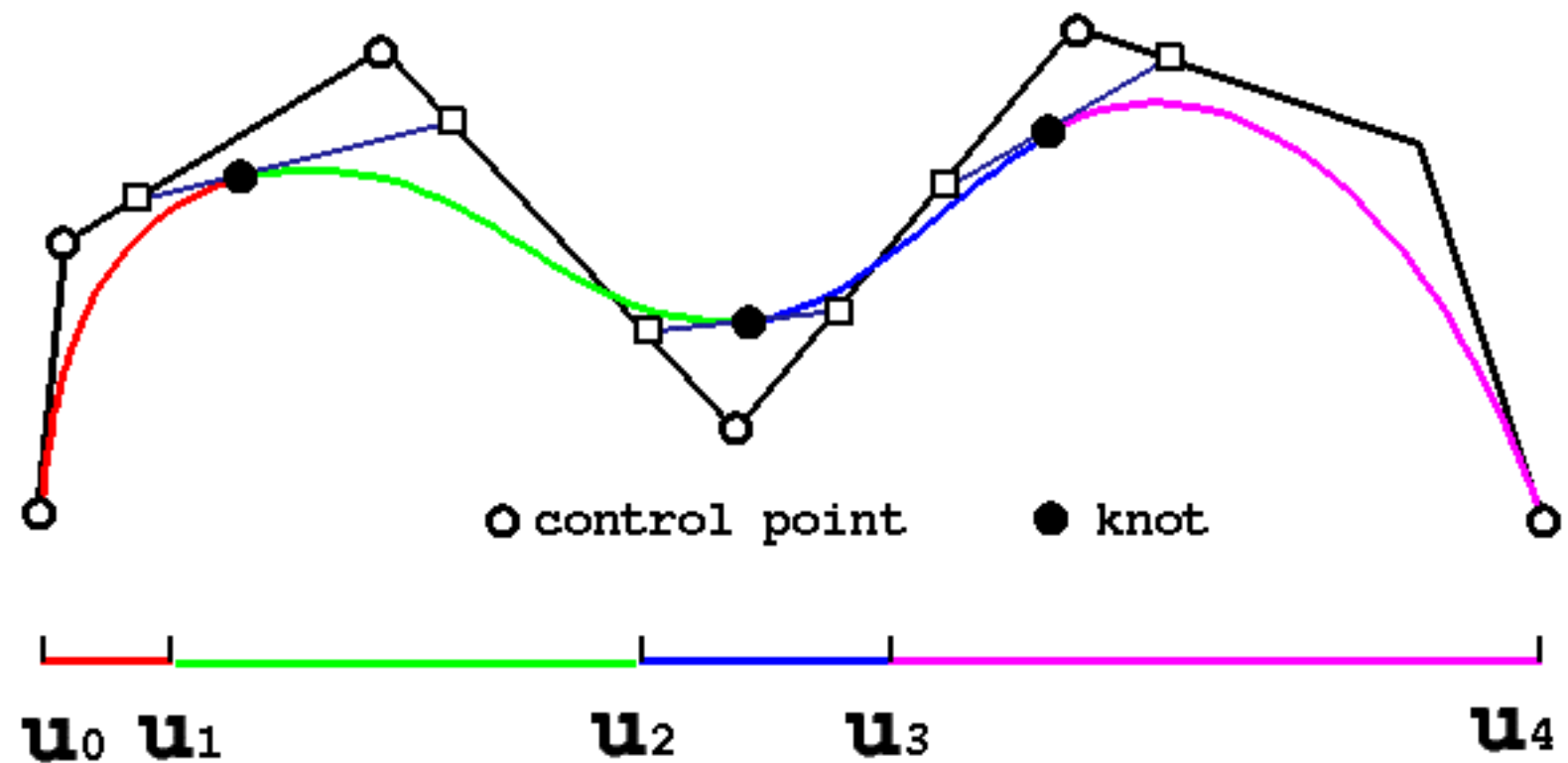


Topic 12: Interpolating Curves

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

What are Splines?

- Numeric function that is piecewise-defined by polynomial functions
- Possesses a high degree of smoothness where pieces connect
- These are intuitively called “knots”



History

- Used by engineers in ship building and airplane design before computers were around
- Used to create smoothly varying curves
- Variations in curve achieved by the use of weights (like control points)



Applications

- Specify smooth camera path in scene along spline curve
- Rollercoaster tracks
- Curved smooth bodies and shells (planes, boats, etc)

Motivation and Goal

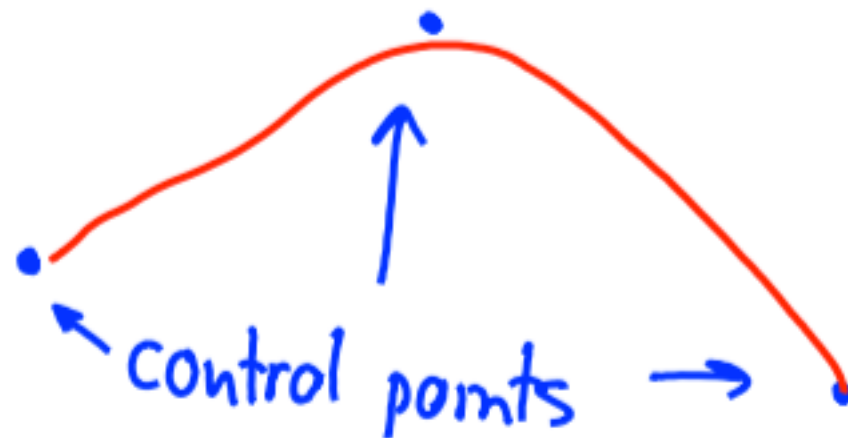
- Expand the capabilities of shapes beyond lines and conics, simple analytic functions and to allow design constraints.

Design issues

- Create curves that can have constraints specified
- Have natural and intuitive interaction
- Controllable smoothness
- Control (local vs global)
- Analytic derivatives that are easy to compute
- Compactly represented
- Other geometric properties (planarity, tangent/curvature control)

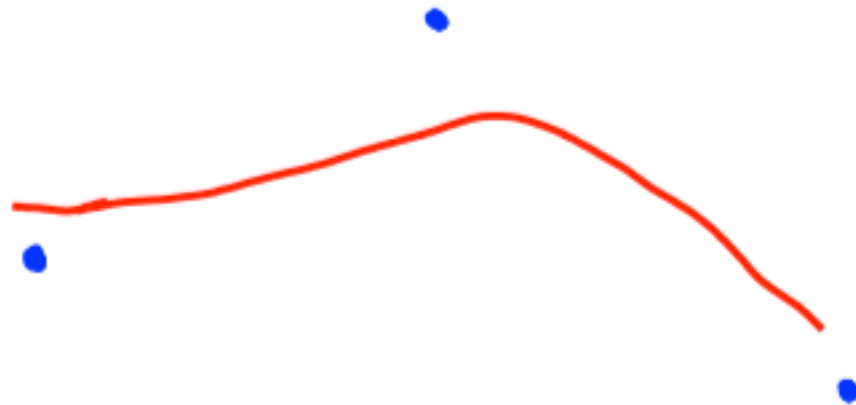
Interpolation

- Interpolating splines: pass through all the data points (control points).
Example: Hermite splines



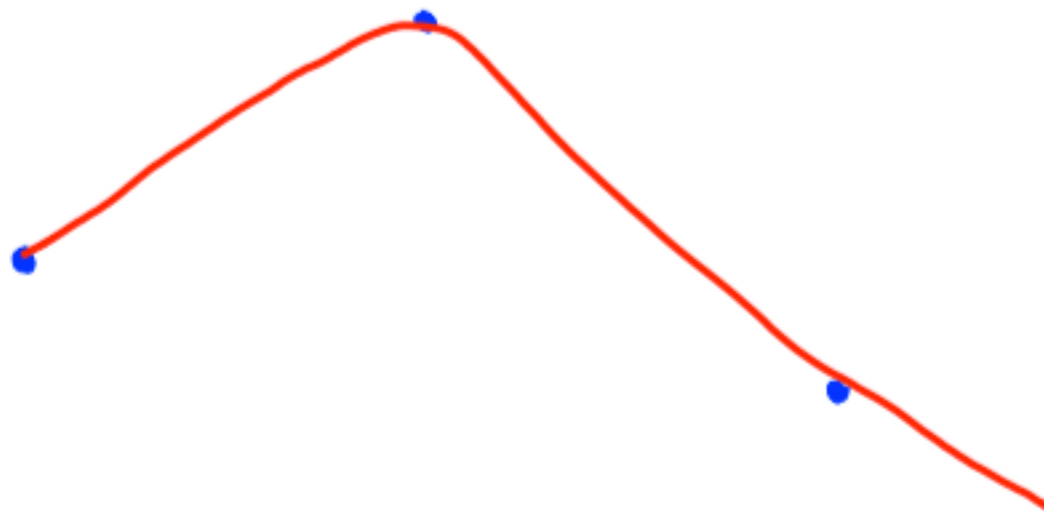
Approximation

- Curve approximates but does not go through all of the control points.
- Comes close to them.



Extrapolation

- Extend the curve beyond the domain of the control points



Local properties

- Continuity
- Position at a specific place on the curve
- Direction at a specific place on the curve
- Curvature

Global properties

- Closed or open curve
- Self intersection
- Length

Local vs Global Control

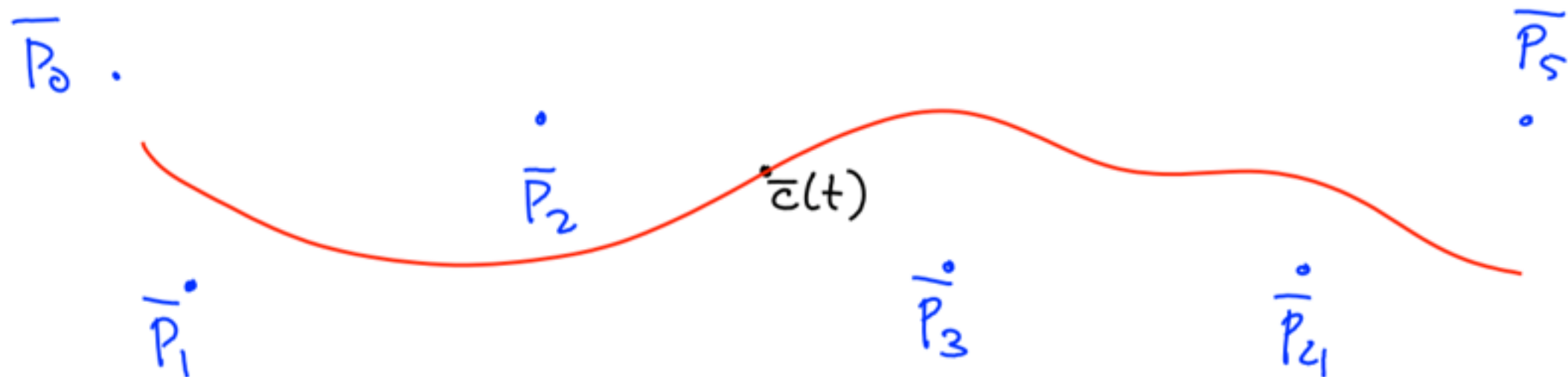
- Local control changes curve only locally while maintaining some constraints
- Modifying point on curve affects local part of curve or entire curve

Parametric and Geometric Continuity

- When piecing together smooth curves, consider the degrees of smoothness at the joints.
- Parametric Continuity: differentiability of the parametric representation (C^0 , C^1 , C^2 , ...)
- Geometric Continuity: smoothness of the resulting displayed shape ($G^0=C^0$, G^1 =tangent-cont., G^2 =curvature-cont.)

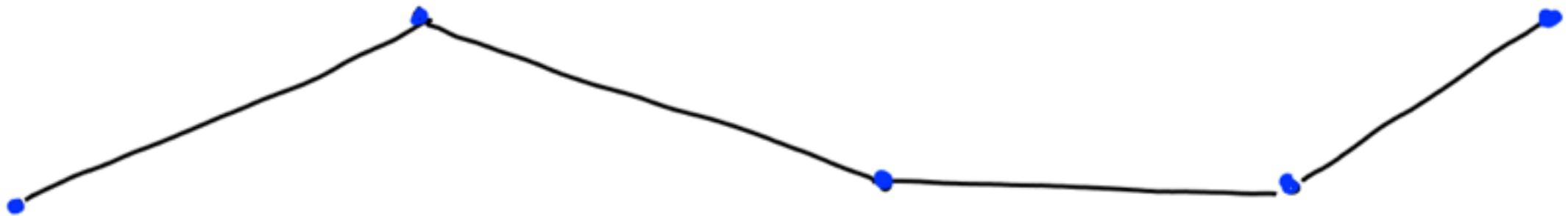
2D Curve Design: General Problem Statement

- Given N control points, P_i , $i = 0 \dots n - 1$, $t \in [0, 1]$ (by convention)
- Define a curve $c(t)$ that interpolates / approximates them
- Compute its derivatives (and tangents, normals etc)



Linear Interpolation

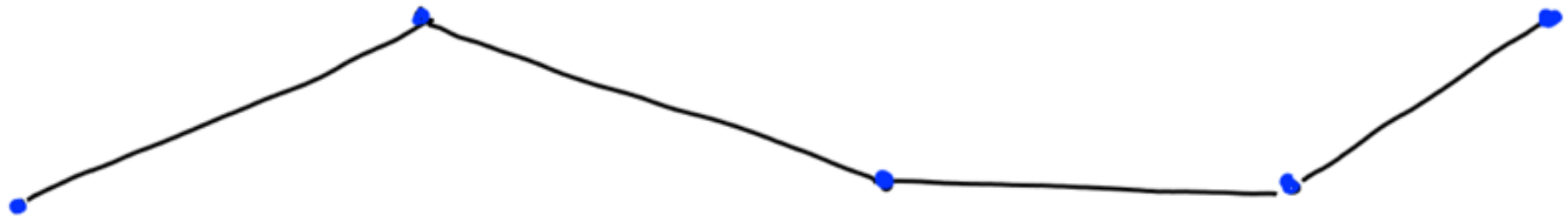
- The simplest possible interpolation technique
- Create a piecewise linear curve that connects the control points



- Q: What is the disadvantage of such a technique?

Linear Interpolation

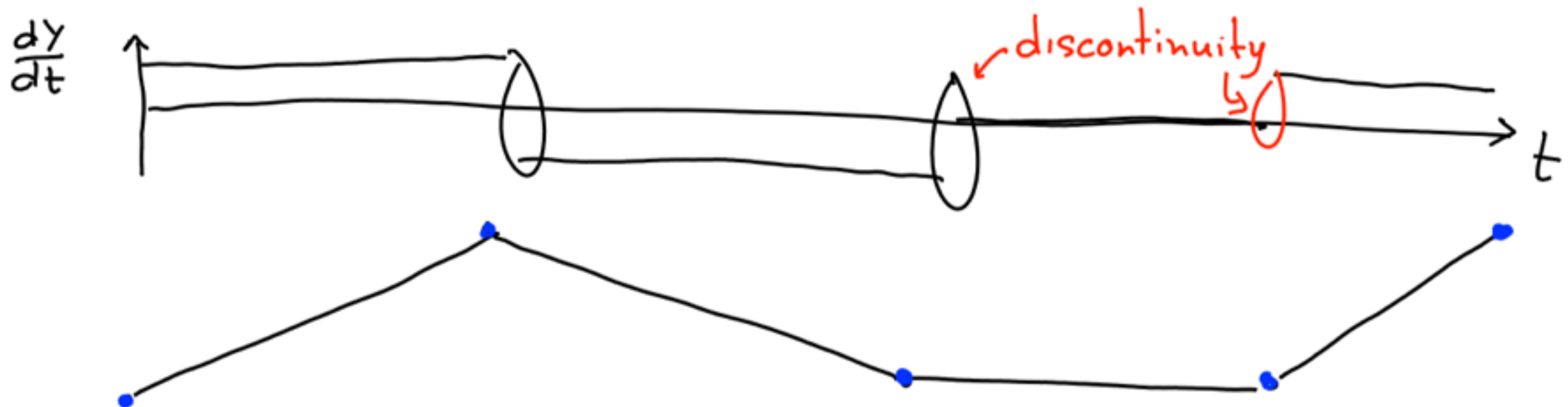
- The simplest possible interpolation technique
- Create a piecewise linear curve that connects the control points



- Q: What is the disadvantage of such a technique?
- A: The curves may be continuous but its derivatives are not...

Linear Interpolation

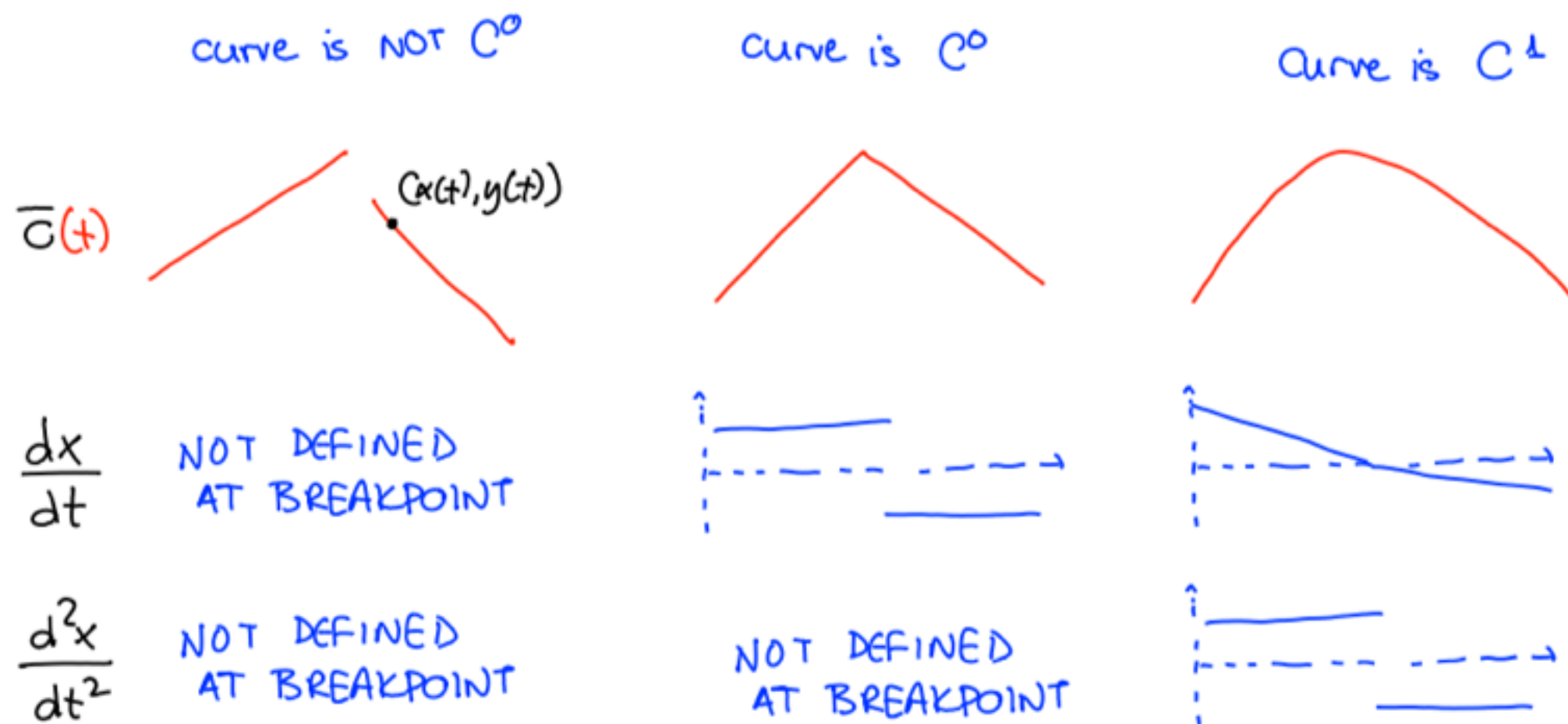
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- Q: What is the disadvantage of such a technique?
- A: The curves may be continuous but its derivatives are not...

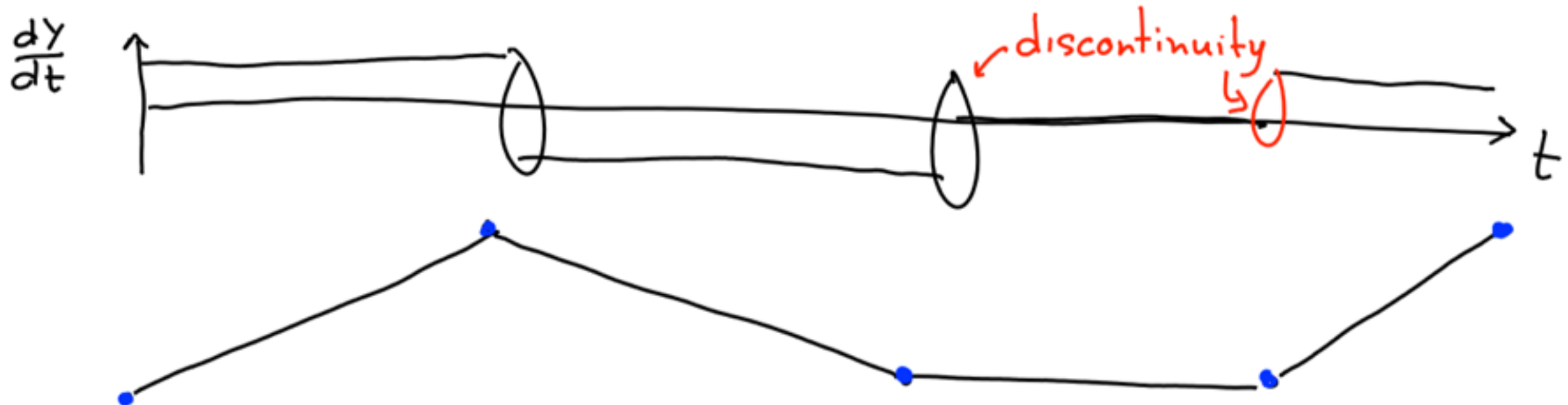
C^n continuity

- Definition: a function is called C^n if its n^{th} order derivative is continuous everywhere



Linear Interpolation

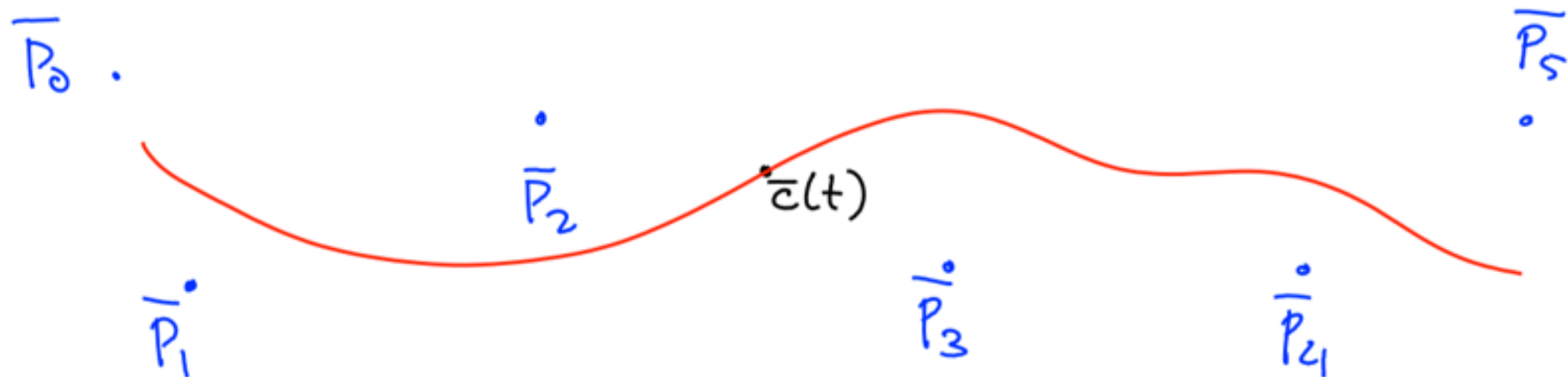
- The simplest possible interpolation technique
- Create a piecewise linear curve that connects the control points



- Q: What is the disadvantage of such a technique?
- Curve has only C^0 continuity

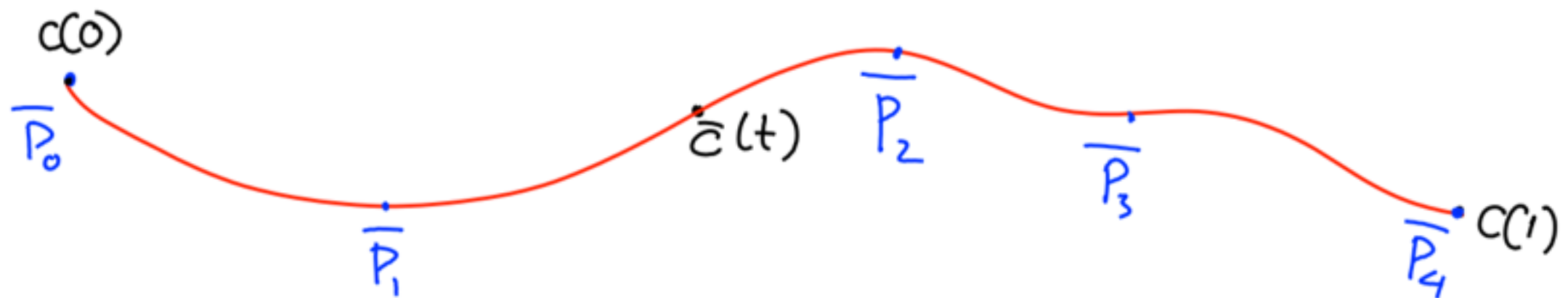
2D Curve Design: General Problem Statement

- Given N control points, P_i , $i = 0 \dots n-1$, $t \in [0, 1]$ (by convention)
- Define a curve $c(t)$ that interpolates / approximates them
- Compute its derivatives (and tangents, normals etc)
- We will seek functions that are at least C^1



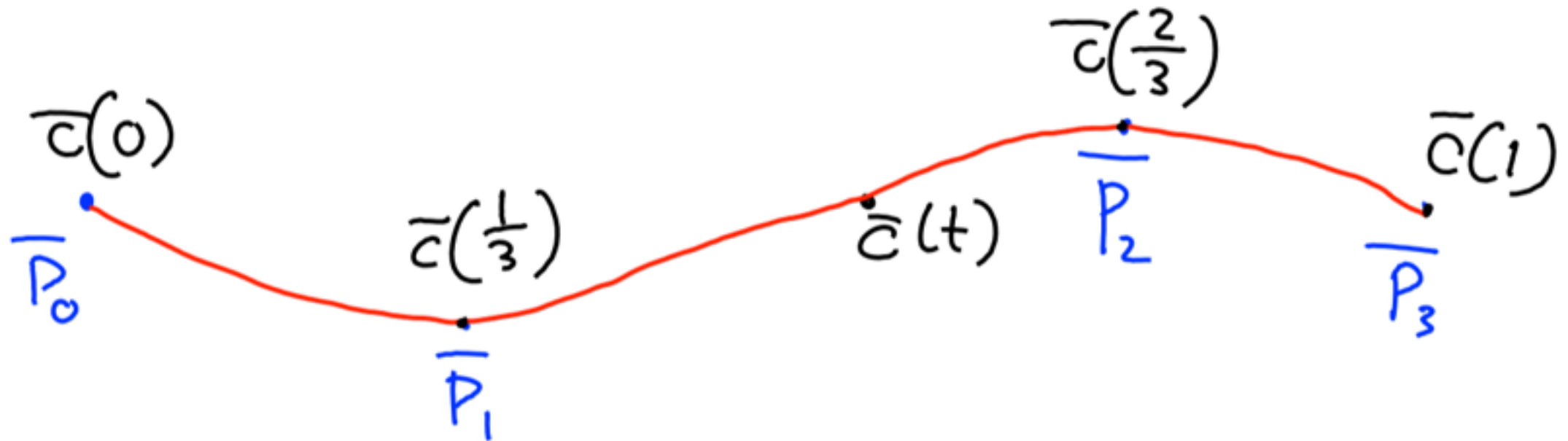
Polynomial Interpolation

- Given N control points, P_i , $i = 0 \dots n-1$, $t \in [0, 1]$ (by convention)
 - Define $(N-1)$ -order polynomial $x(t)$, $y(t)$ such that $x(i/(N-1)) = x_i$, $y(i/(N-1)) = y_i$ for $i = 0, \dots, N-1$
- Compute its derivatives (and tangents, normals etc)



Cubic Interpolation

- Given 4 control points, P_i , $i = (x_i, y_i)$, for $i = 0, \dots, 3$
- Define 3rd-order polynomial $x(t)$, $y(t)$ such that $x(i/3) = x_i$, $y(i/3) = y_i$
- Compute its derivatives (and tangents, normals etc)



Cubic Interpolation: Basic Equations

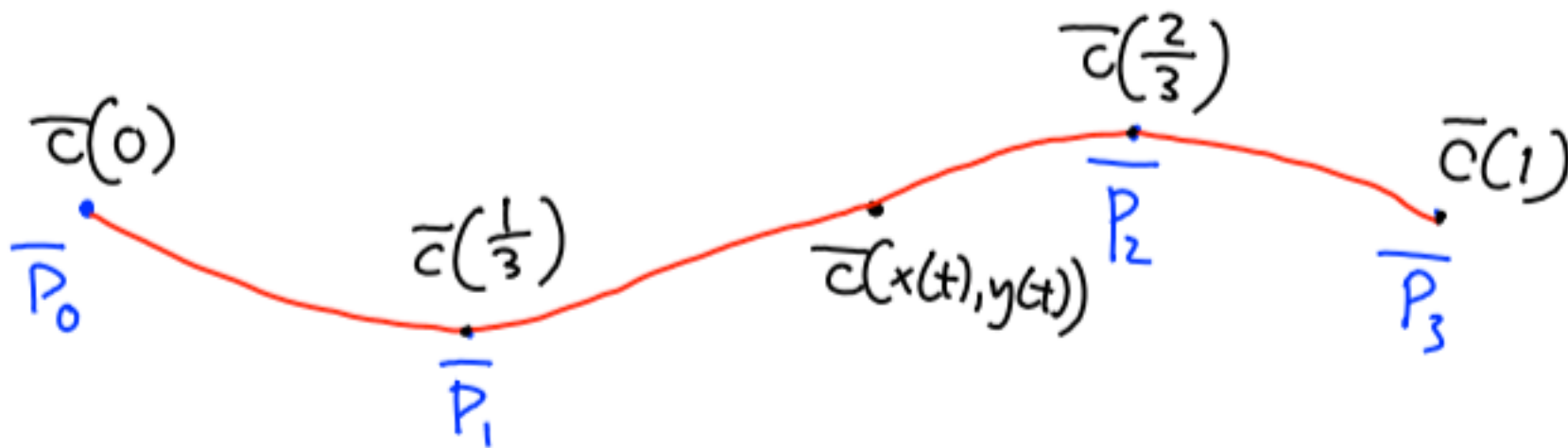
$$\left. \begin{aligned} x(t) &= a_0 + a_1 t + a_2 t^2 + a_3 t^3 \\ y(t) &= b_0 + b_1 t + b_2 t^2 + b_3 t^3 \end{aligned} \right\} \begin{array}{l} \text{-given } \bar{P}_1, \bar{P}_2, \bar{P}_3, \bar{P}_4 \\ \text{compute } a_i, b_i \end{array}$$

Equations for one control point:

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

Equations in matrix form:

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



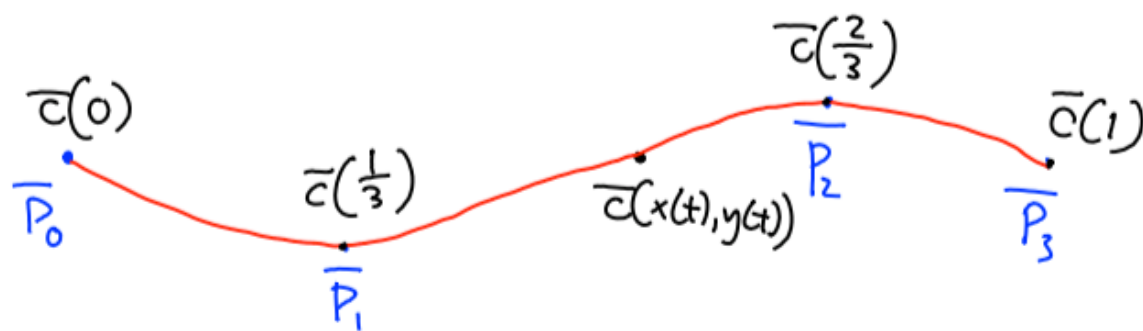
Cubic Interpolation: Computing Coeffs

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

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

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 Coeffs

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

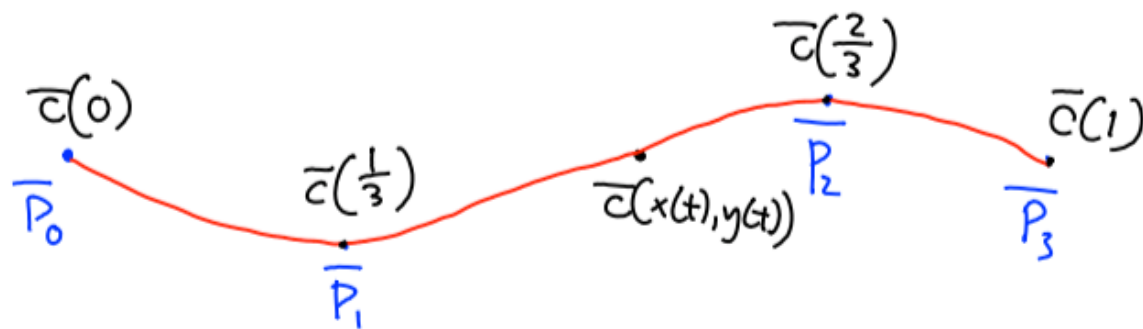
$$\underbrace{\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}}_{\text{known}} = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1/3 & (1/3)^2 & (1/3)^3 \\ 1 & 2/3 & (2/3)^2 & (2/3)^3 \\ 1 & 1 & 1 & 1 \end{bmatrix}}_{\text{known}} \underbrace{\begin{bmatrix} a_1 & b_1 \\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

⇒ solve system
in terms of
unknown
matrix

$$X = A^{-1}C$$

Equations in matrix form:

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

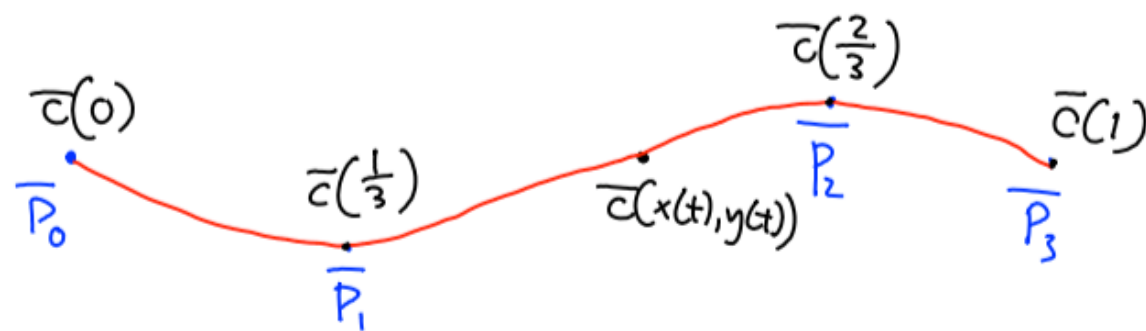


Cubic Interpolation: Computing Coeffs

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

Coefficients of interpolating polynomial computed by:

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



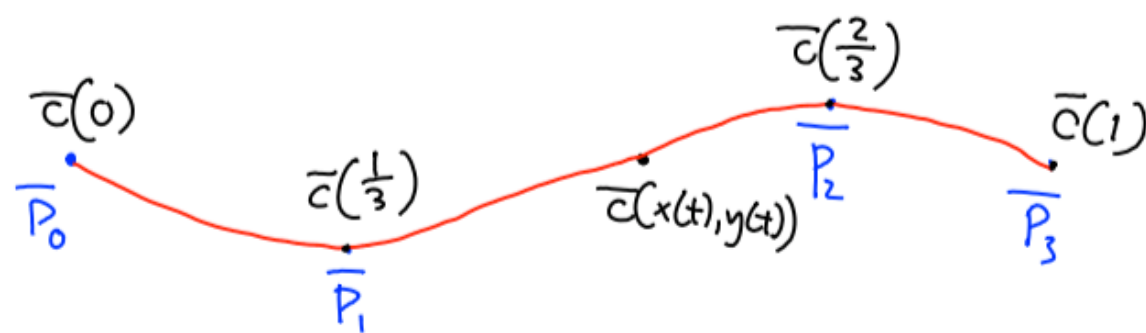
Equations in matrix form:

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

Cubic Interpolation: Evaluating the Polynomial

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

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



Equations in matrix form:

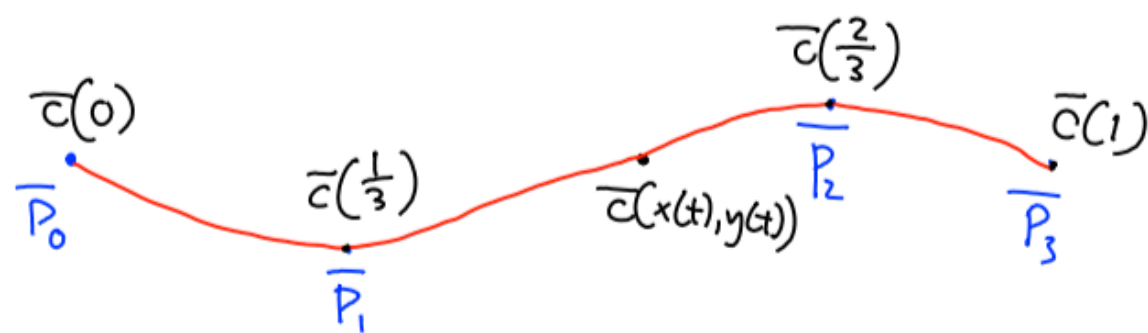
$$\begin{bmatrix} x_1 & y_1 \end{bmatrix} = \begin{bmatrix} 1 & \frac{1}{3} & (\frac{1}{3})^2 & (\frac{1}{3})^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

Cubic Interpolation: What if < 4 Control Points?

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

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

← # control points →



Equations in matrix form:

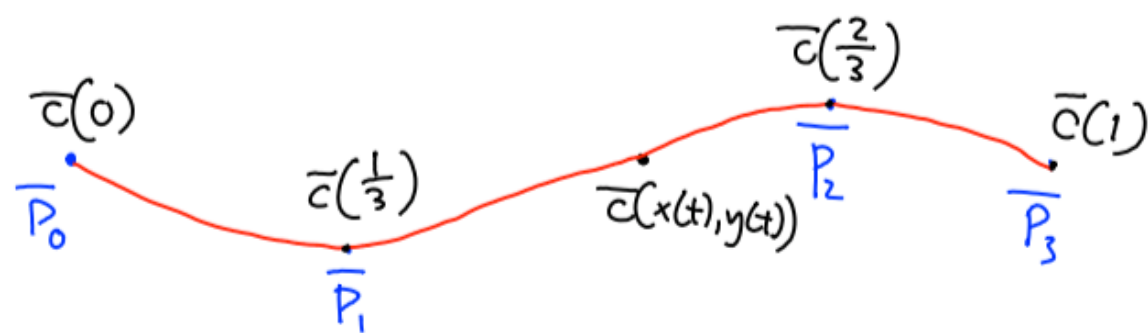
$$\begin{bmatrix} x_1 & y_1 \end{bmatrix} = \begin{bmatrix} 1 & \frac{1}{3} & (\frac{1}{3})^2 & (\frac{1}{3})^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

Cubic Interpolation: What if > 4 Control Points?

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

$$\begin{array}{c} \uparrow \\ \text{degree} \\ +1 \\ \downarrow \end{array} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \\ a_4 & b_4 \end{bmatrix} = \begin{bmatrix} \text{over-determined} \\ \text{linear system} \\ \Rightarrow \\ \text{poly cannot pass} \\ \text{through all pts} \end{bmatrix}^{-1} \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \\ x_4 & y_4 \end{bmatrix}$$

← # control points →



Equations in matrix form:

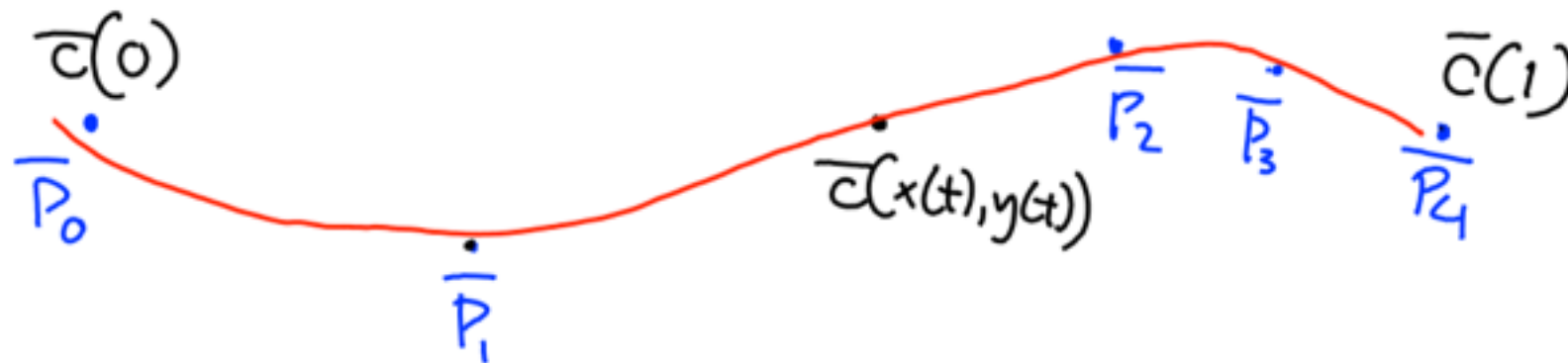
$$\begin{bmatrix} x_1 & y_1 \end{bmatrix} = \begin{bmatrix} 1 & \frac{1}{3} & (\frac{1}{3})^2 & (\frac{1}{3})^3 \end{bmatrix} \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

Exact Interpolation of N points

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

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

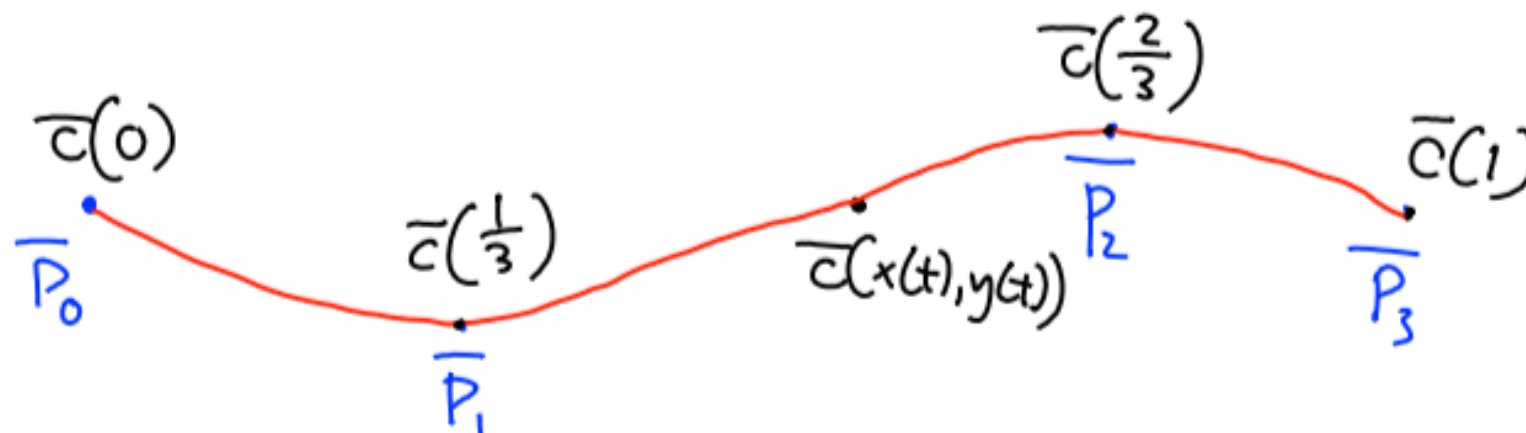
← # control points →



Cubic Interpolation: Evaluating Derivatives

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

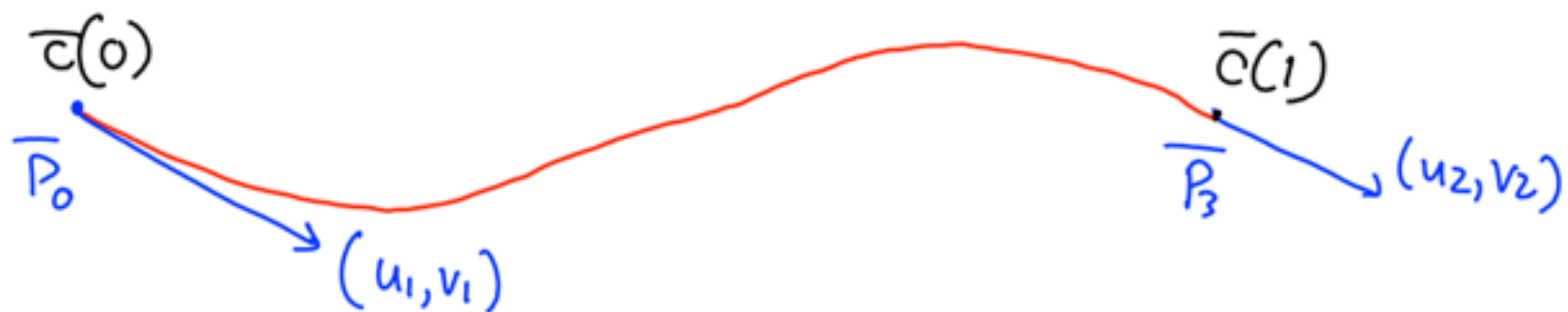
$$\begin{bmatrix} \frac{dx}{dt}(t) & \frac{dy}{dt}(t) \end{bmatrix} = \begin{bmatrix} 1 & 2t & 3t^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



Specifying the Poly via Tangent Constraints

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

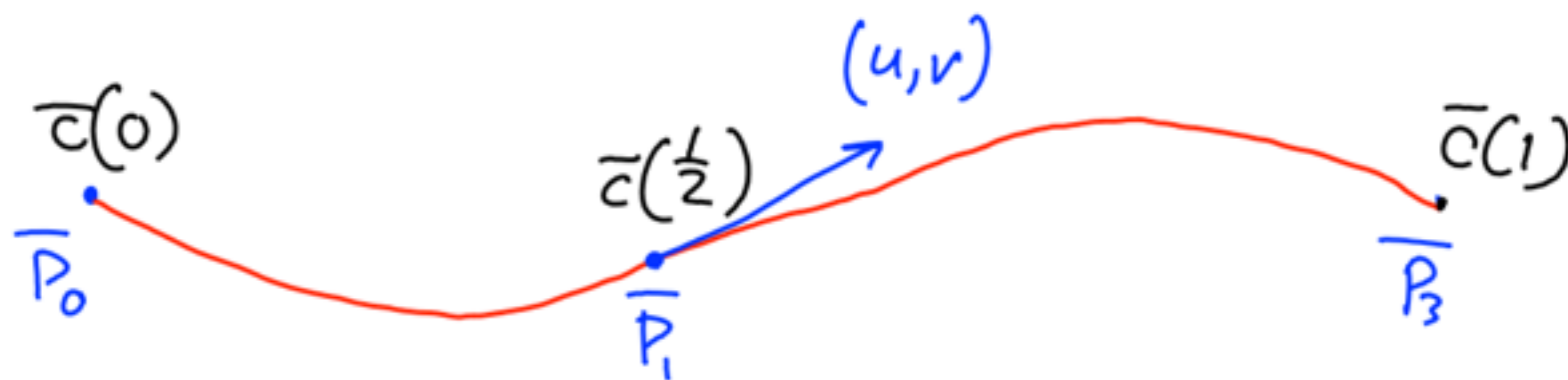
$$\begin{bmatrix} \frac{dx}{dt}(t) & \frac{dy}{dt}(t) \end{bmatrix} = \begin{bmatrix} 1 & 2t & 3t^2 \end{bmatrix} \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$



Specifying the Poly via Tangent Constraints

- Instead of specifying 4 control points, we could specify 3 points and a derivative.
- Replace the 4th pair of equations with

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

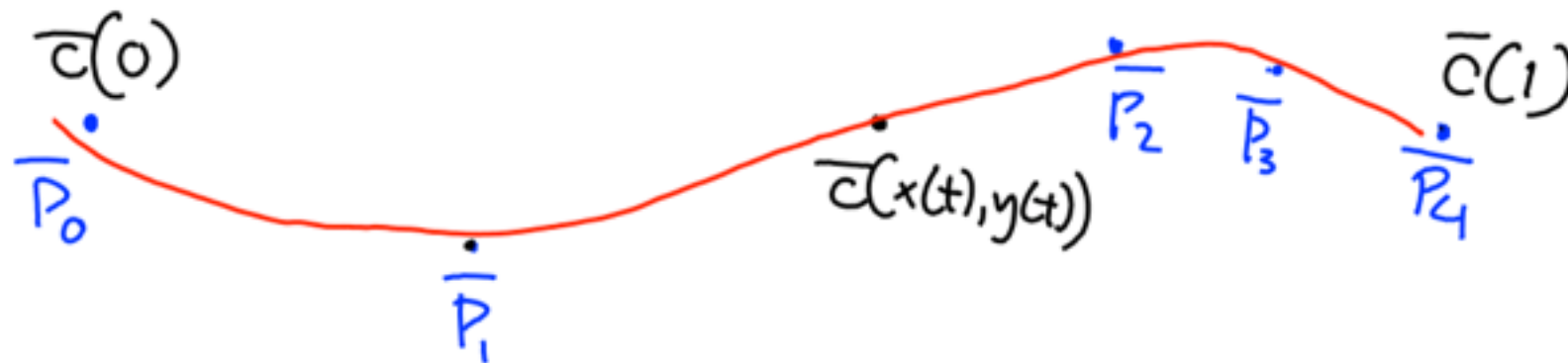


Degree-N Poly Interpolation: Major Drawback

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 and undesirable ways

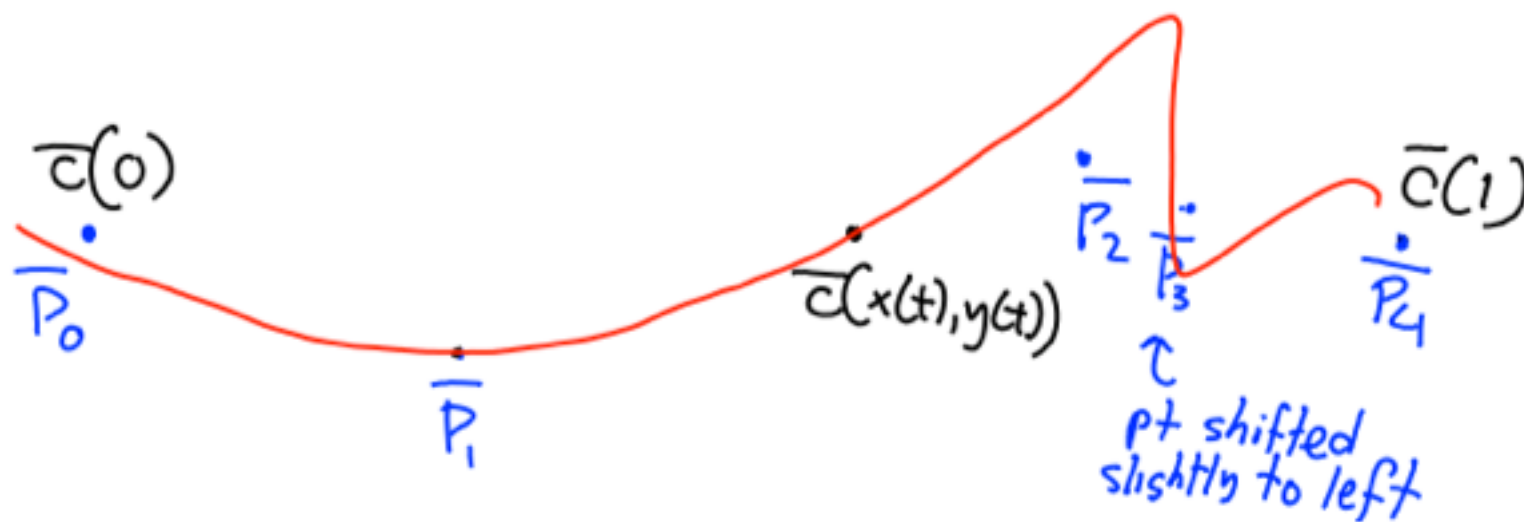


Degree-N Poly Interpolation: Major Drawback

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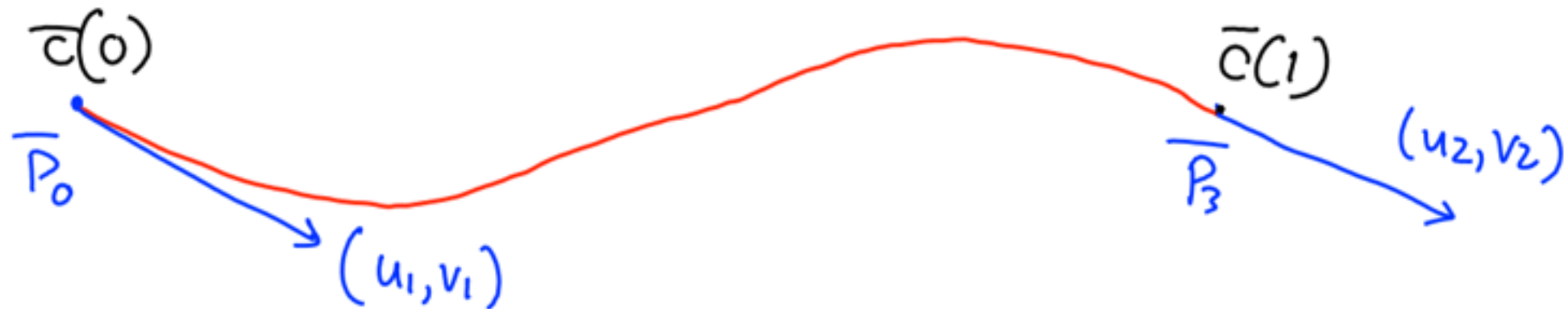
Topic 12: Interpolating Curves

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

Bézier Curves

Properties:

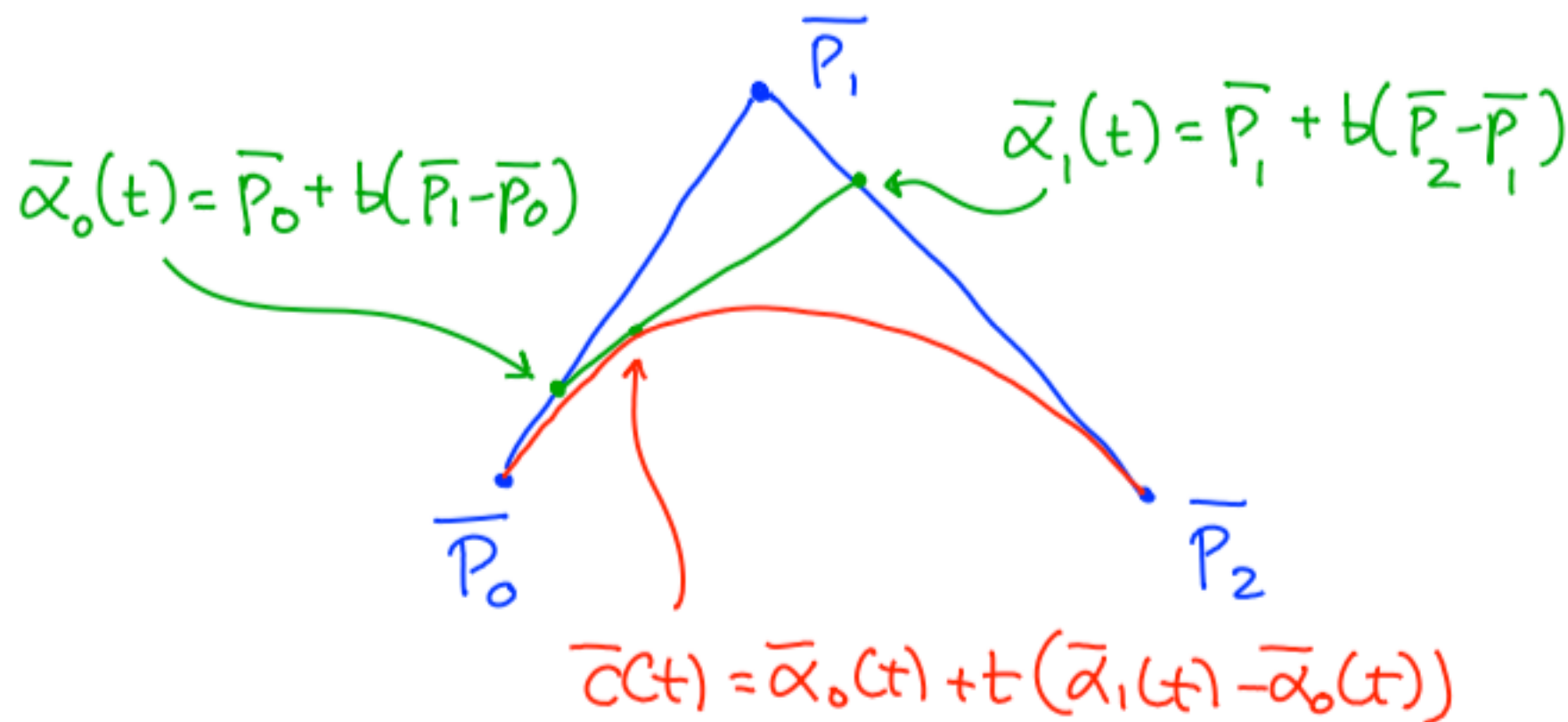
- Polynomial curves defined via endpoints and derivative constraints
- Derivative constraints defined implicitly through extra control points (that are not interpolated)
- They are approximating curves, not interpolating curves



Bézier Curves: Main Idea

Polynomial and 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 $\bar{c}(t)$

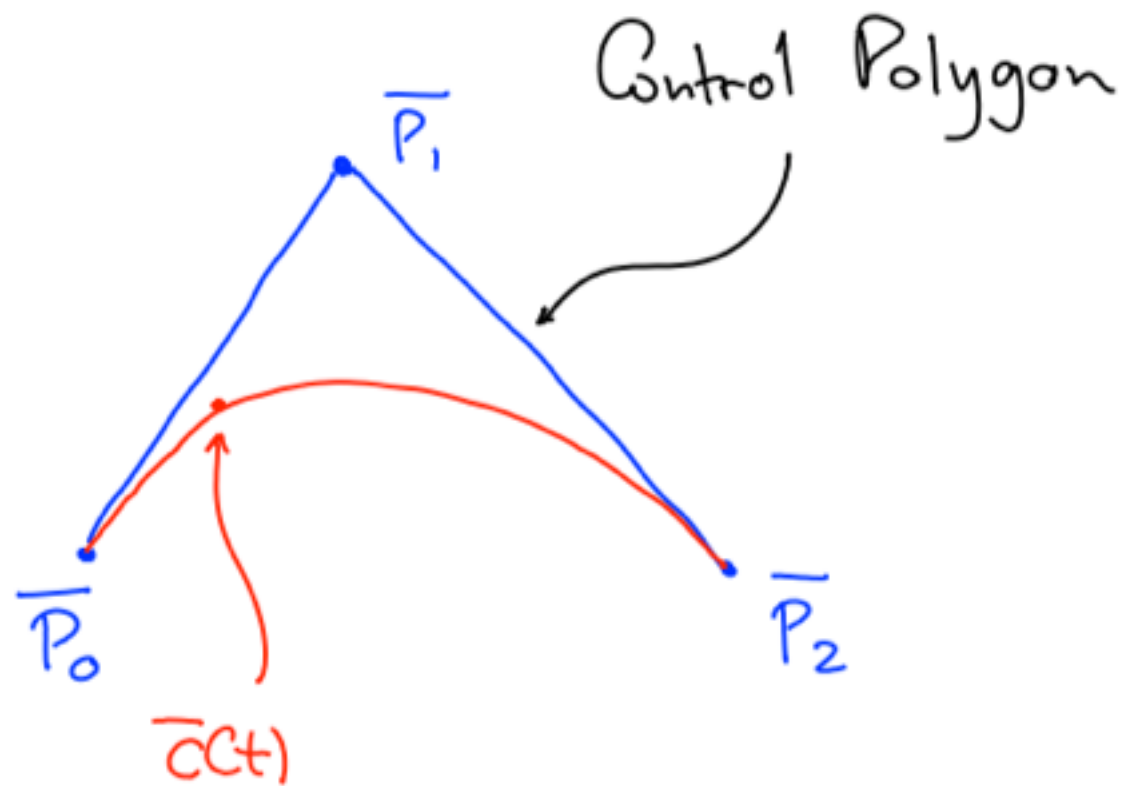
Q: Where have we seen such a cascade before?

Bézier Curves: Control Polygon

A Bézier curve is completely determined by its control polygon

We manipulate the curve by manipulating its polygon

Example: a double cascade



algorithm:

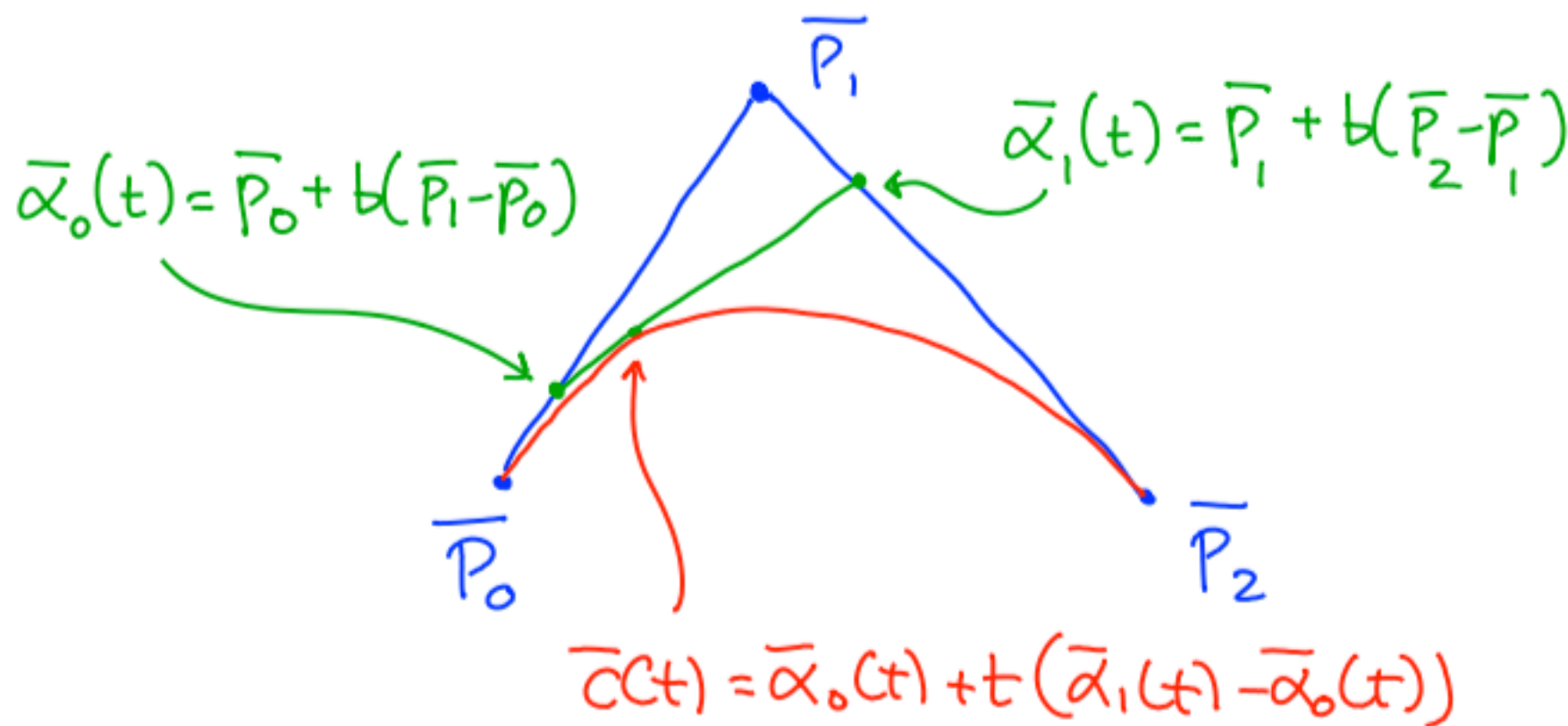
given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t

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 $\bar{c}(t)$

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

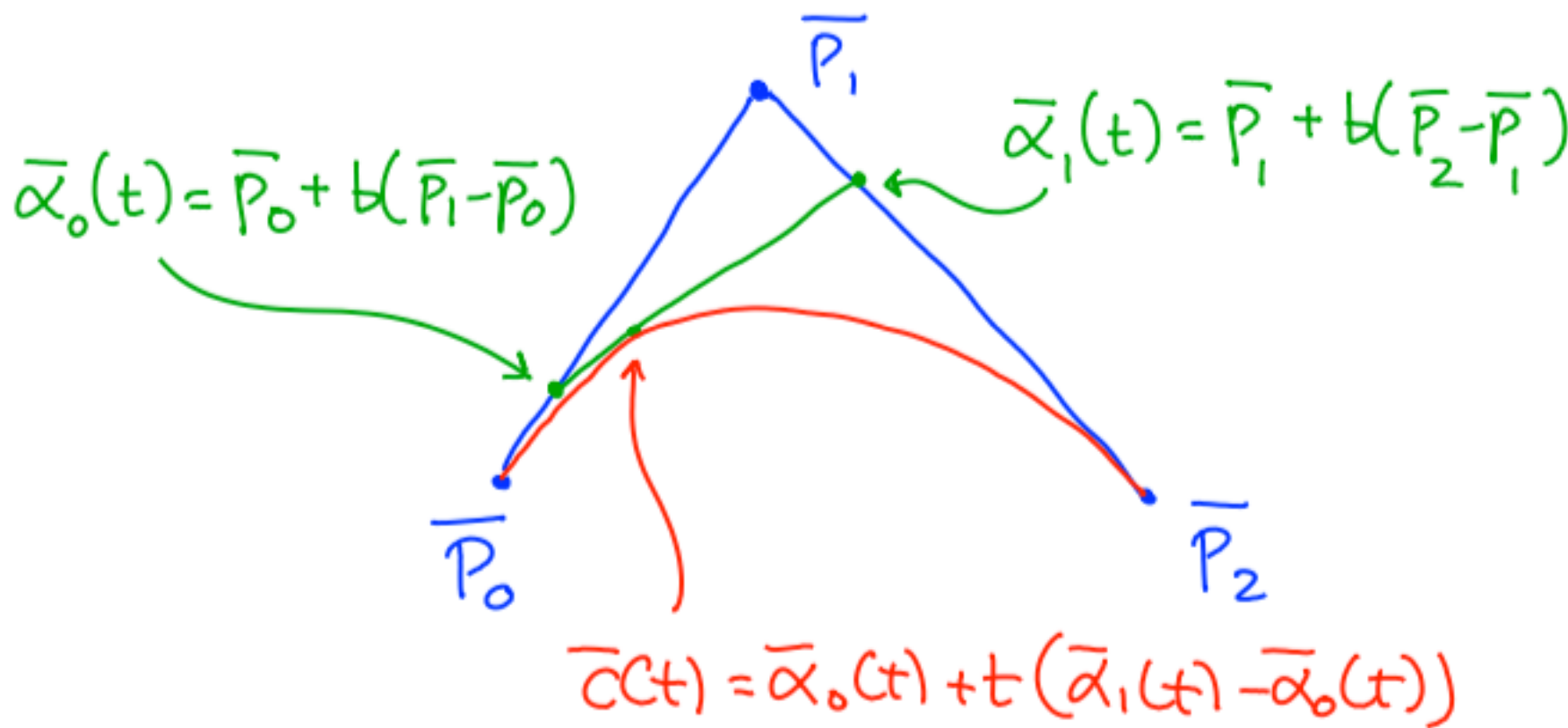
1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
2. linearly interpolate \bar{P}_1, \bar{P}_2 to get $\bar{\alpha}_1(t)$
3. linearly interpolate $\bar{\alpha}_0(t), \bar{\alpha}_1(t)$ to get $\bar{c}(t)$

Derivatives of the Bézier Curve

Computing the polynomial's derivatives:

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

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



algorithm:

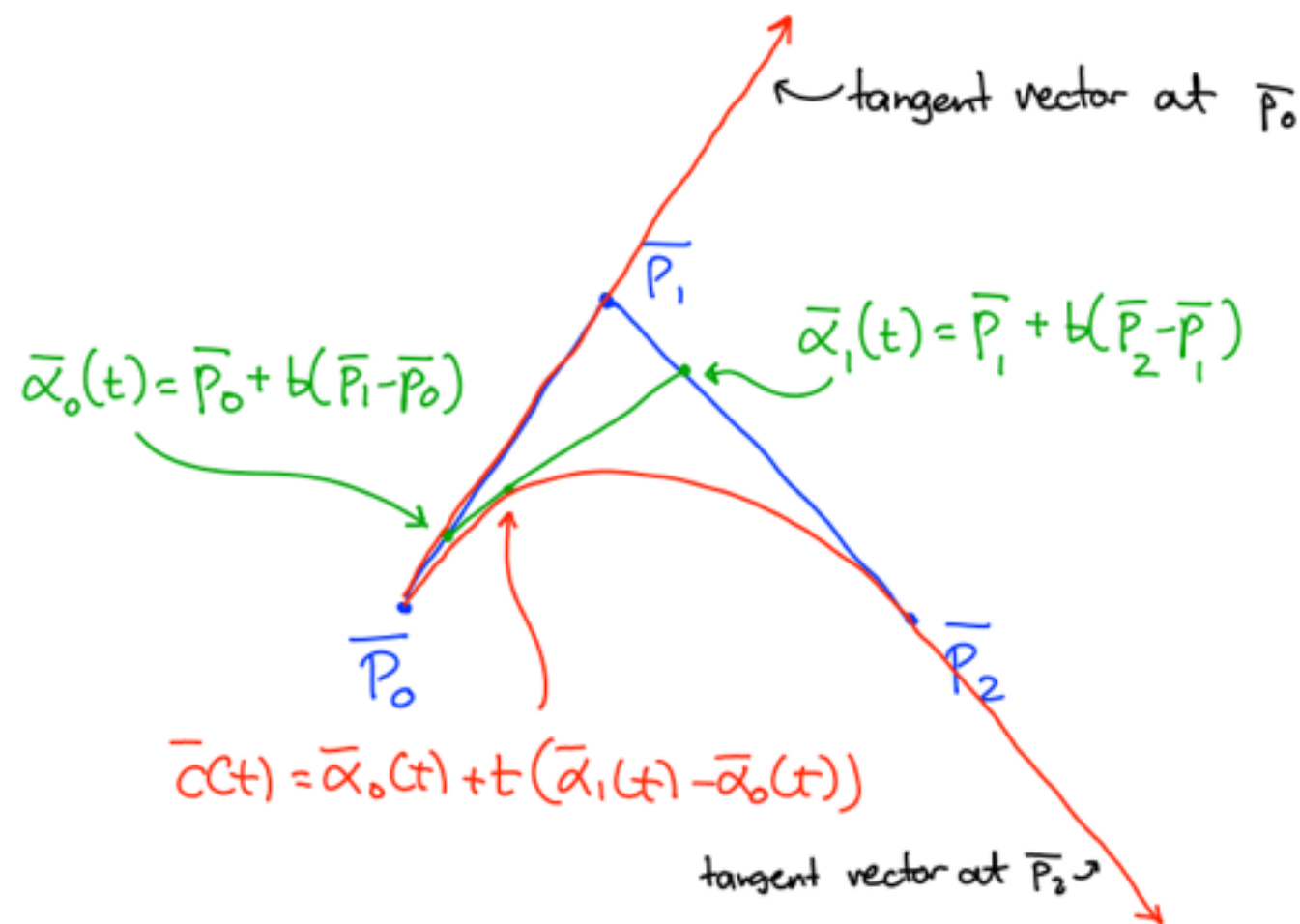
- given $\bar{P}_0, \bar{P}_1, \bar{P}_2$ and t
1. linearly interpolate \bar{P}_0, \bar{P}_1 to get $\bar{\alpha}_0(t)$
 2. linearly interpolate \bar{P}_1, \bar{P}_2 to get $\bar{\alpha}_1(t)$
 3. linearly interpolate $\bar{\alpha}_0(t), \bar{\alpha}_1(t)$ to get $\bar{c}(t)$

Bézier Curves: Endpoints and Tangent Constraints

Computing the polynomial's derivatives:

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

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



General Behaviour

- 1st and 3rd control points define the endpoints.
- 2nd control point defines the tangent vector at the endpoints.

Bézier Curves: Generalization to N+1 points

Expression in compact form:

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

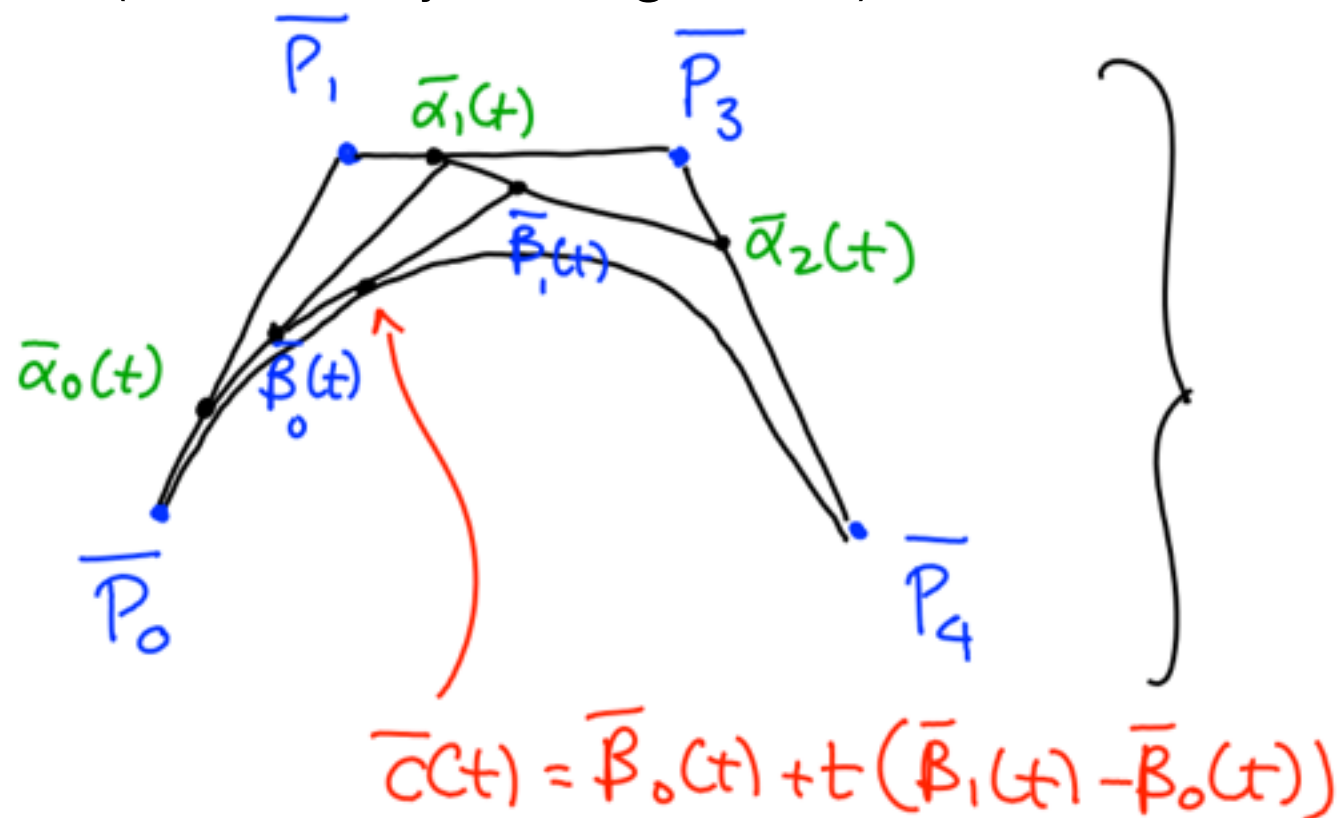
\uparrow
curve
 \uparrow
control pt

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 (De Casteljau's algorithm):



Example for 4 control points and 3 cascades

Bézier Curves: A Different Perspective

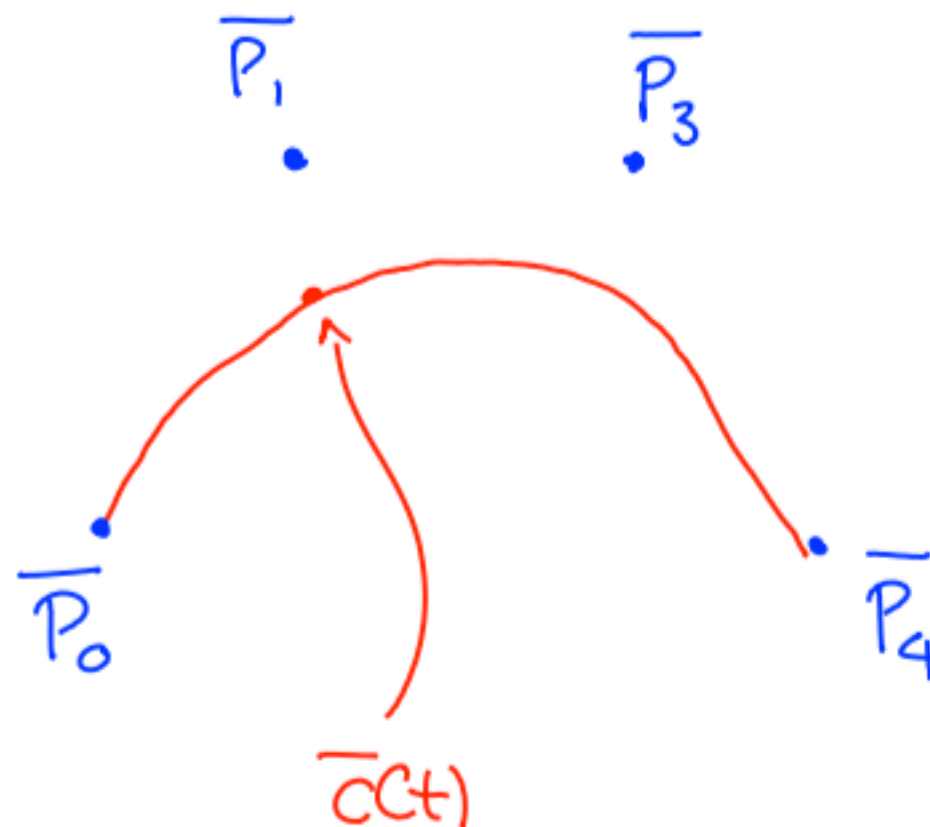
Expression in compact form:

$$\overline{c}(t) = \sum_{i=0}^N \overline{P}_i B_i^N(t) \quad \text{Where:}$$

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 points.
- The blend coefficients depend on t
- They are Bernstein polynomials

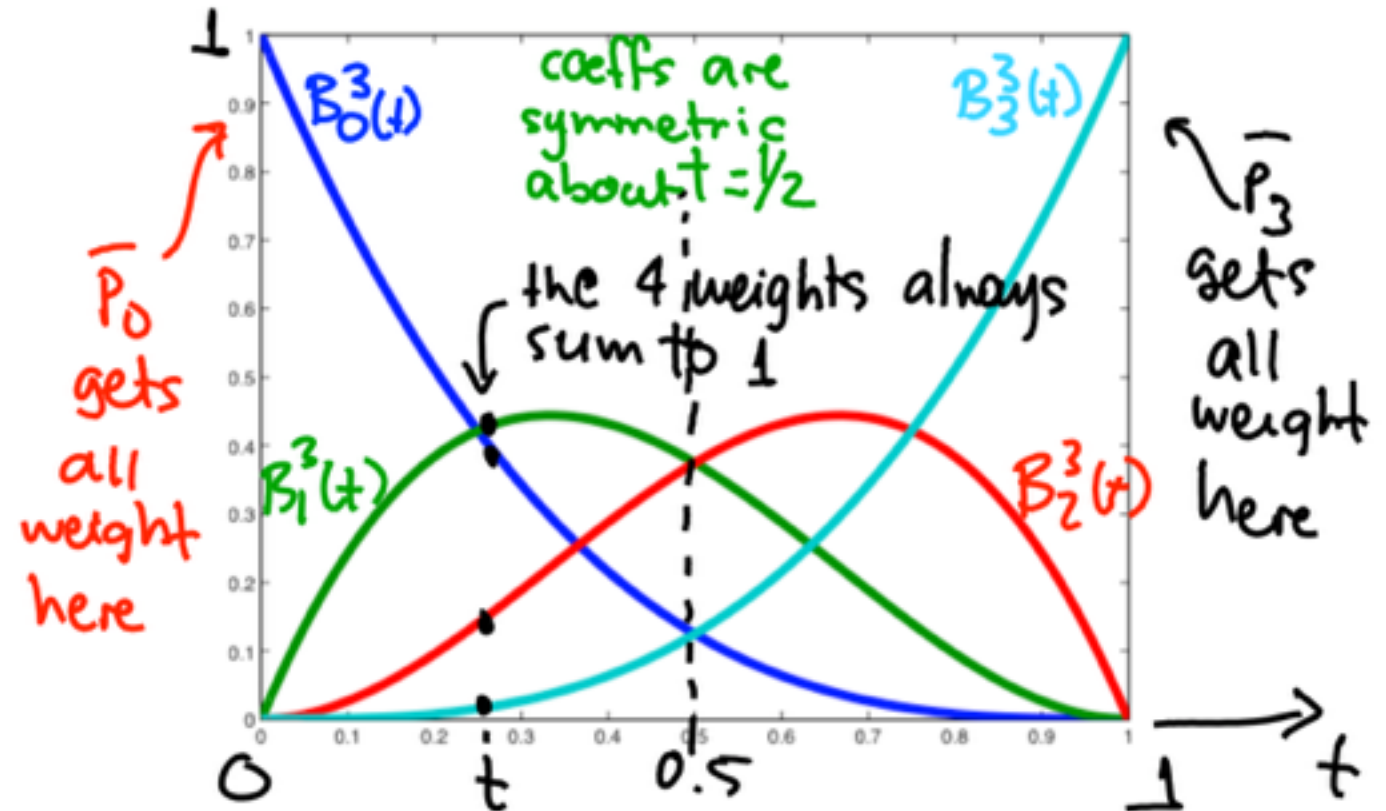
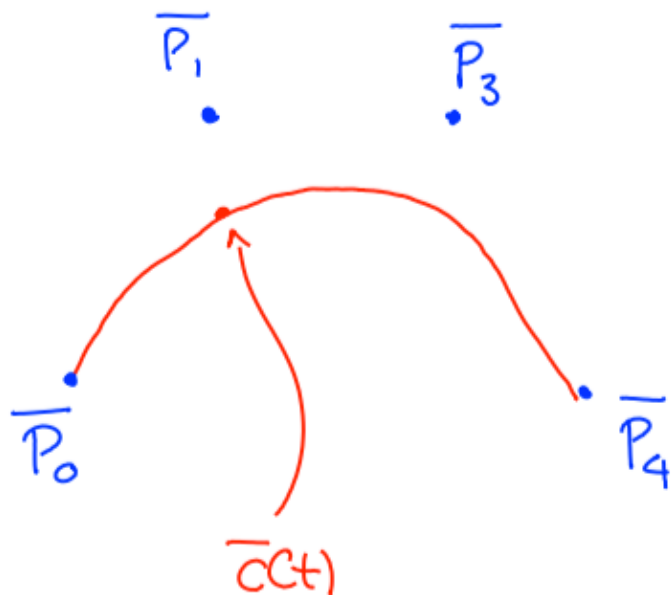
Bézier Curves as “blends” of the Control Points

Expression in compact form:

$$\vec{c}(t) = \sum_{i=0}^3 \vec{P}_i B_i^3(t)$$

↑ curve
↑ control pt

with $\sum_{i=0}^3 B_i^3(t) = 1$ for all t



- Each curve point $c(t)$ is a “blend” of the 4 control points.
- The blend coefficients depend on t
- They are Bernstein polynomials

Bézier Curves: Useful Properties

Expression in compact form:

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

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

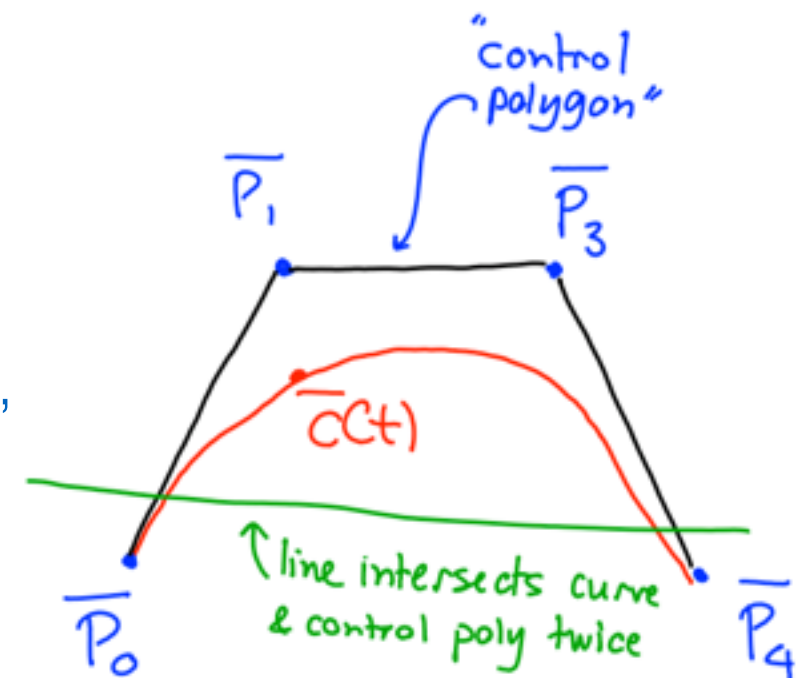
- Transforming a Bézier curve by an affine transform T is equivalent to transforming its control points by T

2. Diminishing Variation

- No line will intersect the curve at more points than the control polygon
 - 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:

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

Where:

called the Bernstein polynomials of degree N

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

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

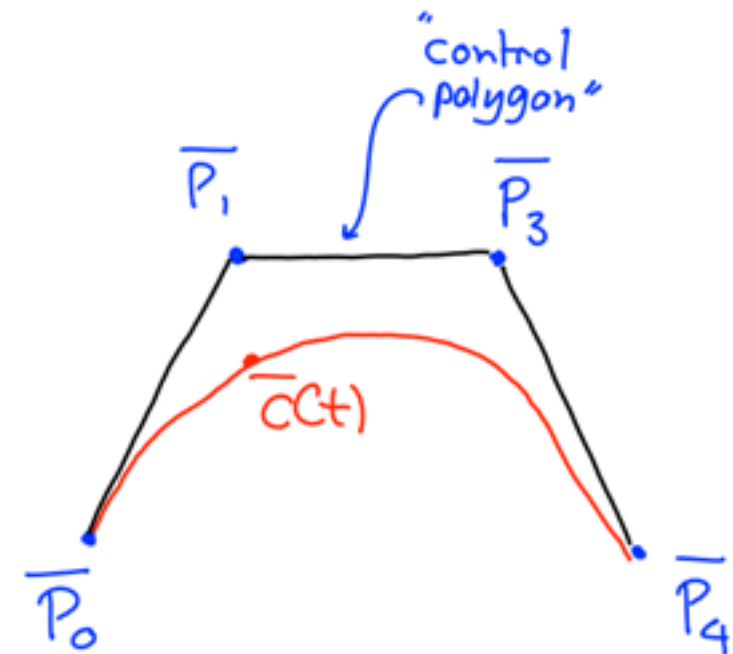
4. Tangents at endpoints are along the 1st and last edges of control polygon:

$$\frac{d}{dt} \bar{c}(t) = \sum_{i=1}^N \bar{P}_i \frac{d}{dt} B_i^N(t)$$

w/ some work

$$= N \sum_{i=0}^{N-1} (\bar{P}_{i+1} - \bar{P}_i) B_i^{N-1}(t)$$

$\stackrel{=}{=} N(\bar{P}_1 - \bar{P}_0)$ for $t=0$
 $\stackrel{=}{=} N(\bar{P}_N - \bar{P}_{N-1})$ for $t=1$



Bézier Curves: Pros and Cons

Expression in compact form:

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

Where:

called the Bernstein polynomials of degree N

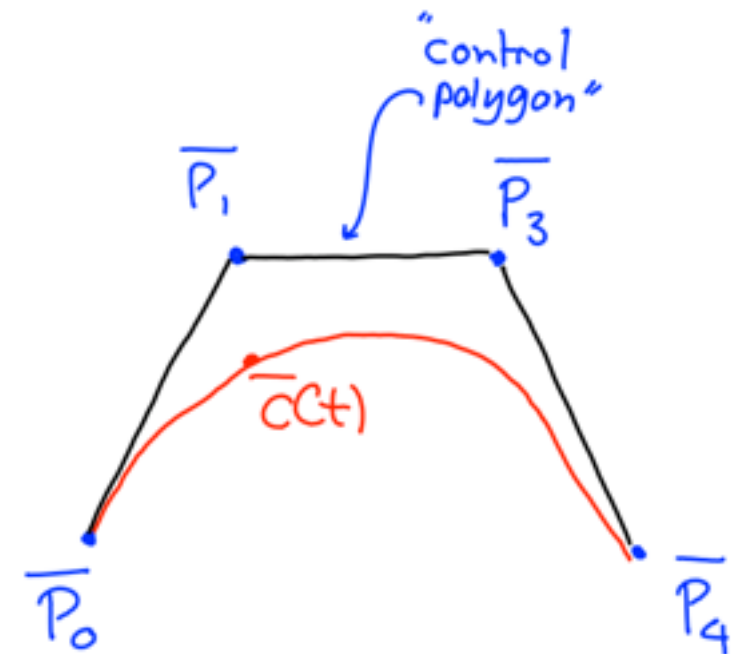
$$B_i^N(t) = \binom{N}{i} (1-t)^{N-i} t^i$$
$$= \frac{N!}{(N-i)! i!} (1-t)^{N-i} t^i$$

Advantages:

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

Disadvantages:

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

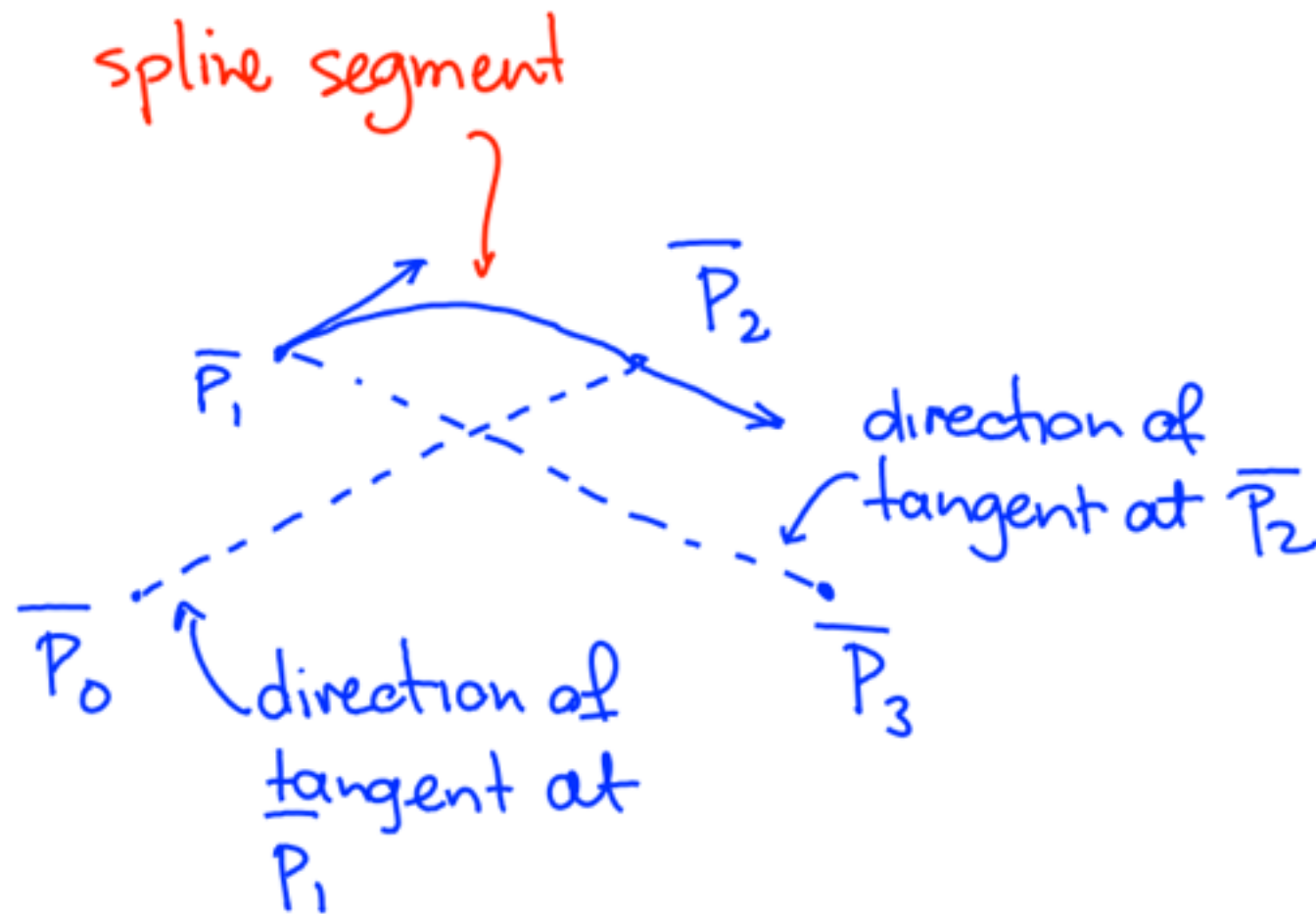


Topic 12: Interpolating Curves

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

Cubic Cardinal Splines: Defining 1st Segment

- Approach:
 1. A user only specifies points p_0, p_1, \dots
 2. Tangent at p_i set to be parallel to vector connecting p_{i-1} and p_{i+1}

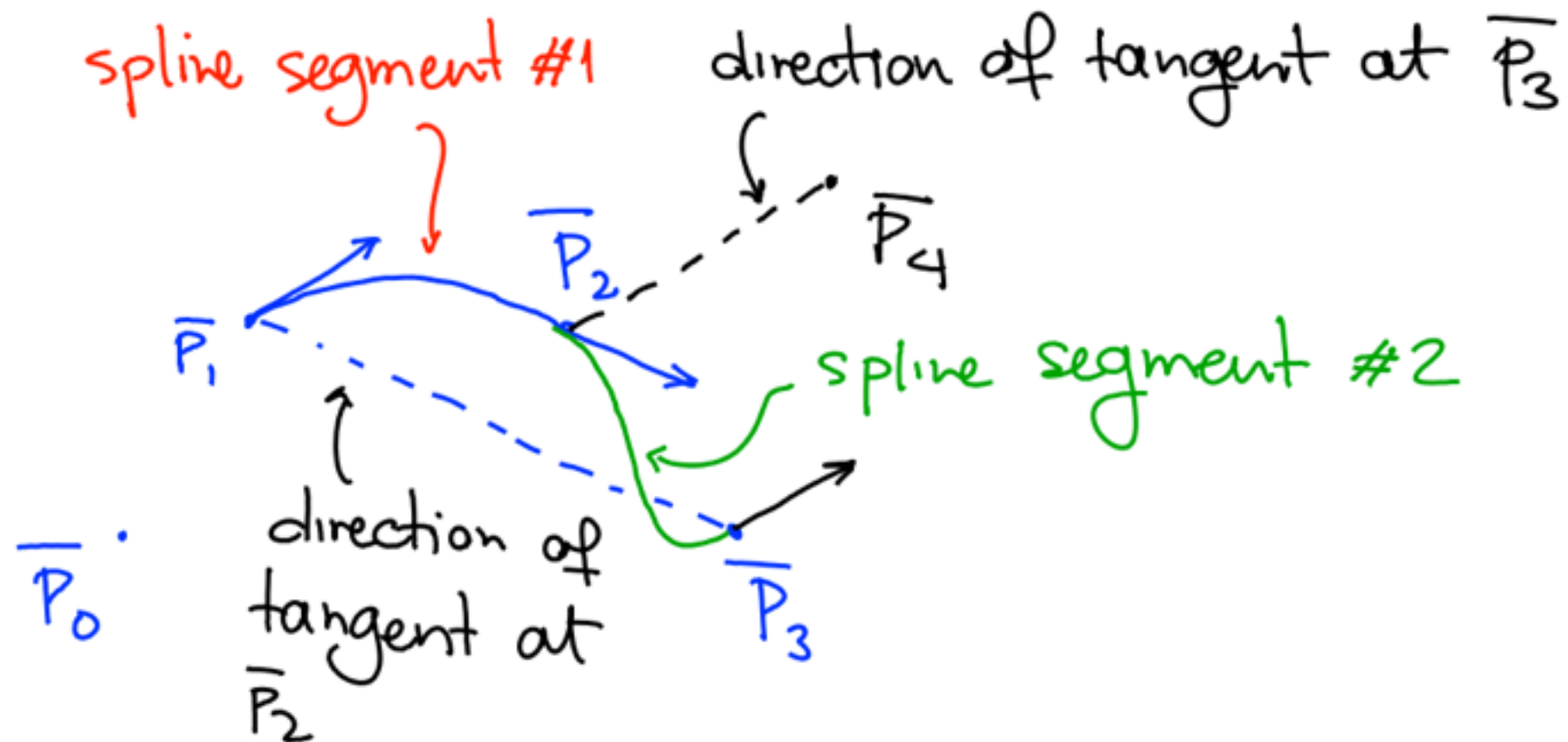


\bar{P}_1, \bar{P}_2 act as endpoint constraints
 \bar{P}_0, \bar{P}_3 define derivative constraints

Cubic Cardinal Splines: Defining 2nd Segment

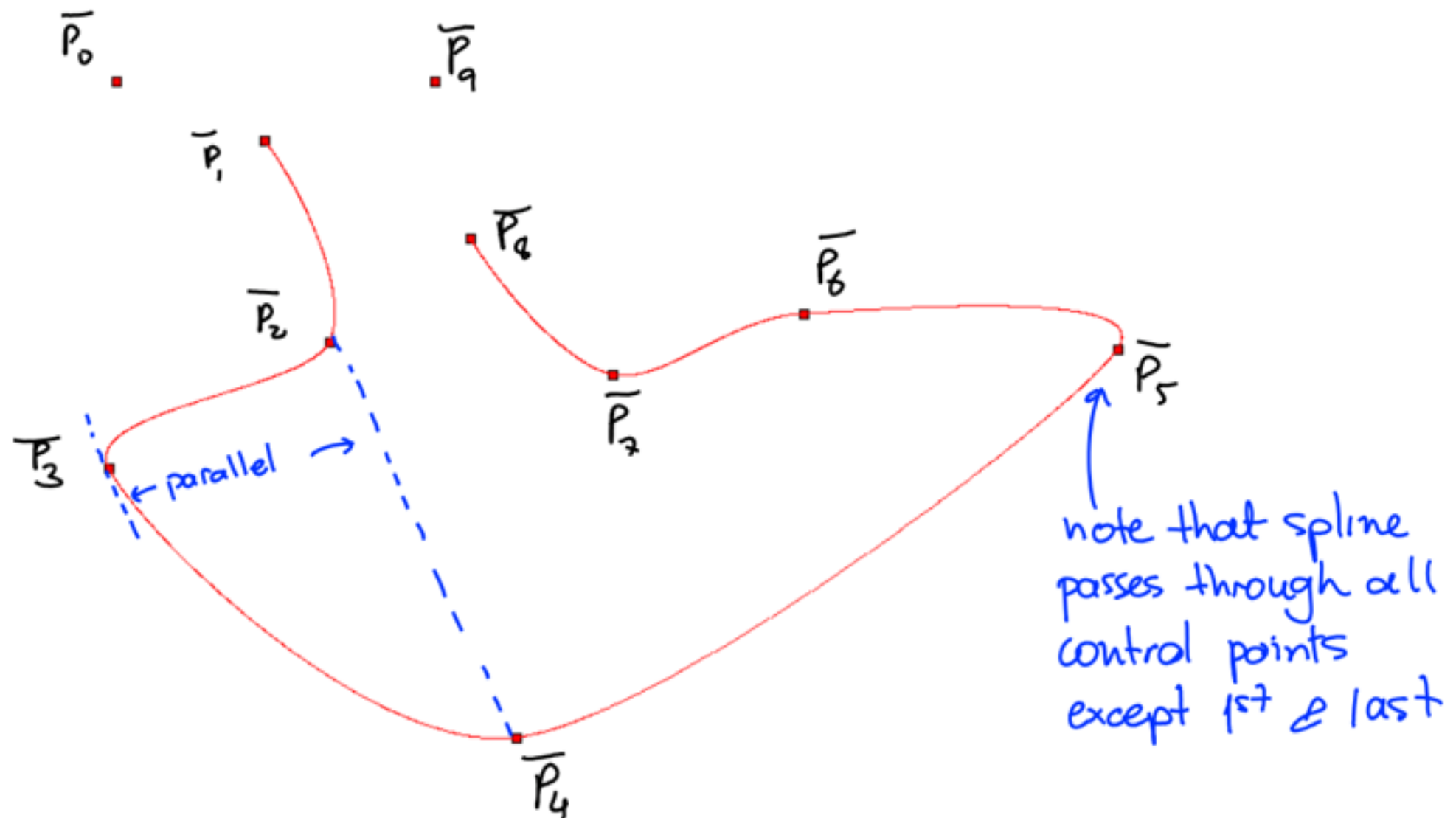
- Approach:
 1. A user only specifies points p_0, p_1, \dots
 2. Tangent at p_i set to be parallel to vector connecting p_{i-1} and p_{i+1}

Example: Adding a fifth point adds a new segment



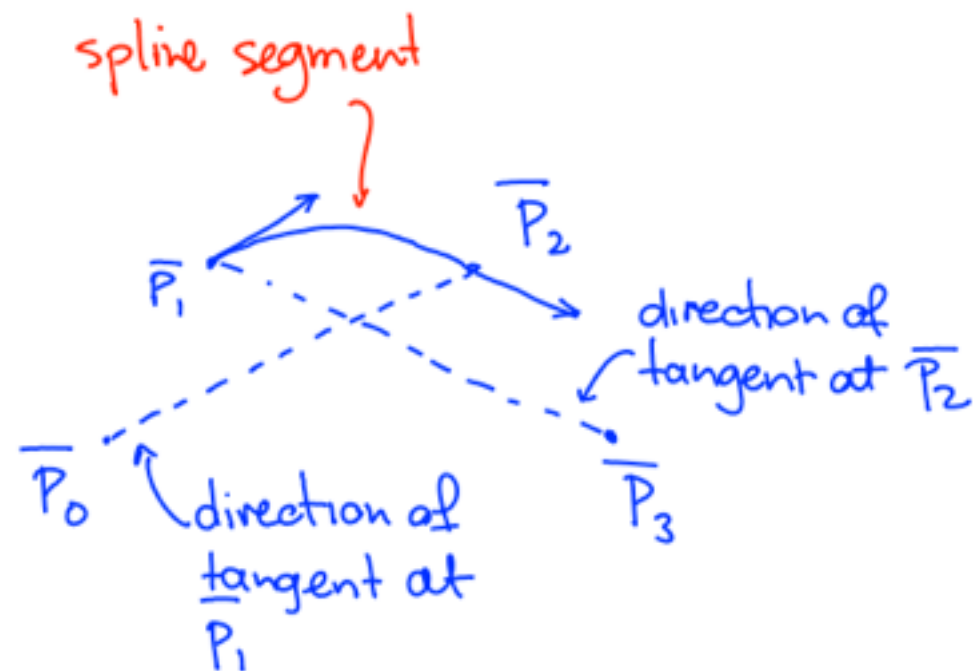
Cubic Cardinal Splines: General Case

- Approach:
 1. A user only specifies points p_0, p_1, \dots
 2. Tangent at p_i set to be parallel to vector connecting p_{i-1} and p_{i+1}



Cubic Cardinal Splines: The Strain Parameter

- Approach:
 1. A user only specifies points p_0, p_1, \dots
 2. Tangent at p_i set to be parallel to vector connecting p_{i-1} and p_{i+1}



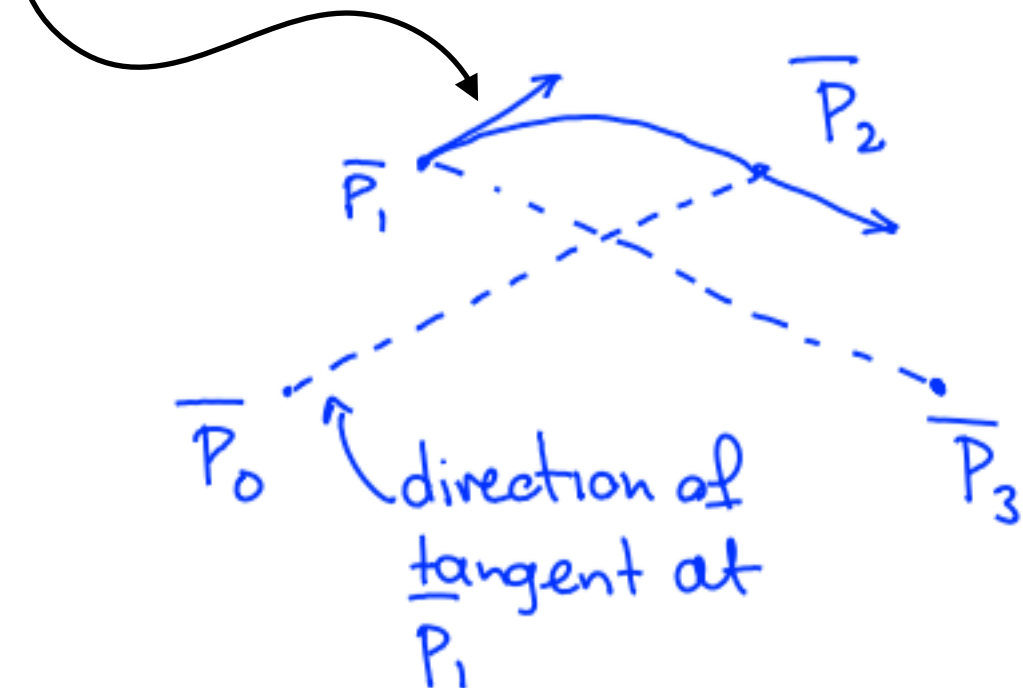
Tangent at $\bar{P}_i = k(\bar{P}_{i+1} - \bar{P}_{i-1})$
called a strain parameter



Catmull-Rom Splines

- Approach:
 1. A user only specifies points p_0, p_1, \dots
 2. Tangent at p_i set to be parallel to vector connecting p_{i-1} and p_{i+1}

length of tangent = 1/2 distance between p_0 and p_2



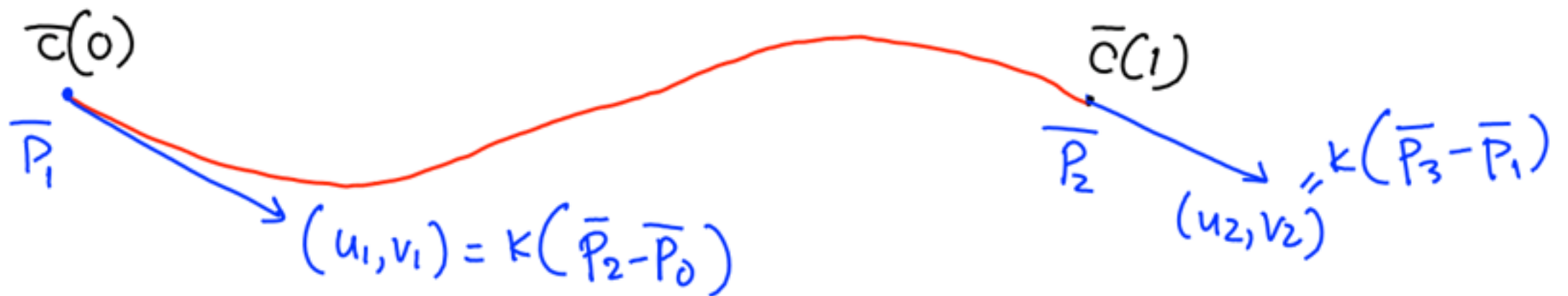
Tangent at $\bar{P}_i = k(\bar{P}_{i+1} - \bar{P}_{i-1})$
called a strain parameter

Note: If $k = 0.5$, 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

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



Cardinal Splines: Solving for the Segment Coeffs

for $t = 0$

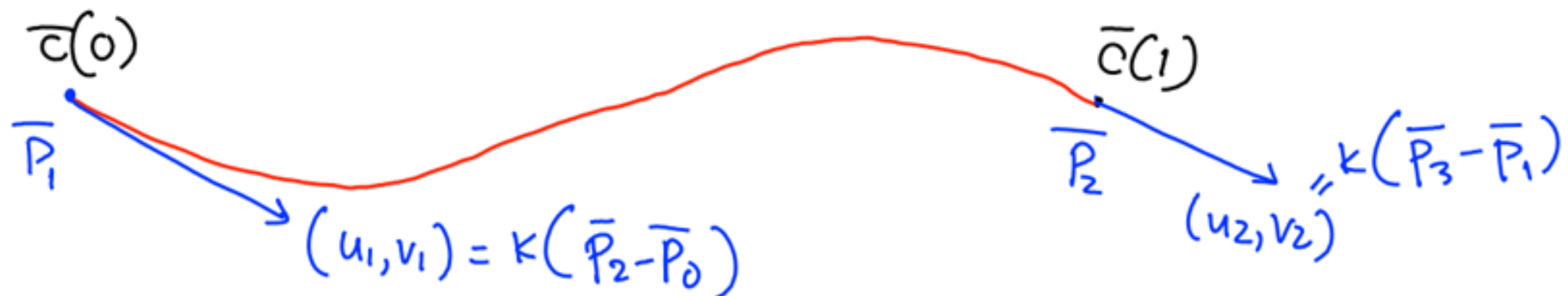
$$[x_1 \ y_1] = [1 \ \cancel{t}_0 \ \cancel{t^2}_0 \ \cancel{t^3}_0] \begin{bmatrix} a_0 & b_0 \\ a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

endpoint constraint

$$[k(x_2 - x_0) \ k(y_2 - y_0)] = [1 \ \cancel{2t}_0 \ \cancel{3t^2}_0] \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}$$

derivative constraint

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

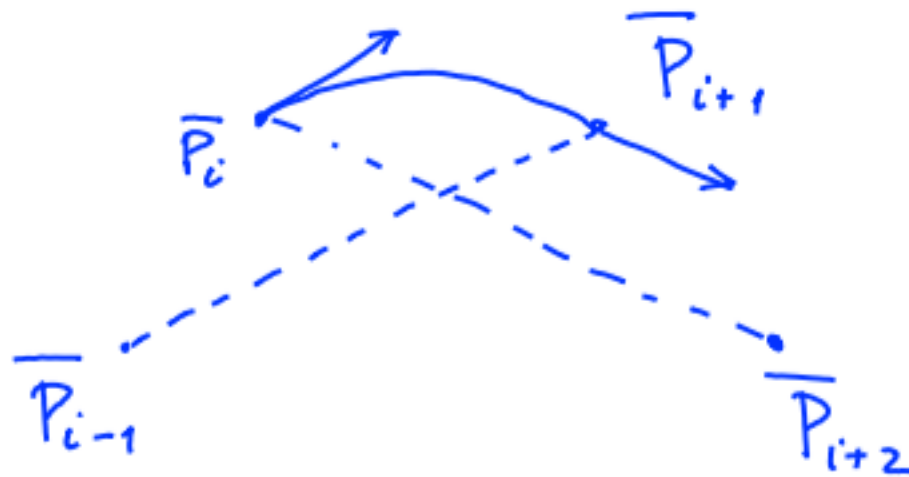


Cubic Cardinal Spline Segment vs Bézier Curve

The two curves are actually equivalent:

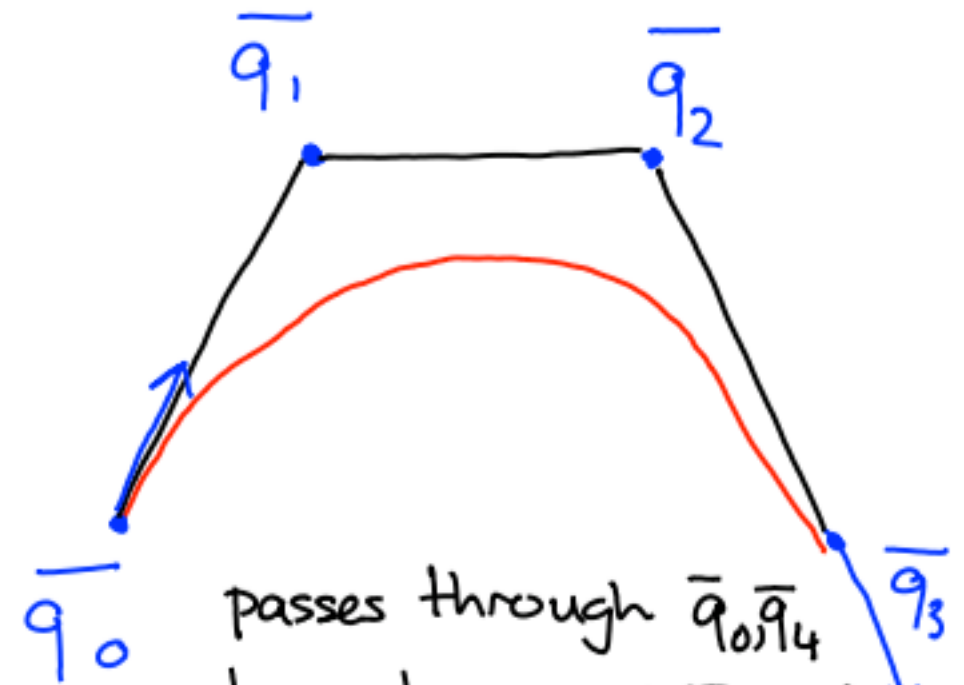
given a cardinal spline, we can compute the control polygon of the equivalent Bézier curve

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 $3(\bar{q}_1 - \bar{q}_0)$
 $3(\bar{q}_3 - \bar{q}_2)$

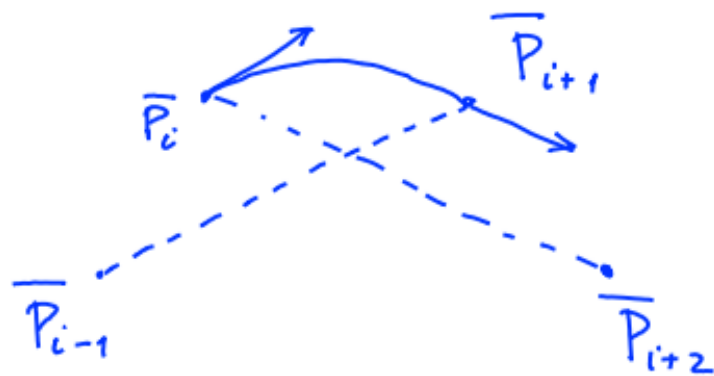
Cubic Cardinal Spline Segment vs Bézier Curve

In order to have $c(t) = r(t)$ for all t , it must be:

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

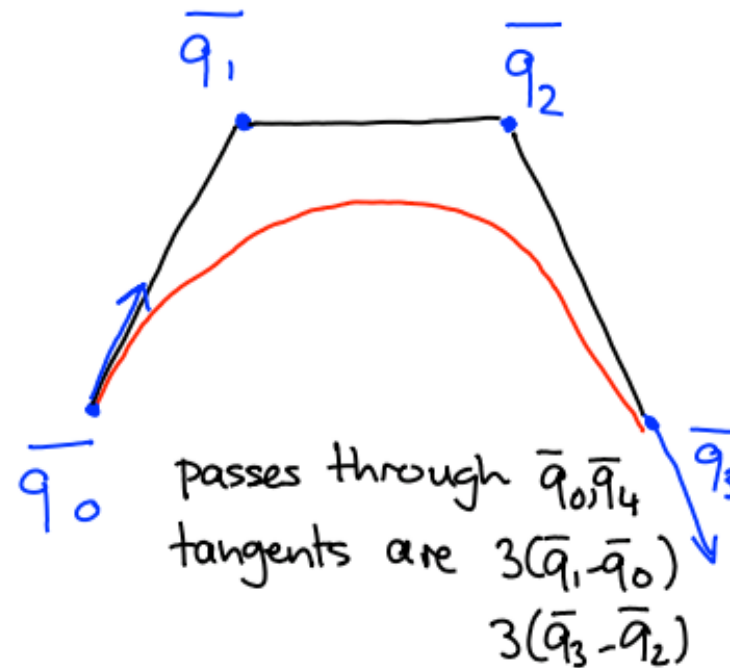
$$K(\bar{p}_{i+1} - \bar{p}_{i-1}) = 3(\bar{q}_1 - \bar{q}_0) \Rightarrow K(\bar{p}_{i+1} - \bar{p}_{i-1}) = 3\bar{q}_1 - 3\bar{q}_0 \stackrel{\bar{p}_i}{=} \\ \Rightarrow \bar{q}_1 = \frac{K}{3}(\bar{p}_{i+1} - \bar{p}_{i-1}) + \bar{p}_i$$

Cardinal Spline Segment



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

Bézier Curve



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

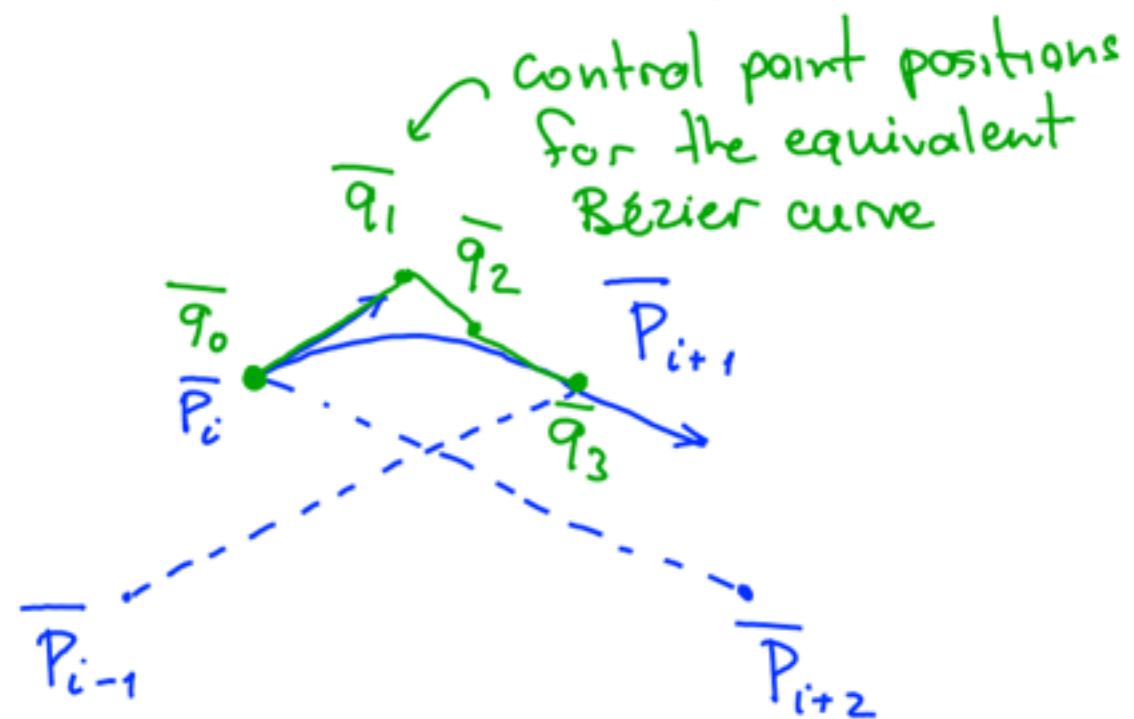
Cubic Cardinal Spline Segment vs Bézier Curve

In order to have $c(t) = r(t)$ for all t , it must be:

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

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

Cardinal Spline Segment



Bézier Curve

