

# **2D Spline Curves**

CS 4620 Lecture 15

# Motivation: smoothness

- In many applications we need smooth shapes
  - that is, without discontinuities

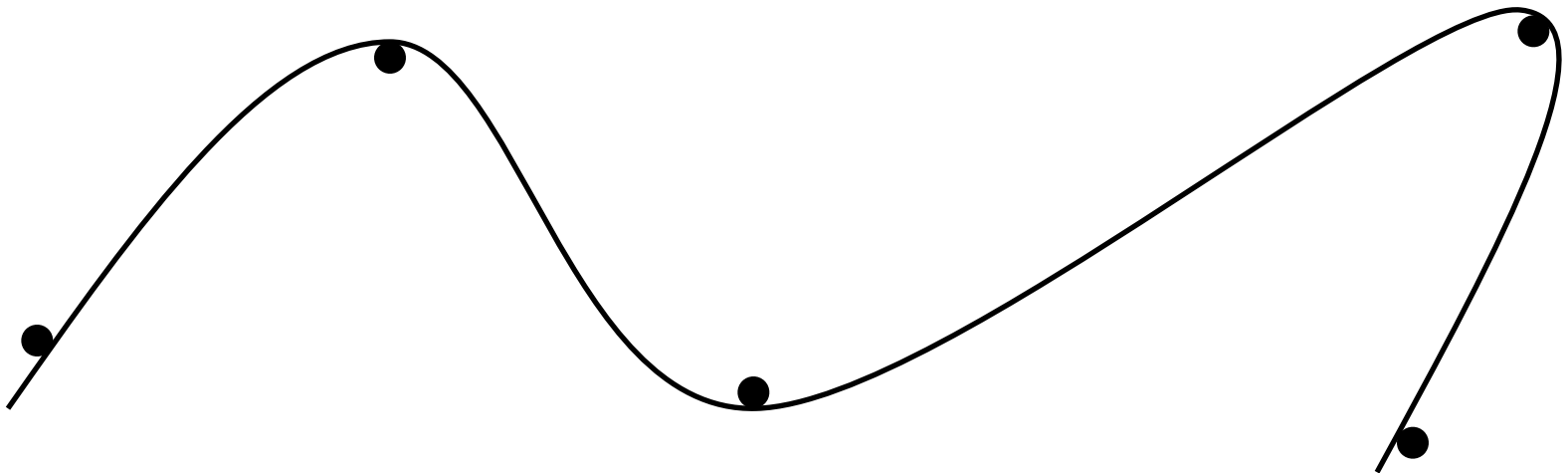


[Boeing]

- So far we can make
  - things with corners (lines, triangles, squares, rectangles, ...)
  - circles, ellipses, other special shapes (only get you so far!)

# Classical approach

- Pencil-and-paper draftsmen also needed smooth curves
- Origin of “spline:” strip of flexible metal
  - held in place by pegs or weights to constrain shape
  - traced to produce smooth contour



# Translating into usable math

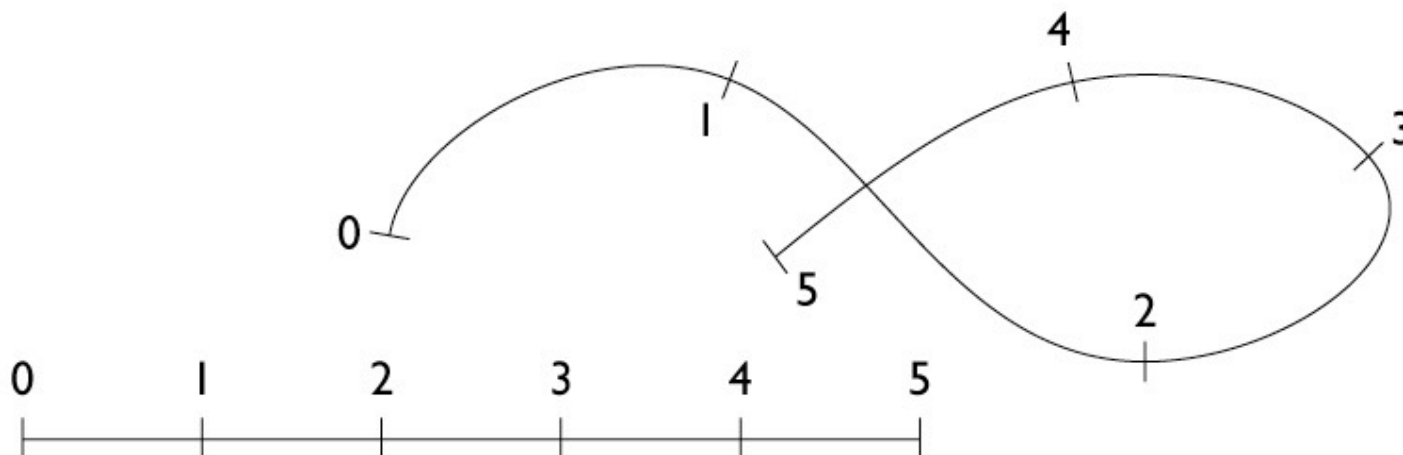
- Smoothness
  - in drafting spline, comes from physical curvature minimization
  - in CG spline, comes from choosing smooth functions
    - usually low-order polynomials
- Control
  - in drafting spline, comes from fixed pegs
  - in CG spline, comes from user-specified *control points*

# Defining spline curves

- At the most general they are parametric curves

$$S = \{\mathbf{f}(t) \mid t \in [0, N]\}$$

- For splines,  $\mathbf{f}(t)$  is piecewise polynomial
  - for this lecture, the discontinuities are at the integers

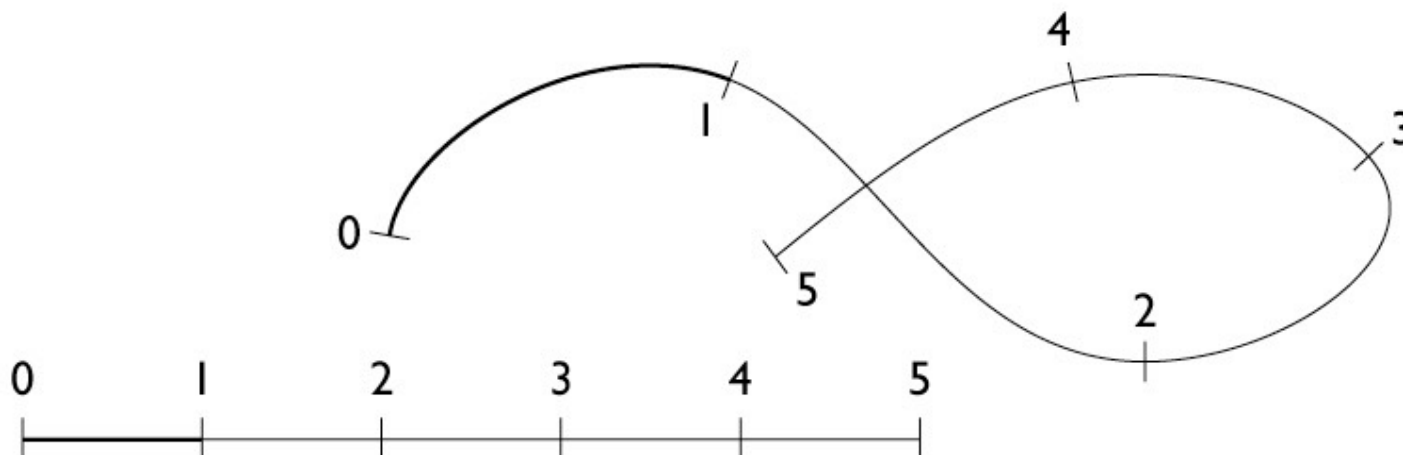


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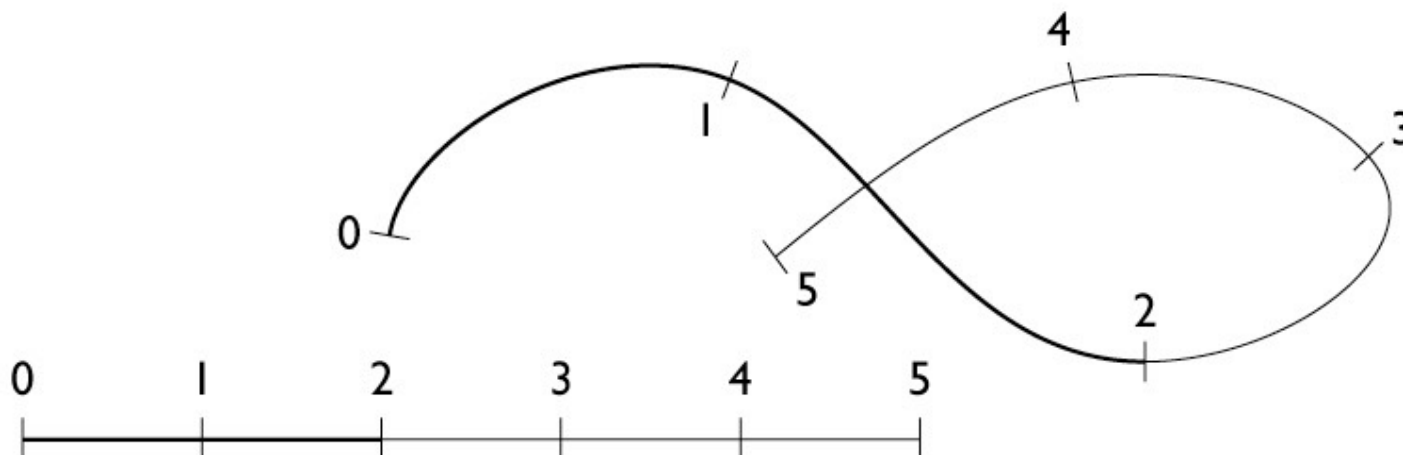


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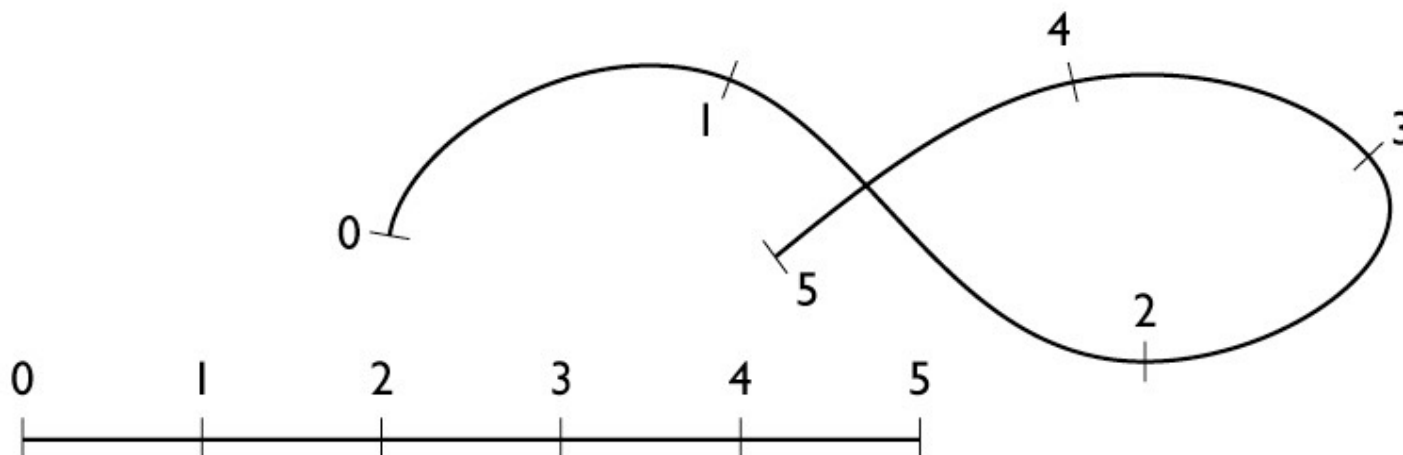


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# Defining spline curves

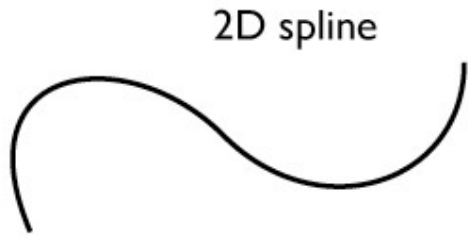
- Generally  $\mathbf{f}(t)$  is a piecewise polynomial
  - for this lecture, the discontinuities are at the integers
  - e.g., a cubic spline has the following form over  $[k, k + 1]$ :

$$x(t) = a_x t^3 + b_x t^2 + c_x t + d_x$$

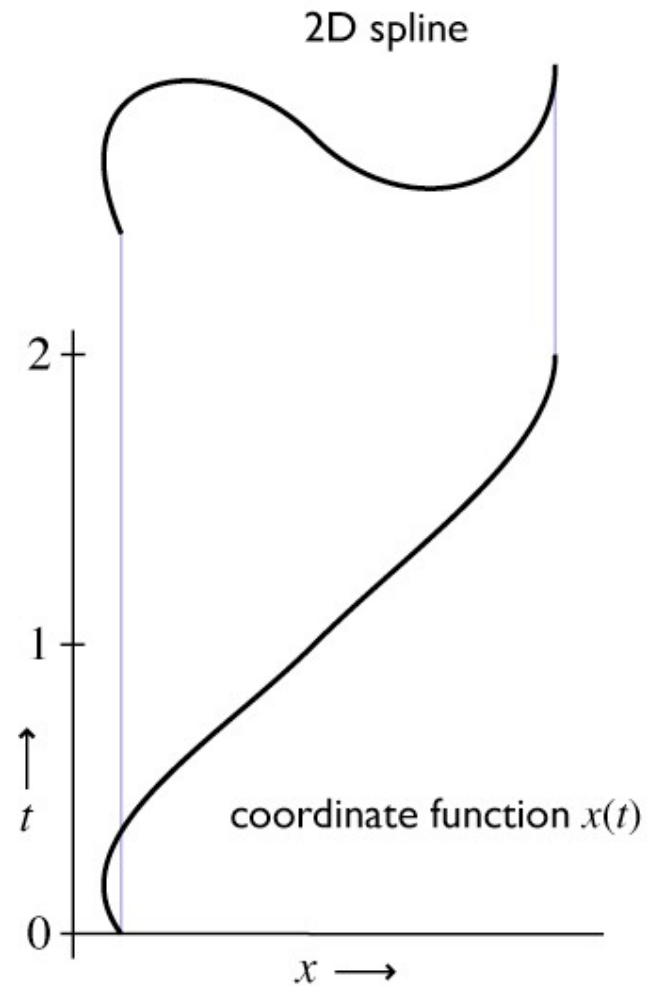
$$y(t) = a_y t^3 + b_y t^2 + c_y t + d_y$$

- Coefficients are different for every interval

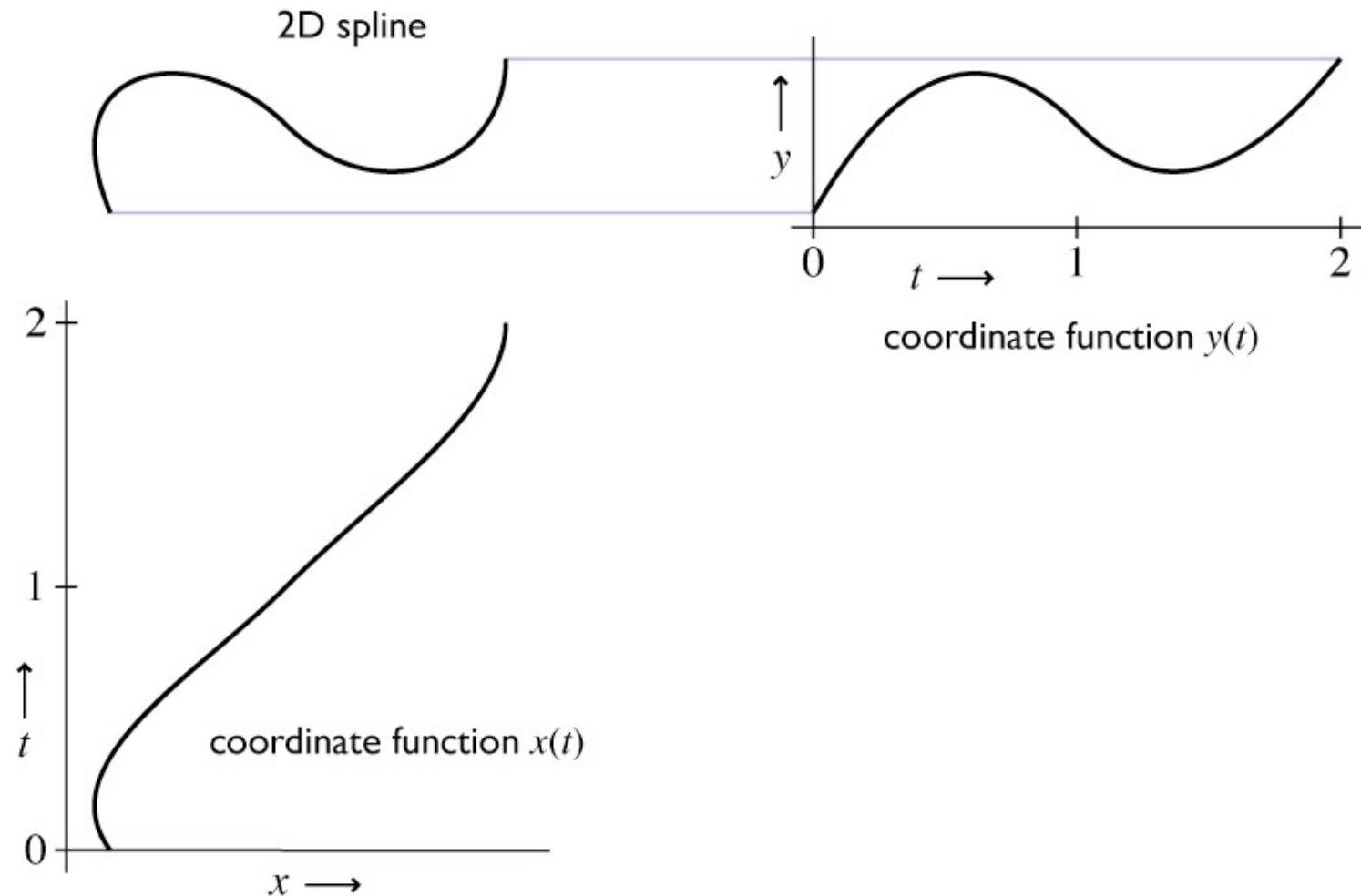
# Coordinate functions



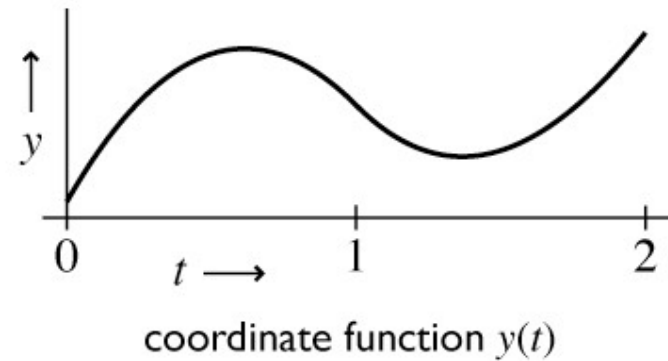
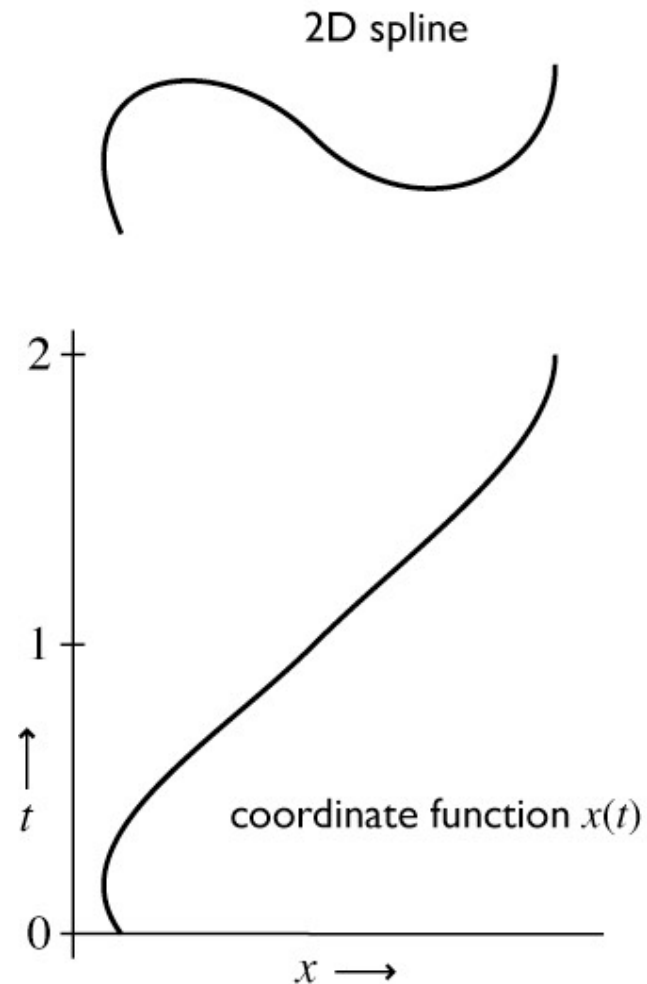
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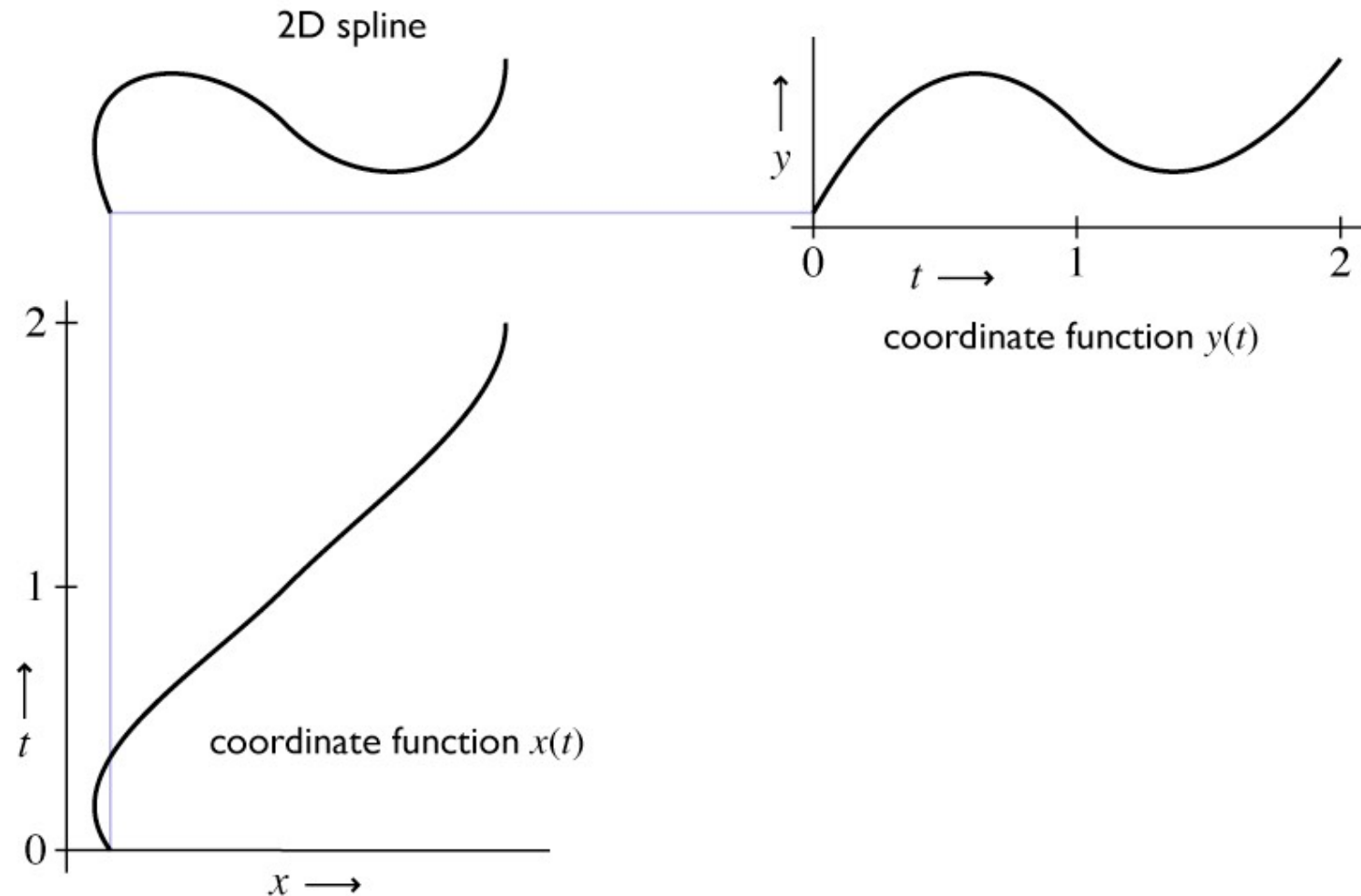
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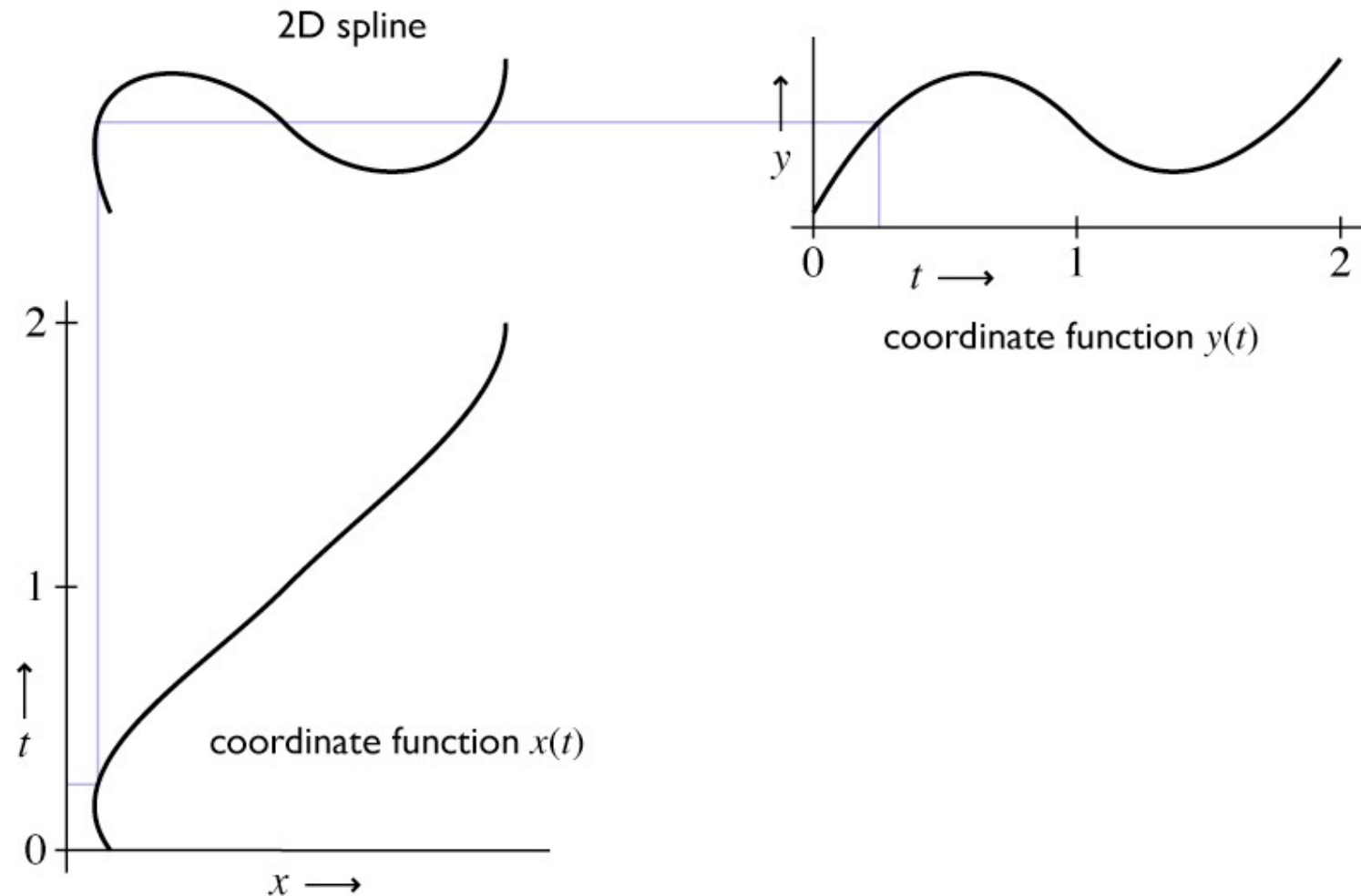
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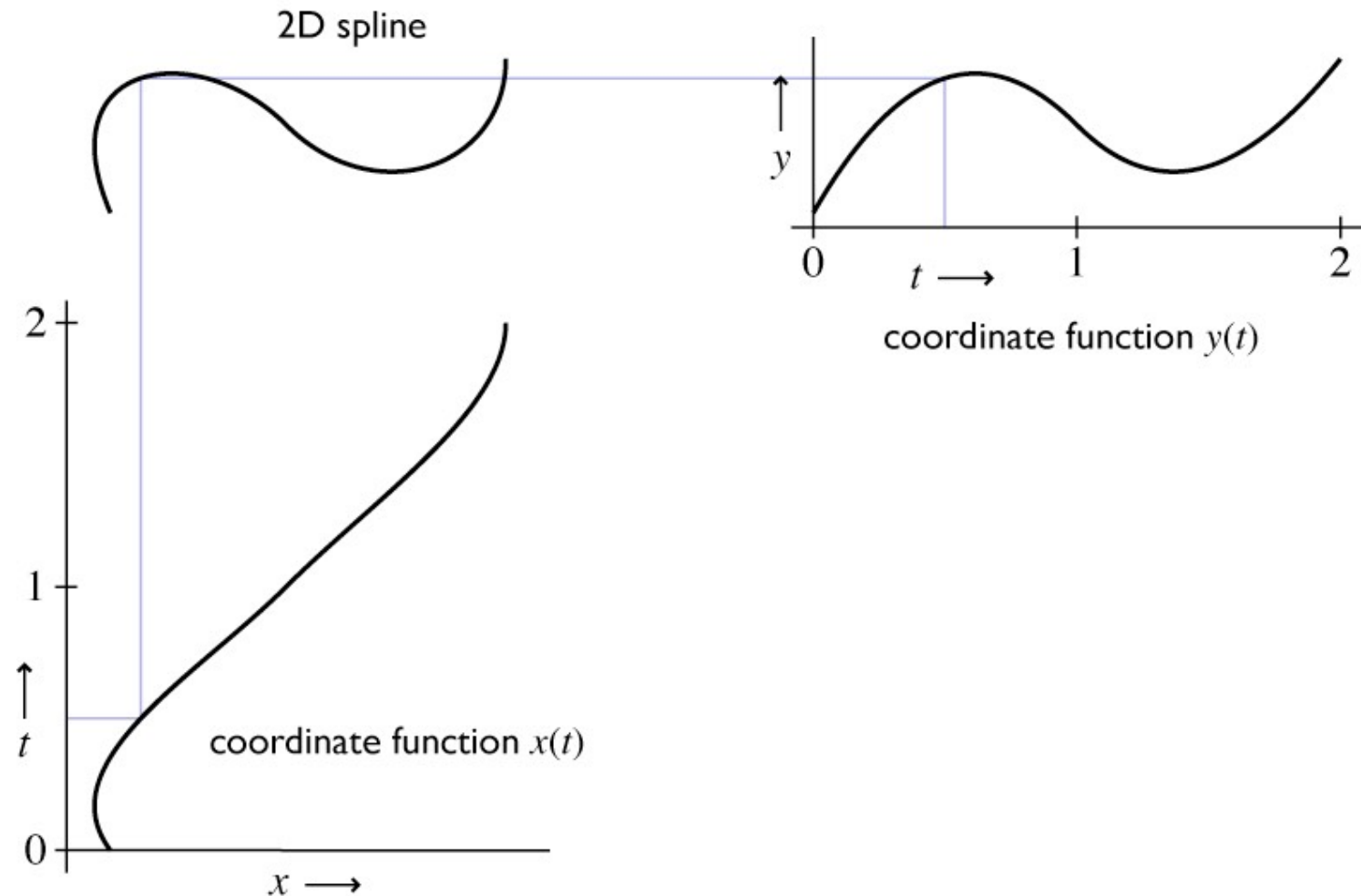
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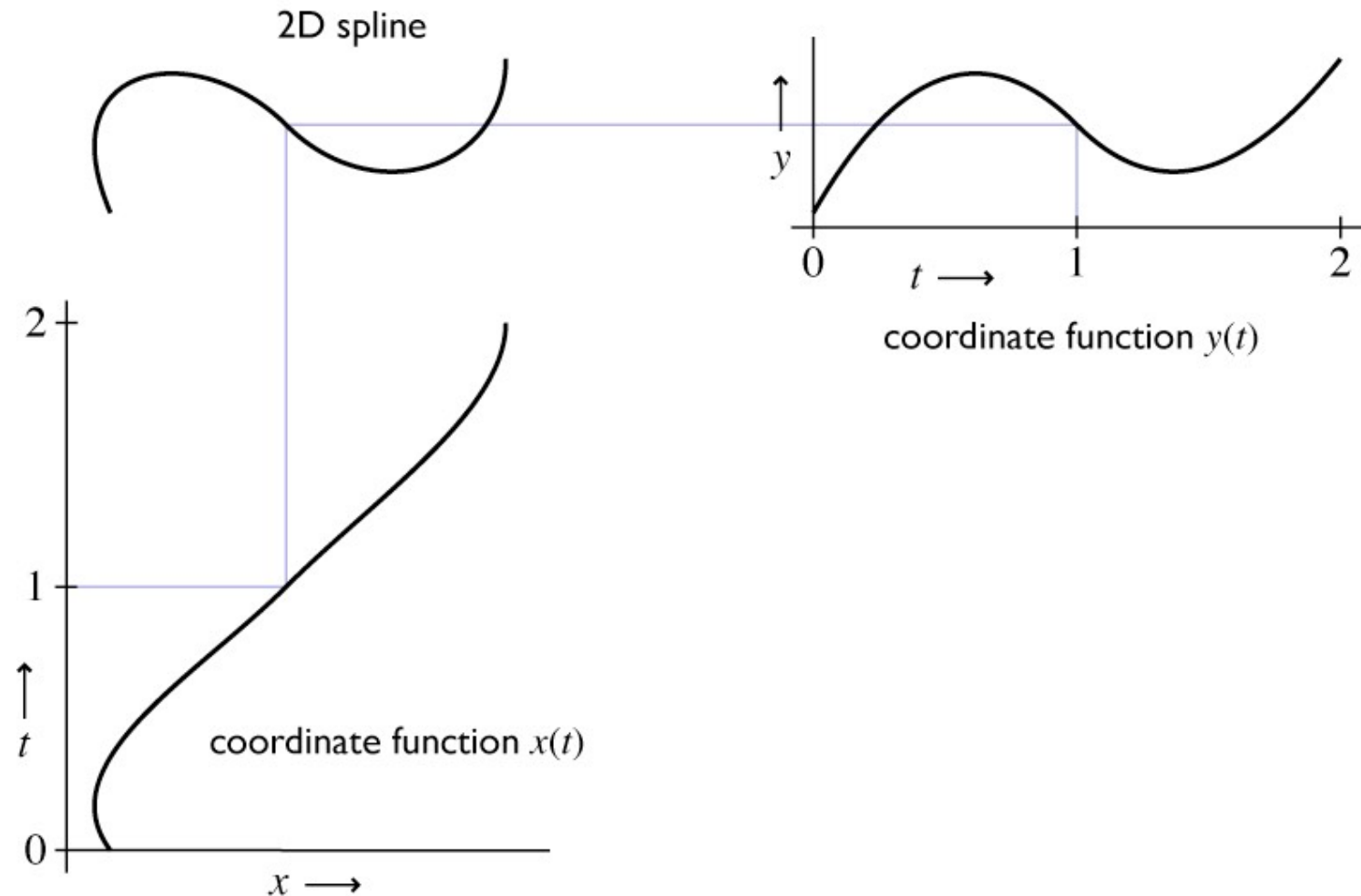


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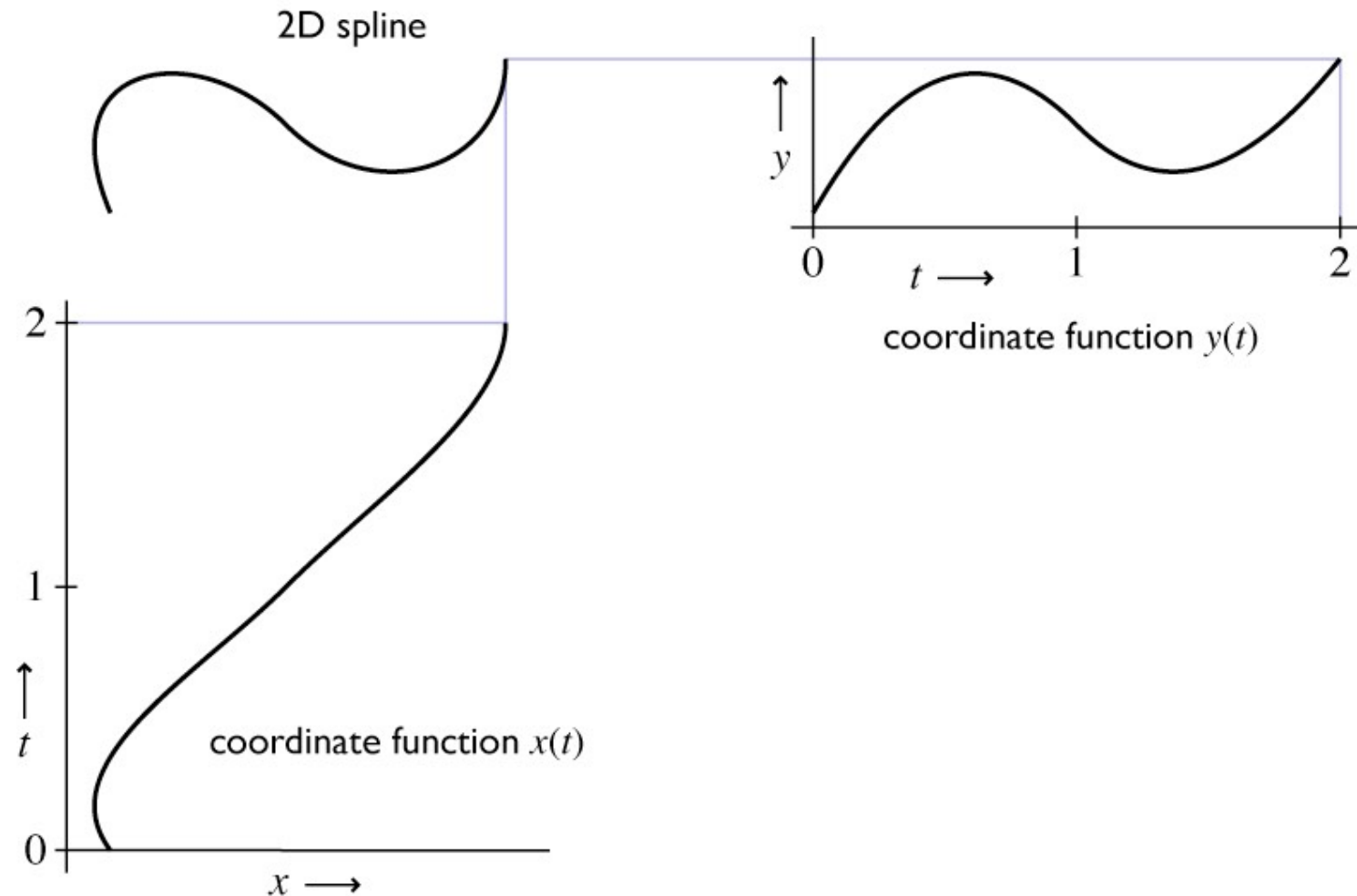




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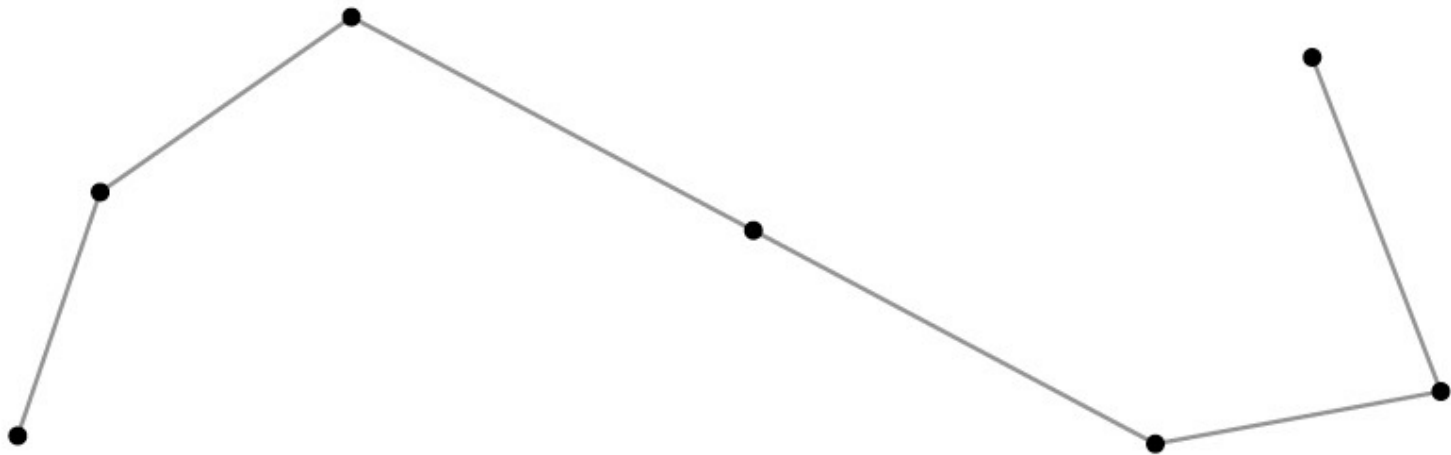


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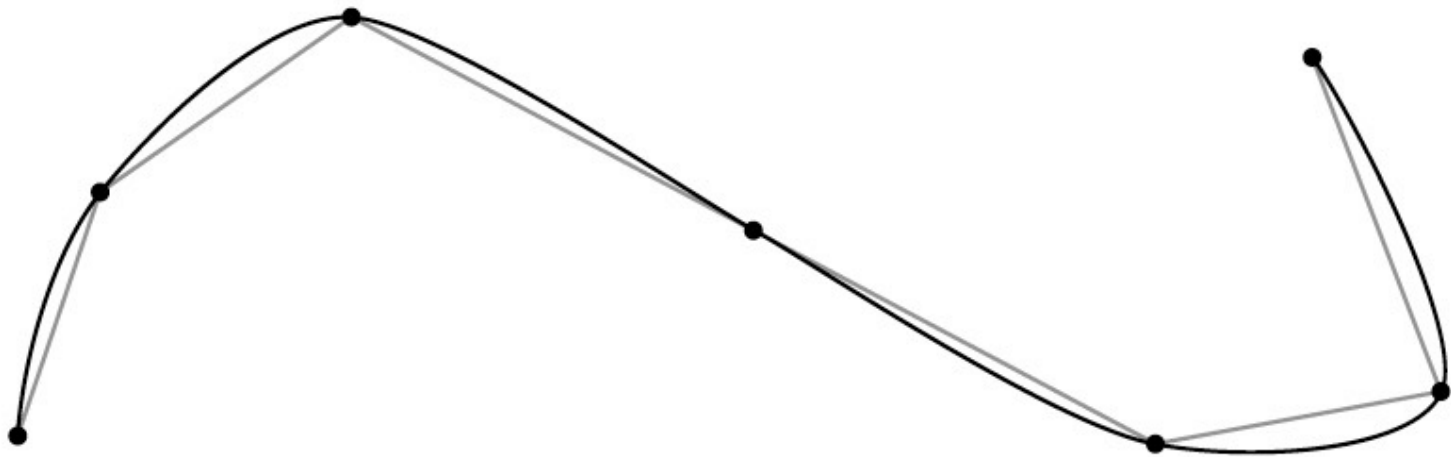
# Control of spline curves

- Specified by a sequence of controls (points or vectors)
- Shape is guided by control points (aka control polygon)
  - interpolating: passes through points
  - approximating: merely guided by points



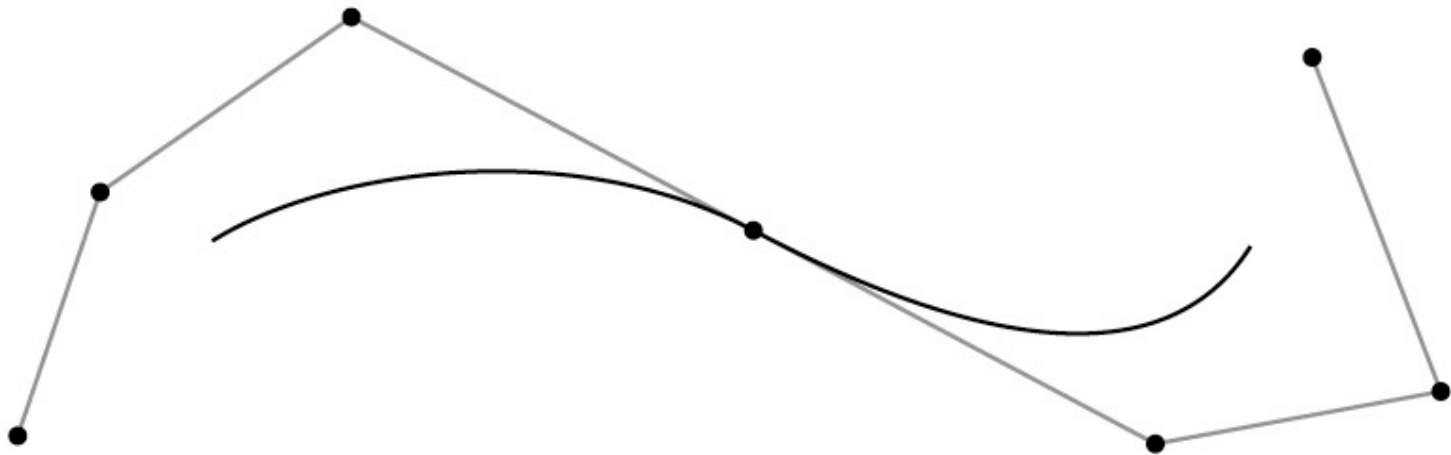
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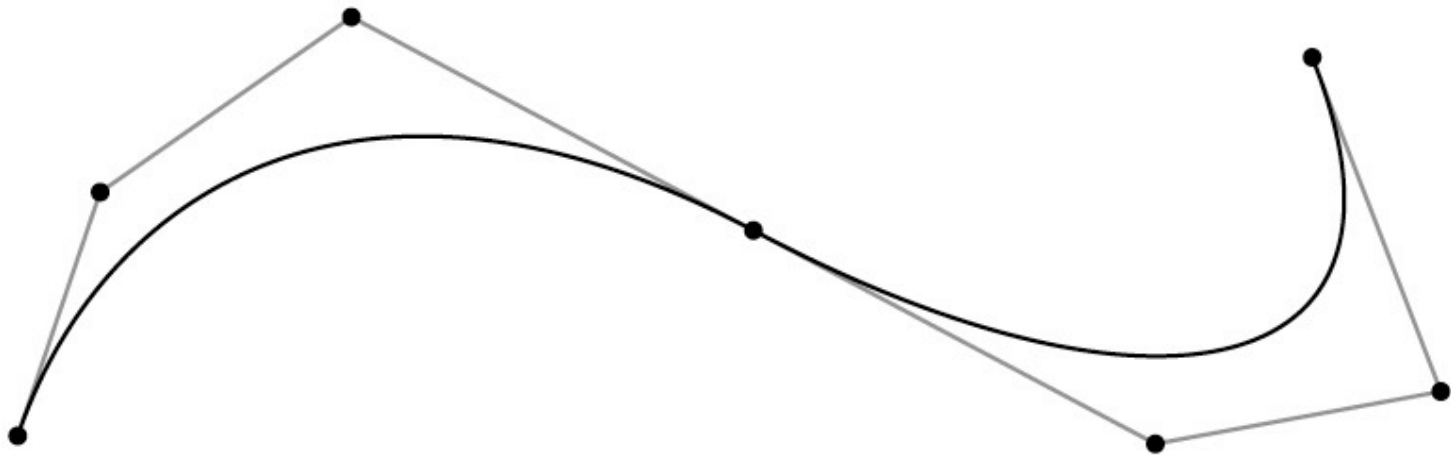
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# How splines depend on their controls

- Each coordinate is separate
  - the function  $x(t)$  is determined solely by the  $x$  coordinates of the control points
  - this means 1D, 2D, 3D, ... curves are all really the same
- Spline curves are **linear** functions of their controls
  - moving a control point two inches to the right moves  $x(t)$  twice as far as moving it by one inch
  - $x(t)$ , for fixed  $t$ , is a linear combination (weighted sum) of the controls'  $x$  coordinates
  - $\mathbf{f}(t)$ , for fixed  $t$ , is a linear combination (weighted sum) of the controls

# Plan

## 1. Spline segments

- how to define a polynomial on  $[0,1]$
- ...that has the properties you want
- ...and is easy to control

## 2. Spline curves

- how to chain together lots of segments
- ...so that the whole curve has the properties you want
- ...and is easy to control

## 3. Refinement and evaluation

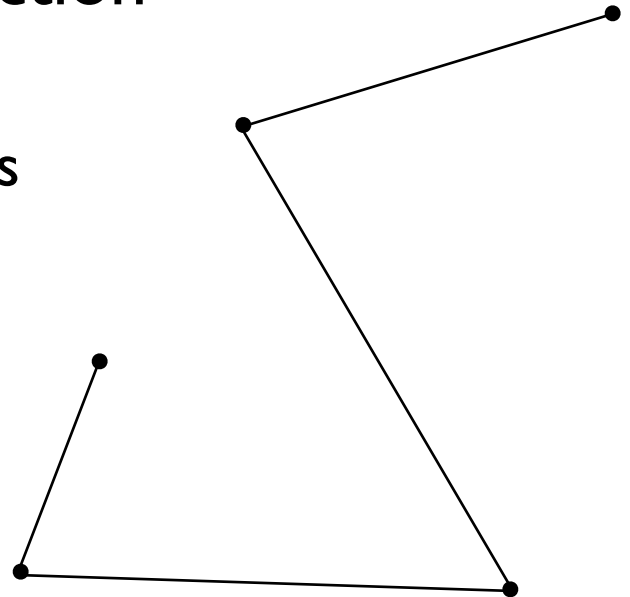
- how to add detail to splines
- how to approximate them with line segments



# Spline Segments

# Trivial example: piecewise linear

- This spline is just a polygon
  - control points are the vertices
- But we can derive it anyway as an illustration
- Each interval will be a linear function
  - $x(t) = at + b$
  - constraints are values at endpoints
  - $b = x_0$  ;  $a = x_1 - x_0$
  - this is linear interpolation



# Trivial example: piecewise linear

- Vector formulation

$$x(t) = (x_1 - x_0)t + x_0$$

$$y(t) = (y_1 - y_0)t + y_0$$

$$\mathbf{f}(t) = (\mathbf{p}_1 - \mathbf{p}_0)t + \mathbf{p}_0$$

- Matrix formulation

$$\mathbf{f}(t) = \begin{bmatrix} t & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \end{bmatrix}$$

# Trivial example: piecewise linear

- Basis function formulation
  - regroup expression by **p** rather than *t*

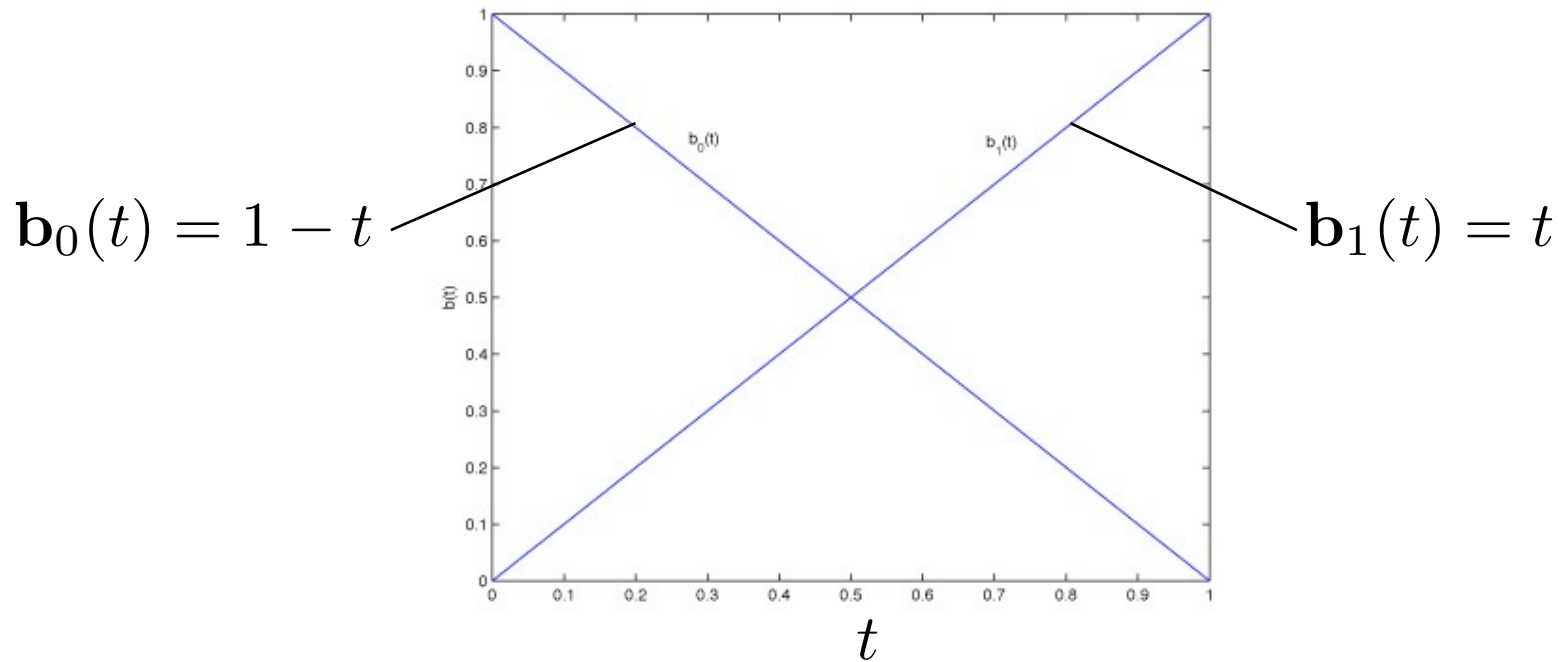
$$\begin{aligned}\mathbf{f}(t) &= (\mathbf{p}_1 - \mathbf{p}_0)t + \mathbf{p}_0 \\ &= (1 - t)\mathbf{p}_0 + t\mathbf{p}_1\end{aligned}$$

- interpretation in matrix viewpoint

$$\mathbf{f}(t) = \left( \begin{bmatrix} t & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \right) \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \end{bmatrix}$$

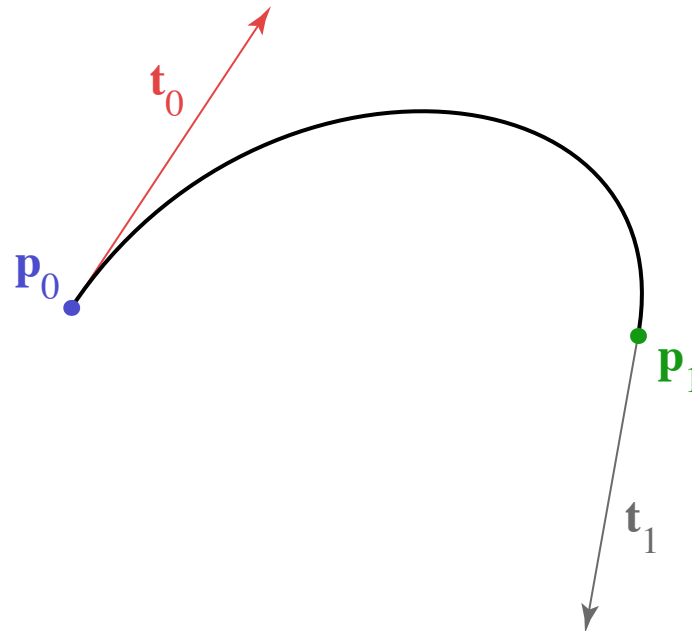
# Trivial example: piecewise linear

- Vector blending formulation: “average of points”
  - blending functions: contribution of each point as  $t$  changes



# Hermite splines

- Less trivial example
- Form of curve: piecewise cubic
- Constraints: endpoints and tangents (derivatives)



# Hermite splines

- Solve constraints to find coefficients

$$x(t) = at^3 + bt^2 + ct + d$$

$$x'(t) = 3at^2 + 2bt + c$$

$$x(0) = x_0 = d$$

$$x(1) = x_1 = a + b + c + d$$

$$x'(0) = x'_0 = c$$

$$x'(1) = x'_1 = 3a + 2b + c$$

$$d = x_0$$

$$c = x'_0$$

$$a = 2x_0 - 2x_1 + x'_0 + x'_1$$

$$b = -3x_0 + 3x_1 - 2x'_0 - x'_1$$

# Matrix form of spline

$$\mathbf{f}(t) = \mathbf{a}t^3 + \mathbf{b}t^2 + \mathbf{c}t + \mathbf{d}$$

$$\begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \begin{bmatrix} \times & \times & \times & \times \\ \times & \times & \times & \times \\ \times & \times & \times & \times \\ \times & \times & \times & \times \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix}$$

$$\mathbf{f}(t) = b_0(t)\mathbf{p}_0 + b_1(t)\mathbf{p}_1 + b_2(t)\mathbf{p}_2 + b_3(t)\mathbf{p}_3$$



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# Hermite splines

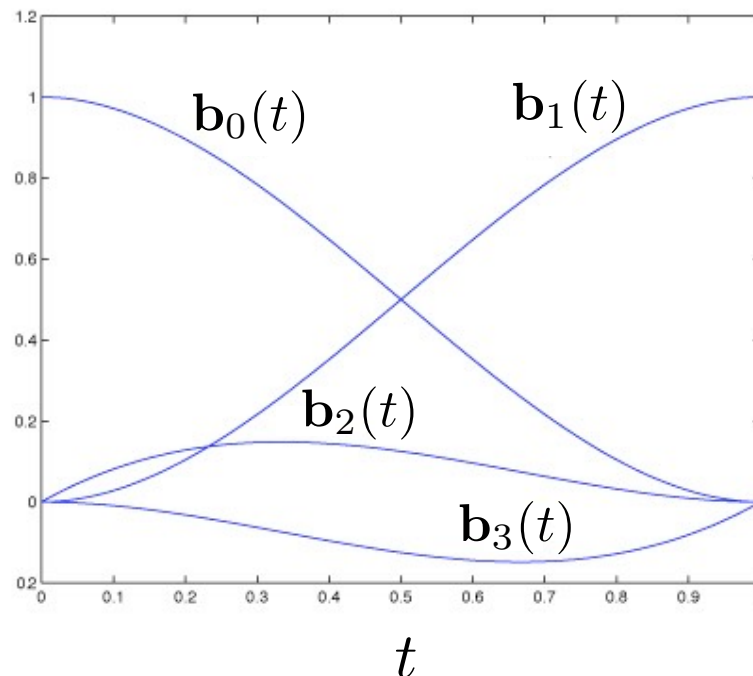
- Matrix form is much simpler

$$\mathbf{f}(t) = \begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{t}_0 \\ \mathbf{t}_1 \end{bmatrix}$$

- coefficients = rows
- basis functions = columns
  - note **p** columns sum to  $[0 \ 0 \ 0 \ 1]^\top$

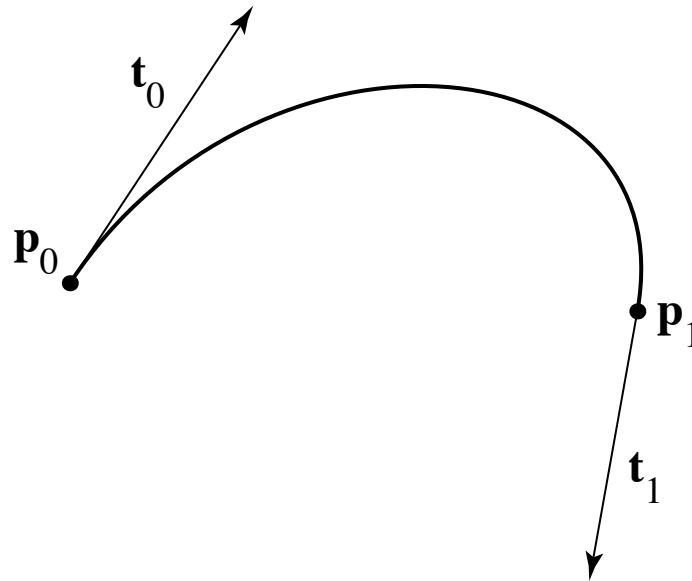
# Hermite splines

- Hermite blending functions



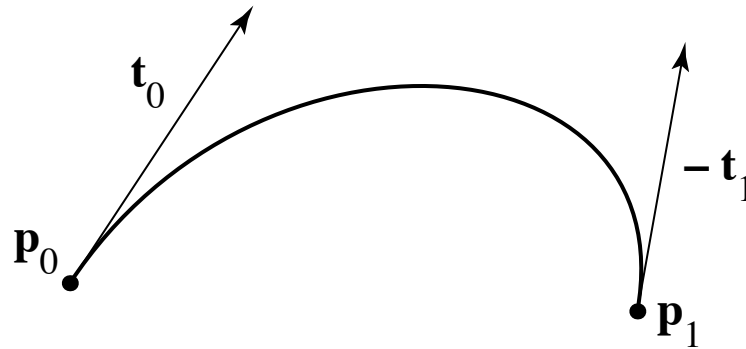
# Hermite to Bézier

- Mixture of points and vectors is awkward
- Specify tangents as differences of points



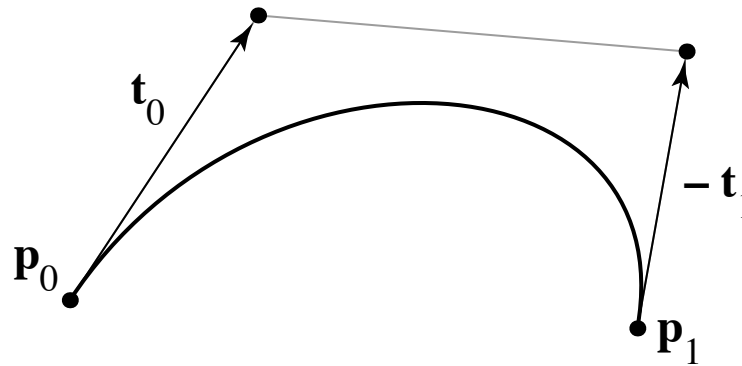
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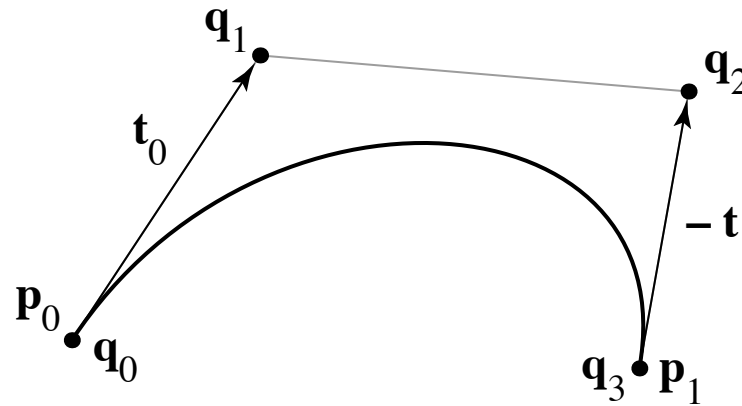
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# Hermite to Bézier

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- Specify tangents as differences of points



I'm calling these points **q** just for this slide and the next one.

- note derivative is defined as 3 times offset
  - reason is illustrated by linear case



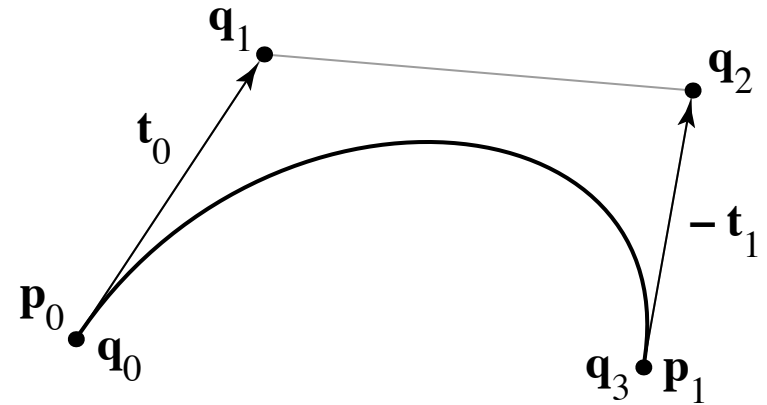
# Hermite to Bézier

$$\mathbf{p}_0 = \mathbf{q}_0$$

$$\mathbf{p}_1 = \mathbf{q}_3$$

$$\mathbf{t}_0 = 3(\mathbf{q}_1 - \mathbf{q}_0)$$

$$\mathbf{t}_1 = 3(\mathbf{q}_3 - \mathbf{q}_2)$$



$$\begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{v}_0 \\ \mathbf{v}_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -3 & 3 & 0 & 0 \\ 0 & 0 & -3 & 3 \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \mathbf{q}_2 \\ \mathbf{q}_3 \end{bmatrix}$$

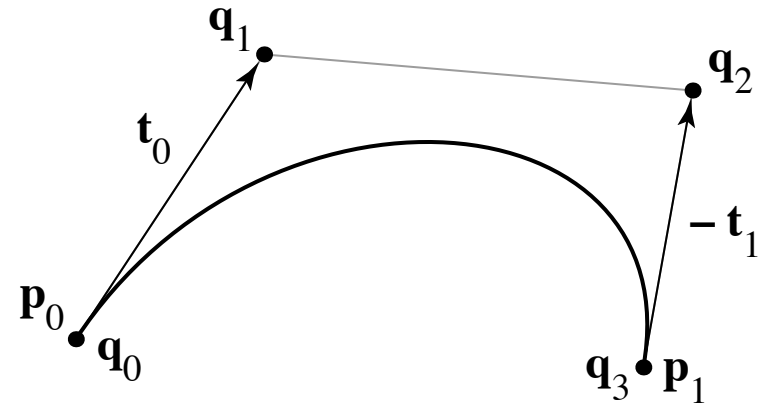
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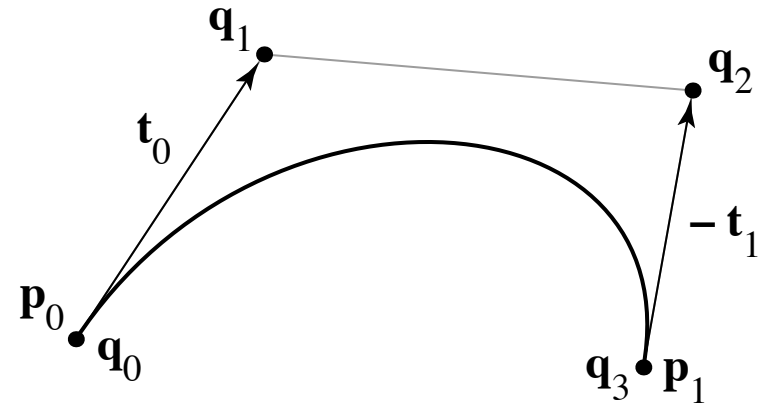
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$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 3 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \mathbf{q}_2 \\ \mathbf{q}_3 \end{bmatrix}$$

# Bézier matrix

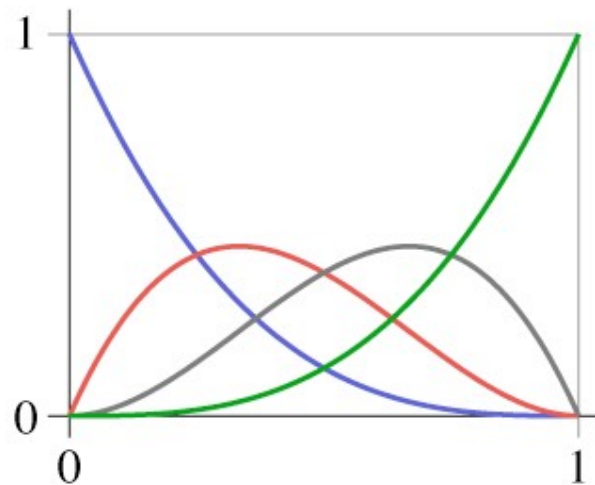
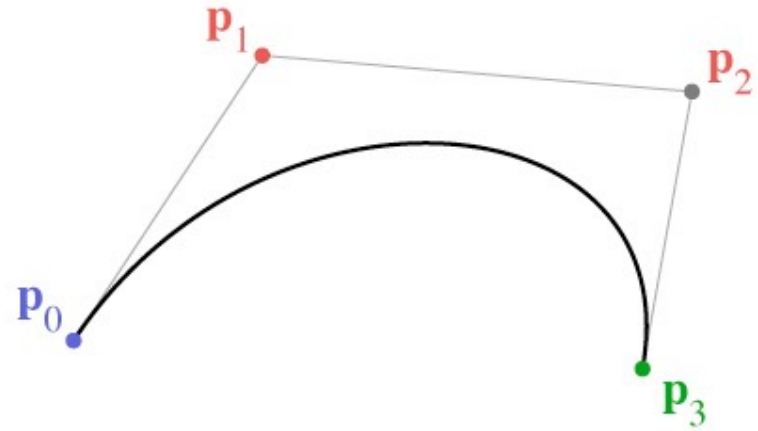
$$\mathbf{f}(t) = \begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \begin{bmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 3 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix}$$

- note that these are the Bernstein polynomials

$$b_{n,k}(t) = \binom{n}{k} t^k (1-t)^{n-k}$$

and that defines Bézier curves for any degree

# Bézier basis



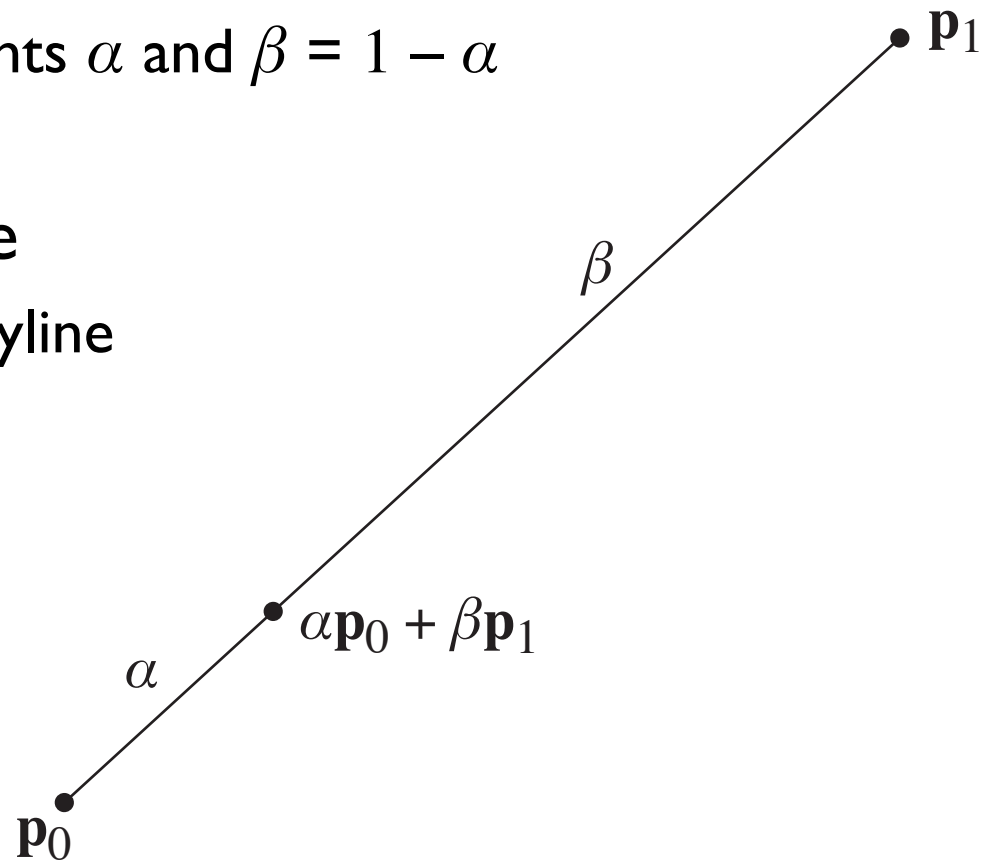
# Another way to Bézier segments

- A really boring spline segment:  $f(t) = p_0$ 
  - it only has one control point
  - the curve stays at that point for the whole time
- Only good for building a *piecewise constant* spline
  - a.k.a. a set of points

•  $p_0$

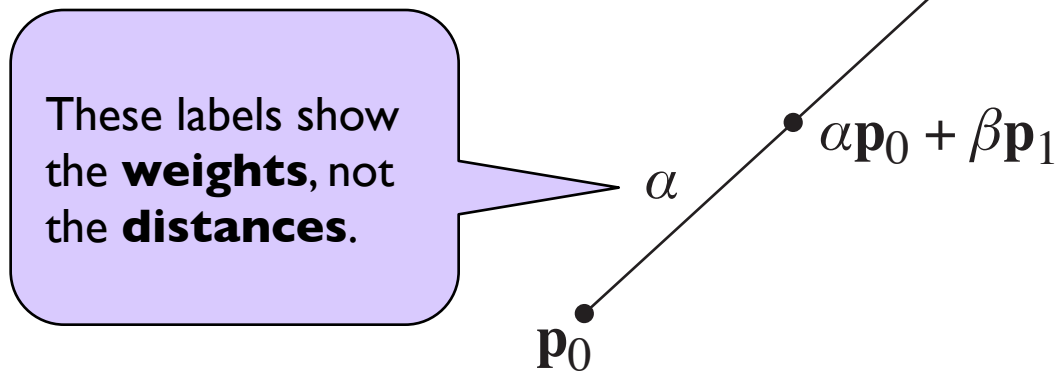
# Another way to Bézier segments

- A piecewise linear spline segment
  - two control points per segment
  - blend them with weights  $\alpha$  and  $\beta = 1 - \alpha$
- Good for building a piecewise linear spline
  - a.k.a. a polygon or polyline



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# Another way to Bézier segments

- A linear blend of two piecewise linear segments
  - three control points now
  - interpolate on both segments using  $\alpha$  and  $\beta$
  - blend the results with the same weights
- makes a quadratic spline segment
  - finally, a curve!

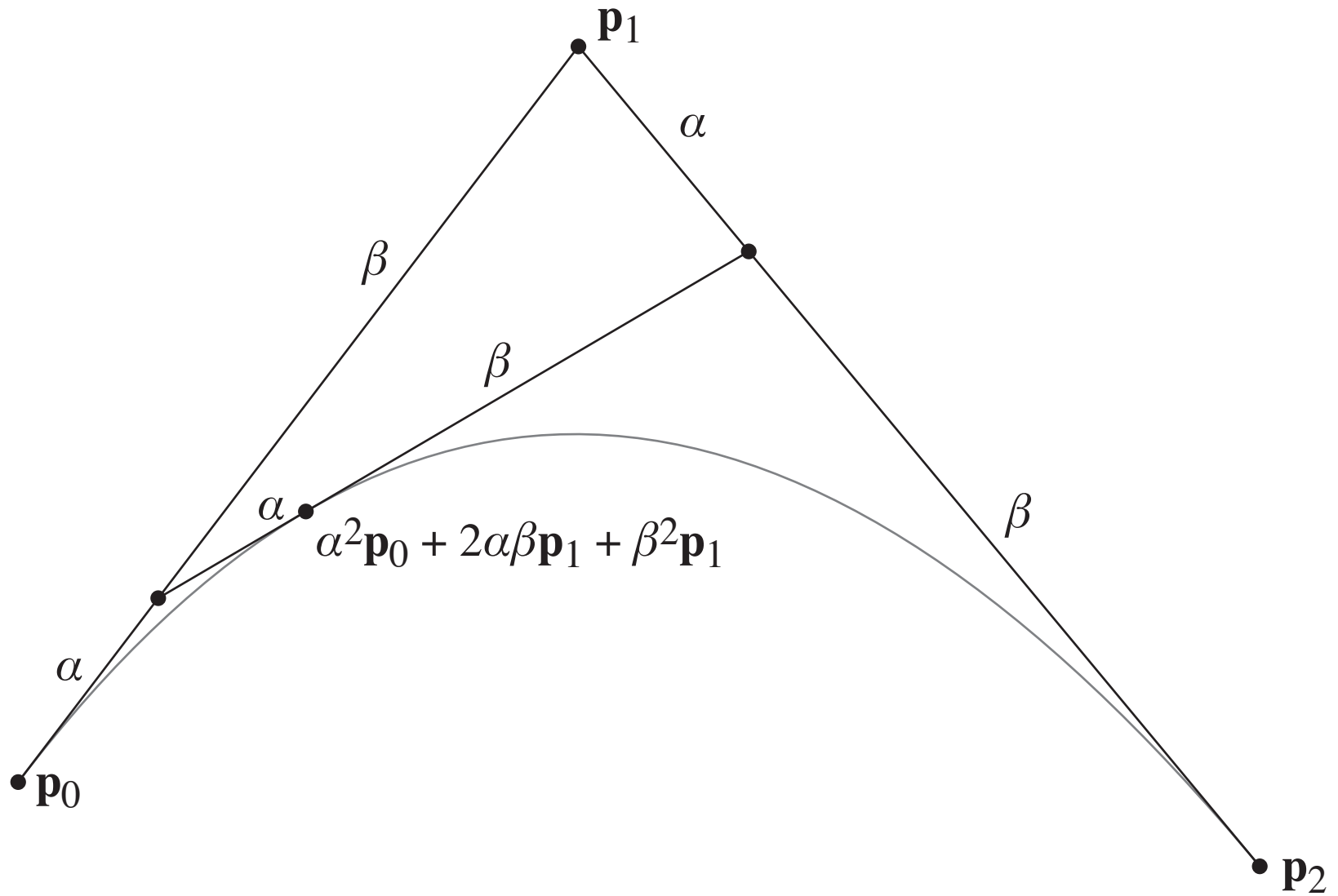
$$\mathbf{p}_{1,0} = \alpha \mathbf{p}_0 + \beta \mathbf{p}_1$$

$$\mathbf{p}_{1,1} = \alpha \mathbf{p}_1 + \beta \mathbf{p}_2$$

$$\mathbf{p}_{2,0} = \alpha \mathbf{p}_{1,0} + \beta \mathbf{p}_{1,1}$$

$$= \alpha \alpha \mathbf{p}_0 + \alpha \beta \mathbf{p}_1 + \beta \alpha \mathbf{p}_1 + \beta \beta \mathbf{p}_2$$

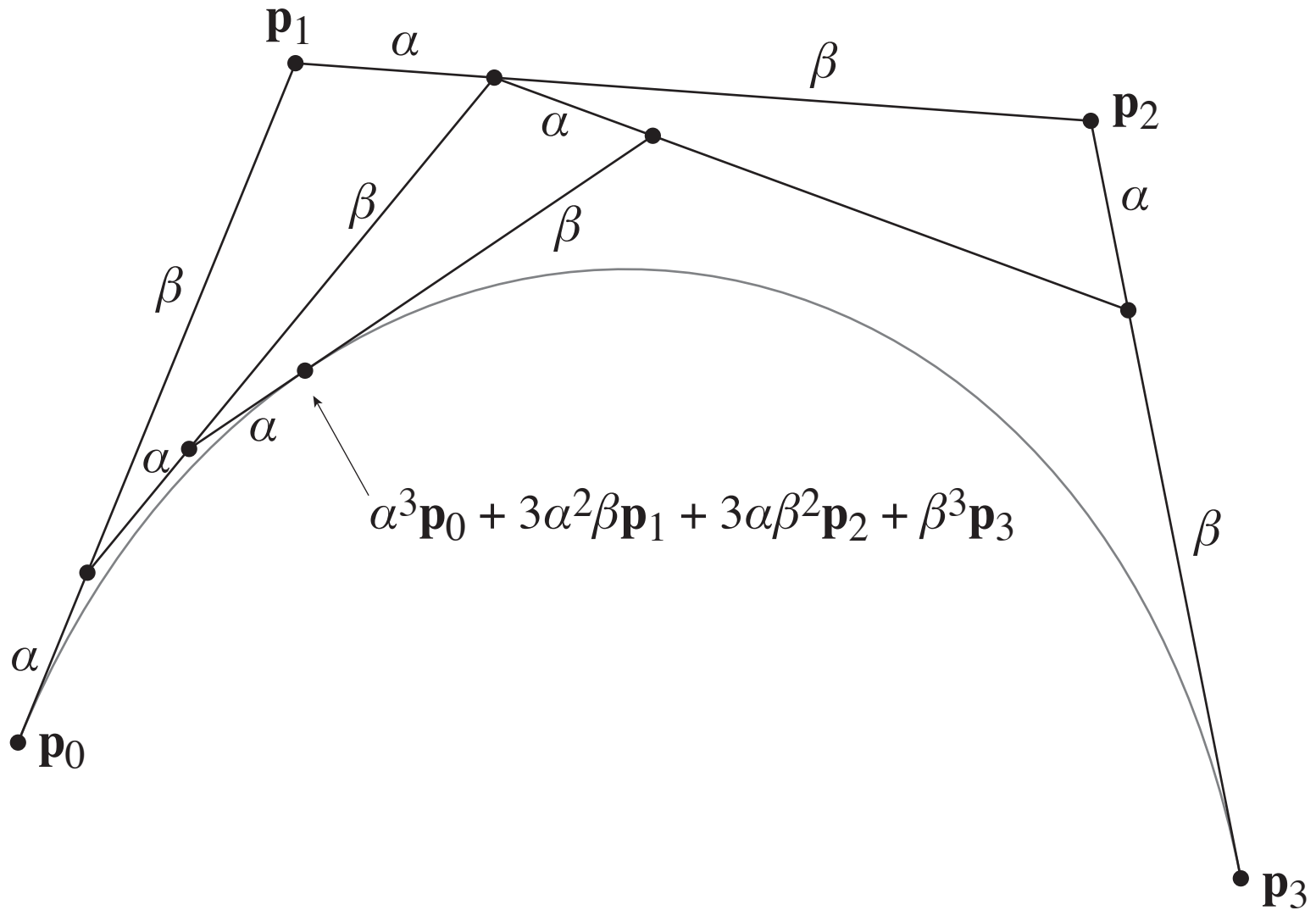
$$= \alpha^2 \mathbf{p}_0 + 2\alpha\beta \mathbf{p}_1 + \beta^2 \mathbf{p}_2$$



# Another way to Bézier segments

- Cubic segment: blend of two quadratic segments
  - four control points now (overlapping sets of 3)
  - interpolate on each quadratic using  $\alpha$  and  $\beta$
  - blend the results with the same weights
- makes a cubic spline segment
  - this is the familiar one for graphics—but you can keep going

$$\begin{aligned}\mathbf{p}_{3,0} &= \alpha \mathbf{p}_{2,0} + \beta \mathbf{p}_{2,1} \\ &= \alpha \alpha \alpha \mathbf{p}_0 + \alpha \alpha \beta \mathbf{p}_1 + \alpha \beta \alpha \mathbf{p}_1 + \alpha \beta \beta \mathbf{p}_2 \\ &\quad \beta \alpha \alpha \mathbf{p}_1 + \beta \alpha \beta \mathbf{p}_2 + \beta \beta \alpha \mathbf{p}_2 + \beta \beta \beta \mathbf{p}_3 \\ &= \alpha^3 \mathbf{p}_0 + 3\alpha^2 \beta \mathbf{p}_1 + 3\alpha \beta^2 \mathbf{p}_2 + \beta^3 \mathbf{p}_3\end{aligned}$$



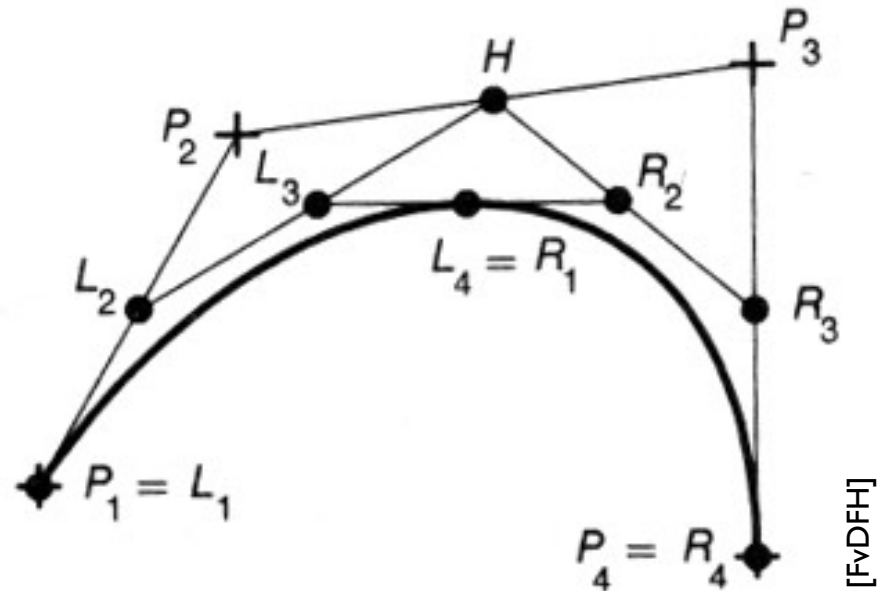
# de Casteljau's algorithm

- A recurrence for computing points on Bézier spline segments:

$$\mathbf{p}_{0,i} = \mathbf{p}_i$$

$$\mathbf{p}_{n,i} = \alpha \mathbf{p}_{n-1,i} + \beta \mathbf{p}_{n-1,i+1}$$

- Cool additional feature:  
also subdivides  
the segment into two  
shorter ones

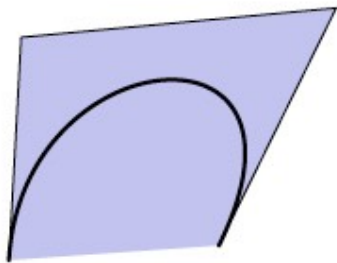


# Cubic Bézier splines

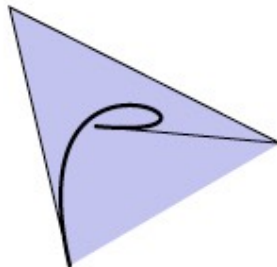
- Very widely used type, especially in 2D
  - e.g. it is a primitive in PostScript/PDF
- Can represent smooth curves with corners
- Nice de Casteljau recurrence for evaluation
- Can easily add points at any position
- Illustrator demo

# Spline segment properties

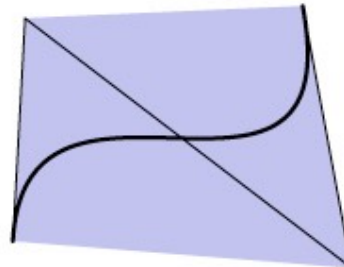
- Convex hull property
  - convex hull = smallest convex region containing points
    - think of a rubber band around some pins
  - some splines stay inside convex hull of control points
    - make clipping, culling, picking, etc. simpler



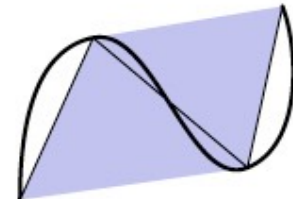
**YES**



**YES**



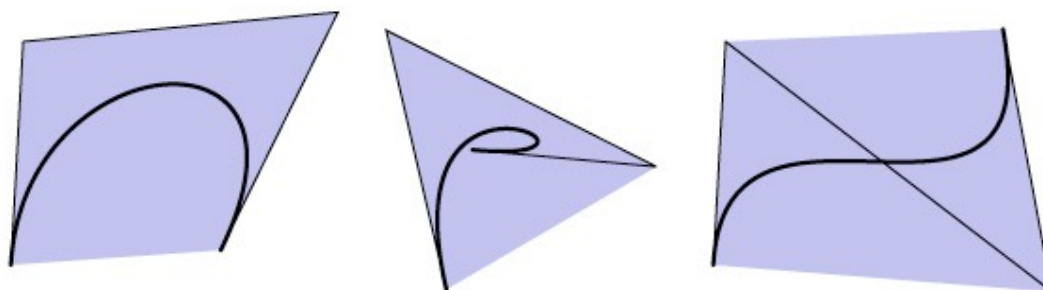
**YES**



**NO**

# Convex hull

- If basis functions are all positive, the spline has the convex hull property
  - we're still requiring them to sum to 1

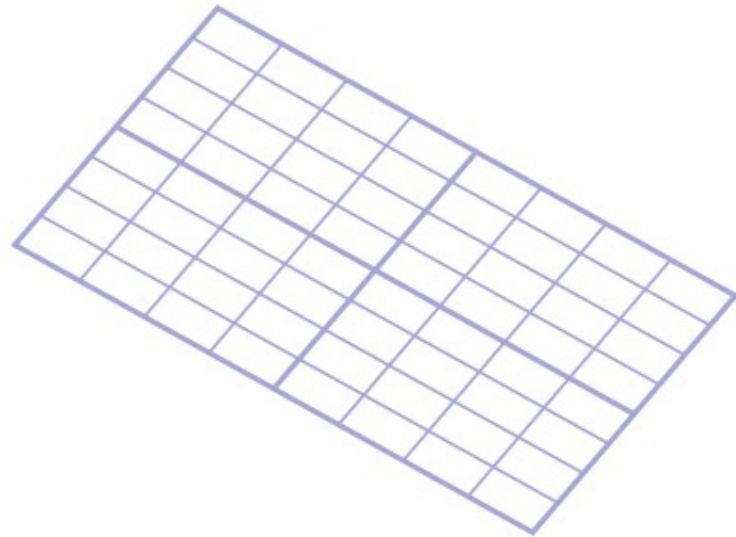
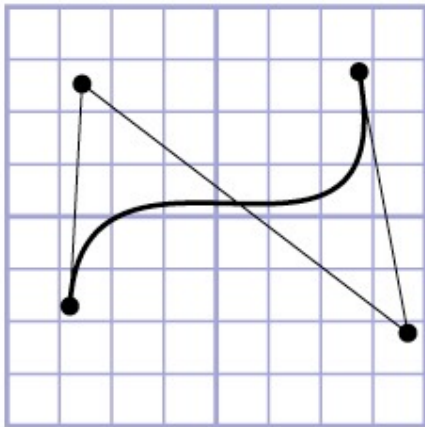


- if any basis function is ever negative, no convex hull prop.
  - proof: take the other three points at the same place



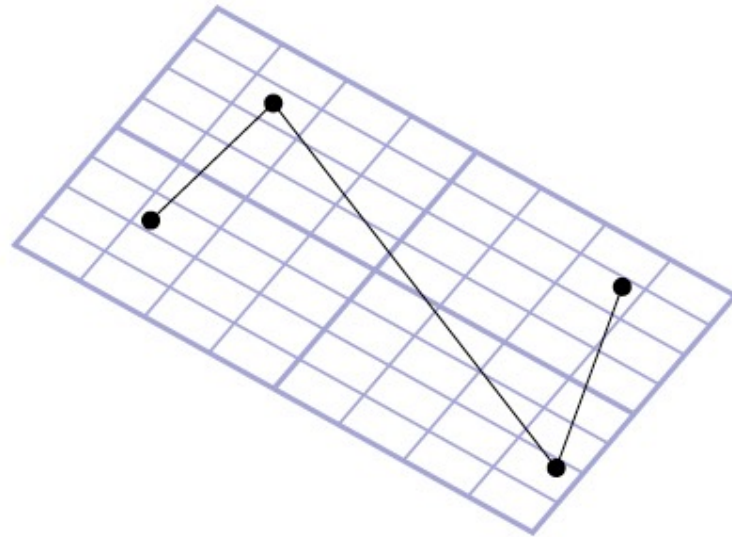
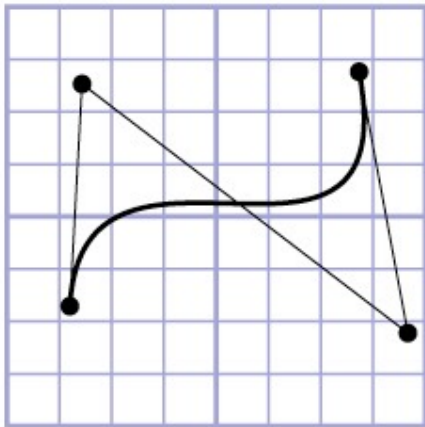
# Affine invariance

- Transforming the control points is the same as transforming the curve
  - true for all commonly used splines
  - extremely convenient in practice...



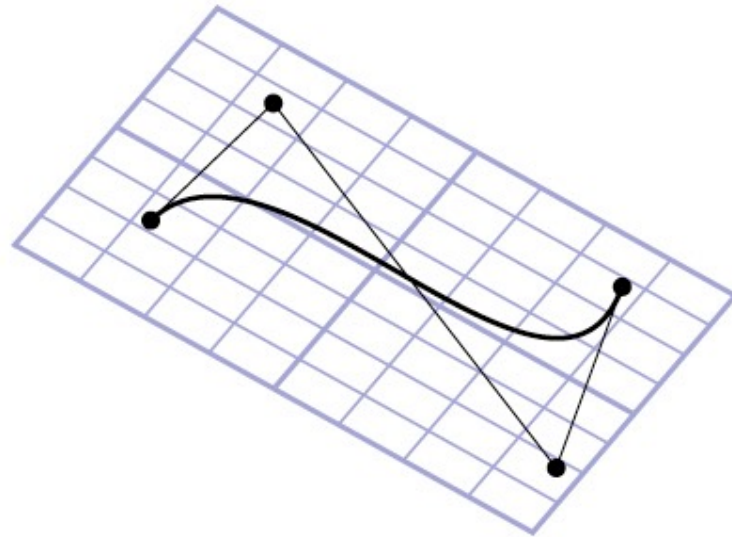
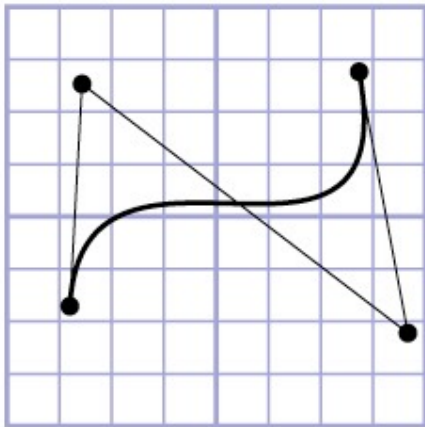
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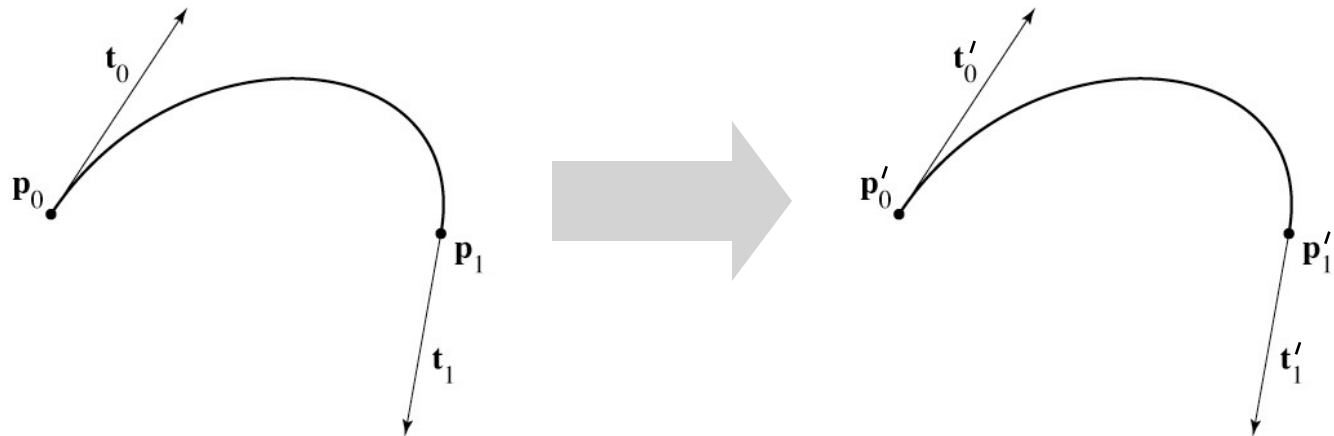
# Affine invariance

- Transforming the control points is the same as transforming the curve
  - true for all commonly used splines
  - extremely convenient in practice...



# Affine invariance

- Basis functions associated with points should always sum to 1



$$\begin{aligned}\mathbf{p}(t) &= b_0\mathbf{p}_0 + b_1\mathbf{p}_1 + b_2\mathbf{v}_0 + b_3\mathbf{v}_1 \\ \mathbf{p}'(t) &= b_0(\mathbf{p}_0 + \mathbf{u}) + b_1(\mathbf{p}_1 + \mathbf{u}) + b_2\mathbf{v}_0 + b_3\mathbf{v}_1 \\ &= b_0\mathbf{p}_0 + b_1\mathbf{p}_1 + b_2\mathbf{v}_0 + b_3\mathbf{v}_1 + (b_0 + b_1)\mathbf{u} \\ &= \mathbf{p}(t) + \mathbf{u}\end{aligned}$$

# Spline Curves

# Chaining spline segments

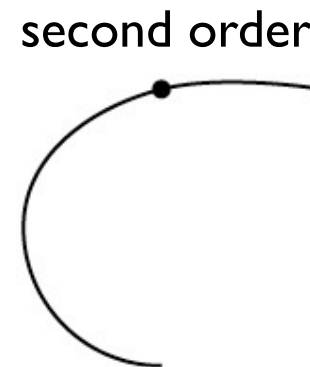
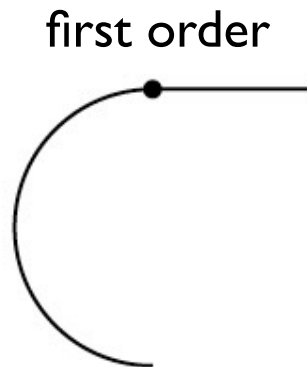
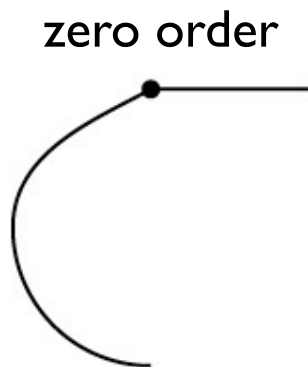
- Can only do so much with a single polynomial
- Can use these functions as segments of a longer curve
  - curve from  $t = 0$  to  $t = 1$  defined by first segment
  - curve from  $t = 1$  to  $t = 2$  defined by second segment

$$\mathbf{f}(t) = \mathbf{f}_i(t - i) \quad \text{for } i \leq t \leq i + 1$$

- To avoid discontinuity, match derivatives at junctions
  - this produces a  $C^1$  curve

# Continuity

- Smoothness can be described by degree of continuity
  - zero-order ( $C^0$ ): position matches from both sides
  - first-order ( $C^1$ ): tangent matches from both sides
  - second-order ( $C^2$ ): curvature matches from both sides
  - $G^n$  vs.  $C^n$

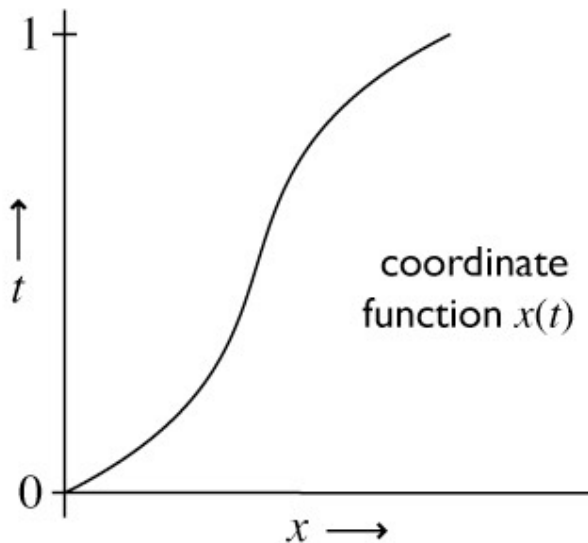
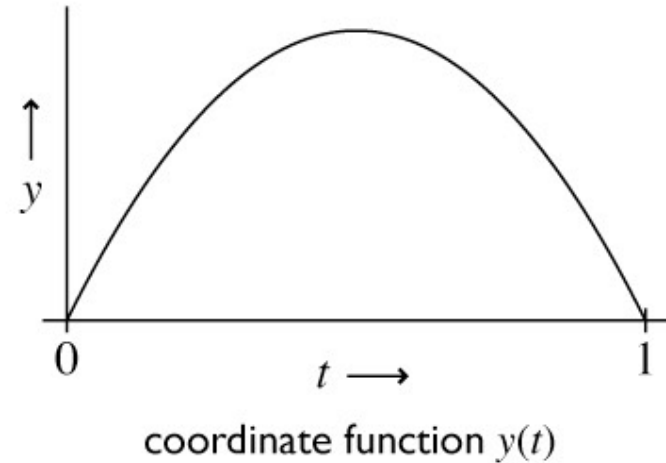
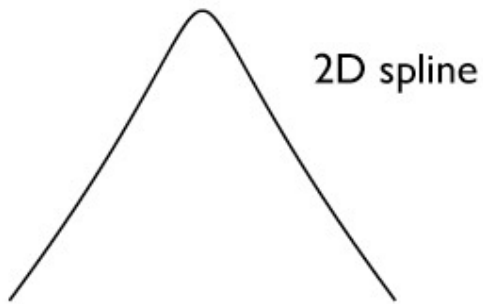


# Continuity

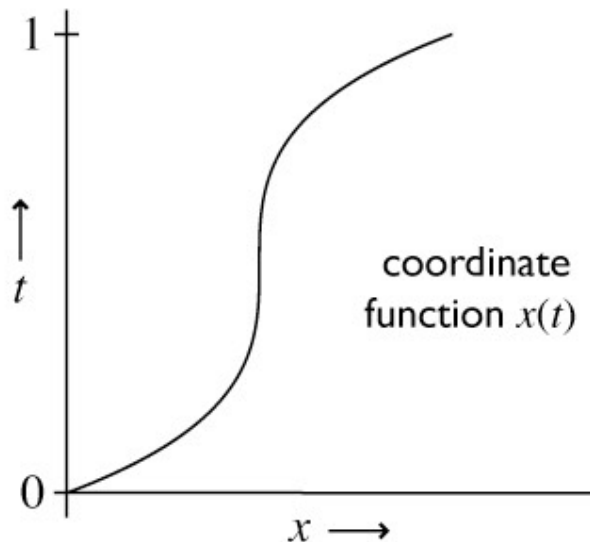
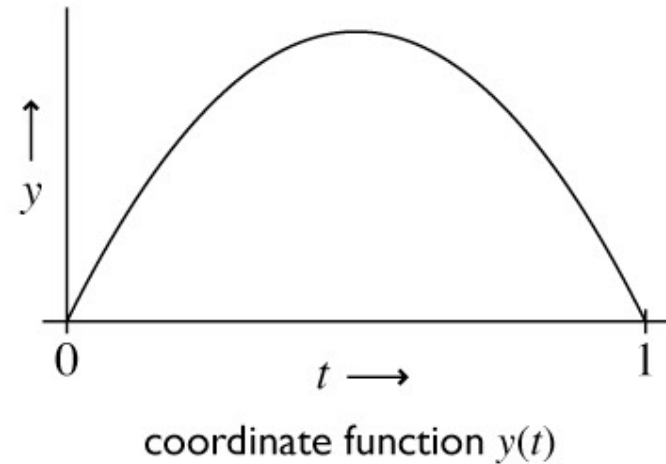
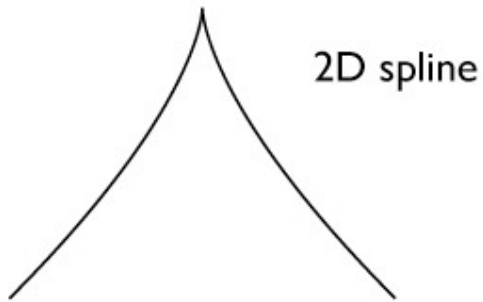
- Parametric continuity ( $C$ ) of spline is continuity of coordinate functions
- Geometric continuity ( $G$ ) is continuity of the curve itself
- Neither form of continuity is guaranteed by the other
  - Can be  $C^1$  but not  $G^1$  when  $\mathbf{p}(t)$  comes to a halt (next slide)
  - Can be  $G^1$  but not  $C^1$  when the tangent vector changes length abruptly



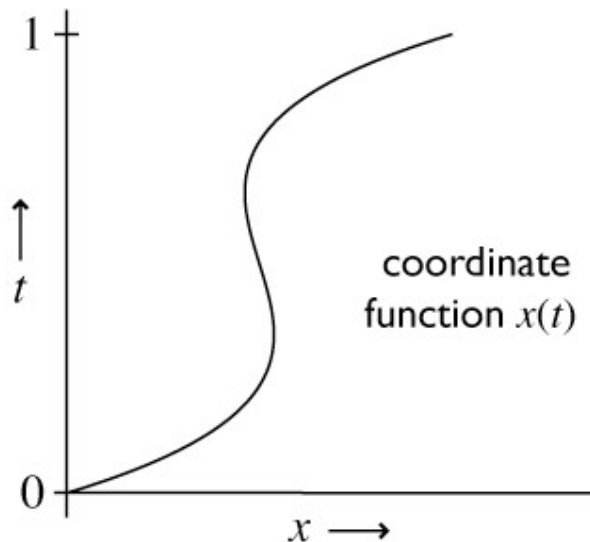
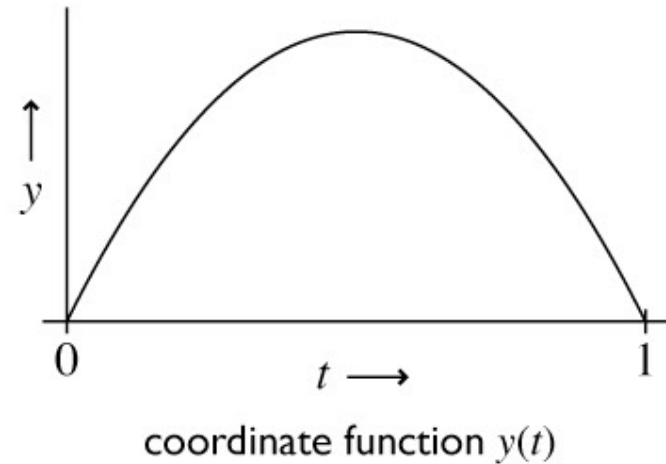
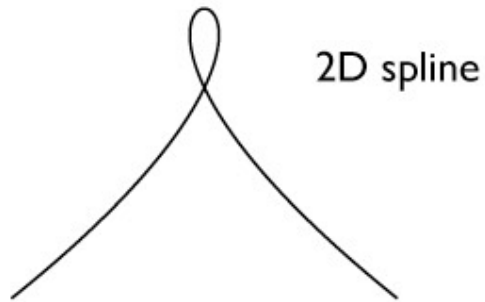
# Geometric vs. parametric continuity



# Geometric vs. parametric continuity

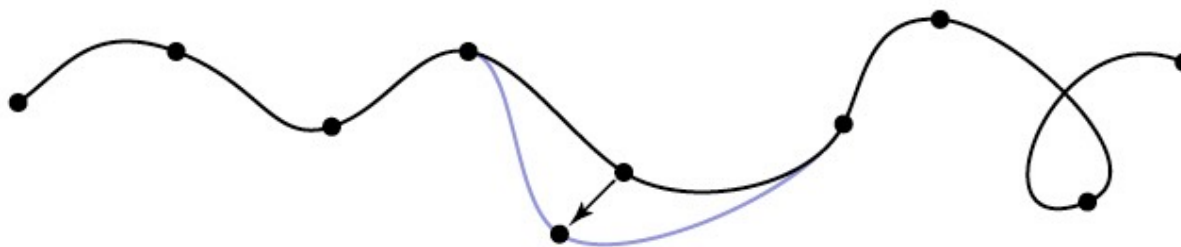


# Geometric vs. parametric continuity



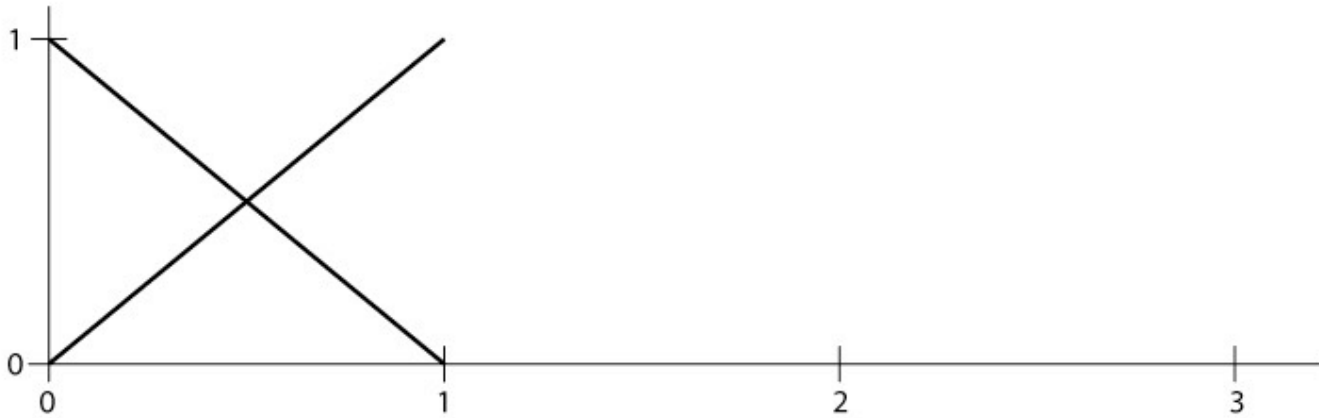
# Control

- Local control
  - changing control point only affects a limited part of spline
  - without this, splines are very difficult to use
  - many likely formulations lack this
    - natural spline
    - polynomial fits



# Trivial example: piecewise linear

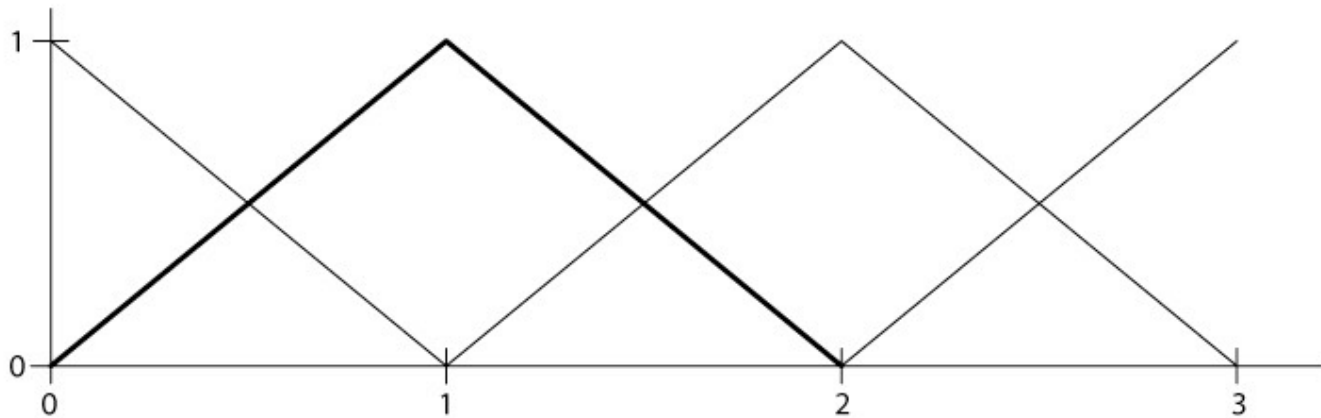
- Basis function formulation: “function times point”
  - basis functions: contribution of each point as  $t$  changes



- can think of them as blending functions glued together
- this is just like a reconstruction filter!

# Trivial example: piecewise linear

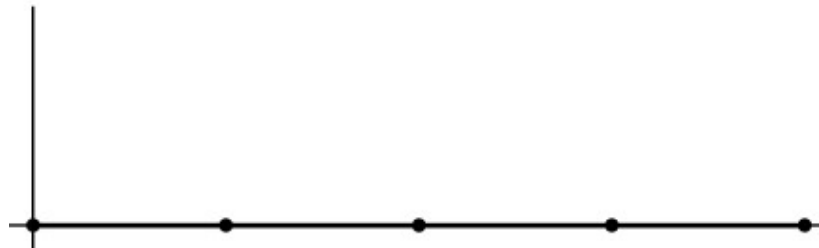
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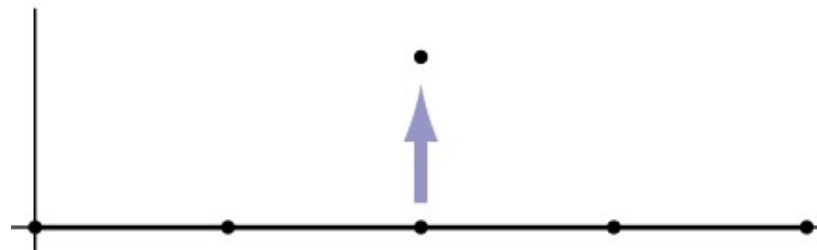
# Seeing the basis functions

- Basis functions of a spline are revealed by how the curve changes in response to a change in one control
  - to get a graph of the basis function, start with the curve laid out in a straight, constant-speed line
    - what are  $x(t)$  and  $y(t)$ ?
  - then move one control straight up



# Seeing the basis functions

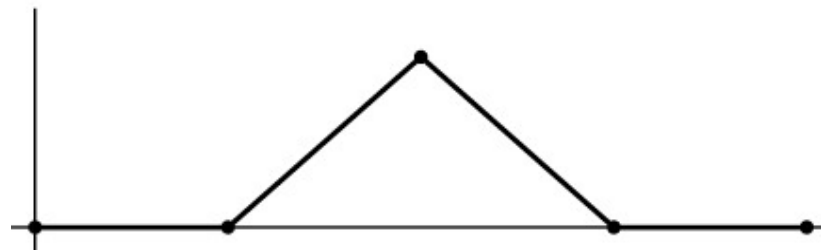
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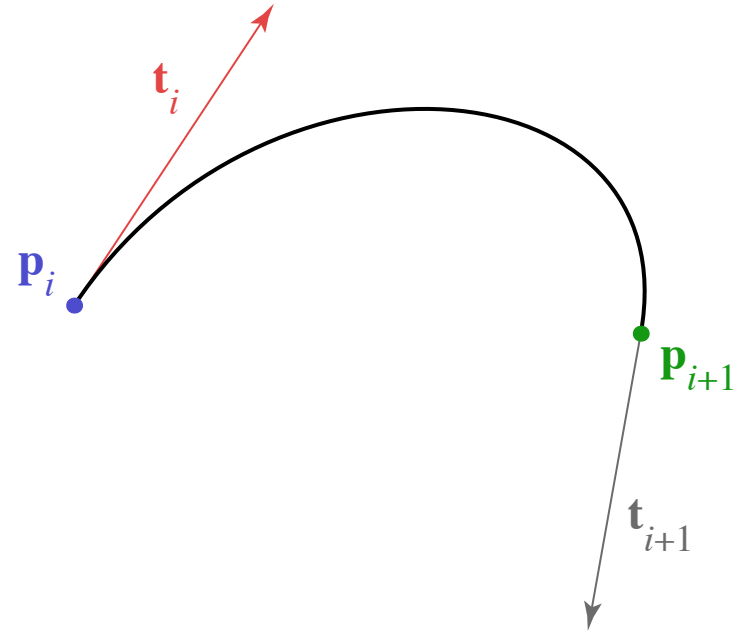
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# Hermite splines

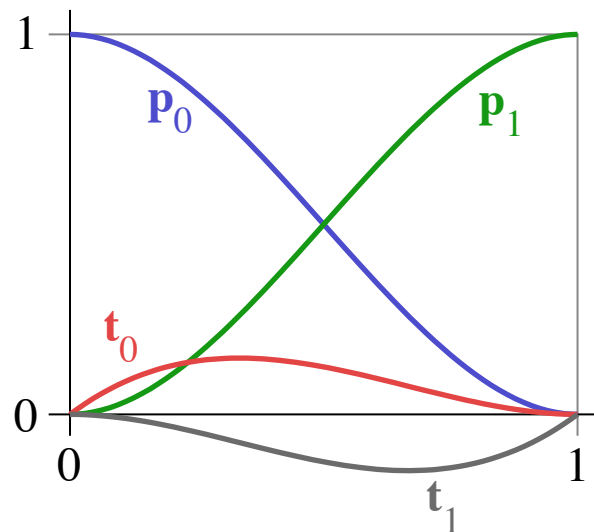
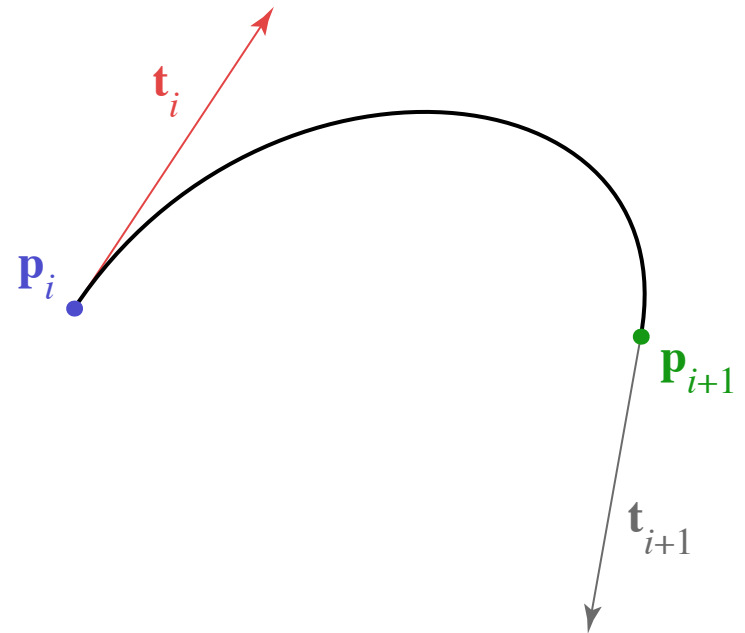
- Constraints are endpoints and endpoint tangents



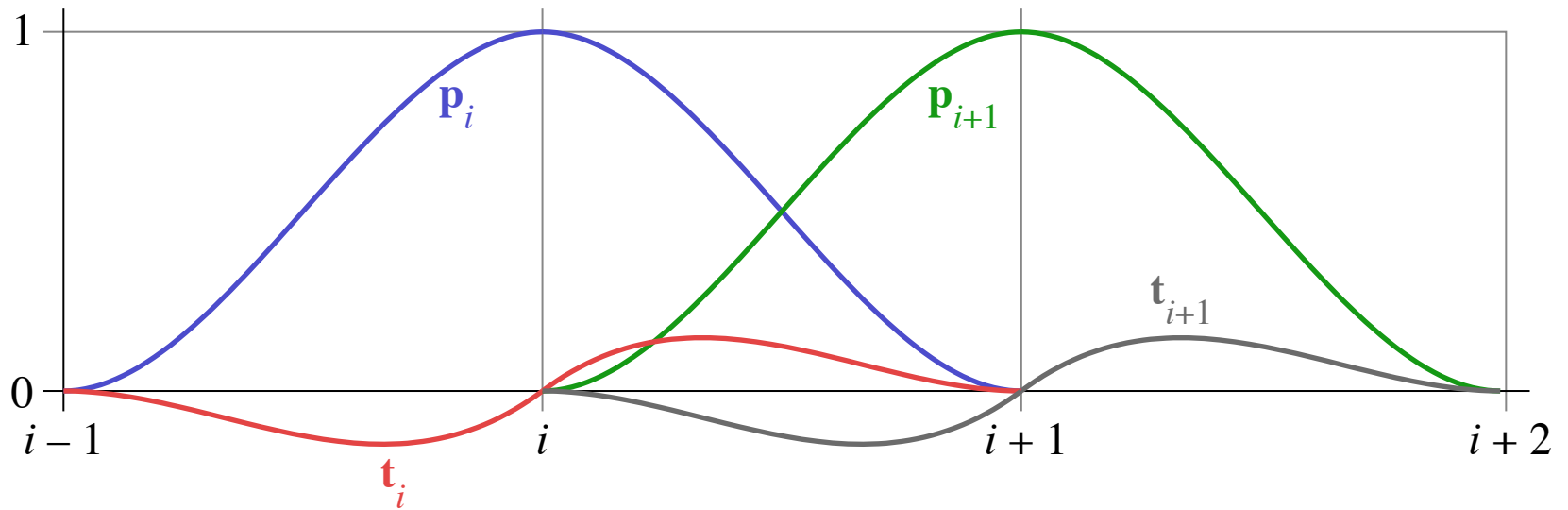
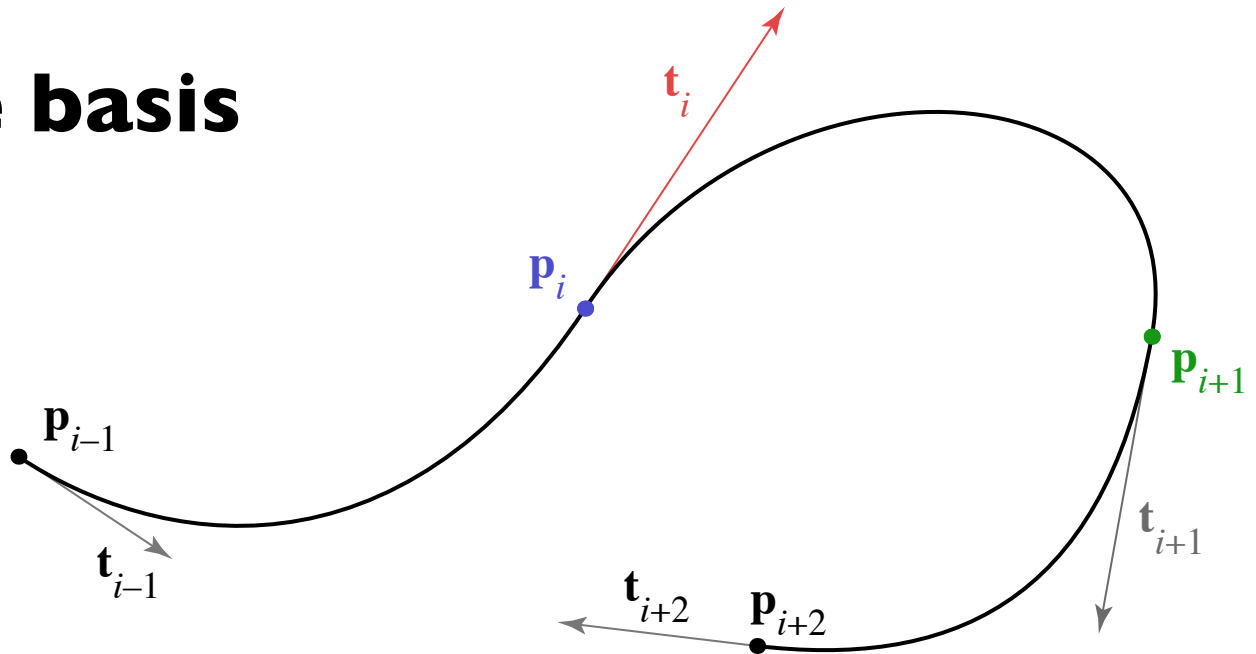
$$\mathbf{f}(t) = \begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 1 & 2 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{p}'_0 \\ \mathbf{p}'_1 \end{bmatrix}$$

# Hermite basis

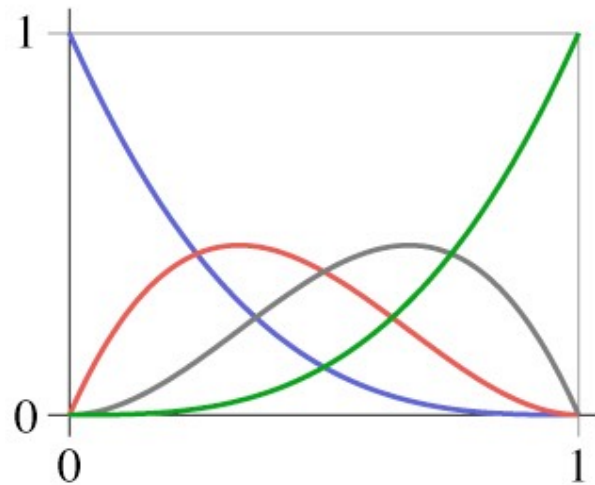
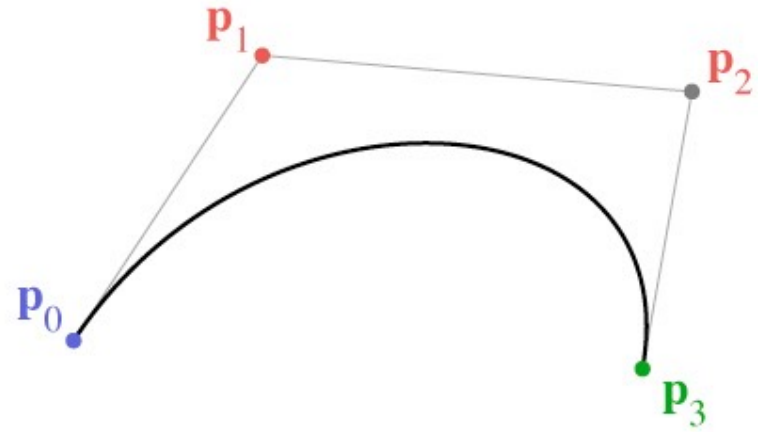
# Hermite basis



# Hermite basis



# Bézier basis

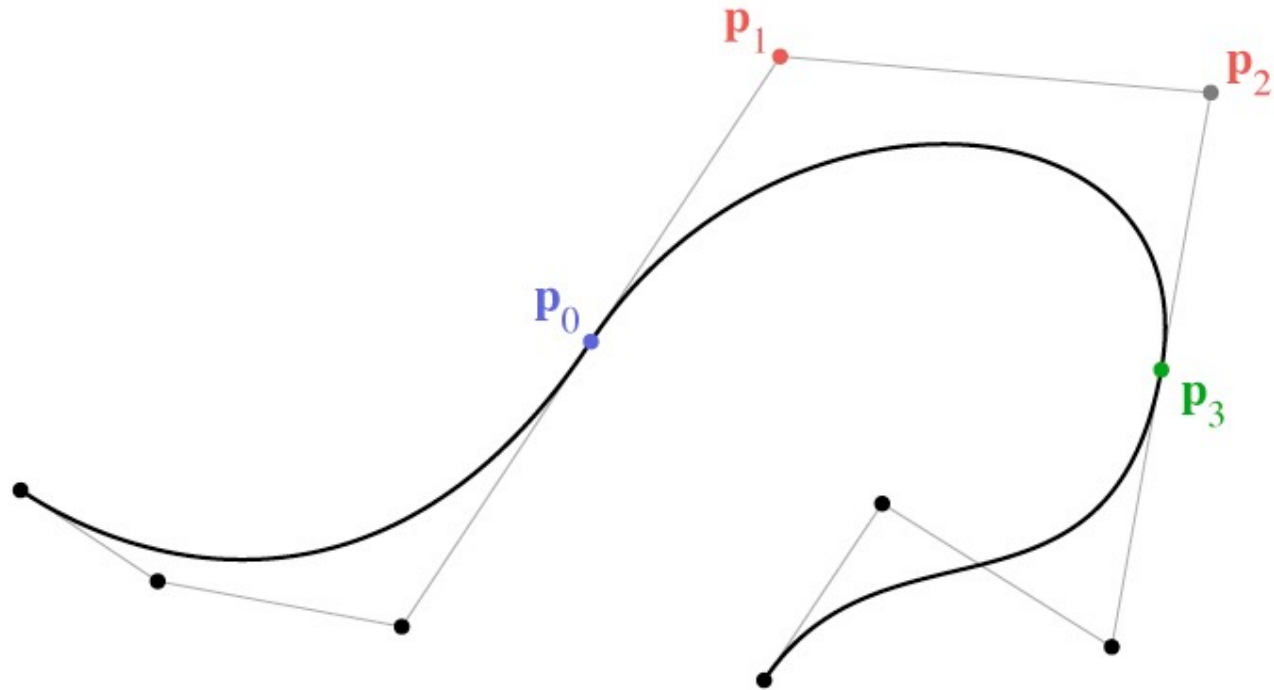


# Chaining Bézier splines

- No continuity built in
- Achieve  $C^1$  using collinear control points

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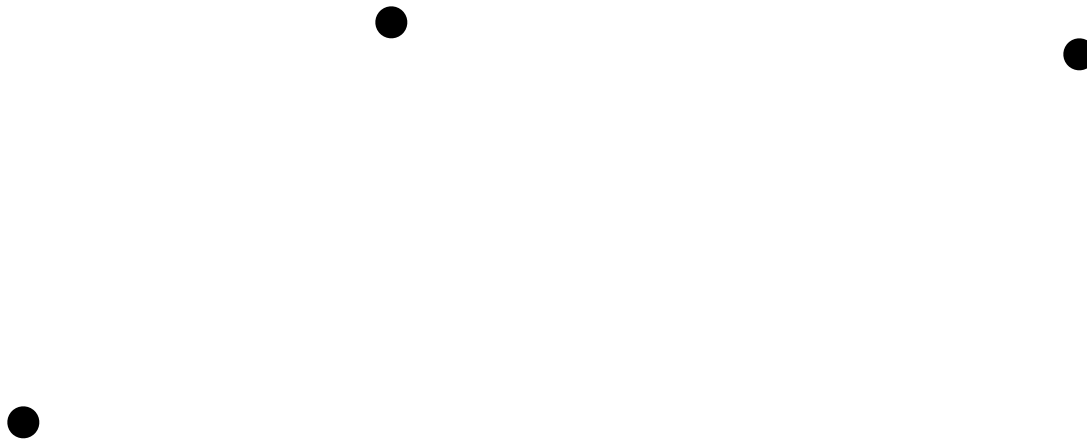


# Chaining spline segments

- Hermite curves are convenient because they can be made long easily
- Bézier curves are convenient because their controls are all points
  - but it is fussy to maintain continuity constraints
  - and they interpolate every 3rd point, which is a little odd
- We derived Bézier from Hermite by defining tangents from control points
  - a similar construction leads to the interpolating *Catmull-Rom* spline

# Hermite to Catmull-Rom

- Have not yet seen any interpolating splines
- Would like to define tangents automatically
  - use adjacent control points

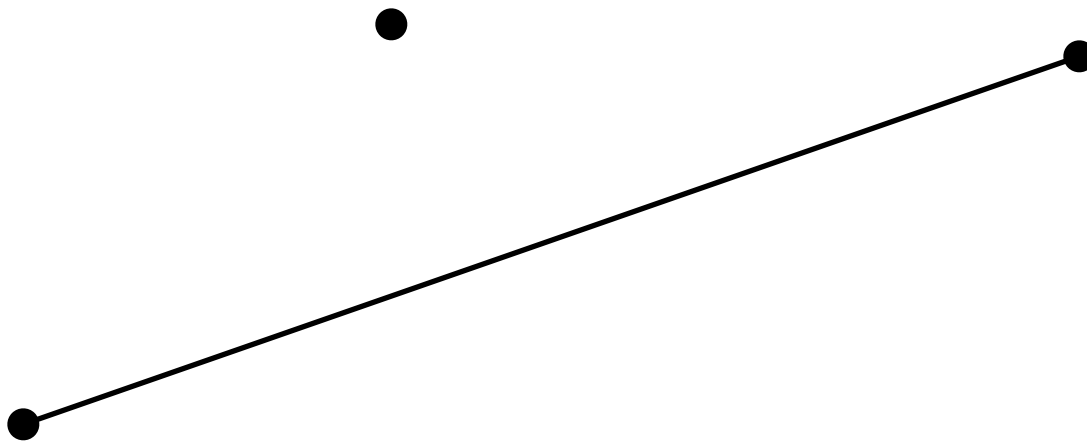


- end tangents: extra points or zero



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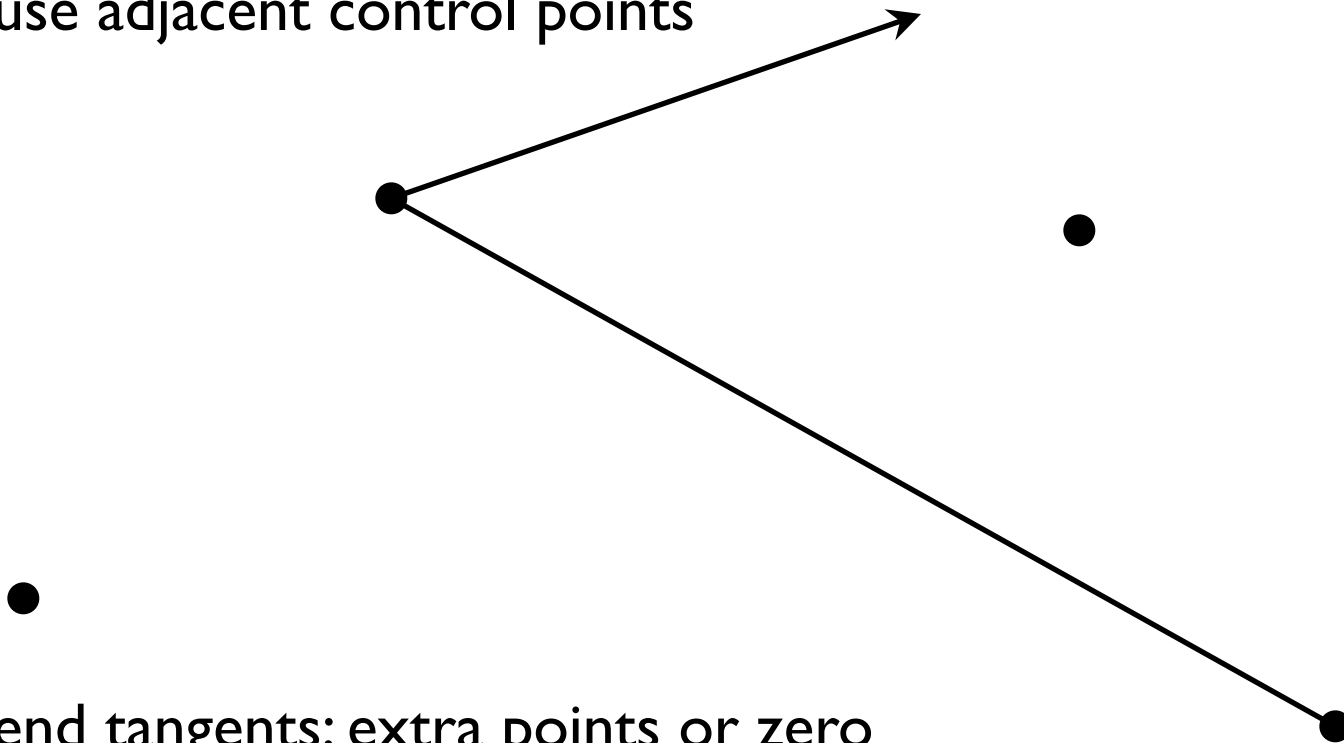


- end tangents: extra points or zero



# Hermite to Catmull-Rom

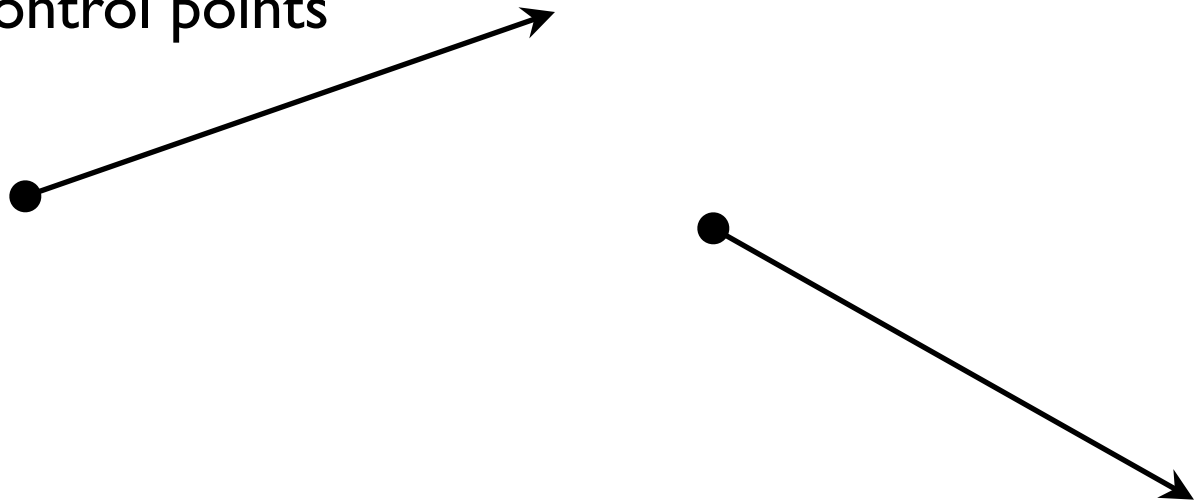
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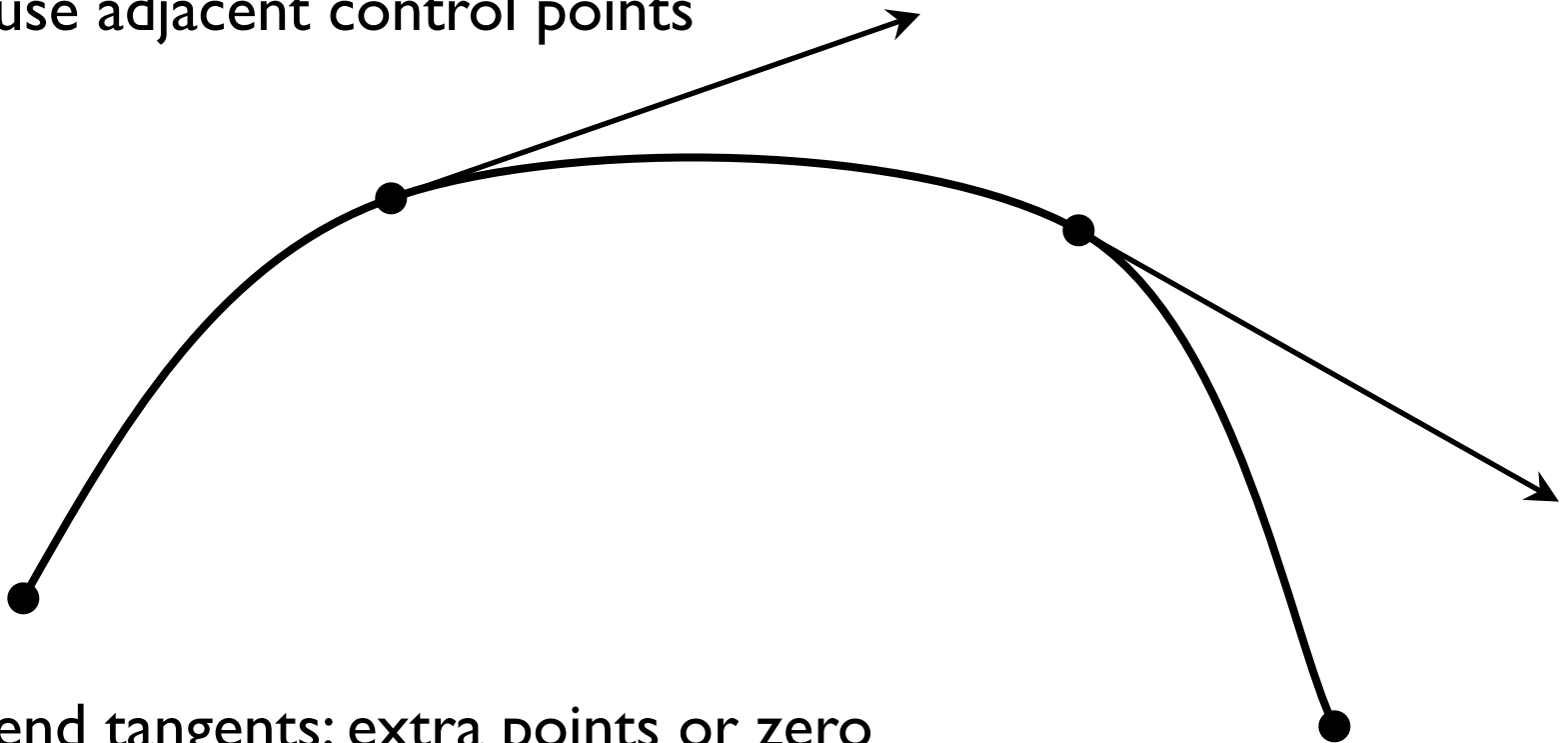


- end tangents: extra points or zero



# Hermite to Catmull-Rom

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- end tangents: extra points or zero

# Hermite to Catmull-Rom

- Tangents are  $(\mathbf{p}_{k+1} - \mathbf{p}_{k-1}) / 2$ 
  - scaling based on same argument about collinear case

$$\mathbf{p}_0 = \mathbf{q}_k$$

$$\mathbf{p}_1 = \mathbf{q}_{k+1}$$

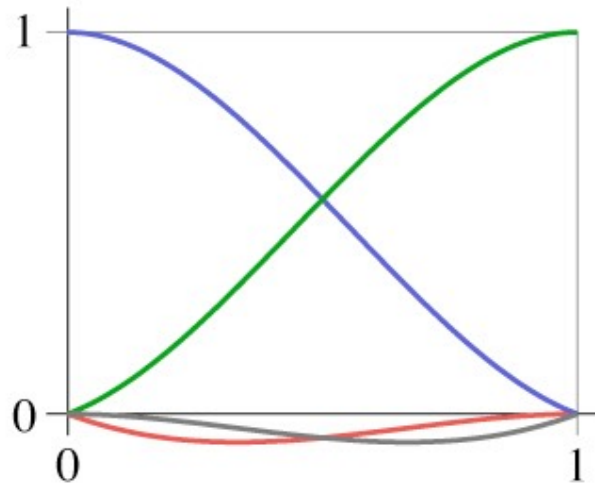
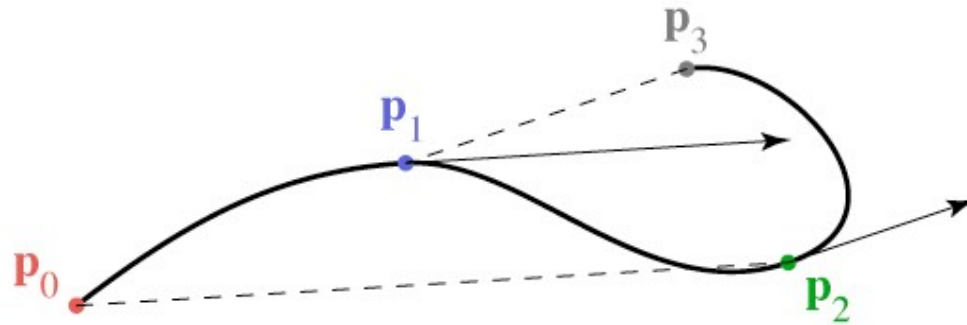
$$\mathbf{v}_0 = 0.5(\mathbf{q}_{k+1} - \mathbf{q}_{k-1})$$

$$\mathbf{v}_1 = 0.5(\mathbf{q}_{k+2} - \mathbf{q}_K)$$

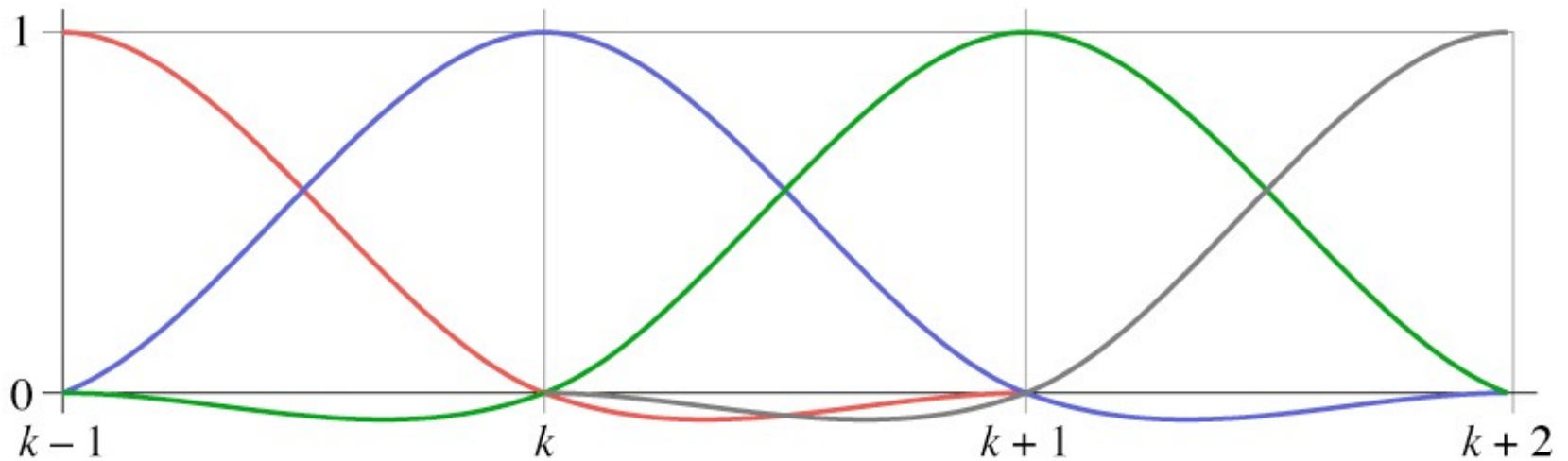
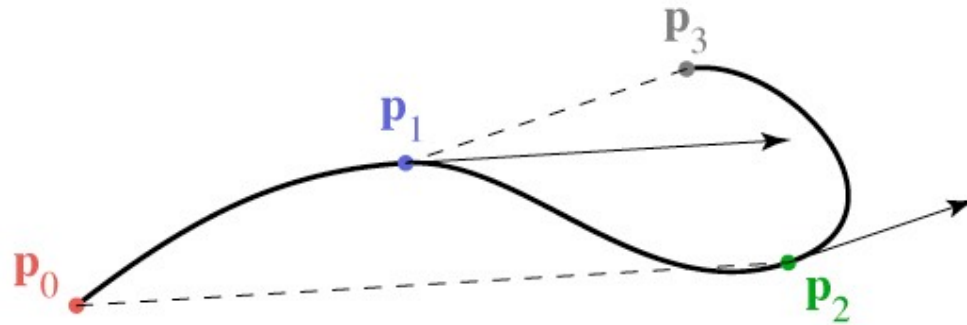
$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -.5 & 0 & .5 & 0 \\ 0 & -.5 & 0 & .5 \end{bmatrix} \begin{bmatrix} \mathbf{q}_{k-1} \\ \mathbf{q}_k \\ \mathbf{q}_{k+1} \\ \mathbf{q}_{k+2} \end{bmatrix}$$



# Catmull-Rom basis



# Catmull-Rom basis



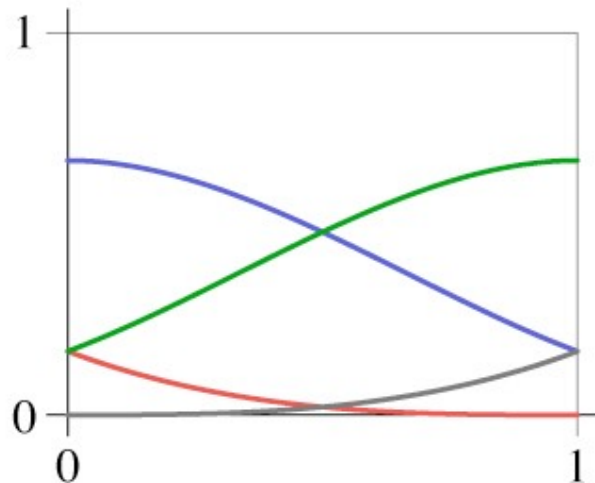
# Catmull-Rom splines

- Our first example of an interpolating spline
- Like Bézier, equivalent to Hermite
  - in fact, all splines of this form are equivalent
- First example of a spline based on just a control point sequence
- Does not have convex hull property

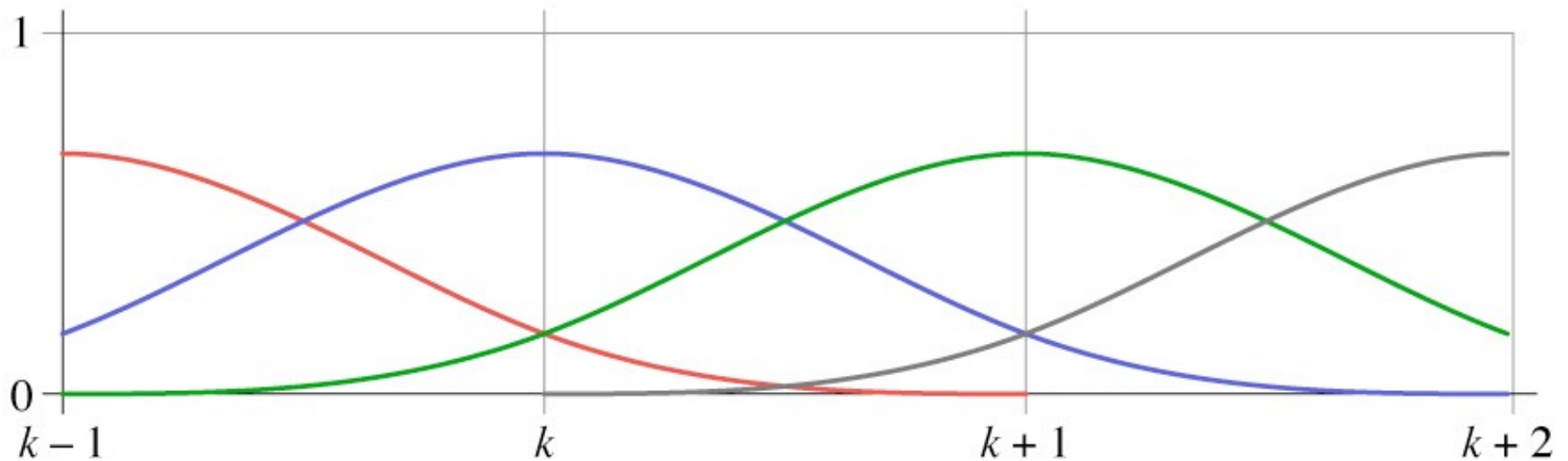
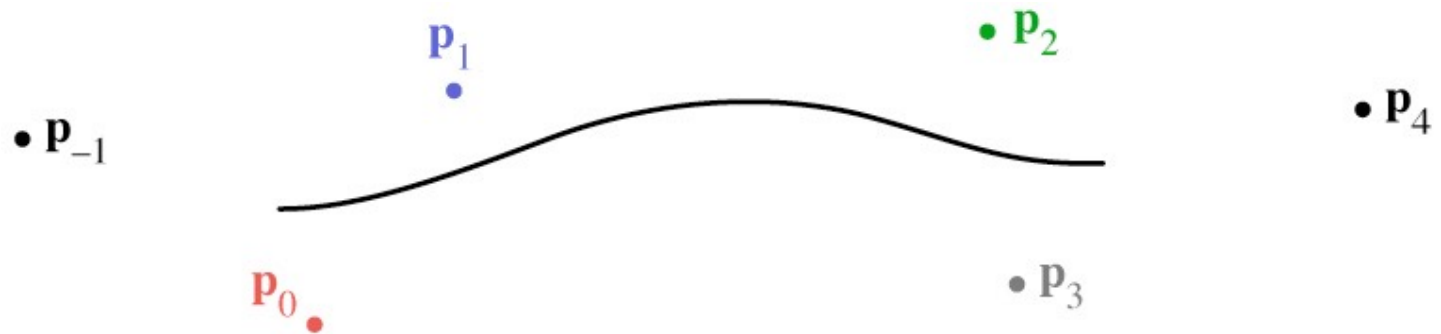
# B-splines

- We may want more continuity than  $C^1$
- We may not need an interpolating spline
- B-splines are a clean, flexible way of making long splines with arbitrary order of continuity
- Various ways to think of construction
  - a simple one is convolution
  - relationship to sampling and reconstruction

# Cubic B-spline basis



# Cubic B-spline basis



# Deriving the B-Spline

- Approached from a different tack than Hermite-style constraints
  - Want a cubic spline; therefore 4 active control points
  - Want  $C^2$  continuity
  - Turns out that is enough to determine everything

# Efficient construction of any B-spline

- B-splines defined for all orders
  - order  $d$ : degree  $d - 1$
  - order  $d$ :  $d$  points contribute to value
- One definition: Cox-deBoor recurrence

$$b_1 = \begin{cases} 1 & 0 \leq u < 1 \\ 0 & \text{otherwise} \end{cases}$$

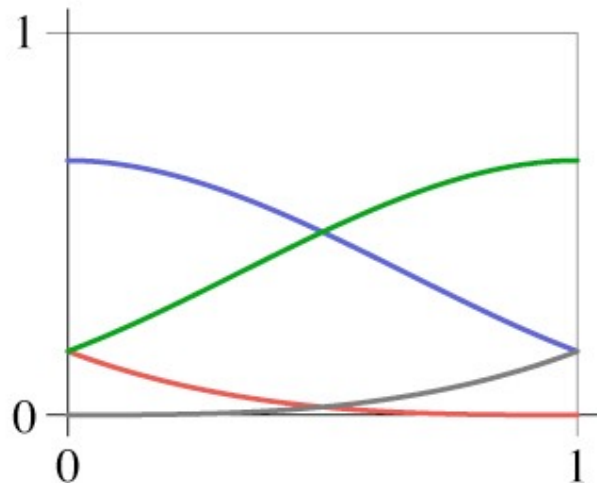
$$b_d = \frac{t}{d-1} b_{d-1}(t) + \frac{d-t}{d-1} b_{d-1}(t-1)$$



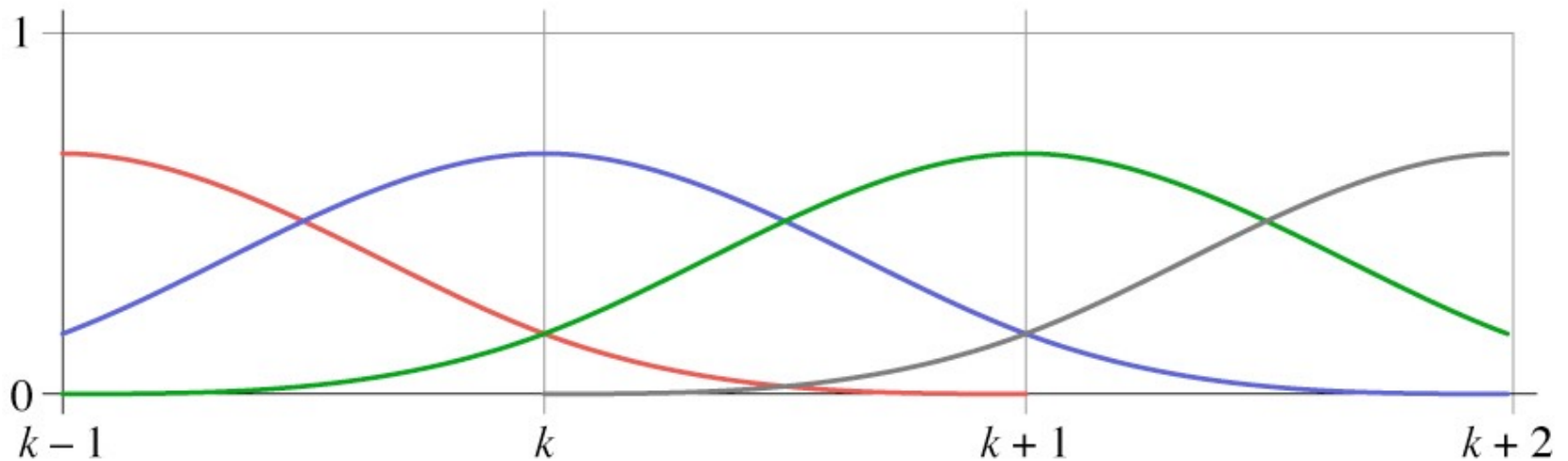
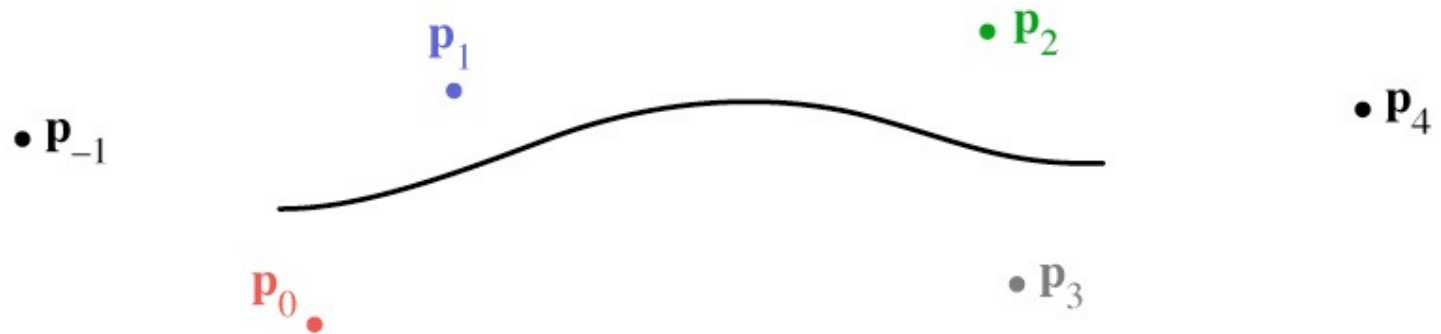
# Cubic B-spline matrix

$$\mathbf{f}_i(t) = \begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \cdot \frac{1}{6} \begin{bmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 0 & 3 & 0 \\ 1 & 4 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_{i-1} \\ \mathbf{p}_i \\ \mathbf{p}_{i+1} \\ \mathbf{p}_{i+2} \end{bmatrix}$$

# Cubic B-spline basis



# Cubic B-spline basis



# **Refinement and Evaluation**

# Converting spline representations

- All the splines we have seen so far are equivalent
  - all represented by geometry matrices

$$\mathbf{p}_S(t) = T(t)M_S P_S$$

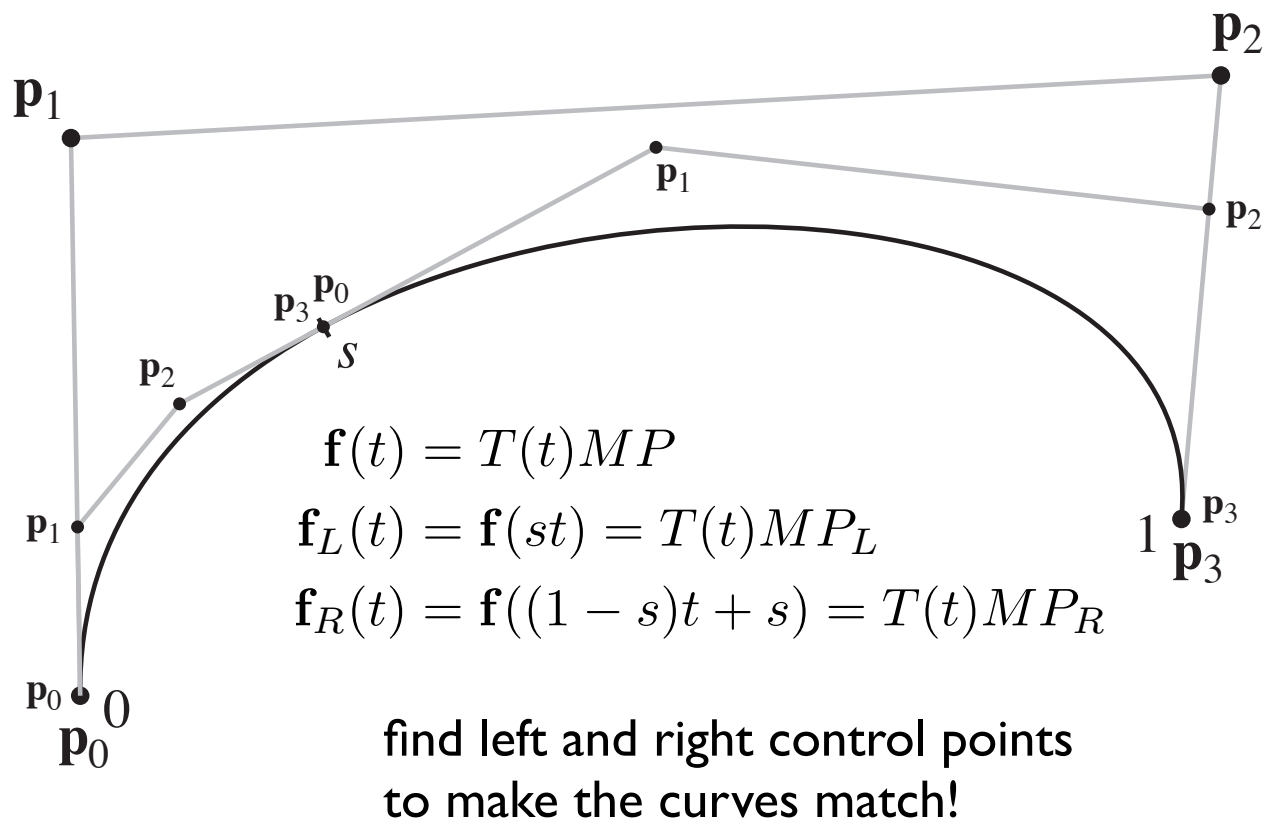
- where  $S$  represents the type of spline
  - therefore the control points may be transformed from one type to another using matrix multiplication

$$P_1 = M_1^{-1} M_2 P_2$$

$$\begin{aligned}\mathbf{p}_1(t) &= T(t)M_1(M_1^{-1}M_2P_2) \\ &= T(t)M_2P_2 = \mathbf{p}_2(t)\end{aligned}$$

# Refinement of splines

- May want to add more control to a curve
- Can add control by splitting a segment into two



# Refinement math

$$\begin{aligned}\mathbf{f}_L(t) &= T(st)MP = T(t)S_LMP \\ &= T(t)M(M^{-1}S_LMP) \\ &= T(t)MP_L\end{aligned}$$

$$P_L = M^{-1}S_LMP$$

$$P_R = M^{-1}S_RMP$$

$$S_L = \begin{bmatrix} s^3 & & & \\ & s^2 & & \\ & & s & \\ & & & 1 \end{bmatrix}$$
$$S_R = \begin{bmatrix} s^3 & & & \\ 3s^2(1-s) & s^2 & & \\ 3s(1-s)^2 & 2s(1-s) & s & \\ (1-s)^3 & (1-s)^2 & (1-s) & 1 \end{bmatrix}$$

# Other types of B-splines

- Nonuniform B-splines
  - discontinuities not evenly spaced
  - allows control over continuity or interpolation at certain points
  - e.g. interpolate endpoints (commonly used case)
- Nonuniform Rational B-splines (NURBS)
  - ratios of nonuniform B-splines:  $x(t) / w(t)$ ;  $y(t) / w(t)$
  - key properties:
    - invariance under perspective as well as affine
    - ability to represent conic sections exactly

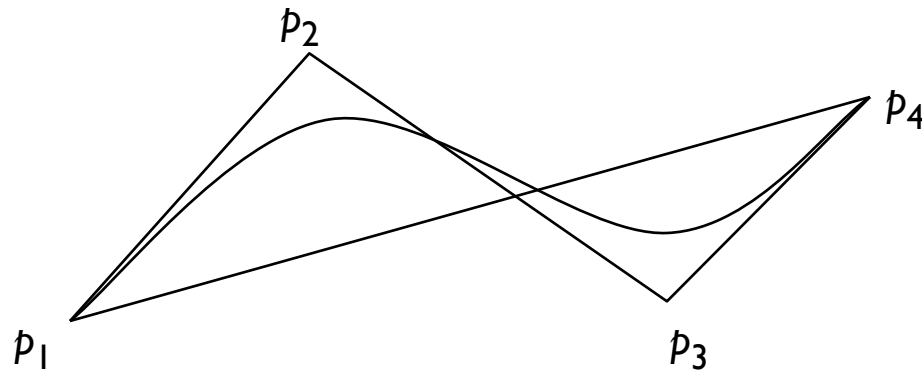
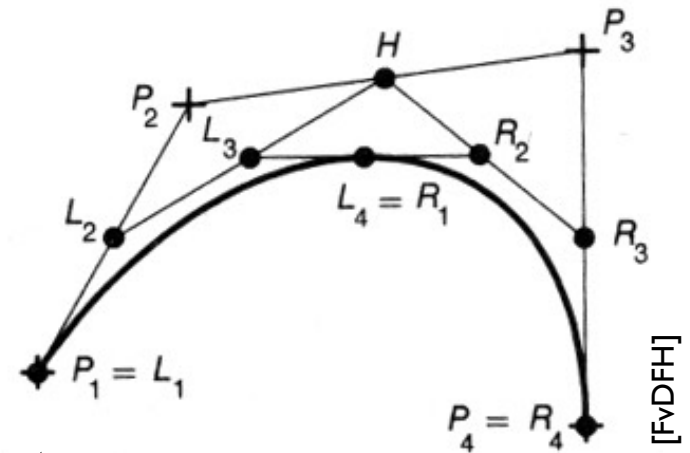


# Evaluating splines for display

- Need to generate a list of line segments to draw
  - generate efficiently
  - use as few as possible
  - guarantee approximation accuracy
- Approaches
  - recursive subdivision (easy to do adaptively)
  - uniform sampling (easy to do efficiently)

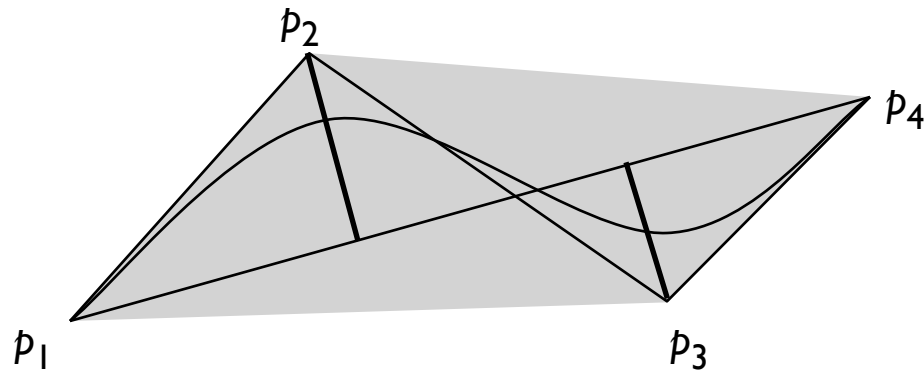
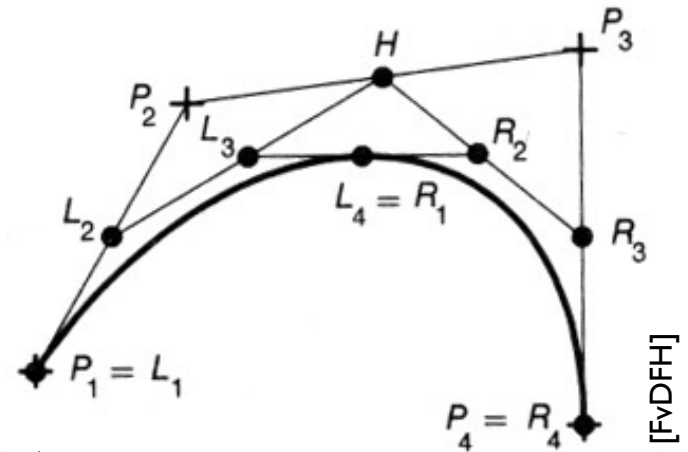
# Evaluating by subdivision

- Recursively split spline
  - stop when polygon is within epsilon of curve
- Termination criteria
  - distance between control points
  - distance of control points from line
  - angles in control polygon



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# Evaluating with uniform spacing

- Forward differencing
  - efficiently generate points for uniformly spaced  $t$  values
  - evaluate polynomials using repeated differences

# Surfaces built from curves

- Parametric spline surfaces
  - extrusions
  - surfaces of revolution
  - generalized cylinders
  - spline patches
- Pause for differential geometry primer...
  - plane and space curves, tangent vectors
  - parametric surfaces, isolines, tangent vectors, normals

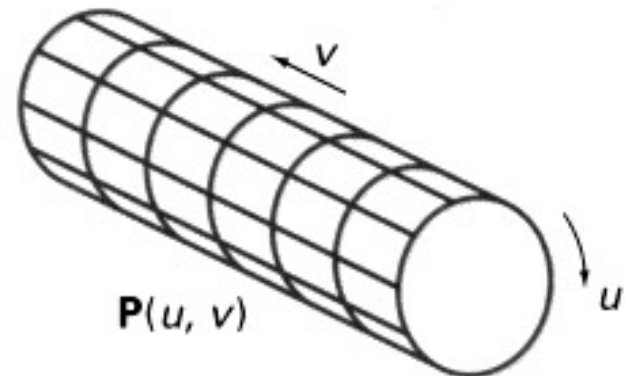
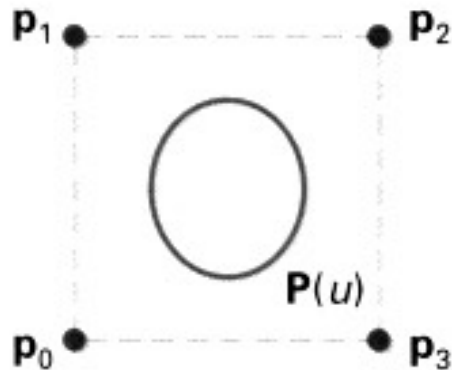
# From curves to surfaces

- So far have discussed spline curves in 2D
  - it turns out that this already provides of the mathematical machinery for several ways of building curved surfaces
- Building surfaces from 2D curves
  - extrusions and surfaces of revolution
- Building surfaces from 2D and 3D curves
  - generalized swept surfaces
- Building surfaces from spline patches
  - generalizing spline curves to spline patches
- Also to think about: generating triangles

# Extrusions

- Given a spline curve  $C \in \mathbb{R}^2$ , define  $S \in \mathbb{R}^3$  by
$$S = C \times [a, b]$$
- This produces a “tube” with the given cross section
  - Circle: cylinder; “L”: shelf bracket; “I”: I beam
- It is parameterized by the spline  $t$  and the interval  $[a, b]$

$$s(t, s) = [c_x(t), c_y(t), s]^T$$

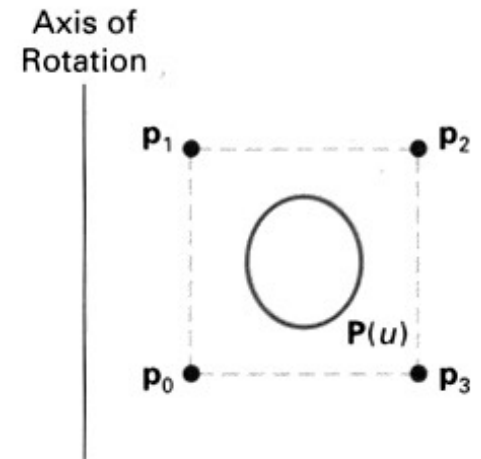


# Surfaces of revolution

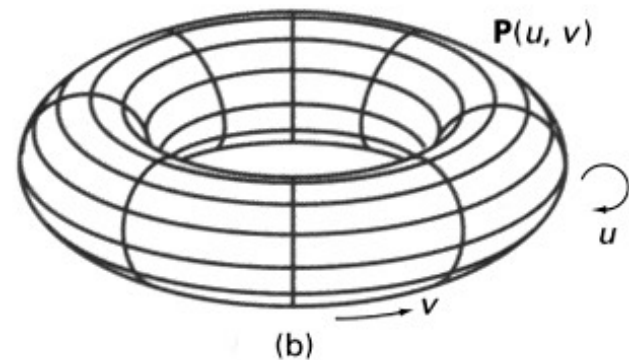
- Take a 2D curve and spin it around an axis
- Given curve  $\mathbf{c}(t)$  in the plane, the surface is defined easily in cylindrical coordinates:

$$\mathbf{s}(t, s) = (r, \phi, z) = (c_x(t), s, c_y(t))$$

- the torus is an example in which the curve  $\mathbf{c}$  is a circle



(a)

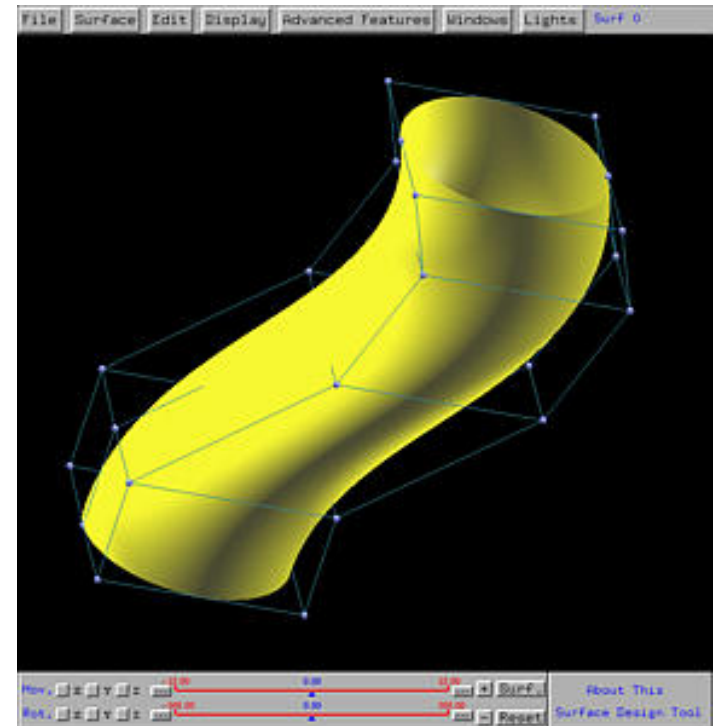
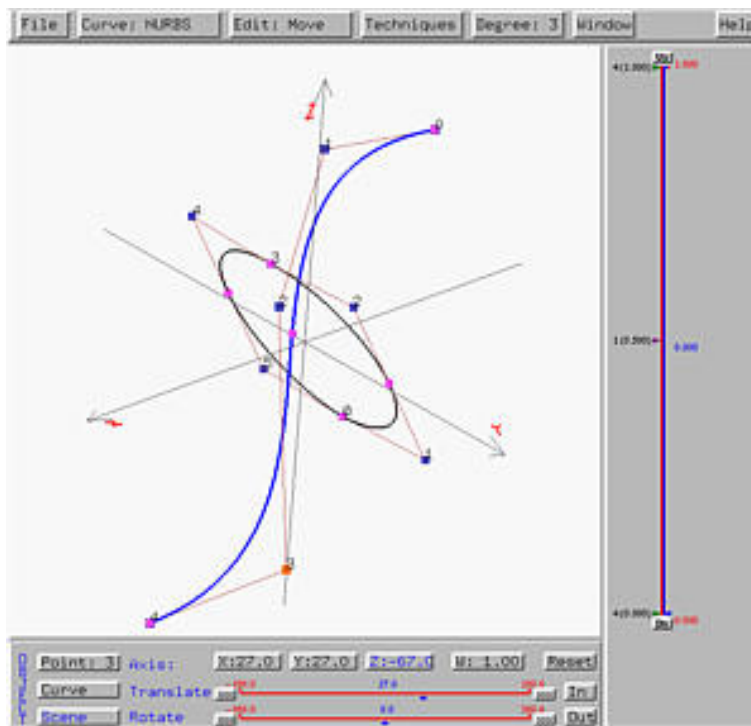


(b)



# Swept surfaces

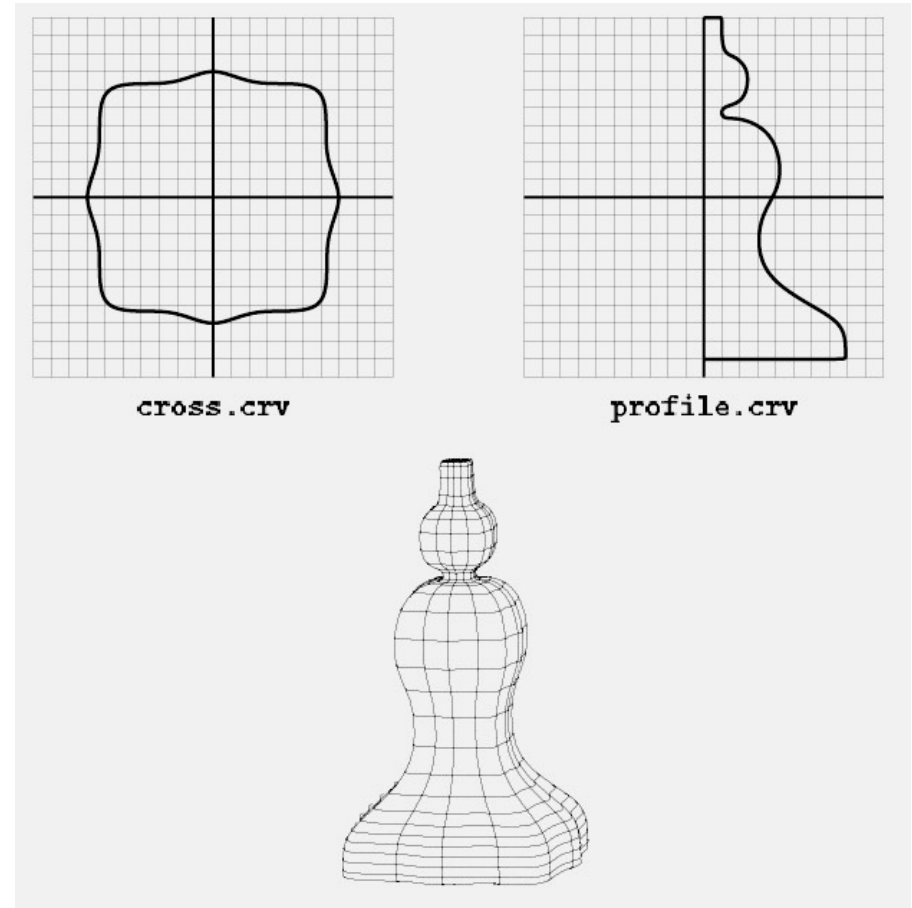
- Surface defined by a *cross section* moving along a *spine*
- Simple version: a single 3D curve for spine and a single 2D curve for the cross section



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# Generalized cylinders

- General swept surfaces
  - varying radius
  - varying cross-section
  - curved axis



[Snyder 1992]