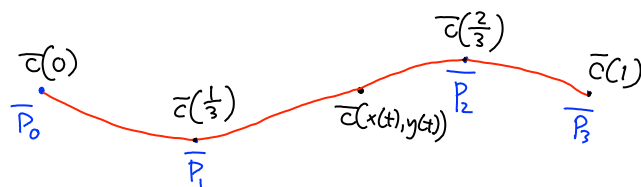
$$\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \\ a_4 & b_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1/3 & (1/3)^2 & (1/3)^3 \\ 1 & 2/3 & (2/3)^2 & (2/3)^3 \\ 1 & 1 & 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}$$



Cubic Interpolation: Evaluating the Polynomial

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{-given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

$$\begin{bmatrix} x(t) & y(t) \end{bmatrix} = \begin{bmatrix} 1 & t & t^2 & t^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



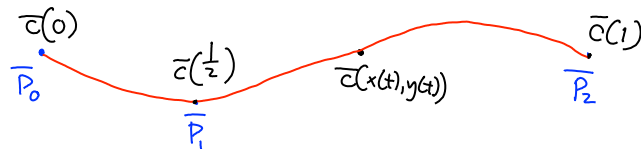
Cubic Interpolation: What If < 4 Control Points?

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

Coefficients of interpolating polynomial computed by

$$\begin{array}{c} \uparrow \text{degree} \\ +1 \\ \downarrow \end{array} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \\ a_4 & b_4 \end{bmatrix} = \begin{bmatrix} \text{more unknowns} \\ \text{than Eqs} \Rightarrow \\ \text{cannot compute} \\ \text{inverse} \end{bmatrix}^{-1} \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}$$

← # control points →



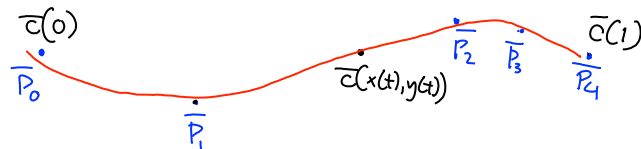
Cubic Interpolation: What If > 4 Control Points?

$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

Coefficients of interpolating polynomial computed by

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← # control points →

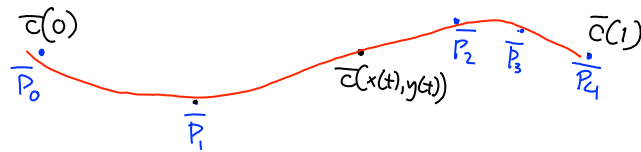


Exact Interpolation of N Points

⇒ To interpolate N points perfectly with a single polynomial we need a polynomial of degree N-1

$$\begin{array}{c} \uparrow \\ \text{degree} \\ +1 \\ \downarrow \end{array} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \\ a_4 & b_4 \end{bmatrix} = \begin{bmatrix} \text{NxN matrix} \\ \# \text{ constraints} = \\ \# \text{ unknown} \\ \text{coeffs} \end{bmatrix}^{-1} \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}$$

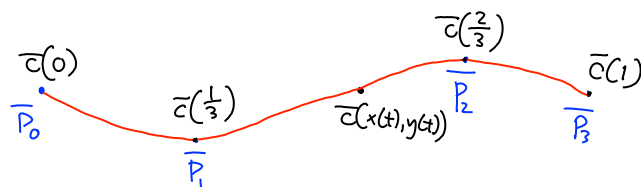
← # control points →



Cubic Interpolation: Evaluating Derivatives

$$\begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ \frac{dx}{dt}(t) &= a_1 + 2a_2 t + 3a_3 t^2 \end{aligned}$$

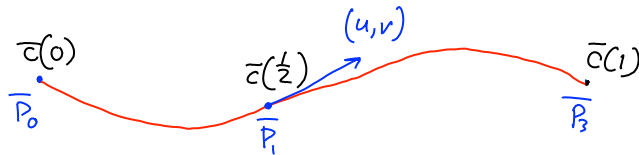
$$\left[\frac{dx}{dt}(t) \quad \frac{dy}{dt}(t) \right] = [1 \quad 2t \quad 3t^2] \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



Specifying the Poly via Tangent Constraints

Instead of specifying 4 control points, we could specify 3 points and a derivative
⇒ replace 4th pair of eqs with

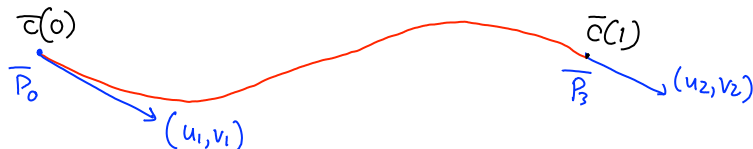
$$\begin{bmatrix} u & v \end{bmatrix} = \begin{bmatrix} 1 & 1 & 3(\frac{1}{2})^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$
$$\begin{bmatrix} \frac{dx}{dt}(t) & \frac{dy}{dt}(t) \end{bmatrix} = \begin{bmatrix} 1 & 2t & 3t^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



Specifying the Poly via Tangent Constraints

Instead of specifying 4 control points, we could specify 2 points and 2 derivatives
(we will use this later today!)

$$\begin{bmatrix} \frac{dx}{dt}(t) & \frac{dy}{dt}(t) \end{bmatrix} = \begin{bmatrix} 1 & 2t & 3t^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

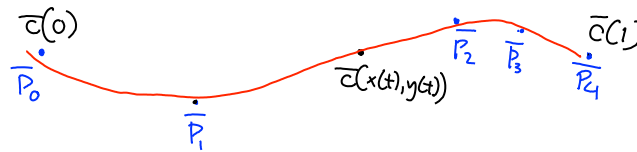


Degree-N Poly Interpolation: Major Drawback

⇒ To interpolate N points perfectly with a single polynomial we need a polynomial of degree $N-1$

Major drawback: It is a global interpolation scheme

i.e. moving one control point changes the interpolation of all points, often in unexpected/unintuitive ways

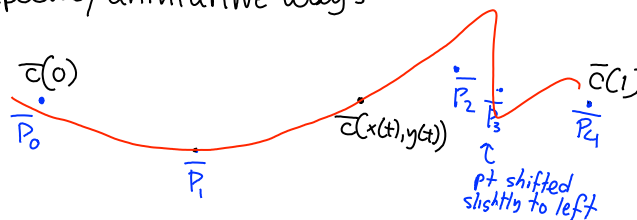


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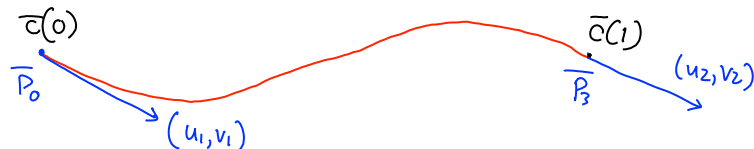
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Bézier Curves

Properties

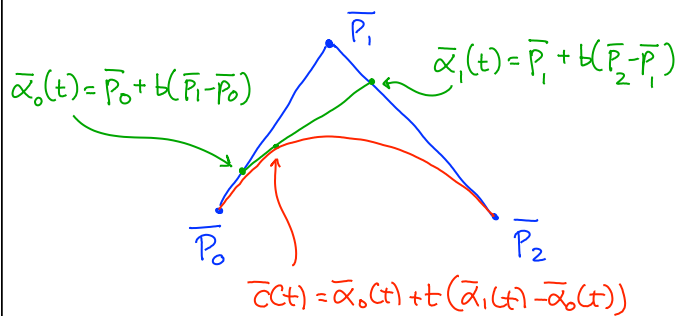
- Polynomial curves defined via endpoints and derivative constraints
 - Derivative constraints defined implicitly through extra control points (that are not interpolated)
- ⇒ they are approximating, not interpolating, curves



Bézier Curves: Main Idea

Polynomial & its derivatives expressed as a cascade of linear interpolations

Example: a double cascade



algorithm:
 given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t

1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
2. linearly interpolate \bar{P}_1, \bar{P}_2 to get $\bar{\alpha}_1(t)$
3. linearly interpolate $\bar{\alpha}_0(t), \bar{\alpha}_1(t)$ to get $\bar{c}(t)$

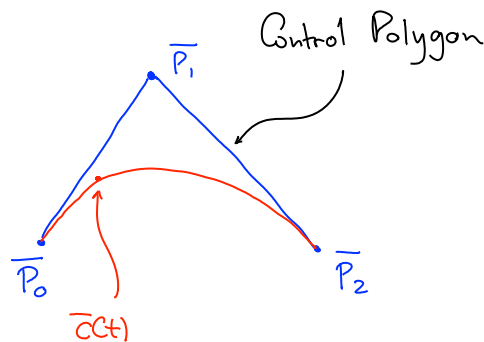
Q: Where have we seen such a cascade before?

Bézier Curves: The Control Polygon

A Bézier curve is completely determined by its control polygon

⇒

We manipulate the curve by manipulating its polygon



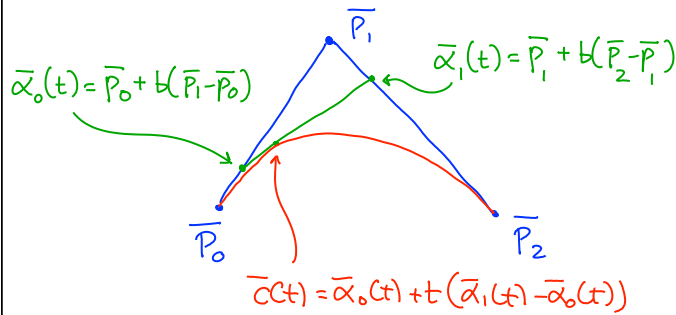
algorithm:
 given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t

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3. linearly interpolate $\bar{\alpha}_0(t), \bar{\alpha}_1(t)$ to get $\bar{c}(t)$

Expressing the Bézier Curve as a Polynomial

Computing the polynomial

$$\begin{aligned} c(t) &= [P_0 + t(\bar{P}_1 - \bar{P}_0)] + t[\bar{P}_1 + t(\bar{P}_2 - \bar{P}_1) - \bar{P}_0 - t(\bar{P}_1 - \bar{P}_0)] \\ &= \bar{P}_0(1-t-t+t^2) + \bar{P}_1(t+t-t^2-t^2) + \bar{P}_2 t^2 \\ &= \bar{P}_0(1-t)^2 + 2\bar{P}_1 t(1-t) + \bar{P}_2 t^2 \end{aligned}$$



algorithm:

- given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t
1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
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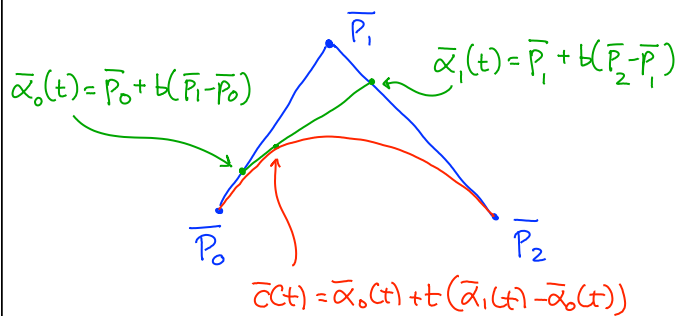
Derivatives of the Bézier Curve

Computing the polynomial's derivatives:

$$\frac{d}{dt} c(t) = -2(1-t)P_0 + 2P_1(1-t) + P_2 \cdot 2t = 2(P_1 - P_0) \text{ at } t=0$$

$$= 2(P_2 - P_1) \text{ at } t=1$$

$$c(t) = \bar{P}_0(1-t)^2 + 2\bar{P}_1 t(1-t) + \bar{P}_2 t^2$$



algorithm:

- given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t
1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
 2. linearly interpolate \bar{P}_1, \bar{P}_2 to get $\bar{\alpha}_1(t)$
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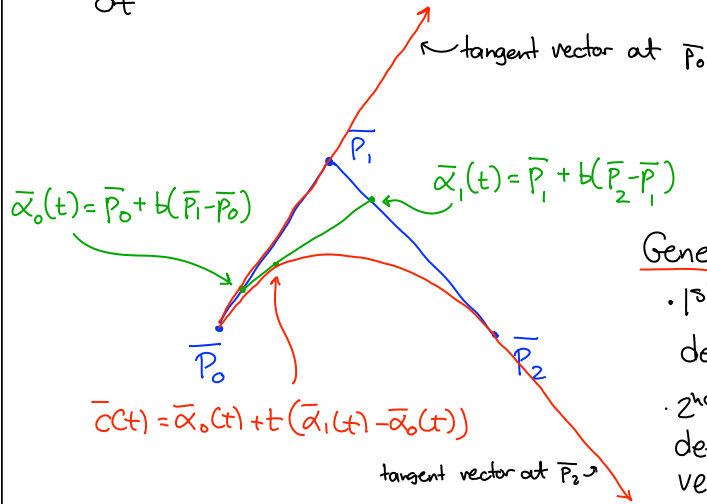
Bézier Curves: Endpoint & Tangent

Constraints

Computing the polynomial's derivatives:

$$\frac{d}{dt}c(t) = -2(1-t)P_0 + 2P_1(1-t) + P_2 \cdot 2t = 2(P_1 - P_0) \text{ at } t=0$$

$$= 2(P_2 - P_1) \text{ at } t=1$$



General behavior:

- 1st & 3rd control pts define the endpoints
- 2nd control point define the tangent vector at endpoints

Bézier Curve Polynomial: Basic Expression

Expression in compact form

$$c(t) = \sum \bar{P}_i B_i^2(t) \quad \text{where}$$

↑ curve ↑ control pt

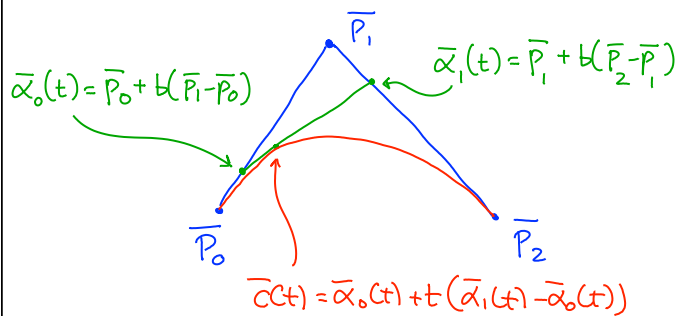
called the Bernstein polynomials of degree 2

$$B_0^2(t) = (1-t)^2$$

$$B_1^2(t) = 2t(1-t)$$

$$B_2^2(t) = t^2$$

$$c(t) = \bar{P}_0(1-t)^2 + 2\bar{P}_1 t(1-t) + \bar{P}_2 t^2$$



algorithm:

- given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t
1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
 2. linearly interpolate \bar{P}_1, \bar{P}_2 to get $\bar{\alpha}_1(t)$
 3. linearly interpolate $\bar{\alpha}_0(t), \bar{\alpha}_1(t)$ to get $\bar{c}(t)$

Bézier Curves: Generalization to N+1 points

Expression in compact form

$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t) \quad \text{where}$$

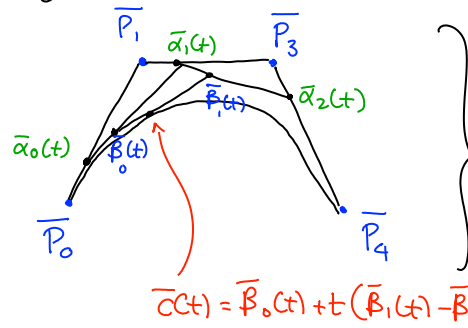
↑ curve ↑ control pt

called the Bernstein Polynomials of degree N

$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i$$

$$= \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$

Curve defined by N linear interpolation cascades



Example for 4 points and 3 cascades

$$\bar{c}(t) = \bar{\beta}_0(t) + t(\bar{\beta}_1(t) - \bar{\beta}_0(t))$$

Bézier Curves: A Different Perspective

Expression in compact form

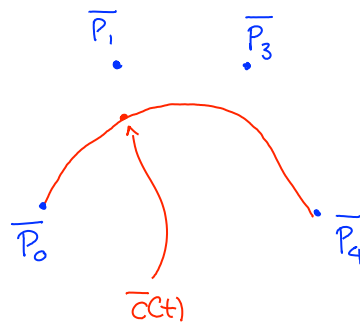
$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t) \quad \text{where}$$

↑ curve ↑ control pt

called the Bernstein Polynomials of degree N

$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i$$

$$= \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$



- Each curve point $c(t)$ is a "blend" of the 4 control pts
- The blend coeffs depend on t
- They are Bernstein polynomials.

Bézier Curves as "Blends" of the Control

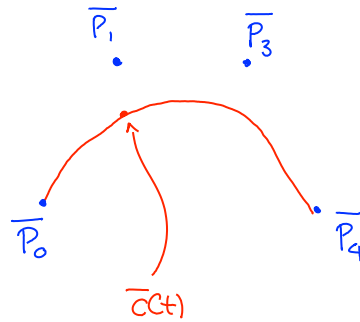
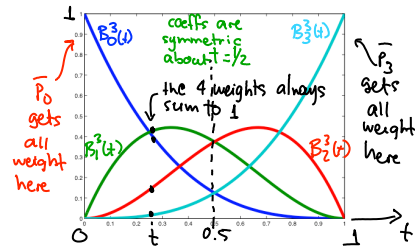
Points

Expression in compact form

$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t)$$

↑ curve ↑ control pt

with $\sum_{i=0}^N B_i^N(t) = 1$ for all t



- Each curve point $c(t)$ is a "blend" of the 4 control pts
- The "blending weights" depend on t
- They are Bernstein polynomials.

Bézier Curves: Useful Properties

Expression in compact form

$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t) \quad \text{where}$$

called the Bernstein polynomials of degree N

$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i$$

$$= \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$

① Affine invariance

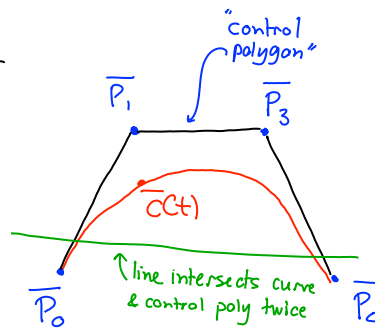
Transforming a Bézier curve by affine transform $T \Leftrightarrow$ transforming its control pts by T

② Diminishing variations

No line will intersect the curve at more pts than the control polygon \Rightarrow curve cannot exhibit "excessive fluctuations"

③ Linear precision

If control poly approximates a line, so will the curve



Bézier Curves: Useful Properties

Expression in compact form

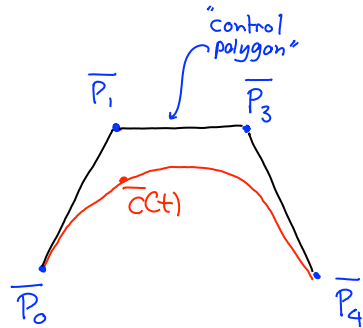
$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t) \quad \text{where}$$

called the Bernstein polynomials of degree N

$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i \\ = \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$

⊕ Tangents at endpoints are along the 1st & last edges of control polygon:

$$\frac{d}{dt} \bar{c}(t) = \sum_{i=1}^N \bar{P}_i \frac{d}{dt} B_i^N(t) \\ \stackrel{\text{w/ some work}}{=} N \sum_{i=0}^{N-1} (\bar{P}_{i+1} - \bar{P}_i) B_i^{N-1}(t) \\ \begin{matrix} N(\bar{P}_1 - \bar{P}_0) & N(\bar{P}_N - \bar{P}_{N-1}) \\ \text{for } t=0 & \text{for } t=1 \end{matrix}$$



Bézier Curves: Pros & Cons

Expression in compact form

$$\bar{c}(t) = \sum_{i=0}^N \bar{P}_i B_i^N(t) \quad \text{where}$$

called the Bernstein polynomials of degree N

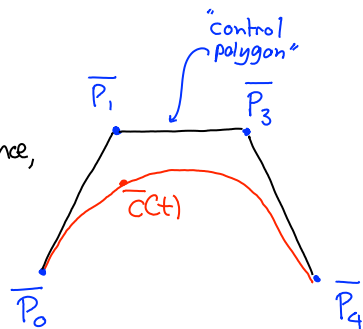
$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i \\ = \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$

Advantages

- Intuitive control for $N \leq 3$
- Derivatives easy to compute
- Nice properties (affine invariance, diminishing variations)

Disadvantages

- Scheme is still global (curve function of all control pts)



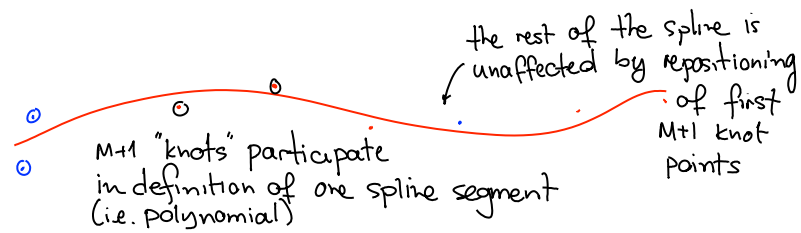
Topic 15:

Interpolating Curves

- Intro to curve interpolation & approximation
- Polynomial interpolation
- Bézier curves
- Cardinal splines

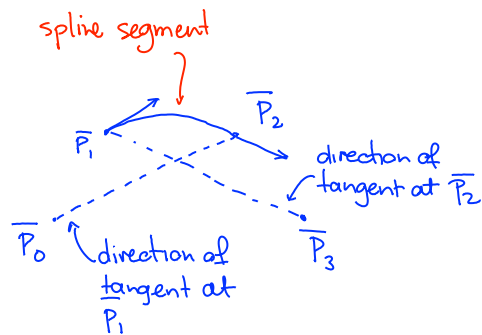
Splines: Local Control + Reduced Continuity

- Idea: provide local control of curve approximating N points by sacrificing C^N continuity
- M -degree spline: a piecewise-polynomial curve of degree M
- Splines often defined to have C^{M-2} continuity at the knot points (aka. control points)



Cubic Cardinal Splines

- Idea: provide local control of curve interpolating $N-2$ points, with C^1 continuity
- Cubic polynomials used as basic building block

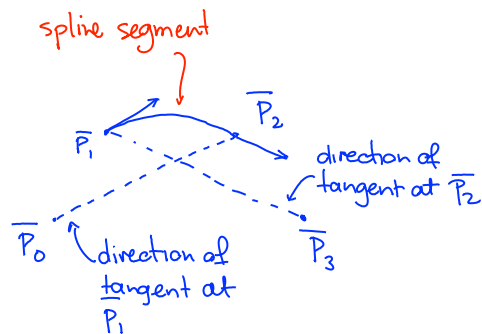


\bar{P}_1, \bar{P}_2 act as endpoint constraints

\bar{P}_0, \bar{P}_3 define derivative constraints

Cubic Cardinal Splines: Specifying the Curve

- Approach:
- ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 - ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

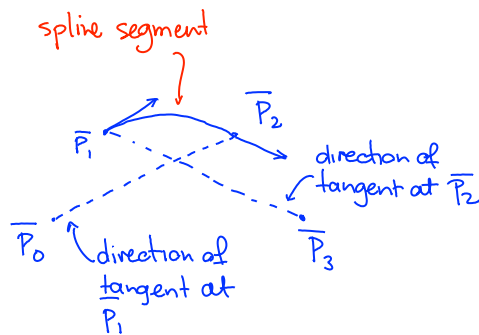


\bar{P}_1, \bar{P}_2 act as endpoint constraints

\bar{P}_0, \bar{P}_3 define derivative constraints

Cubic Cardinal Splines: Defining 1st Segment

- Approach: ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

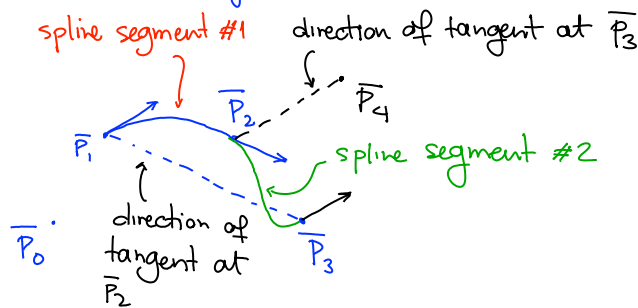


\bar{P}_1, \bar{P}_2 act as endpoint constraints
 \bar{P}_0, \bar{P}_3 define derivative constraints

Cubic Cardinal Splines: Defining 2nd Segment

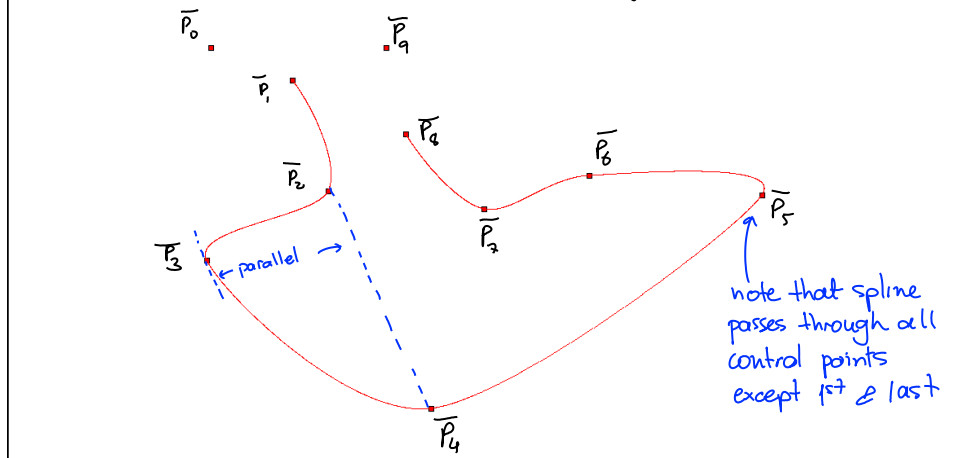
- Approach: ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

Example: Adding a fifth point adds a new segment



Cubic Cardinal Splines: General Case

- Approach: ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

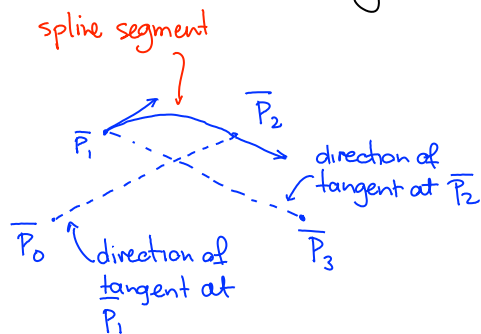


Cardinal Splines: The Strain Parameter

- Approach: ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

tangent at $\bar{P}_i = k(\bar{P}_{i+1} - \bar{P}_{i-1})$

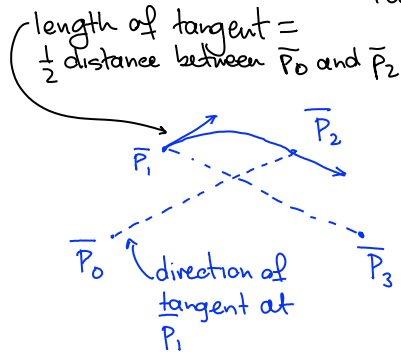
called a strain parameter



Catmull-Rom Splines

- Approach: ① A user only specifies points $\bar{P}_0, \bar{P}_1, \dots$
 ② tangent at \bar{P}_i set to be parallel to vector connecting \bar{P}_{i-1} and \bar{P}_{i+1}

tangent at $\bar{P}_i = k(\bar{P}_{i+1} - \bar{P}_{i-1})$
 called a strain parameter

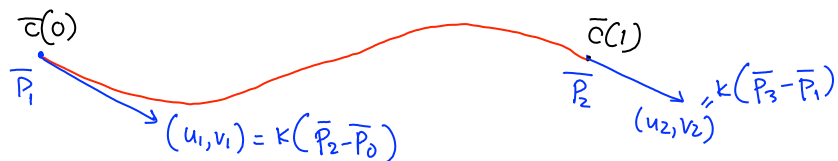


Note: if $k = \frac{1}{2}$
 the spline is called a Catmull-Rom spline

Specifying the Poly via Tangent Constraints

Instead of specifying 4 control points, we could specify 2 points and 2 derivatives

$$\begin{bmatrix} \frac{dx}{dt}(t) & \frac{dy}{dt}(t) \end{bmatrix} = \begin{bmatrix} 1 & 2t & 3t^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



Cardinal Splines: Solving for the Segment

Coeffs

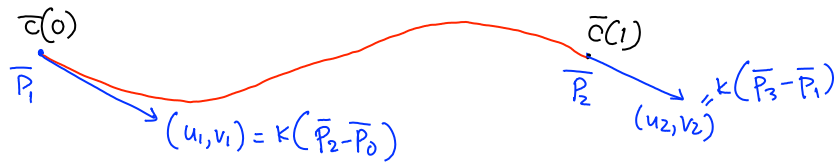
for $t=0$ $[x_1 \ y_1] = [1 \ t \ t^2 \ t^3] \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$

↑
endpoint constraint

$$[k(x_2-x_0) \ k(y_2-y_0)] = [1 \ 2t \ 3t^2] \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

↑
derivative constraint

+ 2 more eqs for other endpoint ($t=1$)

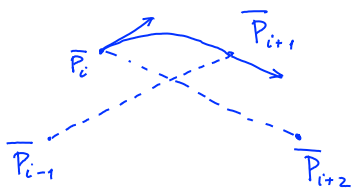


Cubic Cardinal Spline Segment vs. Bézier Curve

Curve

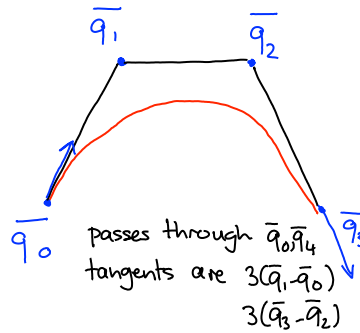
The two curves are actually equivalent:
given a cardinal spline, we can compute the
control polygon of the equivalent Bézier curve

Cardinal Spline Segment



passes through \bar{P}_i, \bar{P}_{i+1}
tangents are $k(\bar{P}_{i+1} - \bar{P}_{i-1})$
 $k(\bar{P}_{i+2} - \bar{P}_i)$

Bézier Curve



passes through \bar{Q}_0, \bar{Q}_4
tangents are $3(\bar{Q}_1 - \bar{Q}_0)$
 $3(\bar{Q}_3 - \bar{Q}_2)$

Control Polygon of a Cardinal Spline Segment

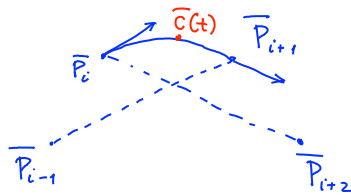
In order to have $\bar{c}(t) = \bar{r}(t)$ for all t , it must be

$$\bar{q}_0 = \bar{p}_i, \quad \bar{q}_4 = \bar{p}_{i+1}$$

$$k(\bar{p}_{i+1} - \bar{p}_{i-1}) = 3(\bar{q}_1 - \bar{q}_0) \Rightarrow k(\bar{p}_{i+1} - \bar{p}_{i-1}) = 3\bar{q}_1 - 3\bar{q}_0$$

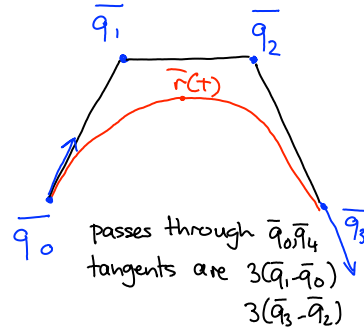
$$\Rightarrow \bar{q}_1 = \frac{k}{3}(\bar{p}_{i+1} - \bar{p}_{i-1}) + \bar{p}_i$$

Cardinal Spline Segment



passes through \bar{p}_i, \bar{p}_{i+1}
 tangents are $k(\bar{p}_{i+1} - \bar{p}_{i-1})$
 $k(\bar{p}_{i+2} - \bar{p}_i)$

Bezier Curve



passes through \bar{q}_0, \bar{q}_4
 tangents are $3(\bar{q}_1 - \bar{q}_0)$
 $3(\bar{q}_3 - \bar{q}_2)$

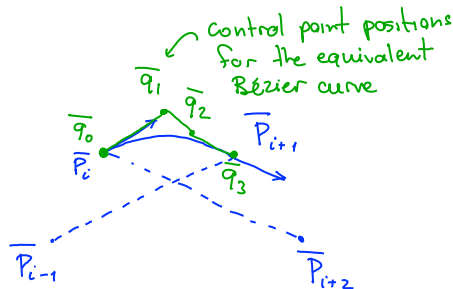
Control Polygon of a Cardinal Spline Segment

In order to have $\bar{c}(t) = \bar{r}(t)$ for all t , it must be

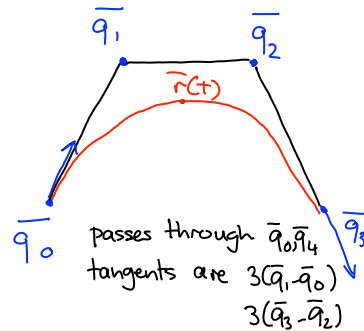
$$\bar{q}_0 = \bar{p}_i, \quad \bar{q}_4 = \bar{p}_{i+1}$$

$$\bar{q}_1 = \frac{k}{3}(\bar{p}_{i+1} - \bar{p}_{i-1}) + \bar{p}_i, \quad \bar{q}_2 = \bar{p}_{i+1} - \frac{k}{3}(\bar{p}_{i+2} - \bar{p}_i)$$

Cardinal Spline Segment



Bezier Curve



passes through \bar{q}_0, \bar{q}_4
 tangents are $3(\bar{q}_1 - \bar{q}_0)$
 $3(\bar{q}_3 - \bar{q}_2)$

Topic 15+:

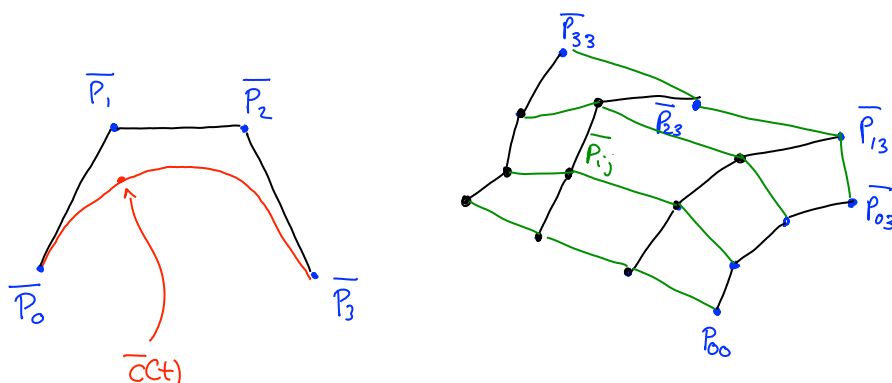
Interpolating Surfaces

- Bézier surface patches

Bézier Surface Patches

Basic idea: Define a 4x4 grid of 3D control points
⇒ Produces a cubic surface patch

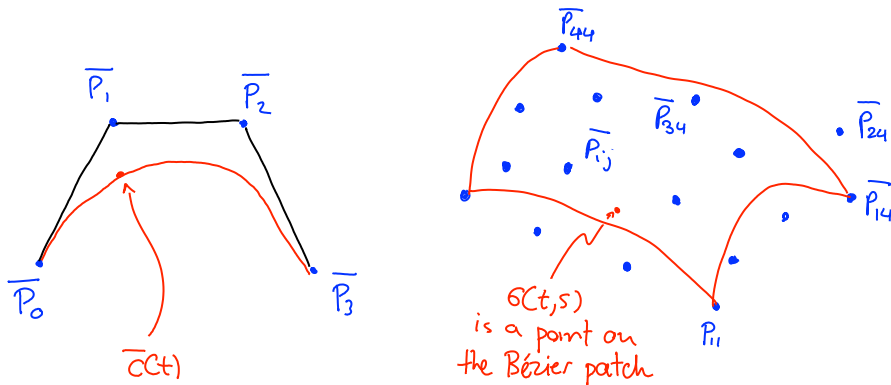
$$G(t,s) = \text{3rd-order poly in } s \text{ and } t \\ s, t \in [0,1]$$



Bézier Surface Patches

Basic idea: Define a 4x4 grid of 3D control points
 \Rightarrow Produces a cubic surface patch

$$G(t,s) = \text{3rd-order poly in } s \text{ and } t \\ \text{with } t, s \in [0,1]$$

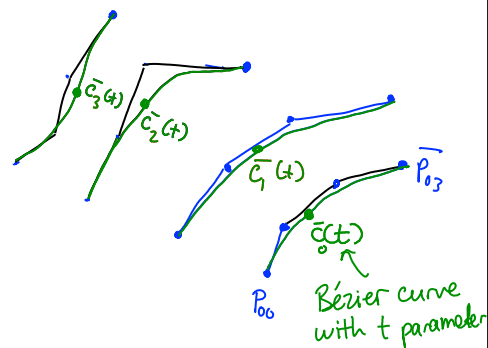


Bézier Surface Patches: Algorithm

Basic idea: ① Define a 4x4 grid of 3D control points
 \Rightarrow defines 4 Bézier curves
 $C_0(t), C_1(t), C_2(t), C_3(t)$ parameterized by t

② Given a value for t
 \Rightarrow obtain 4 points
 $\bar{C}_0(t), \bar{C}_1(t), \bar{C}_2(t), \bar{C}_3(t)$

③ These points become the control points of a Bézier curve parameterized by s



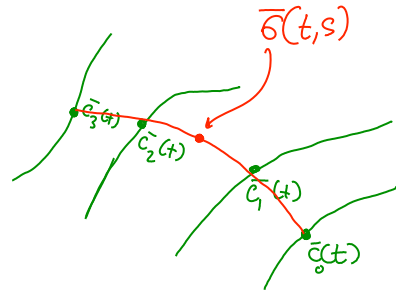
Bézier Surface Patches: Algorithm

Basic idea: ① Define a 4×4 grid of 3D control points
 \Rightarrow defines 4 Bézier curves
 $c_0(t), c_1(t), c_2(t), c_3(t)$ parameterized by t

② Given a value for t
 \Rightarrow obtain 4 points
 $\bar{c}_0(t), \bar{c}_1(t), \bar{c}_2(t), \bar{c}_3(t)$

③ These points become the control points of a Bézier curve parameterized by s

④ The patch is defined by points on that curve, for $s \in [0,1]$



Bézier Surface Patches: Polynomial

Expression

Basic idea: Define a 4×4 grid of 3D control points
 \Rightarrow Produces a cubic surface patch

$$\bar{G}(t,s) = \text{3rd-order poly in } s \text{ and } t \\ s, t \in [0,1]$$

General expression:

$$\bar{G}(t,s) = \sum_{i=0}^3 \sum_{j=0}^3 \bar{P}_{ij} \cdot \underbrace{B_i^3(t)}_{\text{Bernstein poly of degree 3}} \cdot \underbrace{B_j^3(s)}_{\text{Bernstein poly of degree 3}}$$

