CS4620/5620: Lecture 26

Splines

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Announcements

- PPA I grading: Thu and Fri
 - -Signup or send us mail

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Motivation: smoothness

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



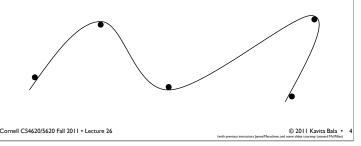
- · So far we can make
 - -things with corners (lines, squares, triangles, ...)
 - -circles and ellipses (only get you so far!)

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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

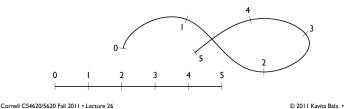
- Smoothness
 - -in drafting spline, comes from physical curvature minimization
 - -in CG spline, comes from choosing smooth functions
 - usually low-order polynomials
- Control
 - in drafting spline, comes from fixed pegs
 - -in CG spline, comes from user-specified control points

Defining spline curves

• At the most general they are parametric curves

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

- Generally p(t) is a piecewise polynomial
 - the discontinuities are at the integers



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Defining spline curves

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

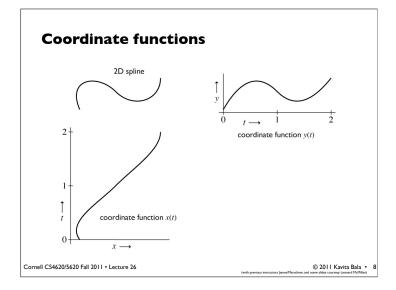
$$x(t) = a_x t^3 + b_x t^2 + c_x t + d_x$$

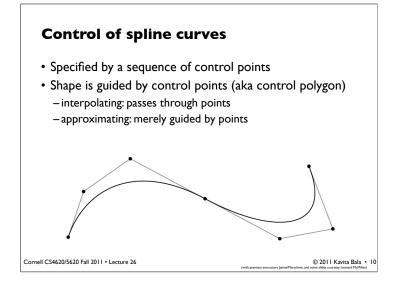
$$y(t) = a_y t^3 + b_y t^2 + c_y t + d_y$$

-Coefficients are different for every interval

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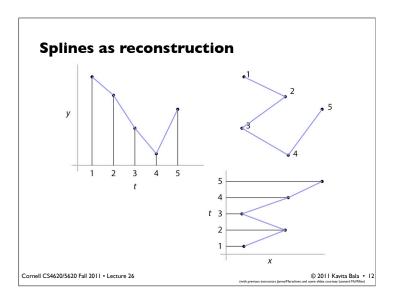


How splines depend on their controls

- Each coordinate is separate
 - the function x(t) is determined solely by the x coordinates of the control points
 - -this means ID, 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

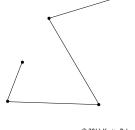
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Trivial example: piecewise linear

- 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



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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

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

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Trivial example: piecewise linear

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

$$\mathbf{p}(t) = (\mathbf{p}_1 - \mathbf{p}_0)t + \mathbf{p}_0$$
$$= (1 - t)\mathbf{p}_0 + t\mathbf{p}_1$$

-interpretation in matrix viewpoint

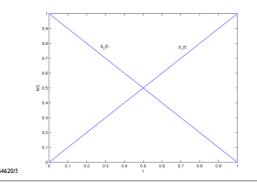
$$\mathbf{p}(t) = \begin{pmatrix} \begin{bmatrix} t & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \end{pmatrix} \begin{bmatrix} \mathbf{p}_0 \\ \mathbf{p}_1 \end{bmatrix}$$

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Trivial example: piecewise linear

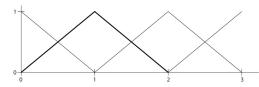
Vector blending formulation: "average of points"
 blending functions: contribution of each point as t changes



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Trivial example: piecewise linear

Basis function formulation: "function times point"
 basis functions: contribution of each point as t changes



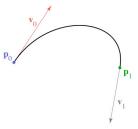
- can think of them as blending functions glued together
- -this is just like a reconstruction filter!

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Hermite splines

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



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Hermite splines

· Solve constraints to find coefficients

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

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

$$x(0) = x_{0} = d$$

$$x(1) = x_{1} = a + b + c + d$$

$$x'(0) = x'_{0} = c$$

$$x'(1) = x'_{1} = 3a + 2b + c$$

$$d = x_{0}$$

$$c = x'_{0}$$

$$a = 2x_{0} - 2x_{1} + x'_{0} + x'_{1}$$

$$b = -3x_{0} + 3x_{1} - 2x'_{0} - x'_{1}$$

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Hermite splines

• Matrix form is much simpler

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

- -coefficients = rows
- -basis functions = columns

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Longer Hermite splines

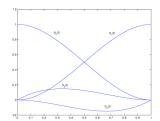
- Can only do so much with one Hermite spline
- Can use these splines as segments of a longer curve
 -curve from t = 0 to t = 1 defined by first segment
 -curve from t = 1 to t = 2 defined by second segment
- To avoid discontinuity, match derivatives at junctions
 -this produces a C¹ curve

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Hermite splines

• Hermite basis functions



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