

2D Spline Curves

CS 4620 Lecture 13

Motivation: smoothness

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

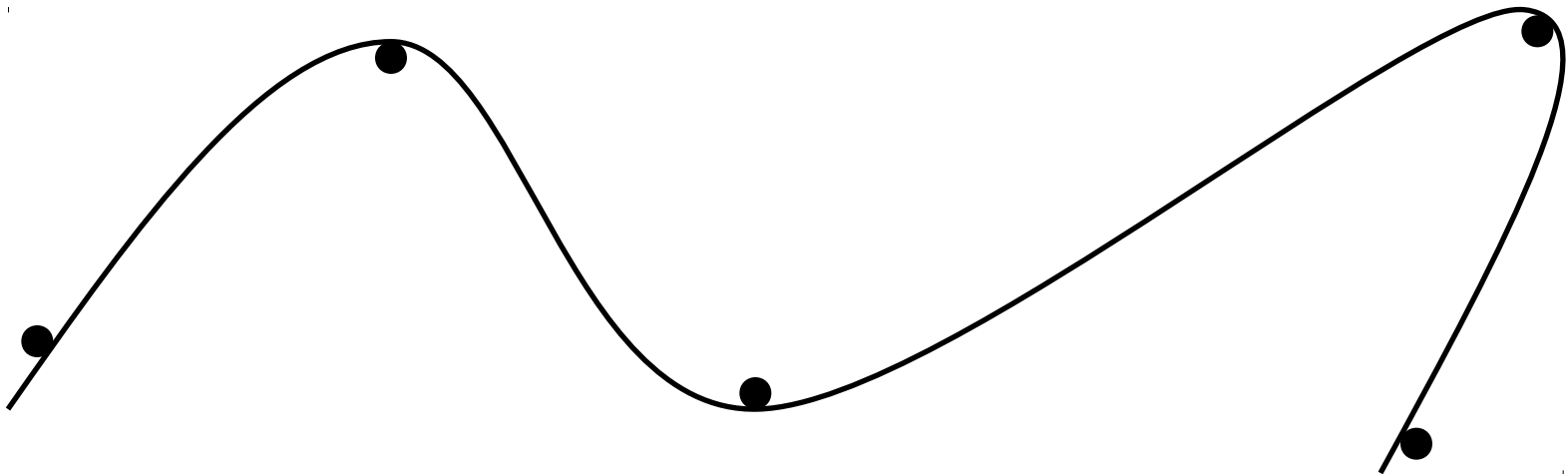


[Boeing]

- So far we can make
 - things with corners (lines, squares, rectangles, ...)
 - circles and ellipses (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



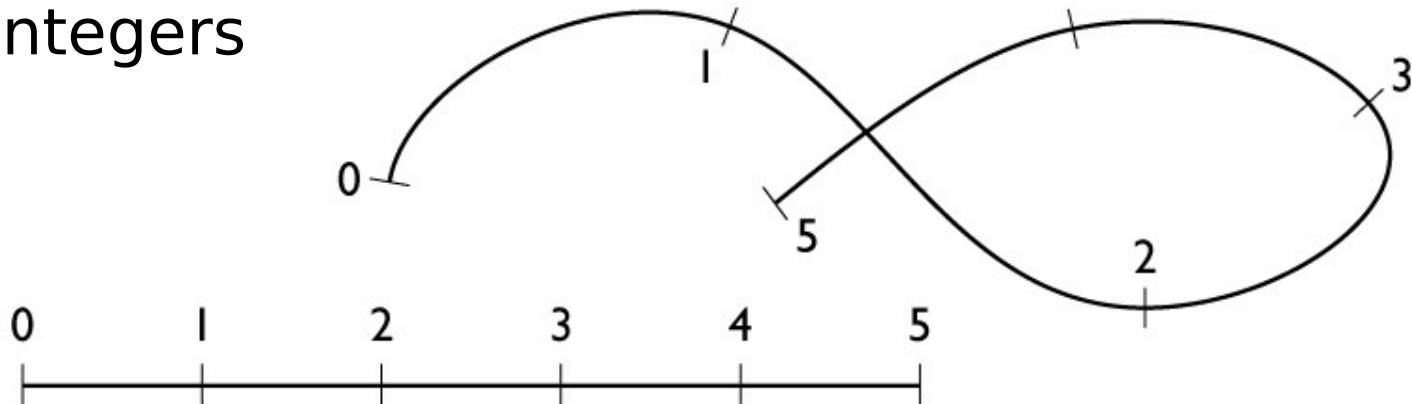
Translating into usable math

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

- Generally $f(t)$ is a piecewise polynomial
 - for this lecture, the discontinuities are at the integers



Defining spline curves

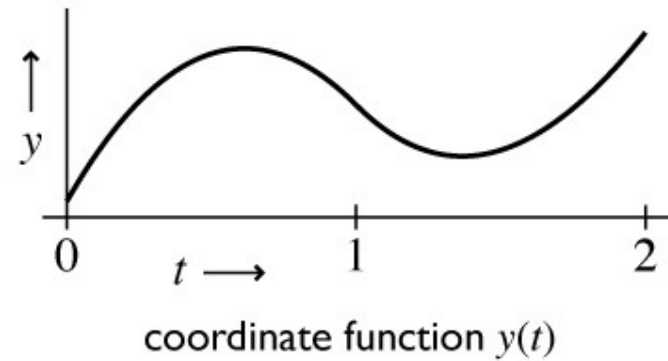
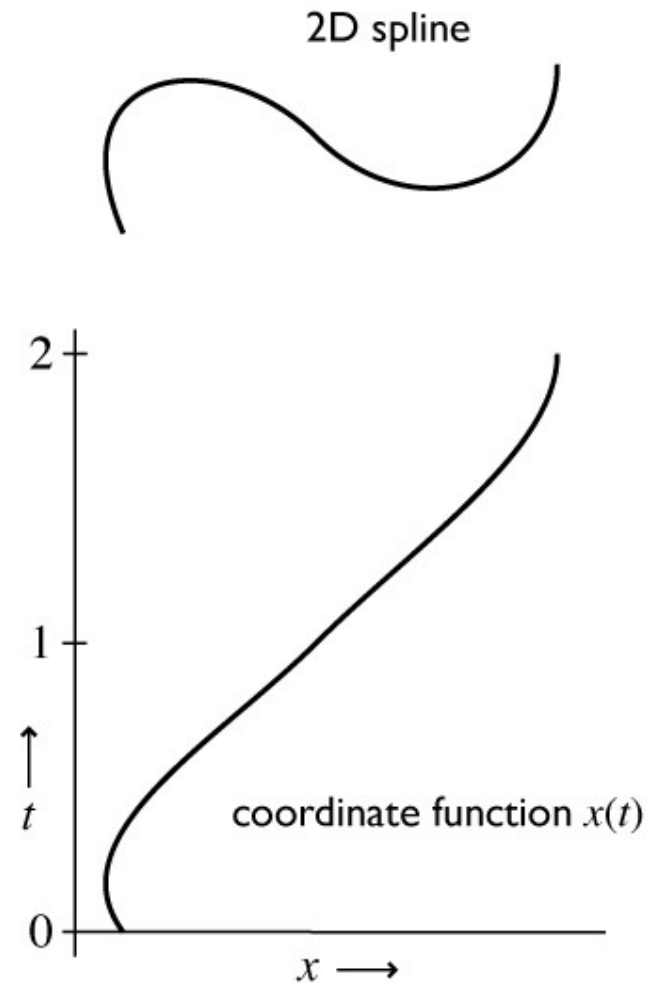
- Generally $f(t)$ is a piecewise polynomial
 - for this lecture, the discontinuities are at the integers
 - e.g., a cubic spline has the following form over $[k, k + 1]$:

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

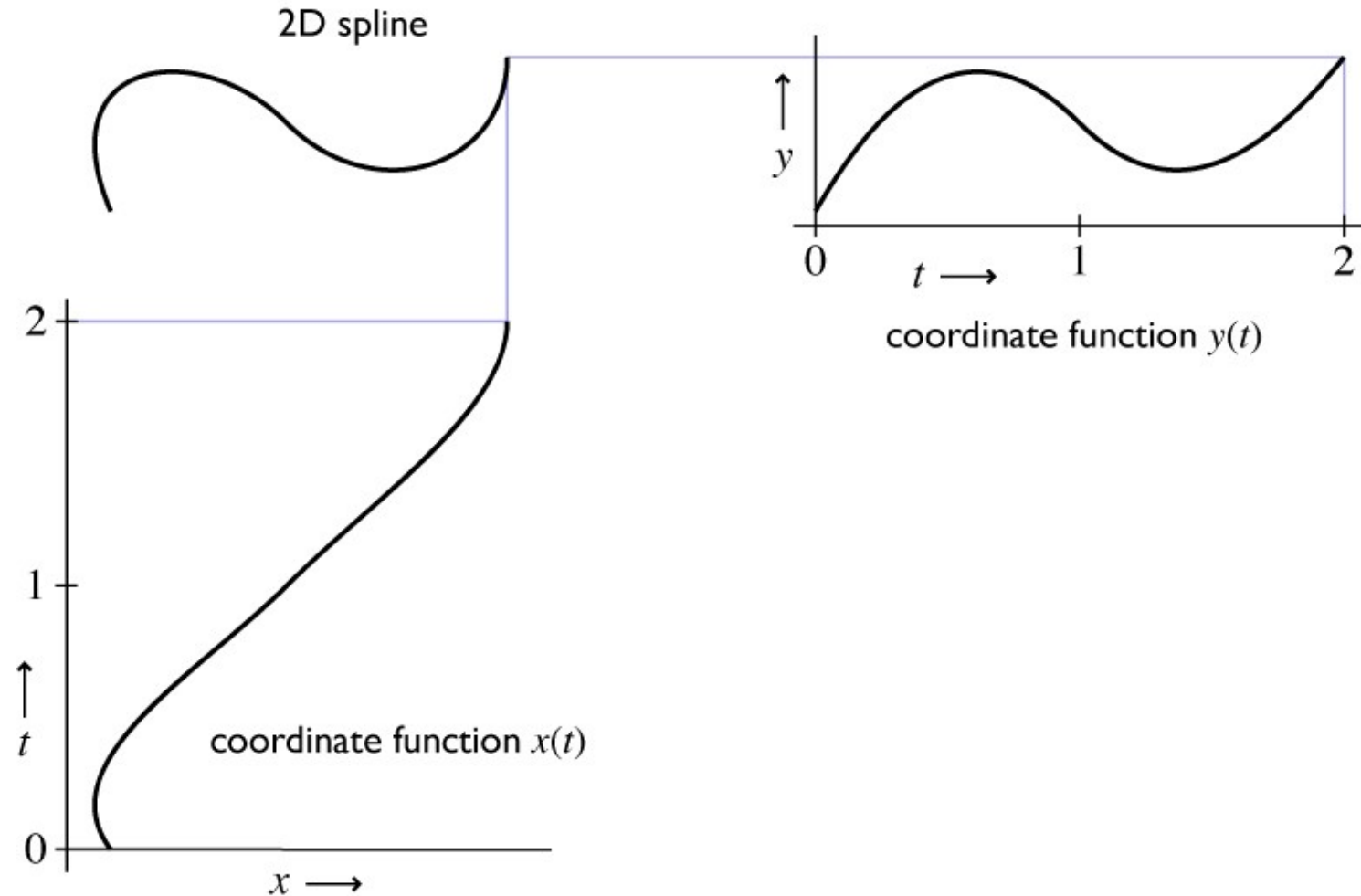
$$y(t) = a_y t^3 + b_y t^2 + c_y t + d_y$$

- Coefficients are different for every interval

Coordinate functions

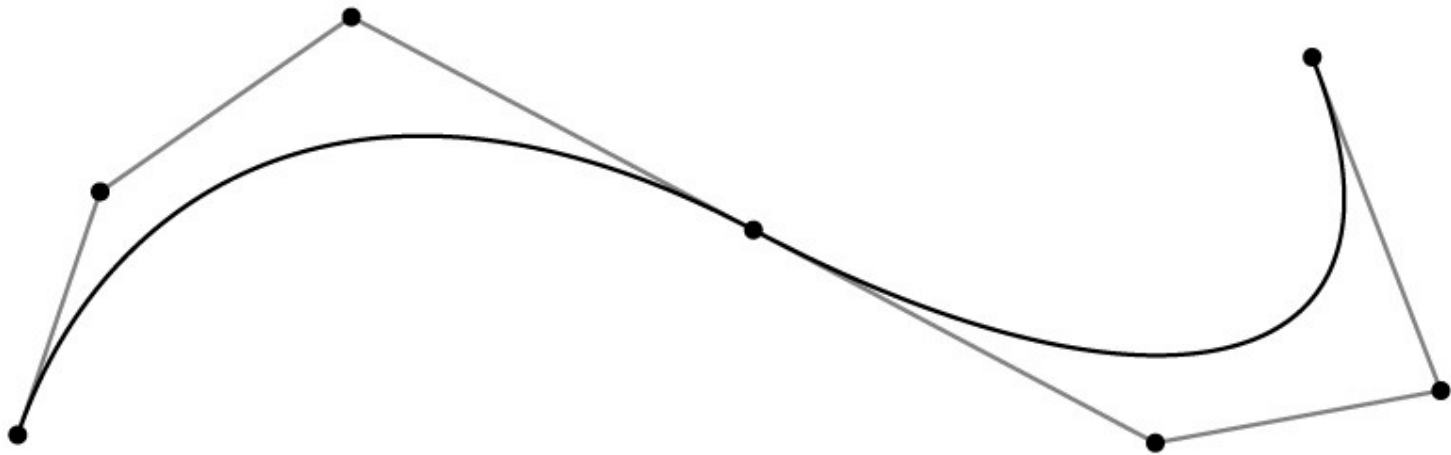


Coordinate functions



Control of spline curves

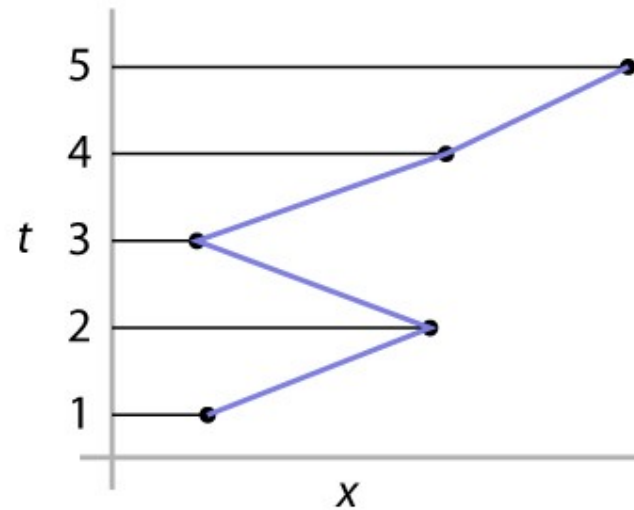
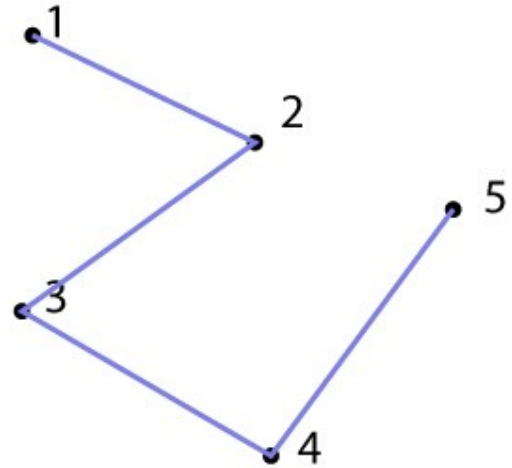
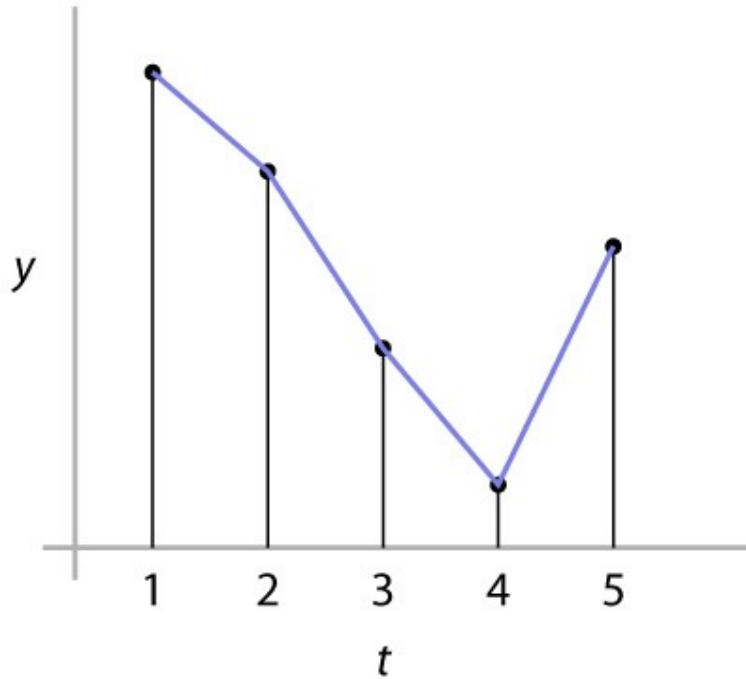
- Specified by a sequence of control points
- Shape is guided by control points (aka control polygon)
 - interpolating: passes through points
 - approximating: merely guided by points



How splines depend on their controls

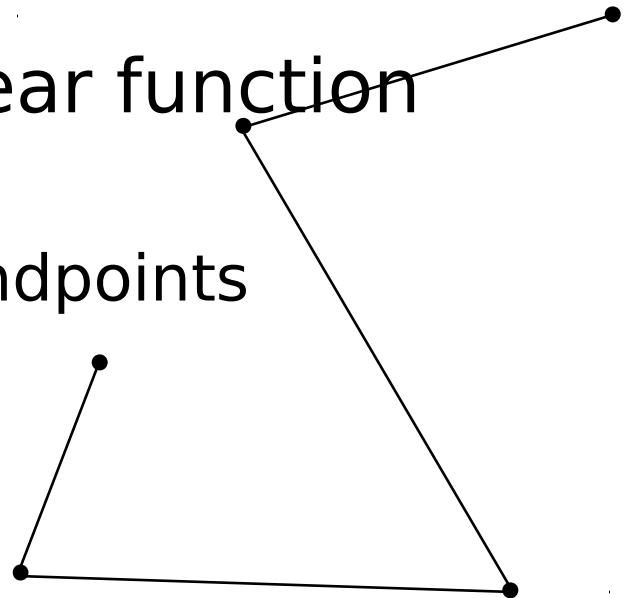
- Each coordinate is separate
 - the function $x(t)$ is determined solely by the x coordinates of the control points
 - this means 1D, 2D, 3D, ... curves are all really the same
- Spline curves are **linear** functions of their controls
 - moving a control point two inches to the right moves $x(t)$ twice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

Splines as reconstruction



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

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

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

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

- Matrix formulation

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

Trivial example: piecewise linear

- Basis function formulation
 - regroup expression by \mathbf{p} rather than t

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

- interpretation in matrix viewpoint

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

Trivial example: piecewise linear

- Basis function formulation
 - regroup expression by \mathbf{p} rather than t

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

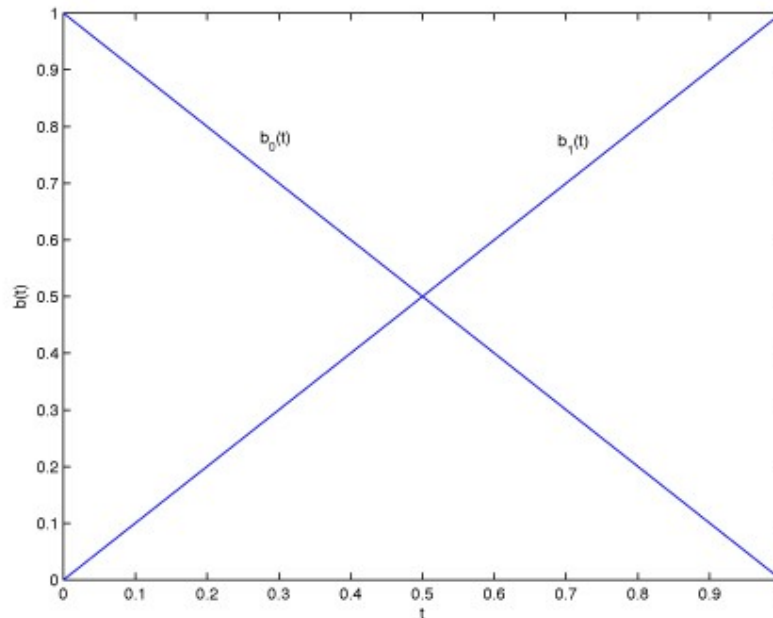
$$\mathbf{p}(t) = \begin{bmatrix} x(t) & y(t) \end{bmatrix} \quad \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \end{bmatrix} = \begin{bmatrix} x_0 & y_0 \\ x_1 & y_1 \end{bmatrix}$$

- interpretation in matrix viewpoint

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

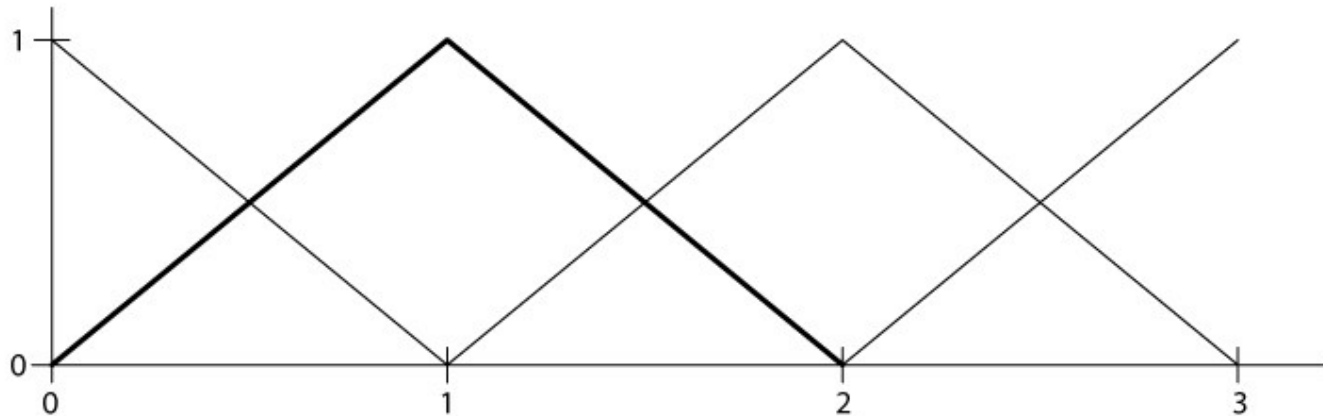
Trivial example: piecewise linear

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



Trivial example: piecewise linear

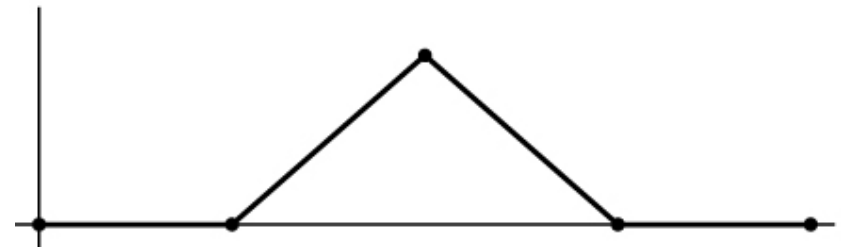
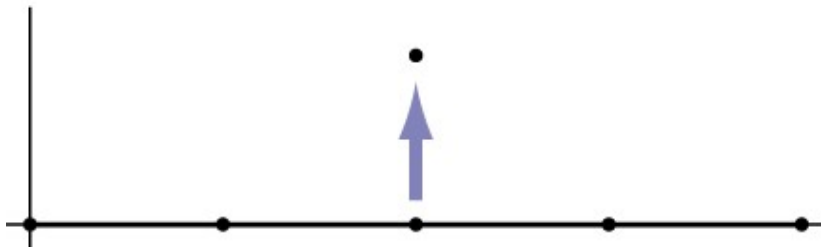
- Basis function formulation: “function times point”
 - basis functions: contribution of each point as t changes



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

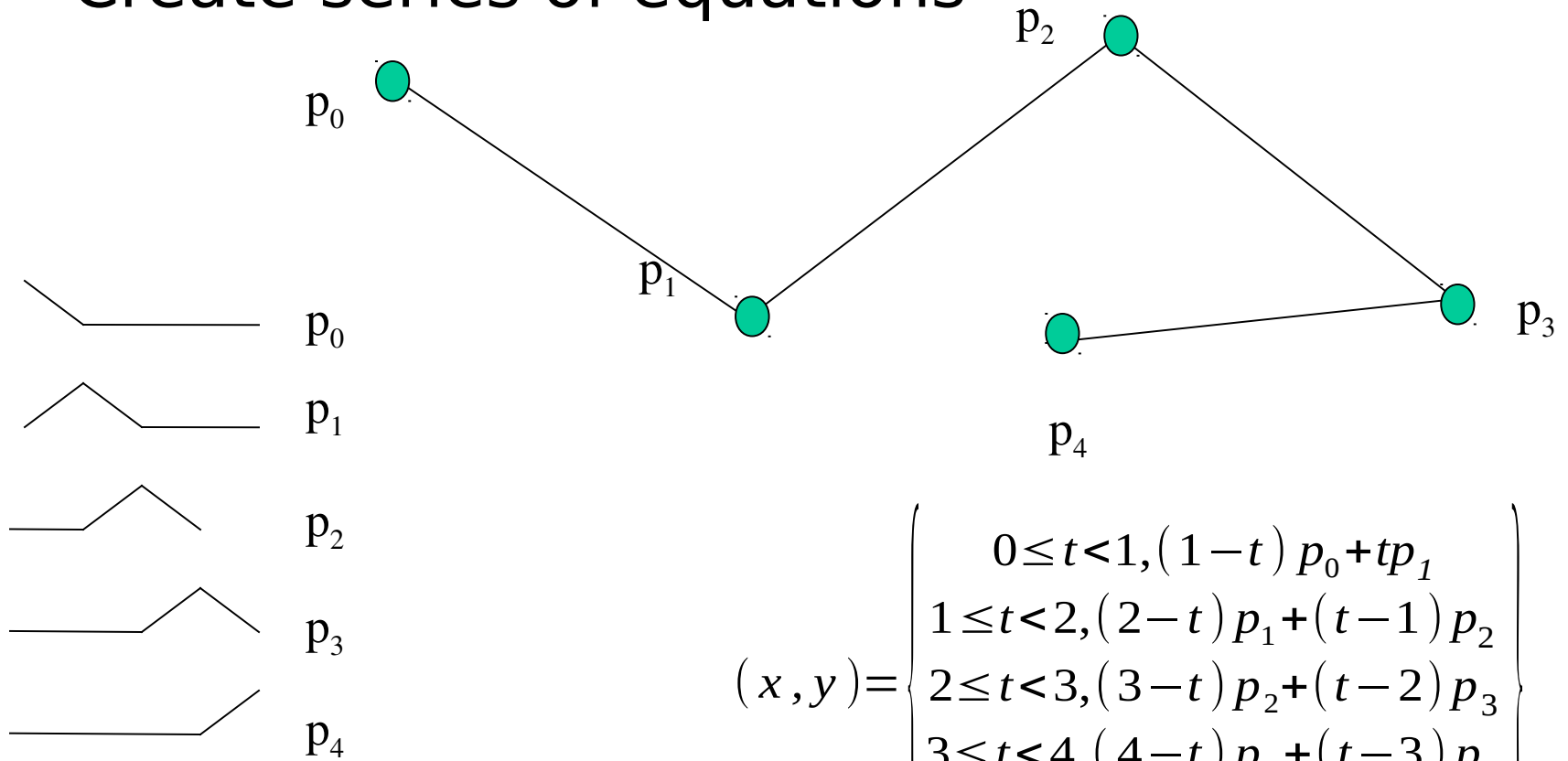
Seeing the basis functions

- 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



Piece these together

Create series of equations

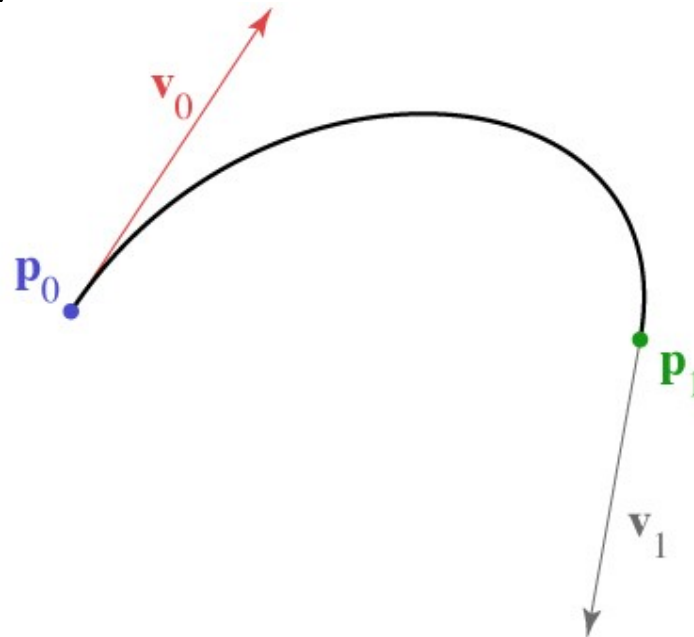


Blending Functions

$$(x, y) = \left\{ \begin{array}{l} 0 \leq t < 1, (1-t)p_0 + tp_1 \\ 1 \leq t < 2, (2-t)p_1 + (t-1)p_2 \\ 2 \leq t < 3, (3-t)p_2 + (t-2)p_3 \\ 3 \leq t < 4, (4-t)p_3 + (t-3)p_4 \\ 4 \leq t < 5, (5-t)p_4 + (t-4)p_5 \end{array} \right.$$

Hermite splines

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



Hermite splines

- Solve constraints to find coefficients

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

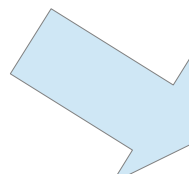
$$x'(t) = 3at^2 + 2bt + c$$

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

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

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

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


$$\begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} x_0 \\ x_1 \\ x'_0 \\ x'_1 \end{bmatrix}$$

Hermite splines

- Solve constraints to find coefficients

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

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

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

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

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

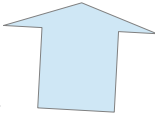
$$x'(1) = x'_1 = 3a + 2b + c$$

$$d = x_0$$

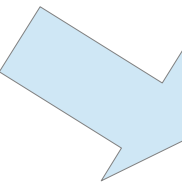
$$c = x'_0$$

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

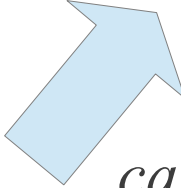
$$b = -3x_0 + 3x_1 - 2x'_0 - x'_1$$



$$\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \\ x'_0 \\ x'_1 \end{bmatrix}$$



$$\begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} x_0 \\ x_1 \\ x'_0 \\ x'_1 \end{bmatrix}$$



calculate A^{-1}

Hermite splines

- Matrix form is much simpler

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

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

Hermite splines

- Matrix form is much simpler

$$\begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{v}_0 \\ \mathbf{v}_1 \end{bmatrix} = \begin{bmatrix} x_0 & y_0 \\ x_1 & y_1 \\ x'_0 & y'_0 \\ x'_1 & y'_1 \end{bmatrix}$$

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

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

A^{-1}

Coefficients = rows

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

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

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

Basis functions = columns

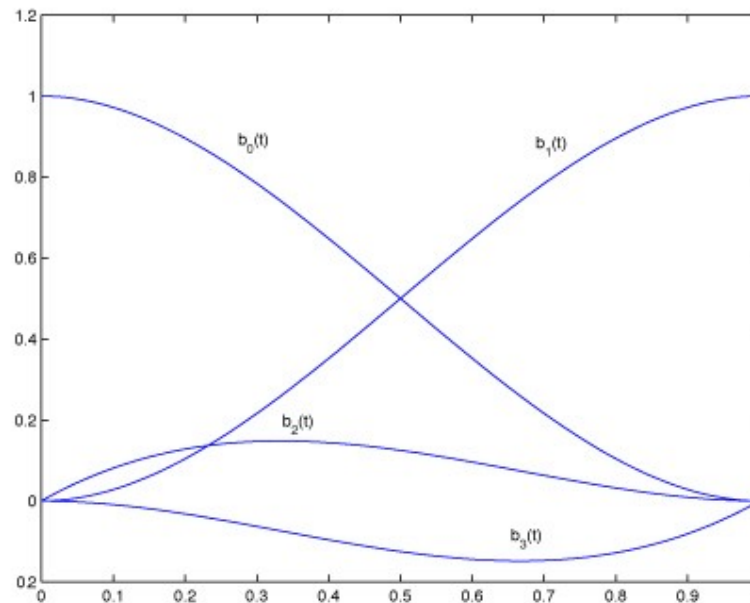
$$\mathbf{p}(t) = at^3 + bt^2 + ct + d$$

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

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

Hermite splines

- Hermite blending functions

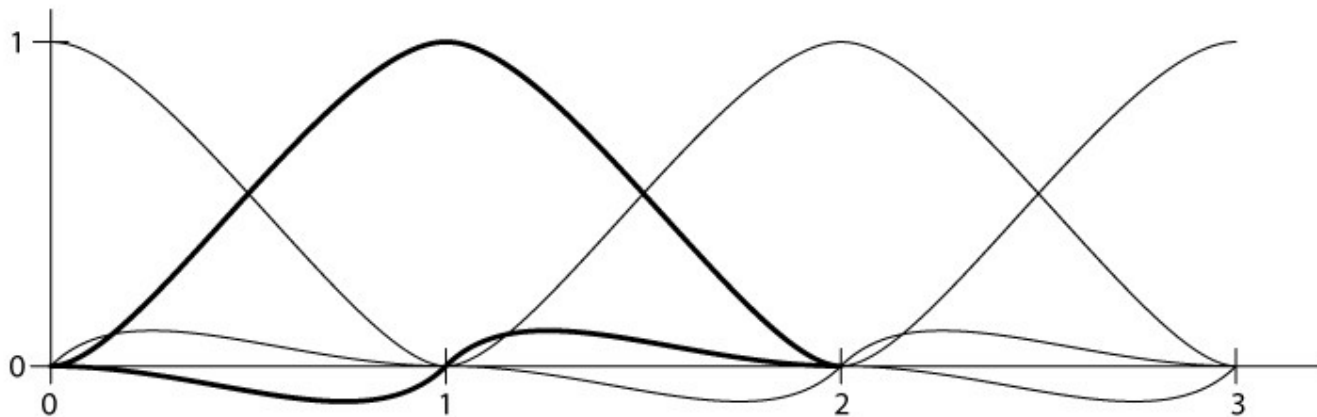


Longer Hermite splines

- Can only do so much with one Hermite spline
- Can use these splines 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
- To avoid discontinuity, match derivatives at junctions
 - this produces a C^1 curve

Hermite splines

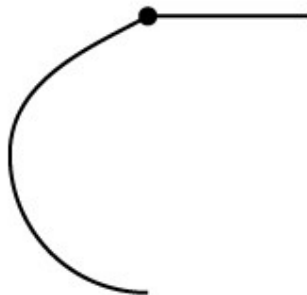
- Hermite basis functions



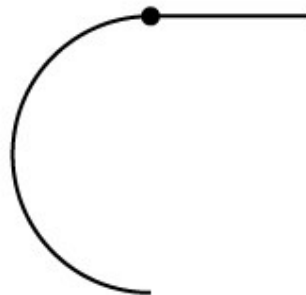
Continuity

- 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

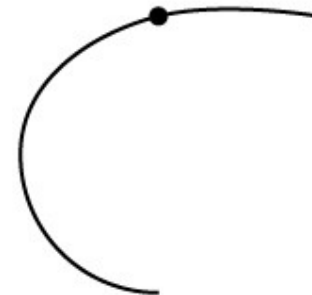
zero order



first order



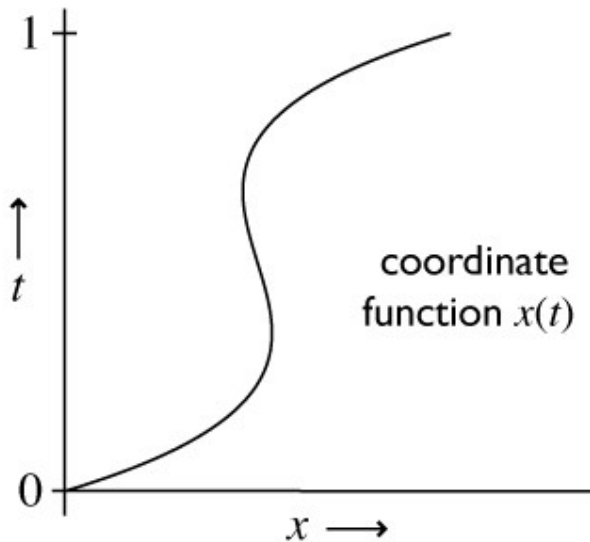
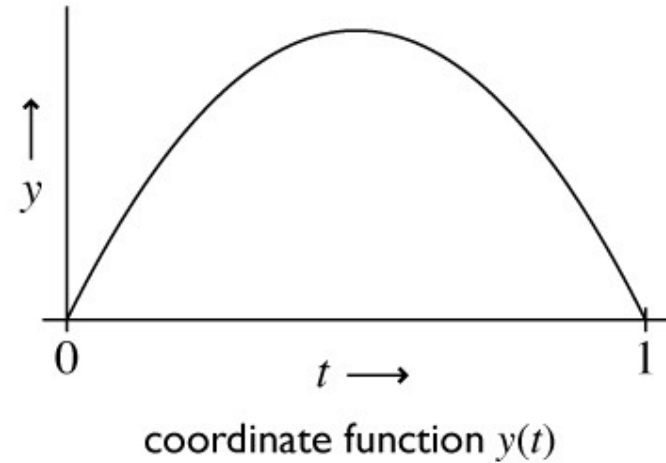
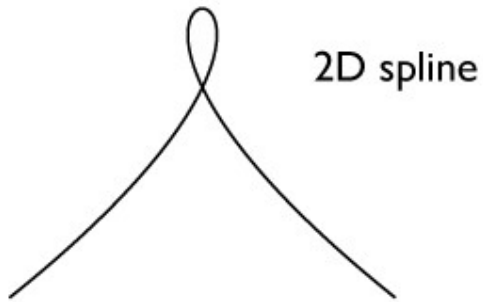
second order



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



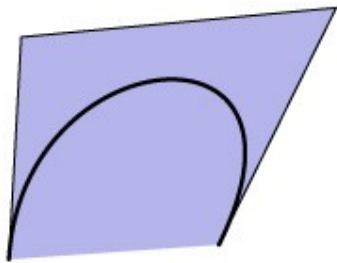
Control

- 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

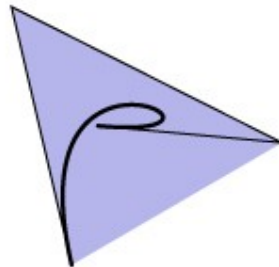


Control

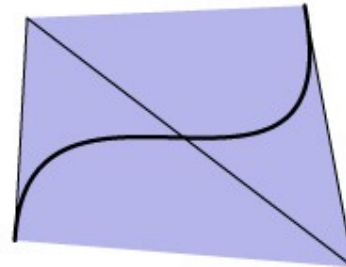
- Convex hull property
 - convex hull = smallest convex region containing points
 - think of a rubber band around some pins
 - some splines stay inside convex hull of control points



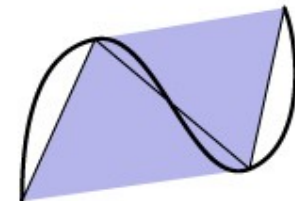
YES



YES



YES

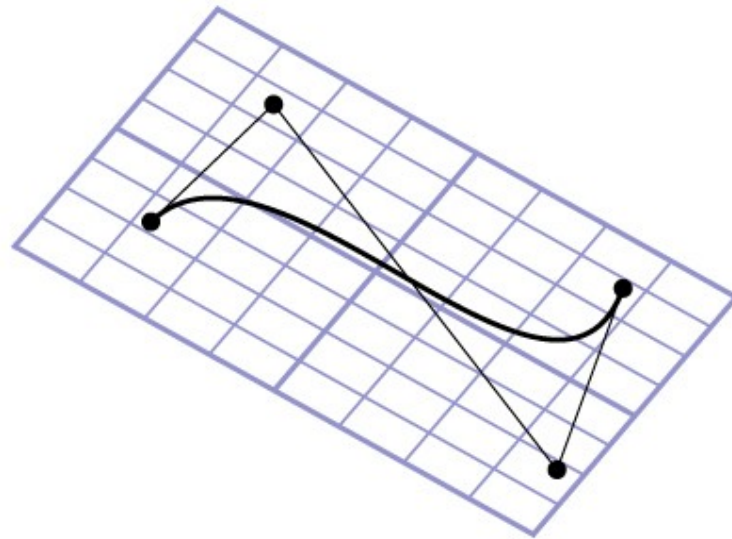
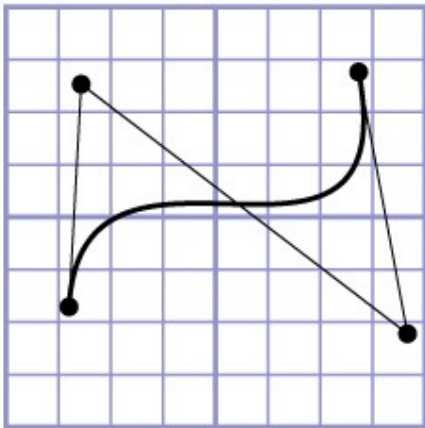


NO

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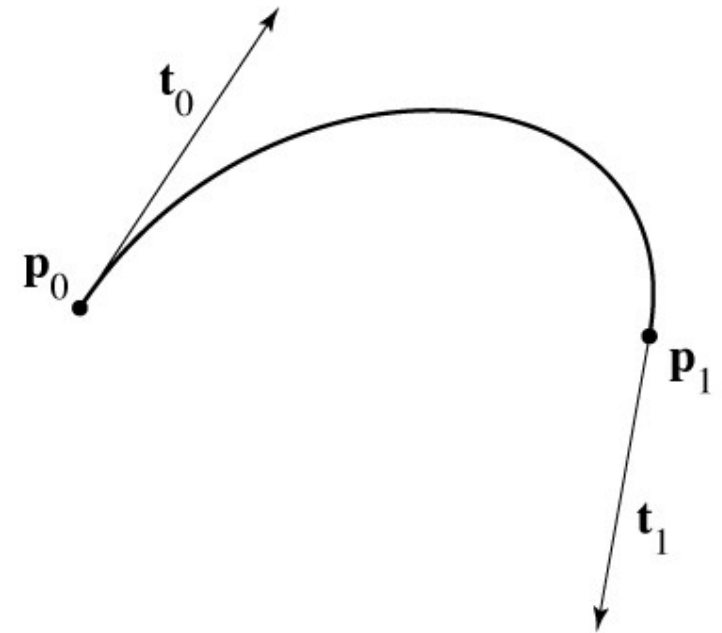
Affine invariance

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



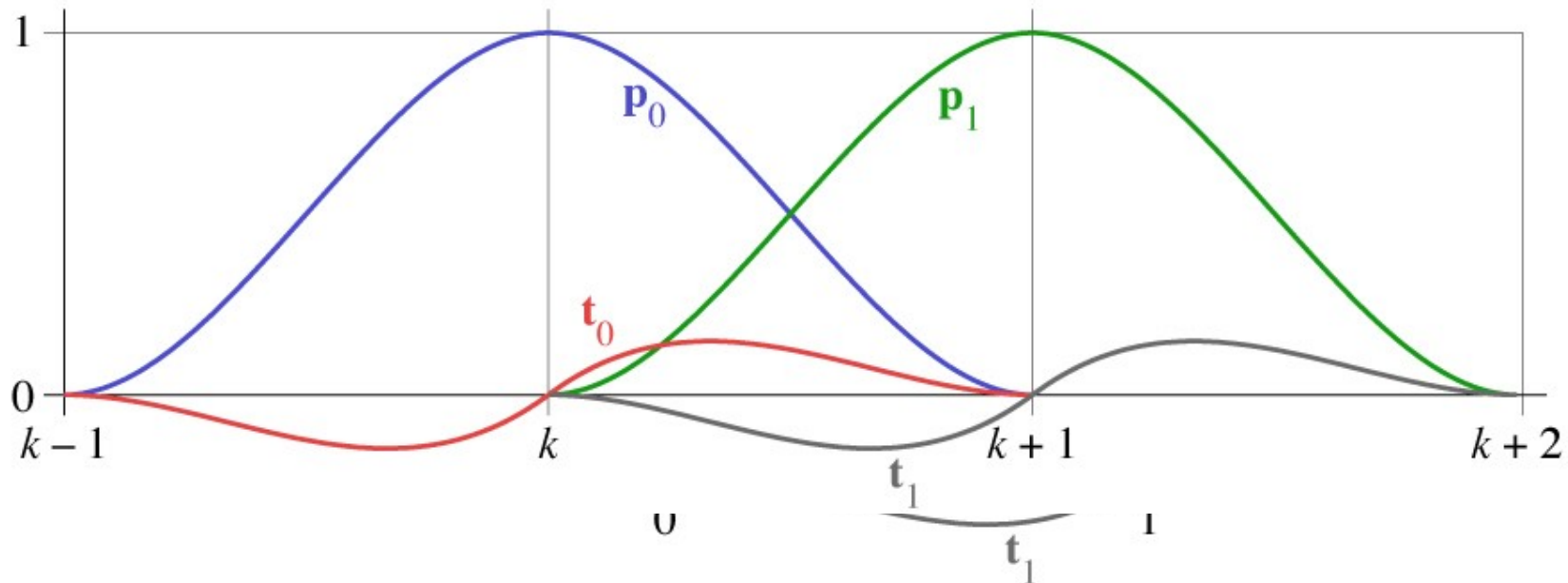
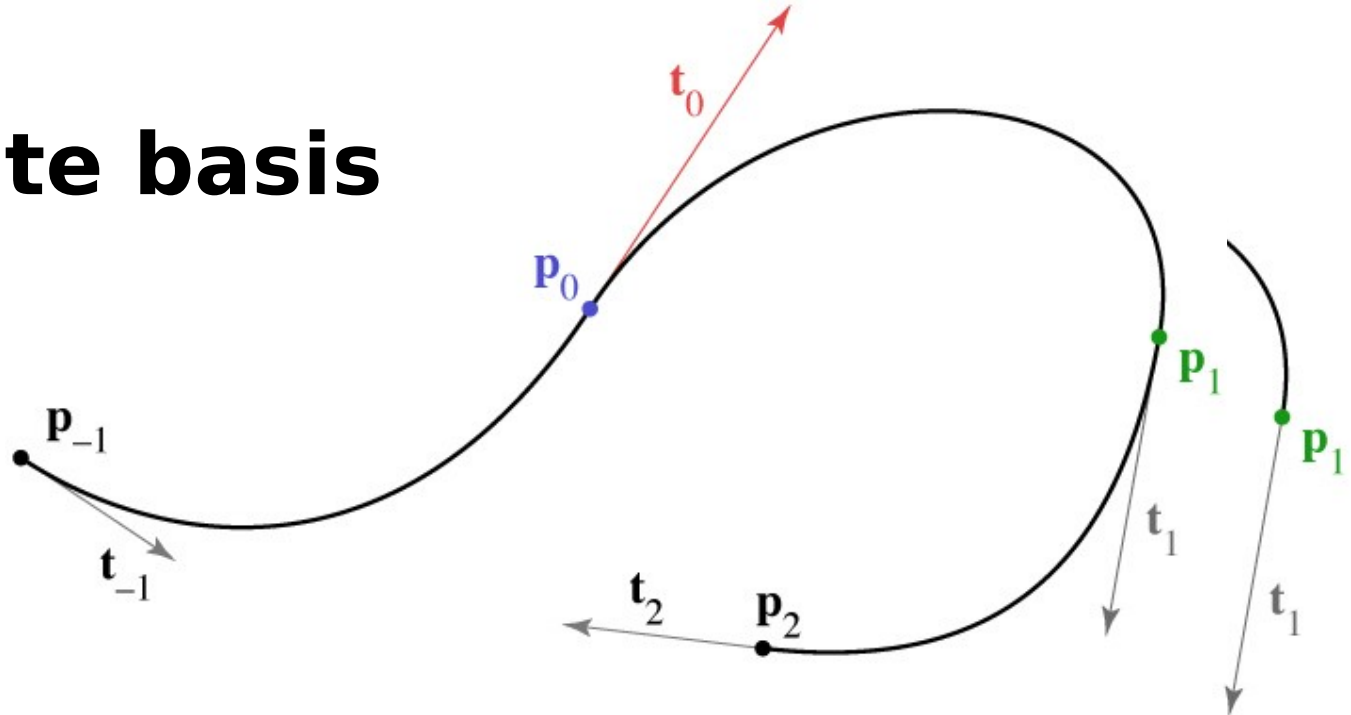
Hermite splines

- Constraints are endpoints and endpoint tangents



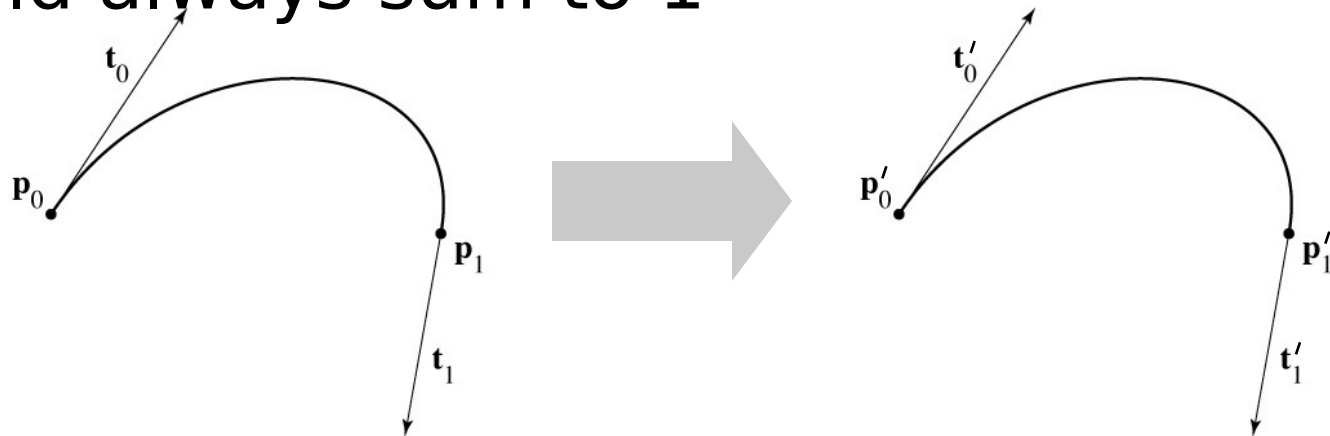
$$\mathbf{p}(t) = \begin{bmatrix} t^3 & t^2 & t & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \\ \mathbf{v}_0 \\ \mathbf{v}_1 \end{bmatrix}$$

Hermite basis



Affine invariance

- Basis functions associated with points should always sum to 1

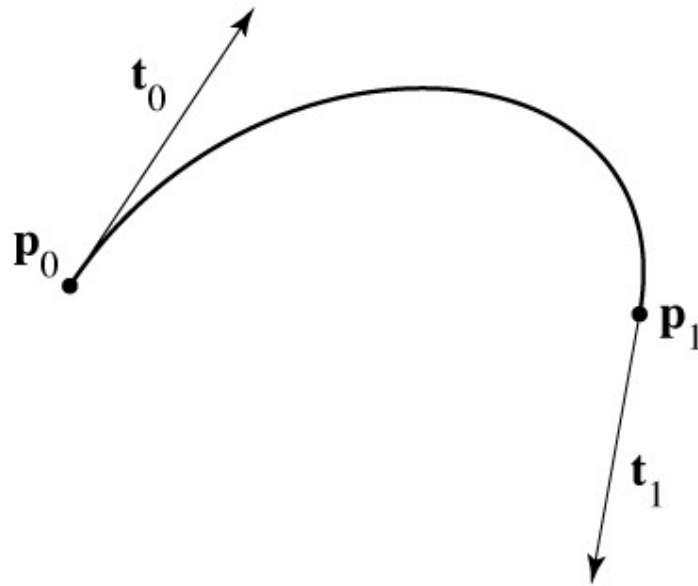


$$\mathbf{p}(t) = b_0\mathbf{p}_0 + b_1\mathbf{p}_1 + b_2\mathbf{v}_0 + b_3\mathbf{v}_1$$

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

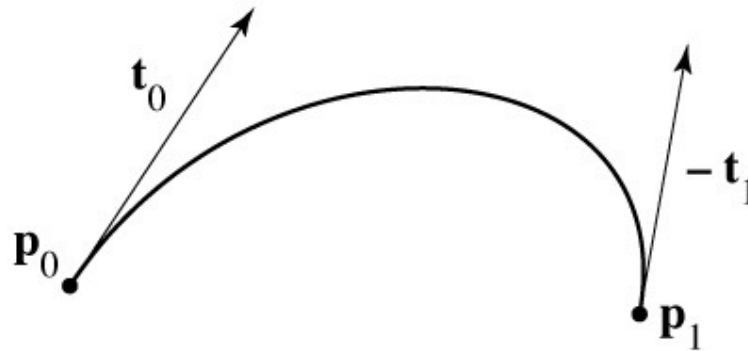
Hermite to Bézier

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



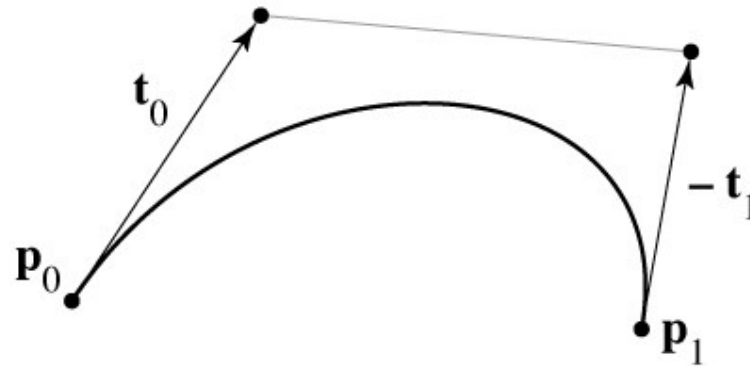
Hermite to Bézier

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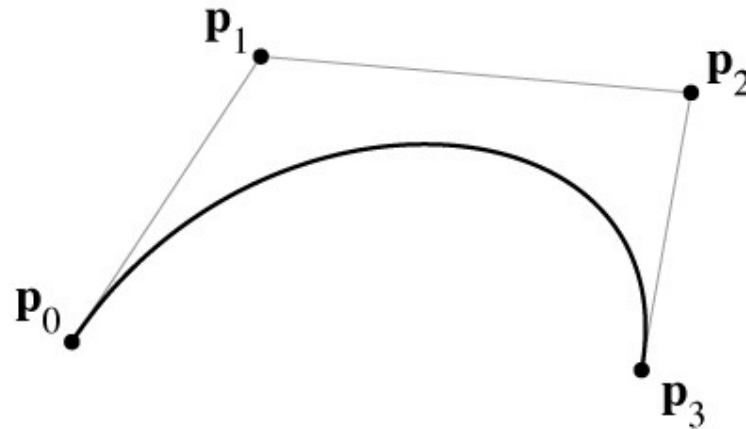
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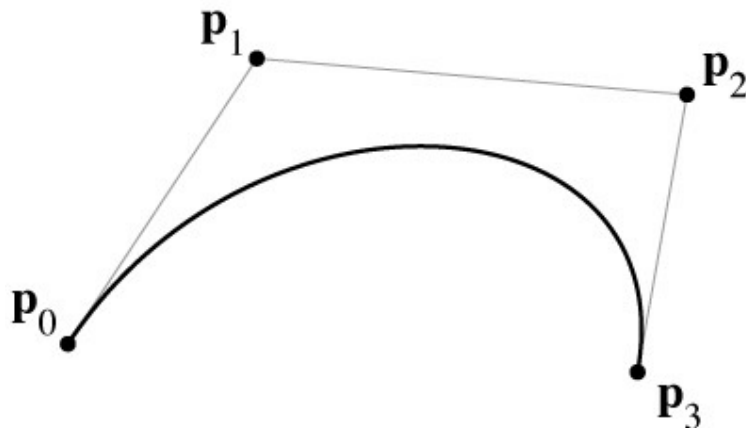
Hermite to Bézier

- Mixture of points and vectors is awkward
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Hermite to Bézier

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



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

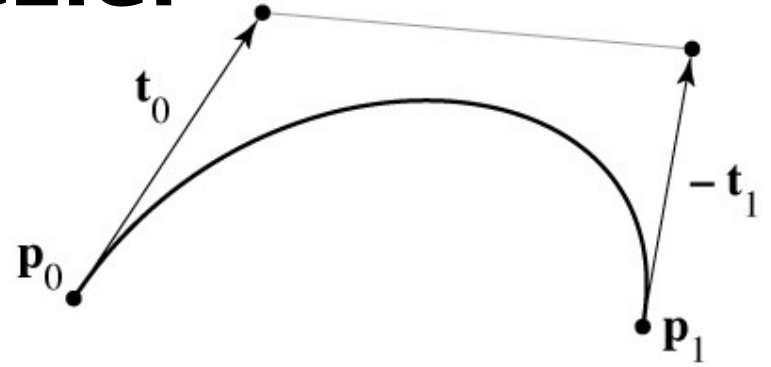
Hermite to Bézier

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

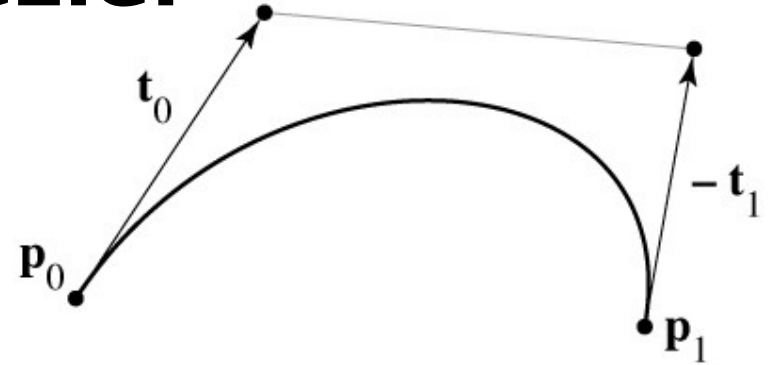
$$\mathbf{p}_1 = \mathbf{q}_3$$

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

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



Hermite to Bézier



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

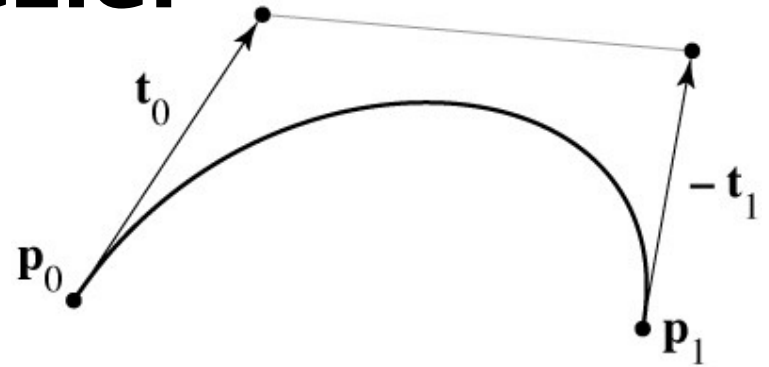
$$\mathbf{p}_1 = \mathbf{q}_3$$

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

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

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

Hermite to Bézier



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

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

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

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

$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -3 & 3 & 0 & 0 \\ 0 & 0 & -3 & 3 \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \mathbf{q}_2 \\ \mathbf{q}_3 \end{bmatrix}$$

Hermite matrix

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

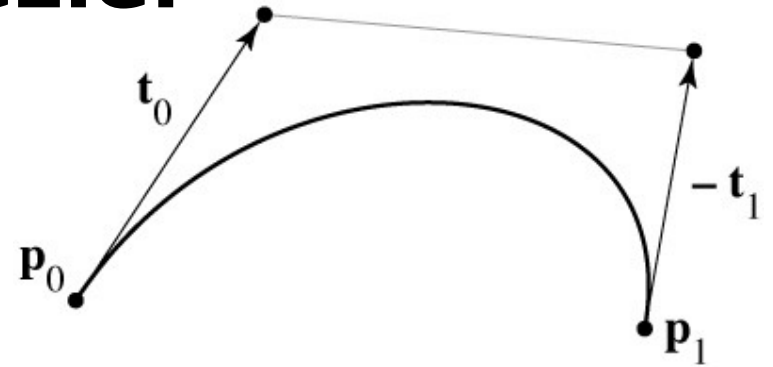
$$\mathbf{p}_0 = \mathbf{q}_0$$

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

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

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

Hermite to Bézier



$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -3 & 3 & 0 & 0 \\ 0 & 0 & -3 & 3 \end{bmatrix} \begin{bmatrix} \mathbf{q}_0 \\ \mathbf{q}_1 \\ \mathbf{q}_2 \\ \mathbf{q}_3 \end{bmatrix}$$

Bézier matrix

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

- note that these are the Bernstein polynomials

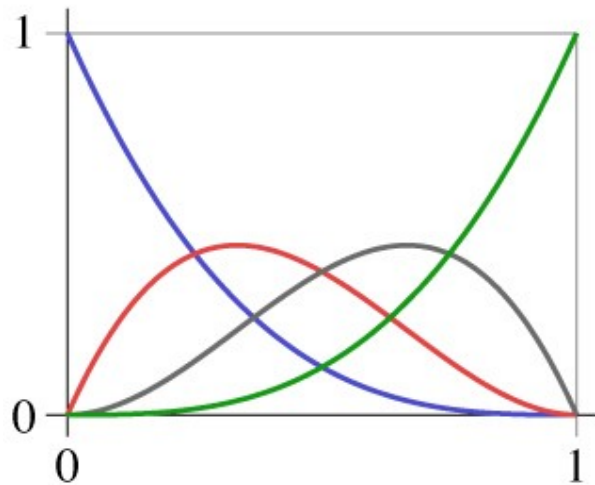
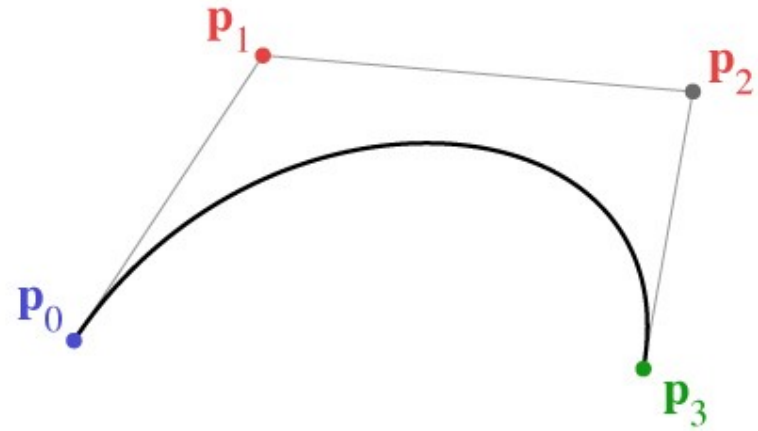
$$C(n,k) t^k (1 - t)^{n - k}$$

and that defines Bézier curves for any degree

Apply Constraint Matrix

$$\begin{bmatrix} 1 & u & u^2 & u^3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ -3 & 3 & 0 & 0 \\ 3 & -6 & 3 & 0 \\ -1 & 3 & -3 & 1 \end{bmatrix} \begin{bmatrix} p_0 \\ p_1 \\ p_2 \\ p_3 \end{bmatrix}$$
$$\begin{bmatrix} (1-u)^3 & 3u(u-1)^2 & 3u^2(u-1) & u^3 \end{bmatrix} \begin{bmatrix} p_0 \\ p_1 \\ p_2 \\ p_3 \end{bmatrix}$$
$$(1-u)^3 p_0 + 3u(1-u)^2 p_1 + 3u^2(1-u) p_2 + u^3 p_3$$

Bézier basis



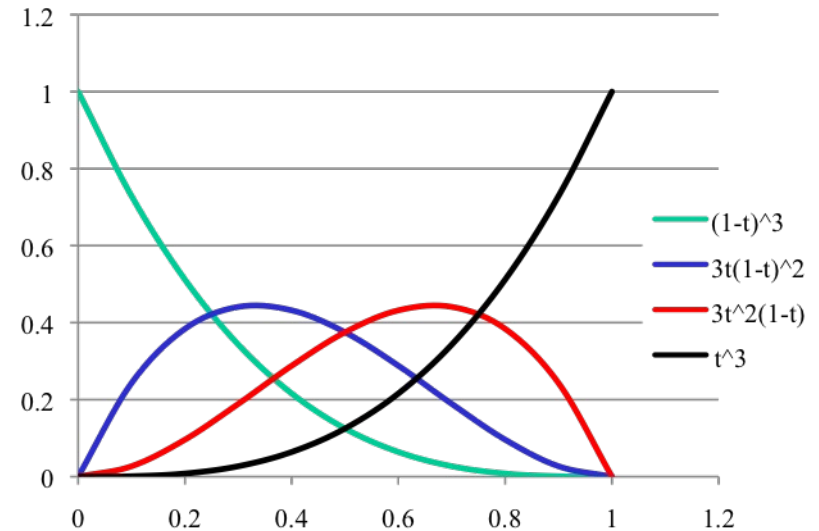
Bezier Polynomials sum to one

$$(1-u) + u = 1$$

$$((1-u) + u)^2 = 1$$

$$((1-u) + u)^3 = 1$$

$$(1-u)^3 + 3u(1-u)^2 + 3u^2(1-u) + u^3 = 1$$

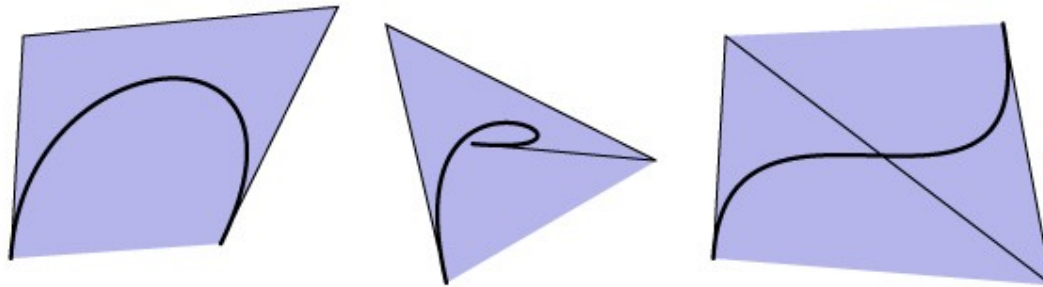


So each point on the curve is a convex sum of the control points

Thus the curve lies inside the convex hull of the control points

Convex hull

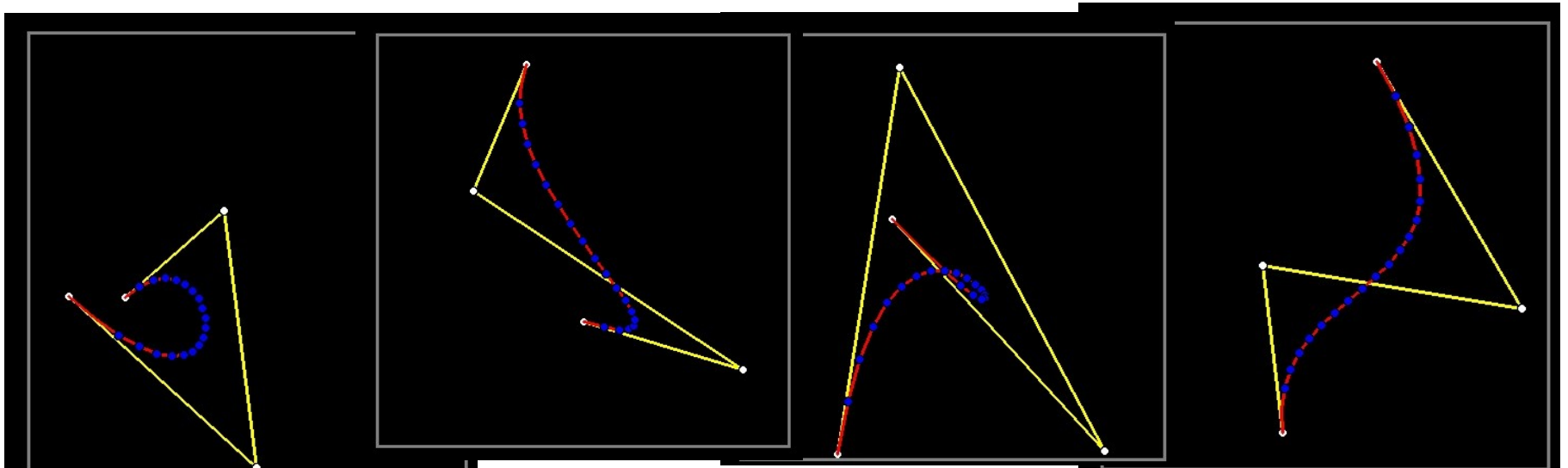
- If basis functions are all positive, the spline has the convex hull property
 - we're still requiring them to sum to 1



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

Convex Hull

Check that the curve remains inside the convex hull of the control points in our examples

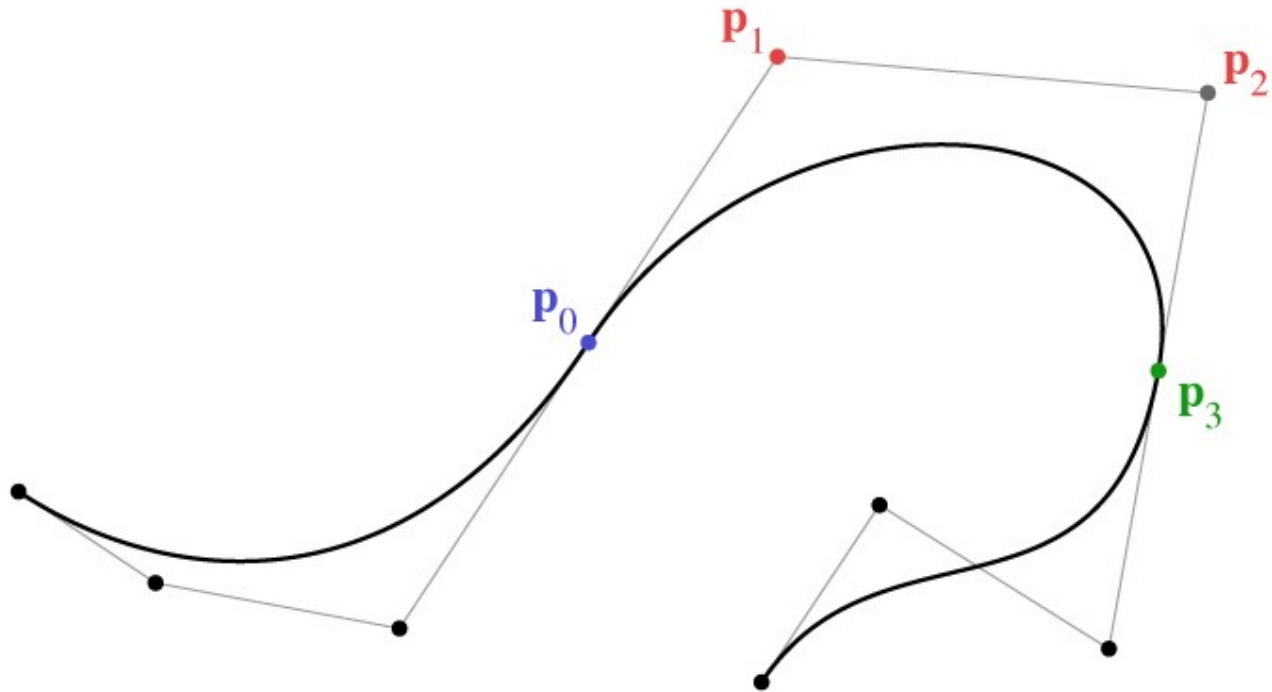


Chaining spline segments

- Hermite curves are convenient because they can be made long easily
- Bézier curves are convenient because their controls are all points and they have nice properties
 - and they interpolate every 4th 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

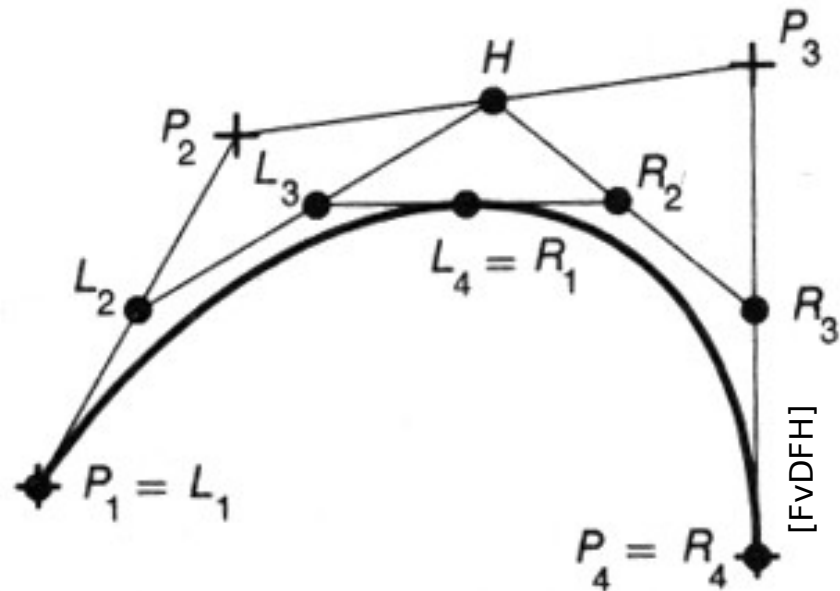
Chaining Bézier splines

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



Subdivision

- A Bézier spline segment can be split into a two-segment curve:



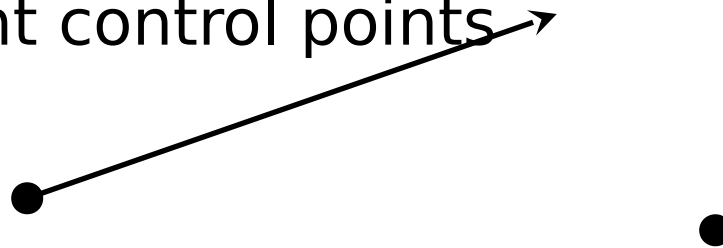
- de Casteljau's algorithm
- also works for arbitrary t

Cubic Bézier splines

- Very widely used type, especially in 2D
 - e.g. it is a primitive in PostScript/PDF
- Can represent C^1 and/or G^1 curves with corners
- Can easily add points at any position
- Illustrator demo

Hermite to Catmull-Rom

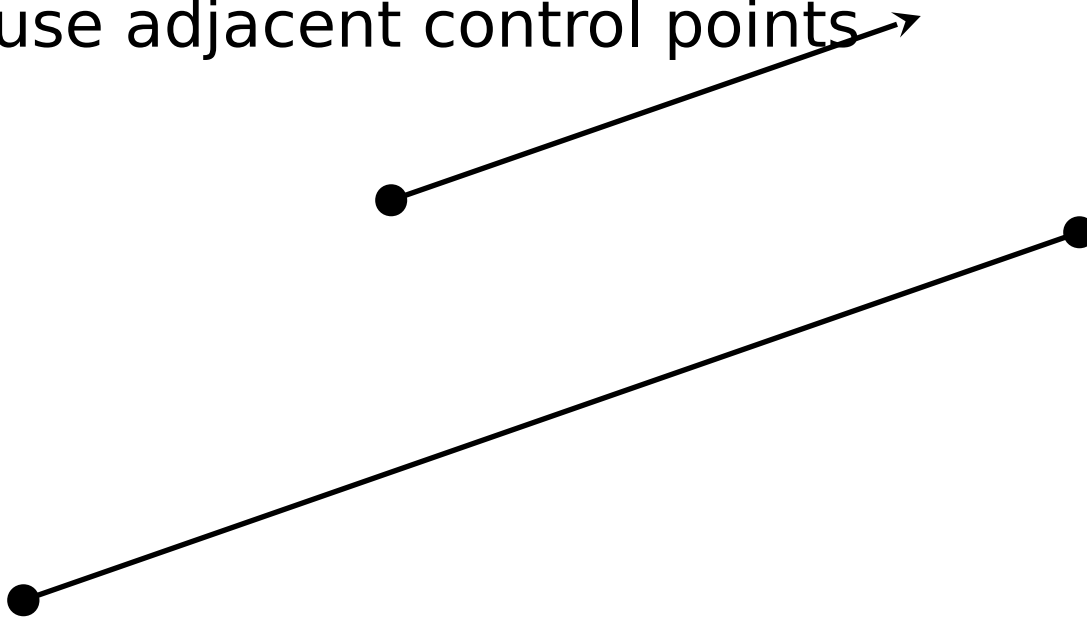
- Have not yet seen any interpolating splines
- Would like to define tangents automatically
 - use adjacent control points



- end tangents: extra points or zero

Hermite to Catmull-Rom

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

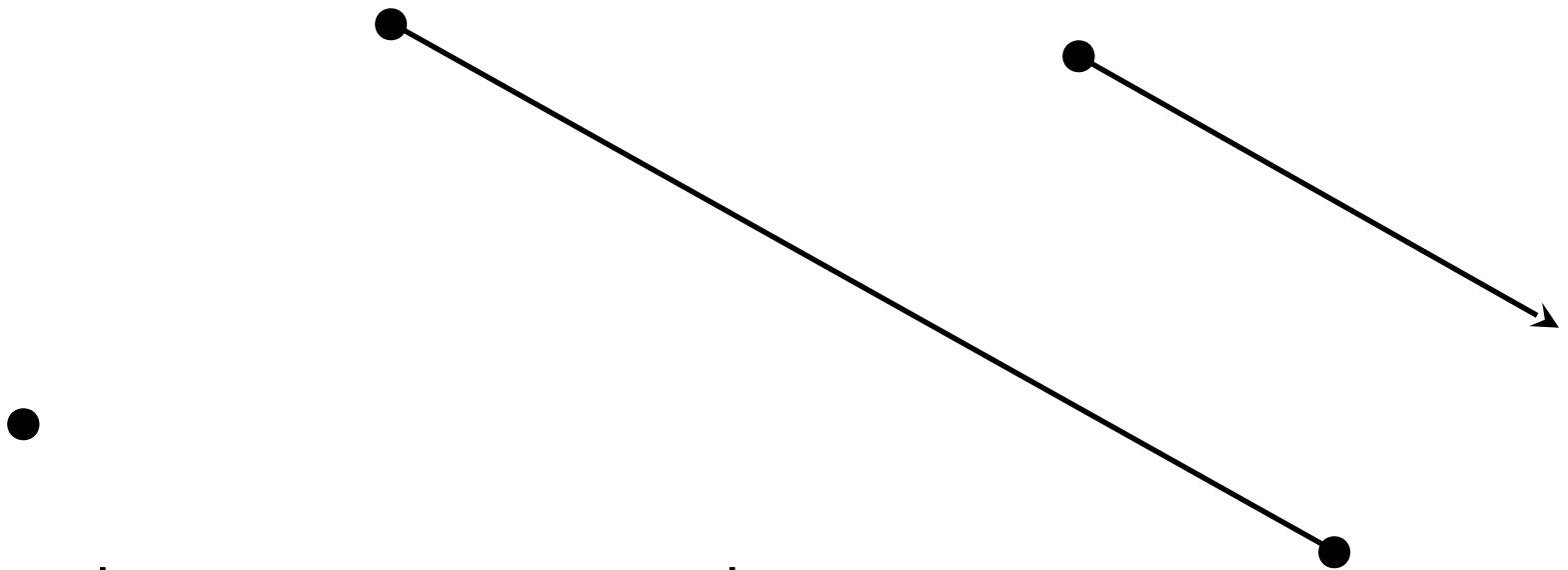


- end tangents: extra points or zero



Hermite to Catmull-Rom

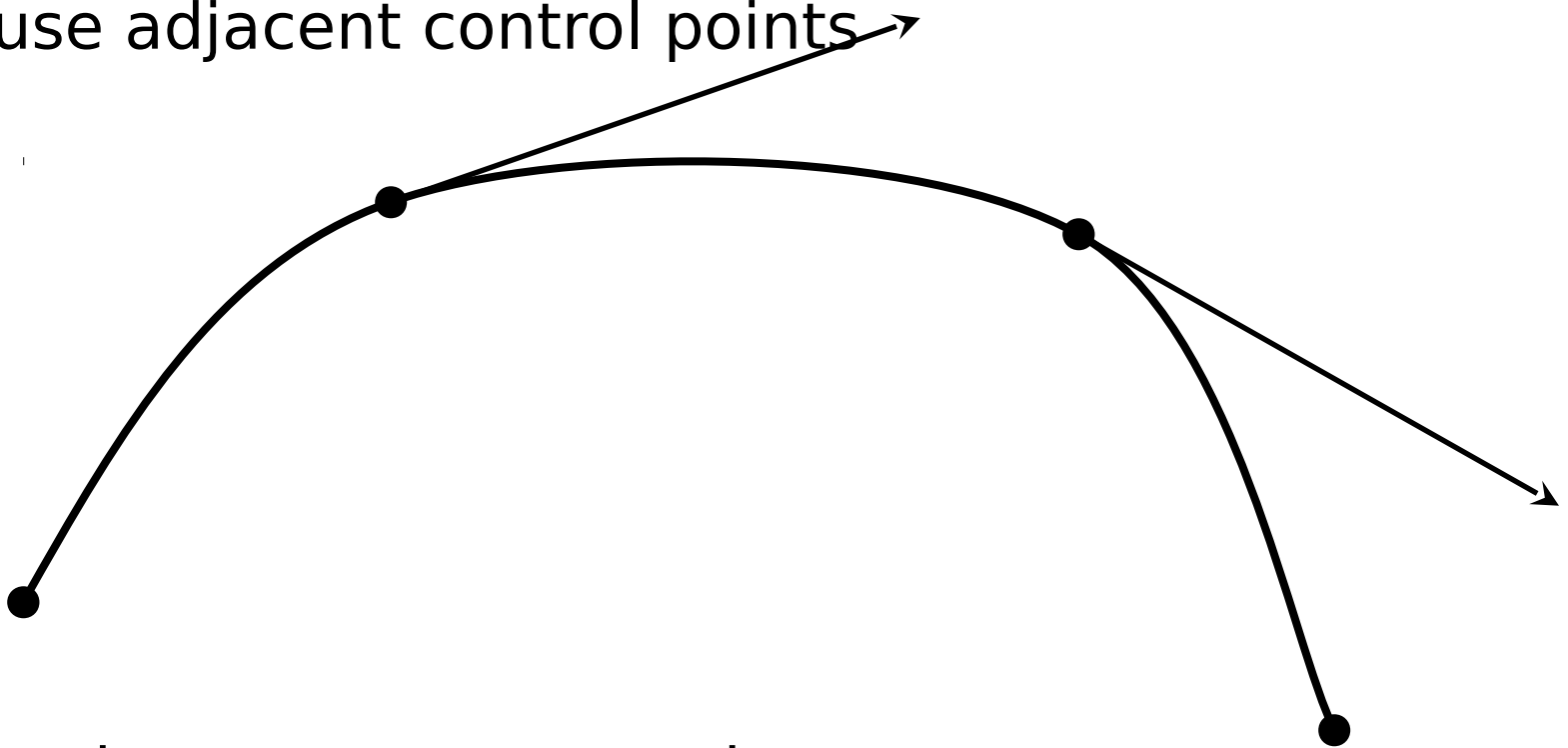
- Have not yet seen any interpolating splines
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 - use adjacent control points



- end tangents: extra points or zero

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

Hermite to Catmull-Rom

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

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

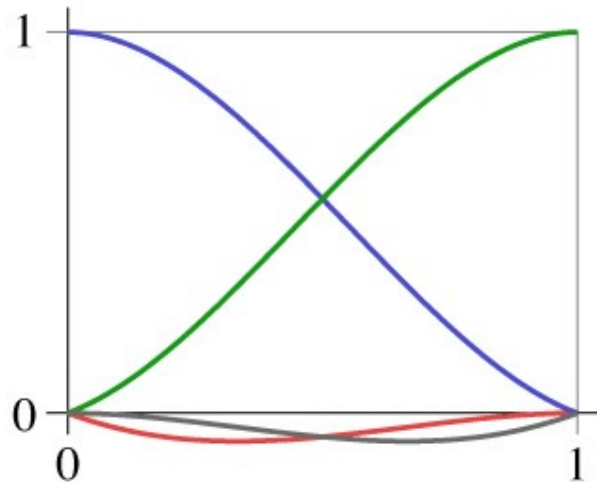
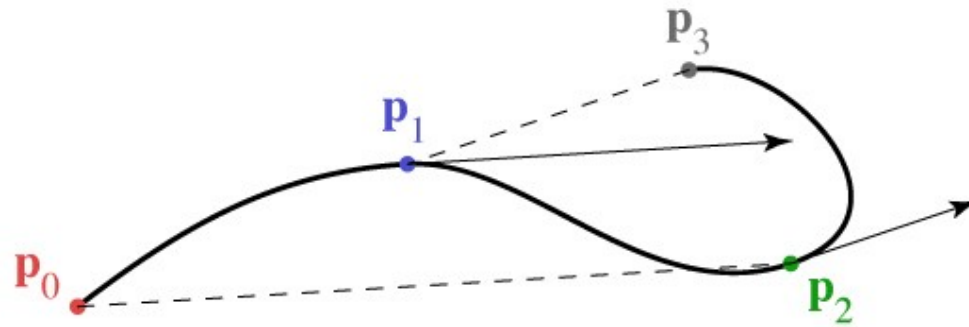
$$\mathbf{p}_1 = \mathbf{q}_{k+1}$$

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

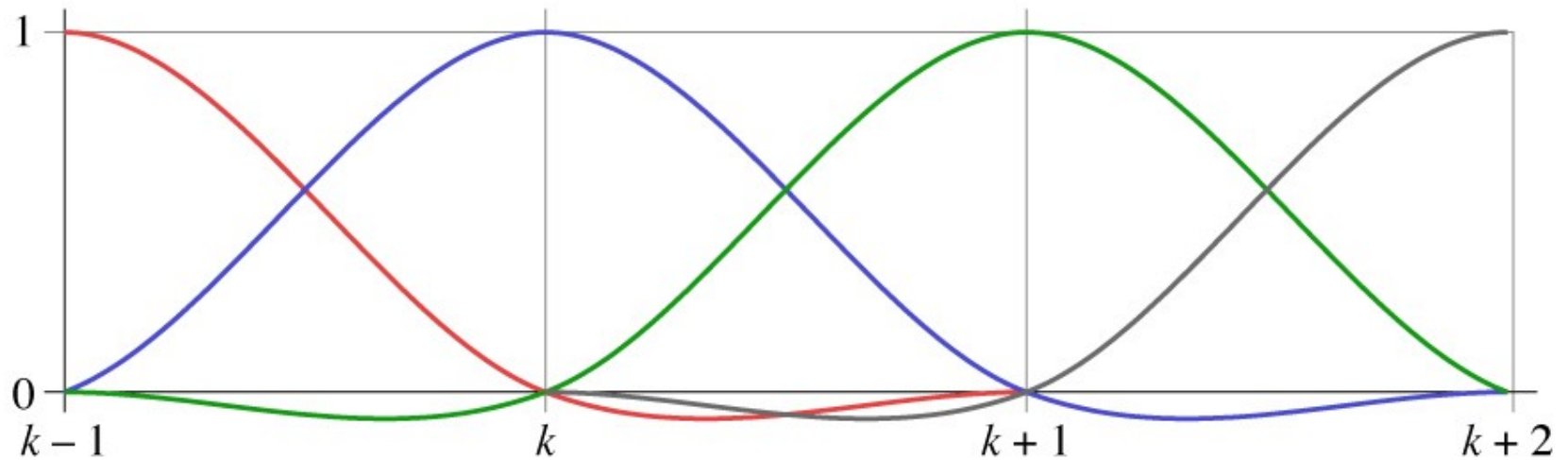
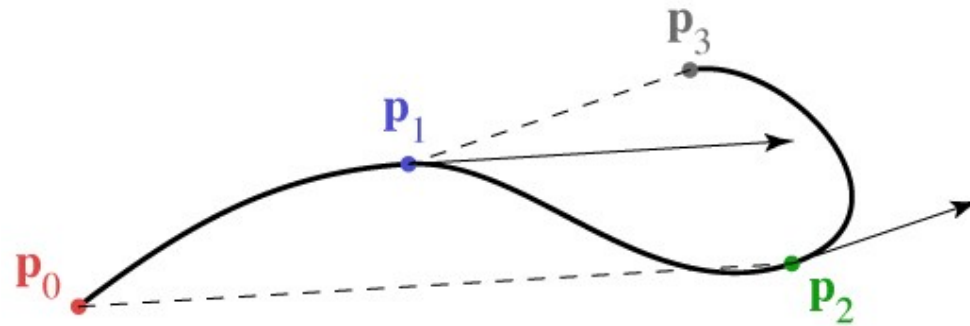
$$\mathbf{v}_1 = 0.5(\mathbf{q}_{k+2} - \mathbf{q}_K)$$

$$\begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -0.5 & 0 & 0.5 & 0 \\ 0 & -0.5 & 0 & 0.5 \end{bmatrix} \begin{bmatrix} \mathbf{q}_{k-1} \\ \mathbf{q}_k \\ \mathbf{q}_{k+1} \\ \mathbf{q}_{k+2} \end{bmatrix}$$

Catmull-Rom basis



Catmull-Rom basis

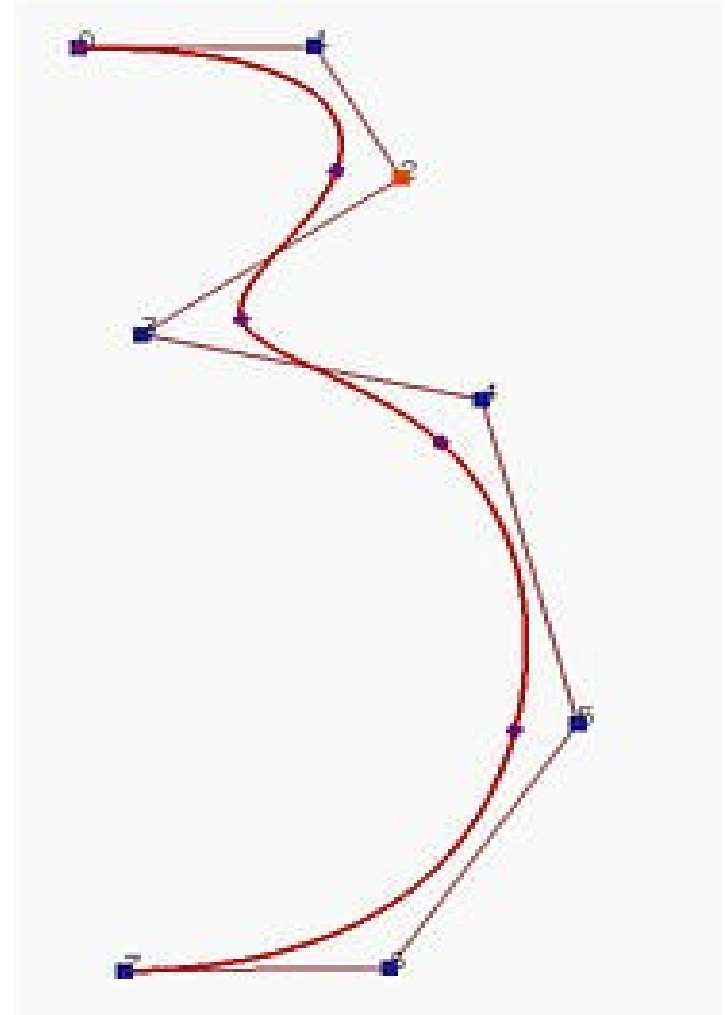


Catmull-Rom splines

- 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-Spline

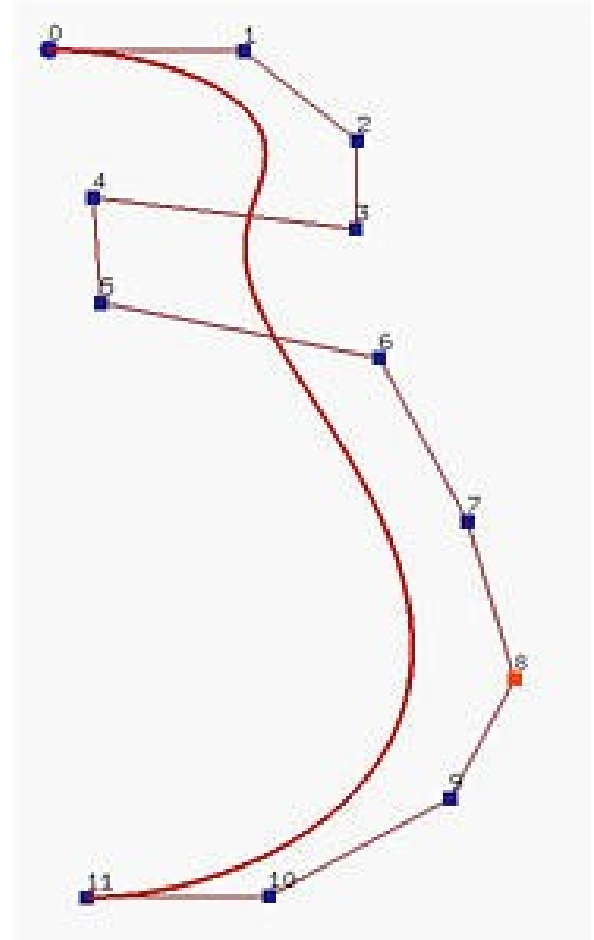
- We may want more continuity than C^1
- We may not need an interpolating spline
- B-splines are a clean, flexible way of making long splines with arbitrary order of continuity



Why B-spline.

1. High-order bezier curve instead?

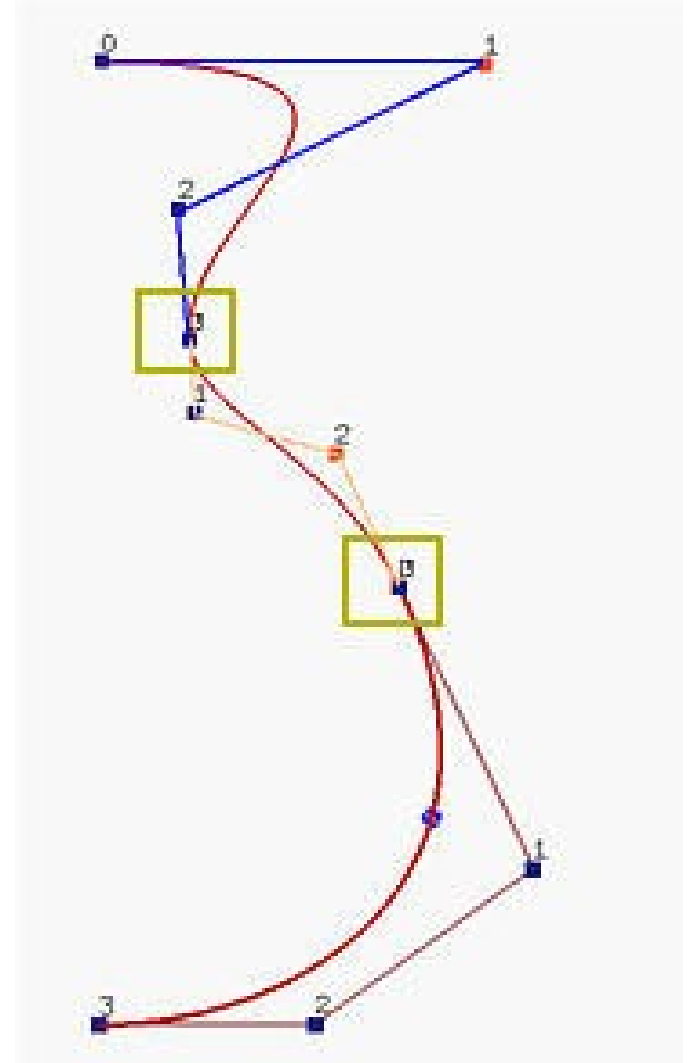
- Recall bezier curve
 - The degree of a Bezier Curve is determined by the number of control points
 - E. g. bezier curve degree 11
 - difficult to bend the "neck" toward the line segment P_4P_5 .
 - We can add more control points, BUT this will increase the degree of the curve → increase computational burden and smoothness



Why B-Spline.

2. Chaining cubic Bezier curves instead?

- Joint many bezier curves of lower degree together (right figure)
 - You can chain Hermite or Bezier curves
 - Catmull-Rom spline is also in this form
 - Unintuitive to control and sometimes not smooth enough



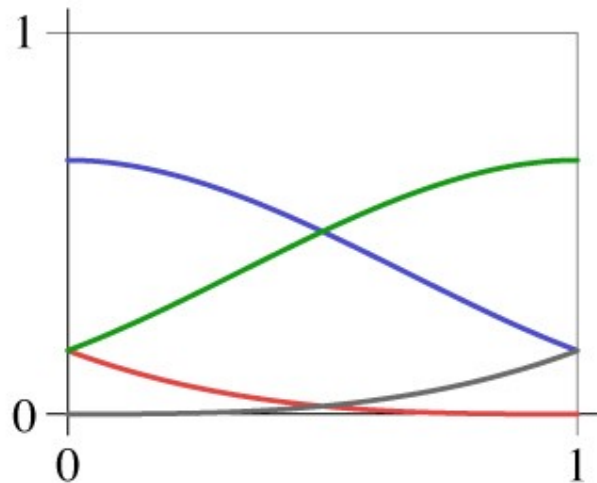
B-splines

- Use 4 points, but approximate only middle two

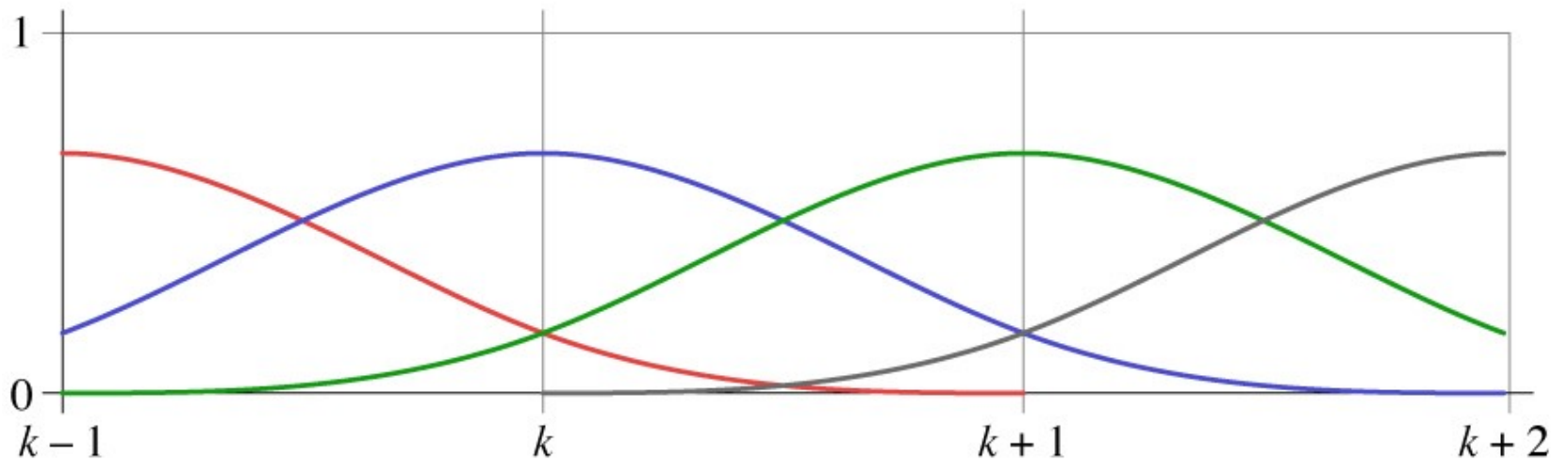
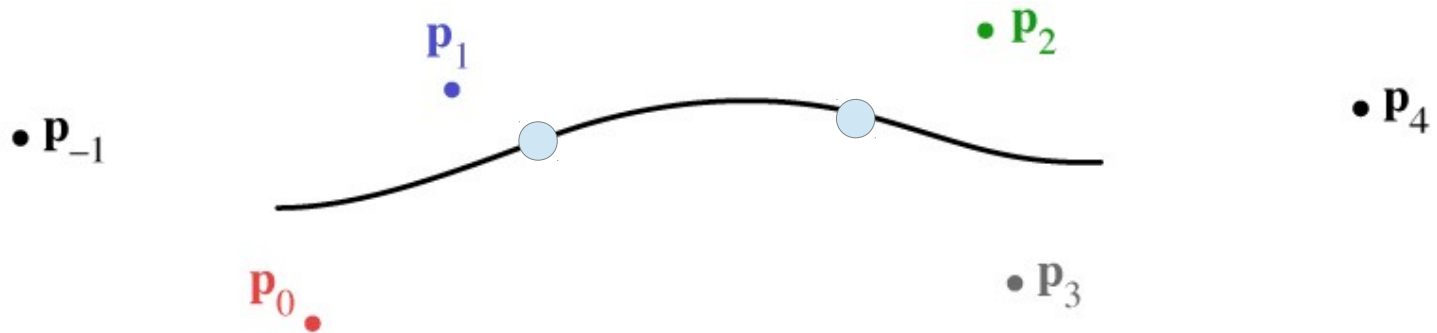


- Draw curve with overlapping segments
 - 0-1-2-3, 1-2-3-4, 2-3-4-5, 3-4-5-6, etc
- Curve may miss all control points
- Smoother at joint points (why? - later)

Cubic B-spline basis



Cubic B-spline basis



Deriving the B-Spline

- Want a cubic spline; therefore 4 active control points
- Want C2 continuity
 - Turns out that is enough to determine everything

Efficient construction of any B-spline

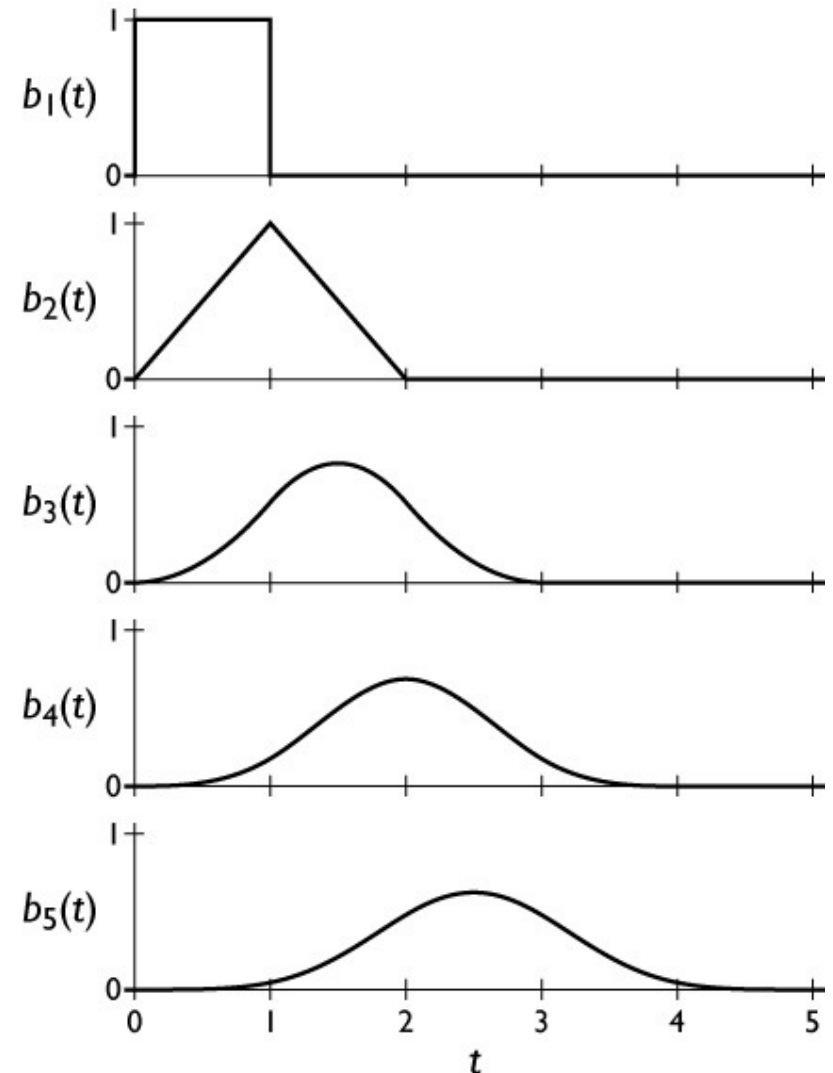
- B-splines defined for all orders
 - order d : degree $d - 1$
 - order d : d points contribute to value
- One definition: Cox-deBoor recurrence

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

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

B-spline construction, alternate view

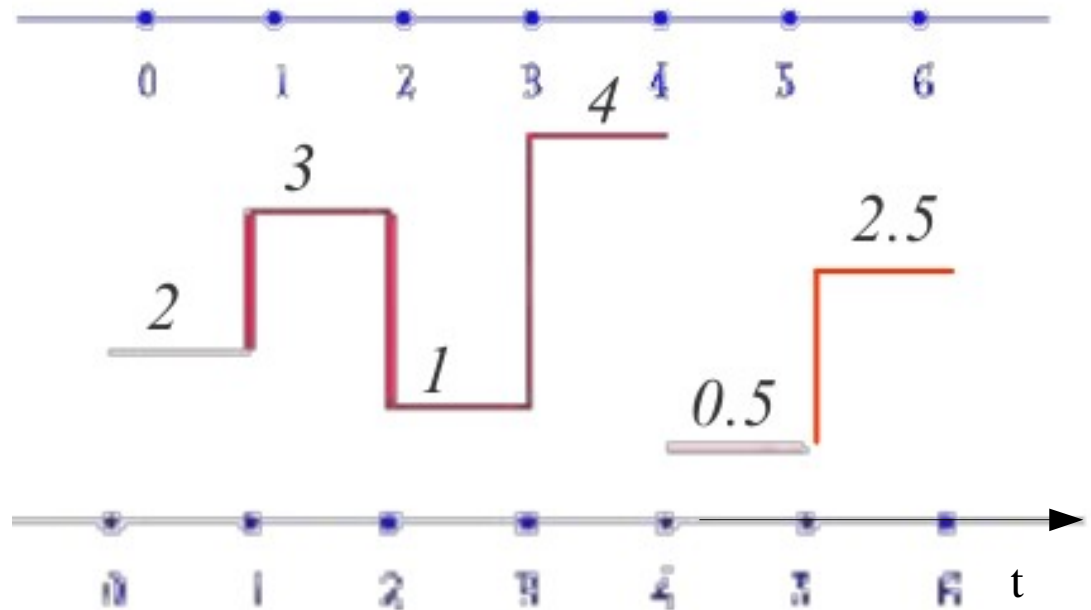
- Recurrence
 - ramp up/down
- Convolution
 - smoothing of basis fn
 - smoothing of curve



B-spline of order 1 using $b_1(t)$

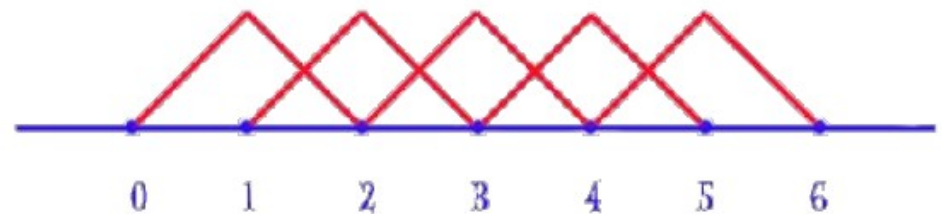
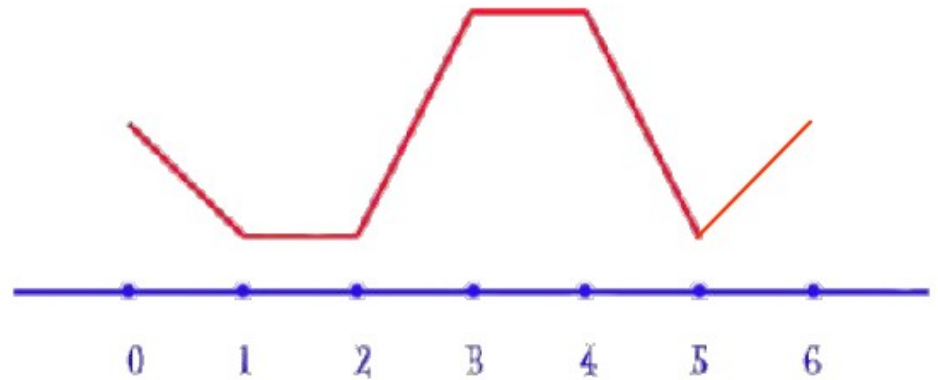
- Order = 1
- Degree = 0
- Discontinuous
- 1 segment basis function

$x(t)$



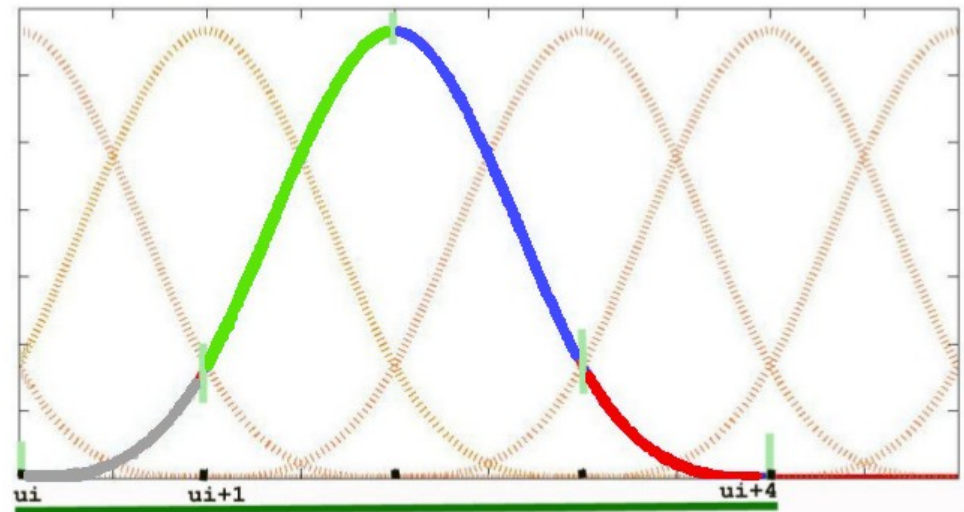
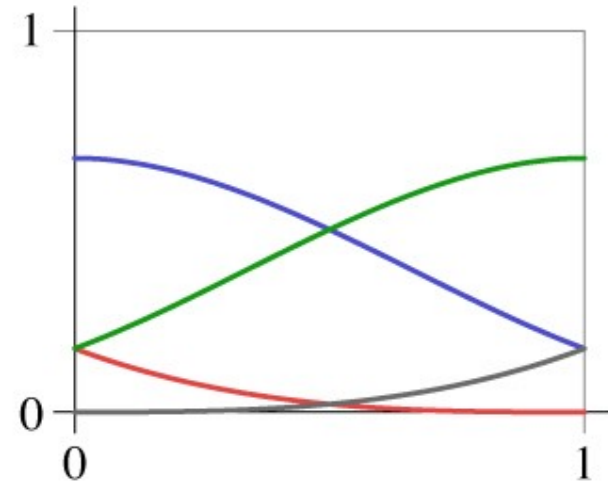
B-spline of order 2 (Linear B-Splines)

- Order = 2
- Degree = 1
- C^0 continuous
- 2 segments



B-spline or order 4 (cubic B-spline)

- Order = 4
- Degree = 3
- C2 continuous
- 4 segments

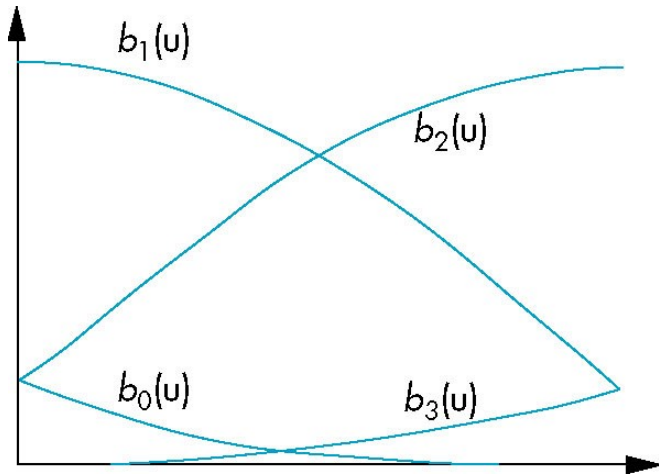


Cubic B-spline matrix

$$\mathbf{p}(t) = [t^3 \quad t^2 \quad t \quad 1] \cdot \frac{1}{6} \begin{bmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 0 & 3 & 0 \\ 1 & 4 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{p}_{k-1} \\ \mathbf{p}_k \\ \mathbf{p}_{k+1} \\ \mathbf{p}_{k+2} \end{bmatrix}$$

$$\mathbf{p}(t) = \mathbf{t}^T \mathbf{M}_s \mathbf{p} = \mathbf{b}(t)^T \mathbf{p}$$

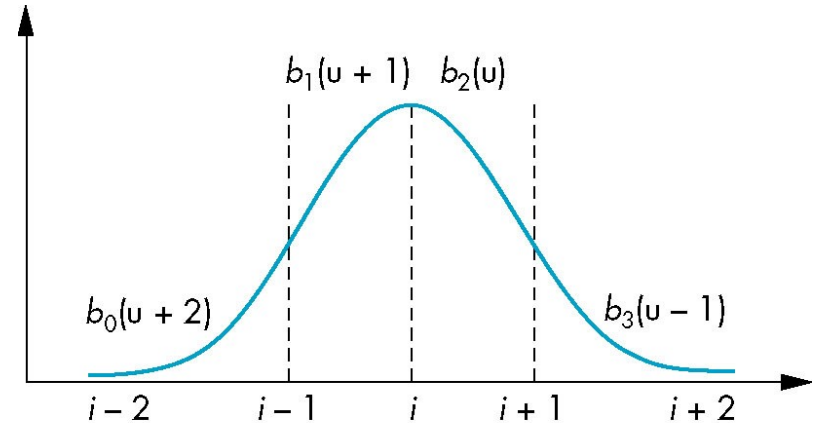
$$\mathbf{b}(t) = \frac{1}{6} \begin{bmatrix} (1-t)^3 \\ 3t^3 - 6t^2 + 4 \\ -3t^3 + 3t^2 + 3t + 1 \\ t^3 \end{bmatrix}$$



Basis Functions

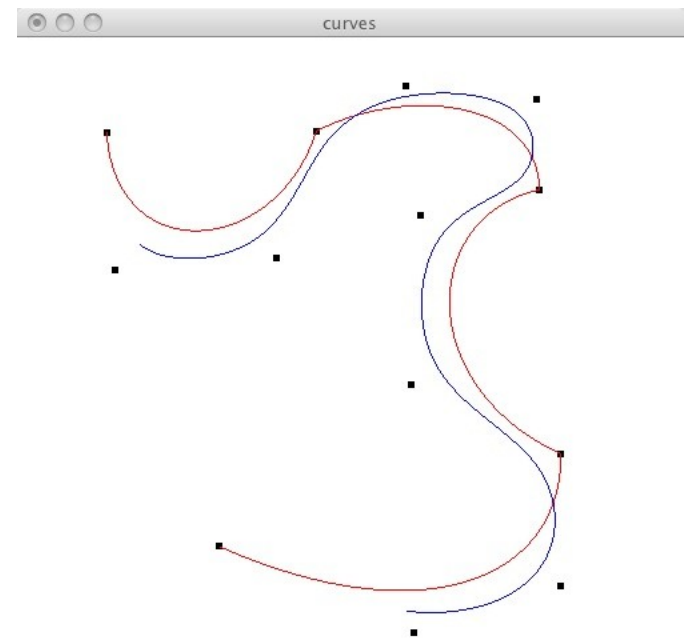
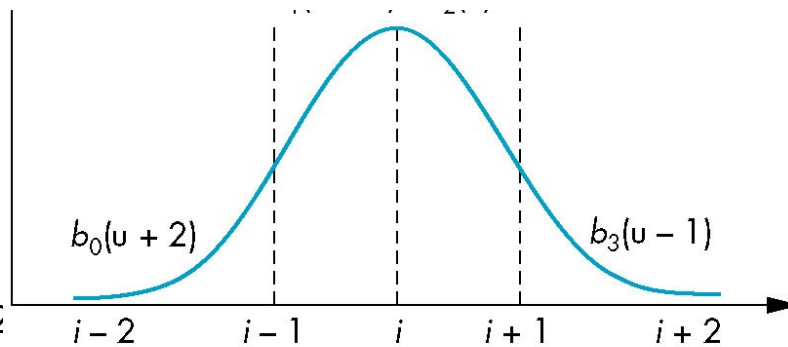
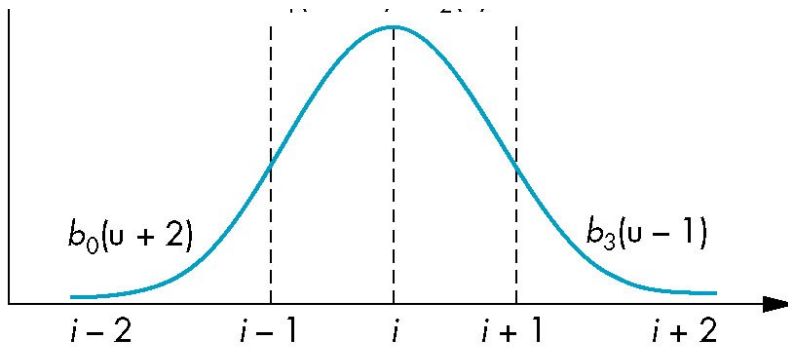
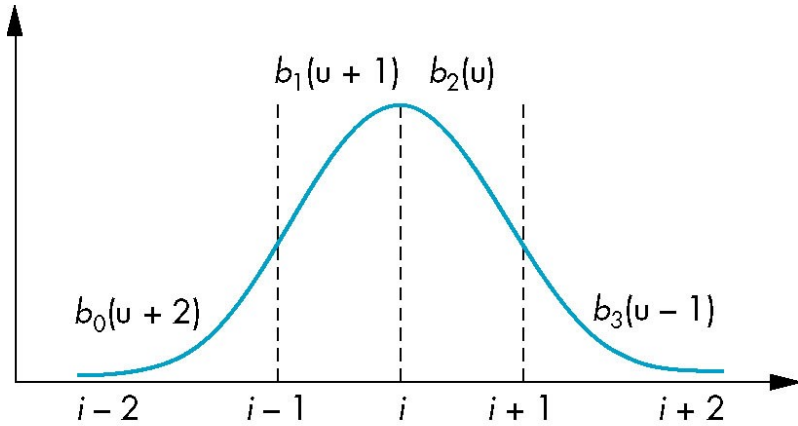
In terms of the blending polynomials

$$B_i(u) = \begin{cases} 0 & u < i-2 \\ b_0(u+2) & i-2 \leq u < i-1 \\ b_1(u+1) & i-1 \leq u < i \\ b_2(u) & i \leq u < i+1 \\ b_3(u-1) & i+1 \leq u < i+2 \\ 0 & u \geq i+2 \end{cases}$$



E. Angel and D. Shreiner: Interactive Computer
Graphics 6E © Addison-Wesley 2012

Basis Functions



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

Non-uniform B-Spline basis function

$$N_{i,k}(u) = (u - u_i) \frac{N_{i,k-1}(u)}{u_{i+k-1} - u_i} + (u_{i+k} - u) \frac{N_{i+1,k-1}(u)}{u_{i+k} - u_{i+1}} \quad (1.1)$$

$$N_{i,1} = \begin{cases} 1 & u_i \leq u \leq u_{i+1} \\ 0 & \text{Otherwise} \end{cases} \quad (1.2)$$

→ In equation (1.1), the denominators can have a value of zero, 0/0 is presumed to be zero.

Type of B-Spline knot vector (the set of parameters t)

Non-periodic knots
(open knots)

- First and last knots are duplicated k times.
- E.g (0,0,0,1,2,2,2)
- Curve pass through the first and last control points

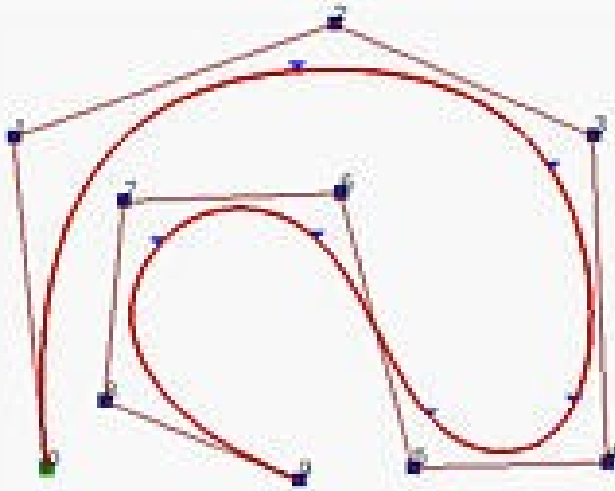
Periodic knots
(non-open knots)

- First and last knots are not duplicated – same contribution.
- E.g (0, 1, 2, 3)
- Curve doesn't pass through end points.
- can used to generate closed curves (when first = last control points)

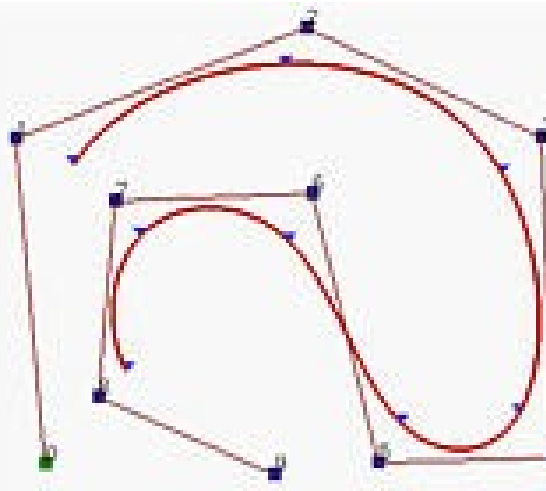
Type of B-Spline knot vector

Non-periodic knots
(open knots)

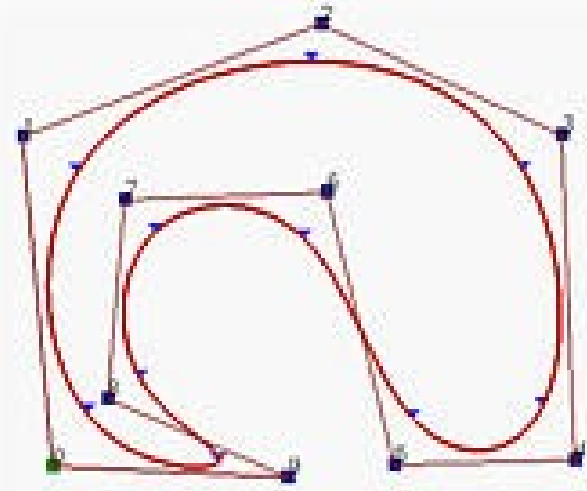
Periodic knots
(non-open knots)



(0,0,0,1,2,2,2)



(0, 1, 2, 3...)



(Closed knots)

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

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

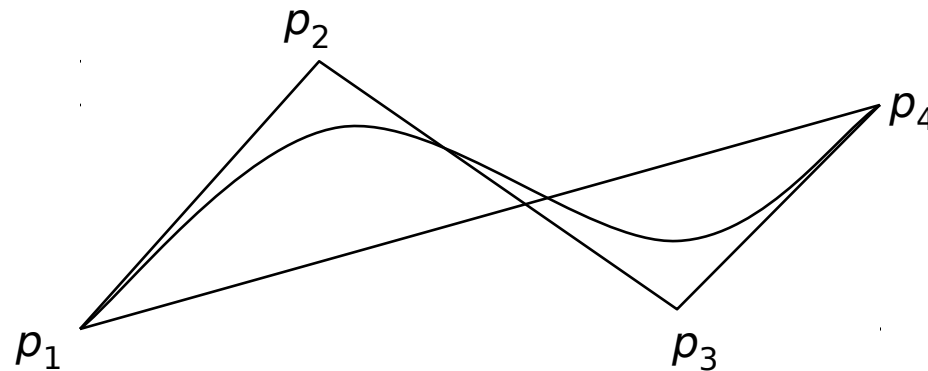
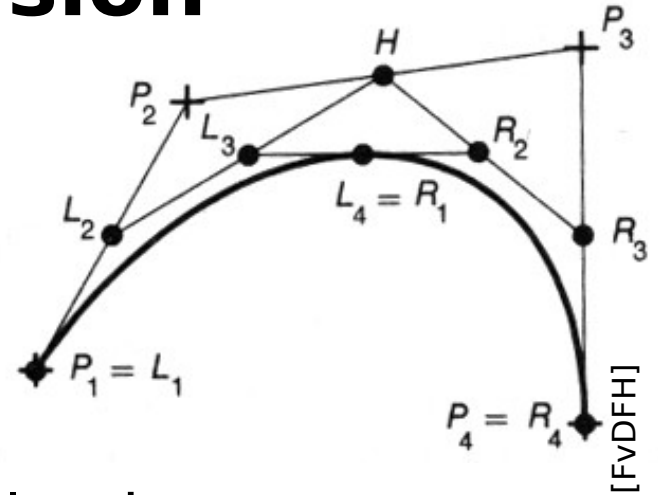
$$\begin{aligned}\mathbf{p}_1(t) &= T(t)M_1(M_1^{-1} M_2 P_2) \\ &= T(t)M_2 P_2 = \mathbf{p}_2(t)\end{aligned}$$

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)

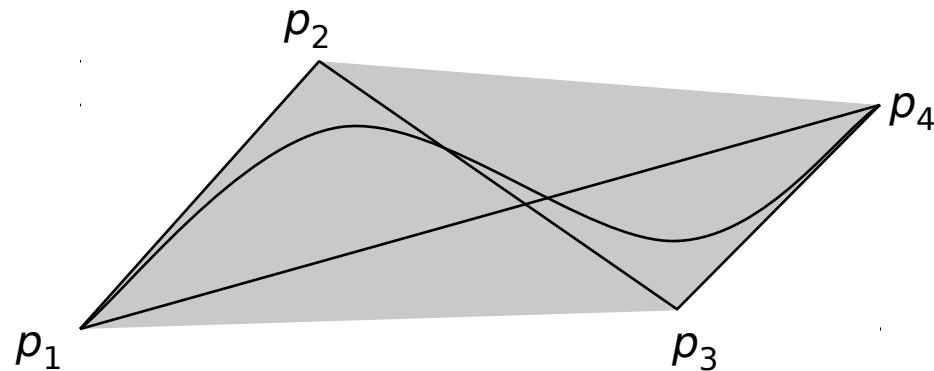
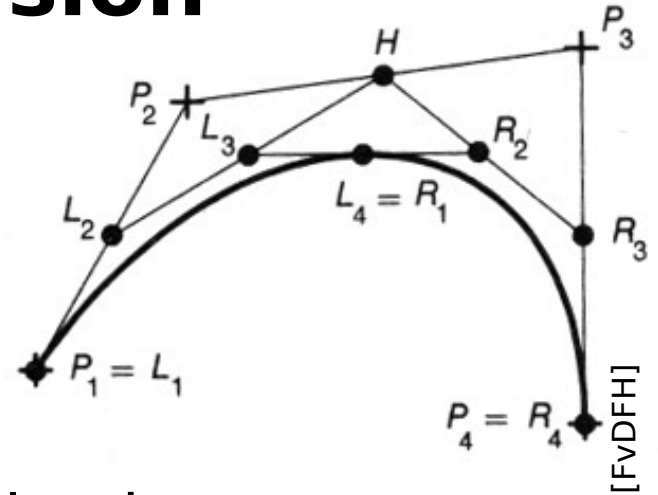
Evaluating by subdivision

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



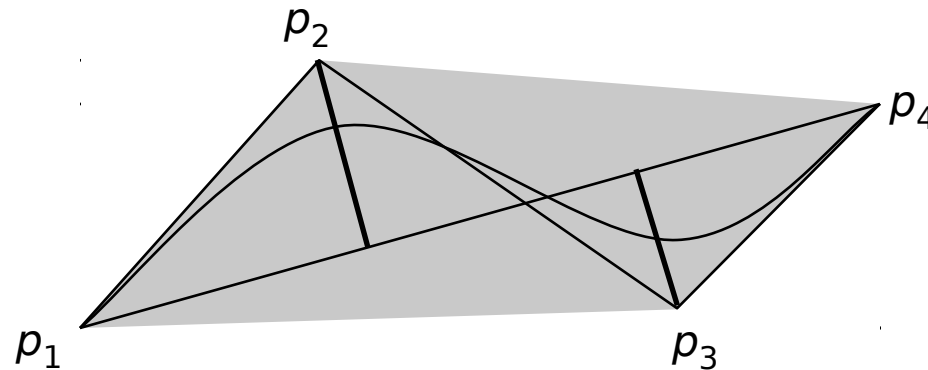
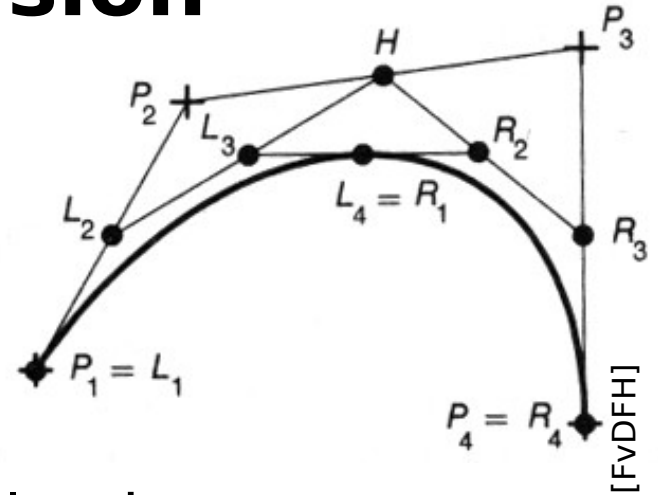
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Evaluating by subdivision

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



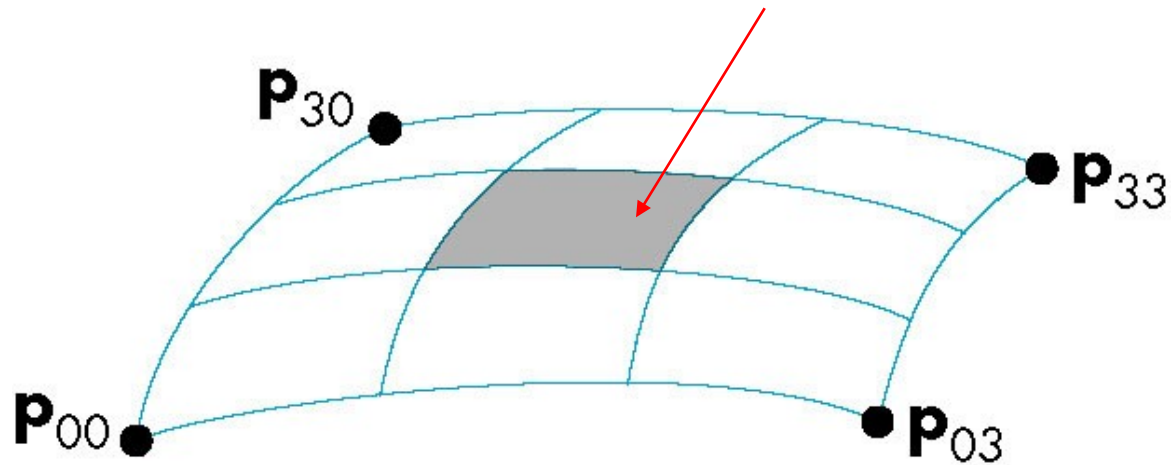
Evaluating with uniform spacing

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

B-Spline Patches

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

defined over only 1/9 of region



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