

The Essentials of CAGD

Chapter 11: Working with B-spline Curves

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www.farinhansford.com/books/essentials-cagd

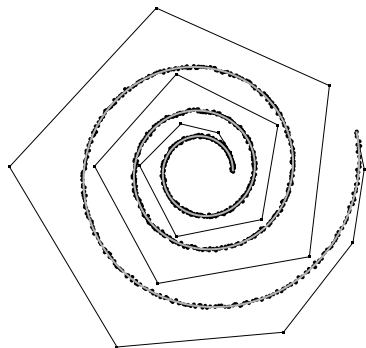
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- 2 Designing with B-spline curves
- 3 Least Squares Approximation
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Introduction to Working with B-spline Curves

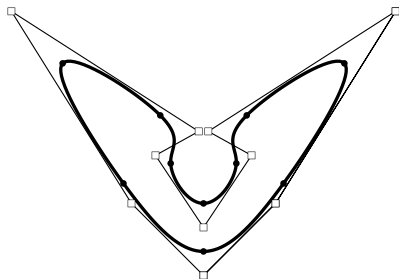


How to use B-spline curves?

B-spline curves popularity due to the many possible ways in which they can be “put to work”

Designing with B-spline curves

Find: a B-spline curve for the character “v” in some fancy font

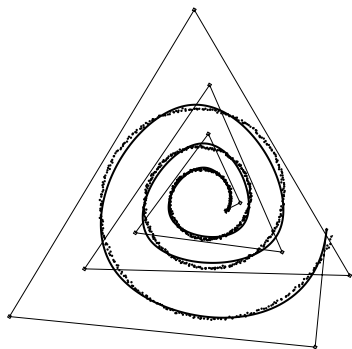


Most basic design process:
Move individual control points until
desired shape achieved

Manual/interactive method ok for
final fine tuning of shape

Initial “guess” can be created faster
with methods in this chapter

Introduction to Working with B-spline Curves



Many applications supply a large number of data points
⇒ from scanning devices

Find a cubic B-spline curve approximating their shape

Most popular method:
least squares approximation

Least Squares Approximation

Cubic B-spline curve defined by

- L polynomial segments
- Assume simple domain knots \Rightarrow number of knots $K = L + 5$
- Knot sequence u_0, \dots, u_{K-1}

Given P data points

- $\mathbf{p}_0, \dots, \mathbf{p}_{P-1}$
- Each \mathbf{p}_i associated with parameter value v_i

Find a cubic B-spline curve $\mathbf{x}(u)$

such that the distances $\|\mathbf{p}_i - \mathbf{x}(v_i)\|$ are small

Least Squares Approximation

B-spline curve

$$\mathbf{x}(u) = \mathbf{d}_0 N_0^3(u) + \dots + \mathbf{d}_{D-1} N_{D-1}^3(u)$$

Given points $\mathbf{p}_i = \mathbf{x}(v_i)$ $i = 0, \dots, P-1$ leading to

$$\mathbf{d}_0 N_0^3(v_0) + \dots + \mathbf{d}_{D-1} N_{D-1}^3(v_0) = \mathbf{p}_0$$

\vdots

$$\mathbf{d}_0 N_0^3(v_{P-1}) + \dots + \mathbf{d}_{D-1} N_{D-1}^3(v_{P-1}) = \mathbf{p}_{P-1}$$

In matrix form:

$$\begin{bmatrix} N_0^3(v_0) & \dots & N_{D-1}^3(v_0) \\ \vdots & & \vdots \\ N_0^3(v_{P-1}) & \dots & N_{D-1}^3(v_{P-1}) \end{bmatrix} \begin{bmatrix} \mathbf{d}_0 \\ \vdots \\ \mathbf{d}_{D-1} \end{bmatrix} = \begin{bmatrix} \mathbf{p}_0 \\ \vdots \\ \mathbf{p}_{P-1} \end{bmatrix}$$

$$\mathbf{M}\mathbf{D} = \mathbf{P}$$

Least Squares Approximation

Linear system $M\mathbf{D} = \mathbf{P}$ is **overdetermined**

– Number P of data points $>$ number D of curve control points

Solution: multiply both sides by M^T :

$$M^T M \mathbf{D} = M^T \mathbf{P}$$

Linear system with D equations in D unknowns

– Square and symmetric coefficient matrix $M^T M$

– Solution straightforward since $M^T M$ invertible

as long as parameter values v_j must be “evenly” distributed in domain knots

Least Squares Approximation

- Parameters to define: – How many segments L should the curve have?
- How to choose the knots u_j
 - How to choose the parameter values v_j ?

No universal answers – suggestions:

- Choose the parameters v_j according to the chord length
- Select $L \approx P/10$
- Choose u_j such that approximately ten v_j fall in each interval domain knot interval $[u_i, u_{i+1}]$

Shape Equations

Possible data point defects:

- noisy
- unevenly distributed

⇒ Least squares approximation might fail to produce nice results

Solution: modify method with shape information

- Accept deviation from data for a better-shaped curve
- Formulate conditions for the control polygon's shape
- Assumption: a polygon is nice if it does not wiggle much

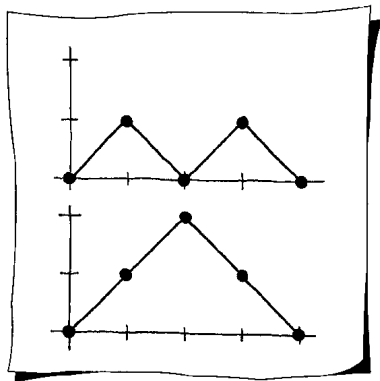
Expressed by computing *second differences* of control points

$$\Delta^2 \mathbf{d}_i = \mathbf{d}_i - 2\mathbf{d}_{i+1} + \mathbf{d}_{i+2}$$

Less wiggle ⇒ smaller sum:

$$S = \|\Delta^2 \mathbf{d}_0\| + \dots + \|\Delta^2 \mathbf{d}_{D-3}\|$$

Shape Equations



Example

Top polygon: $S = 6$

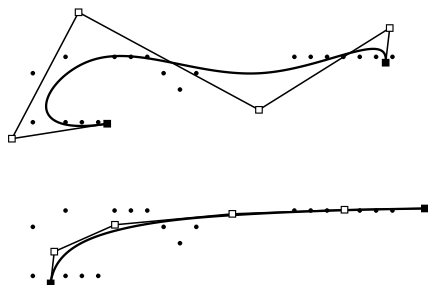
$$\Delta^2 \mathbf{d}_0 = \begin{bmatrix} 0 \\ -2 \end{bmatrix} \quad \Delta^2 \mathbf{d}_1 = \begin{bmatrix} 0 \\ 2 \end{bmatrix} \quad \Delta^2 \mathbf{d}_2 = \begin{bmatrix} 0 \\ -2 \end{bmatrix}$$

Bottom polygon: $S = 2$

$$\Delta^2 \mathbf{d}_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \Delta^2 \mathbf{d}_1 = \begin{bmatrix} 0 \\ -2 \end{bmatrix} \quad \Delta^2 \mathbf{d}_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

\Rightarrow Bottom polygon is smoother

Shape Equations



Least squares approximation:
Add **shape equations** to the
overdetermined system

$$\mathbf{d}_0 - 2\mathbf{d}_1 + \mathbf{d}_2 = \mathbf{0}$$

\vdots

$$\mathbf{d}_{D-3} - 2\mathbf{d}_{D-2} + \mathbf{d}_{D-1} = \mathbf{0}$$

Overdetermined linear system
becomes even more overdetermined

Top: without shape equations
Bottom: with shape equations

Cubic Spline Interpolation

Interpolation: # given data equals # unknown control points

Given P data points

$$\mathbf{p}_0, \dots, \mathbf{p}_{P-1}$$

Interpolate with a
cubic B-spline curve $\mathbf{x}(u)$

End knots of multiplicity three:

$$u_0 = u_1 = u_2$$

$$u_3, \dots, u_{K-4}$$

$$u_{K-3} = u_{K-2} = u_{K-1}$$

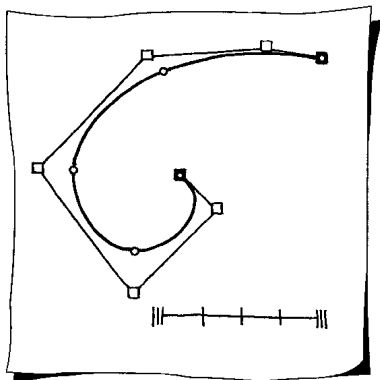
Junction points paired with \mathbf{p}_i

$$\mathbf{x}(u_i) = \mathbf{p}_0, \dots$$

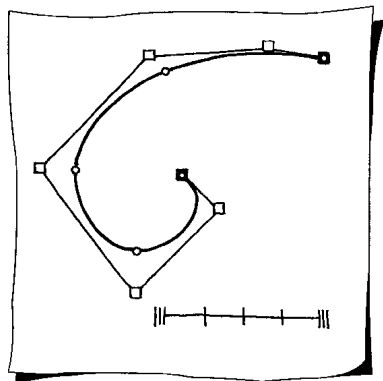
$\Rightarrow P - 1$ curve segments

$\Rightarrow K = P + 4$ knots

$\Rightarrow D = P + 2$ control points



Cubic Spline Interpolation



Example

Given $P = 5$ data points

Need $K = 5 + 4 = 9$ knots

$0, 0, 0, 1, 2, 3, 4, 4, 4$

$\Rightarrow D = 7$ control points

$\mathbf{d}_0, \dots, \mathbf{d}_6$

Cubic Spline Interpolation

Two more data items are needed than the curve has junction points

Solution: add two more data items at the ends of the curve

$$\mathbf{t}_s = \dot{\mathbf{x}}(u_2) \quad \text{start tangent}$$

$$\mathbf{t}_e = \dot{\mathbf{x}}(u_{K-3}) \quad \text{end tangent}$$

These are called **end conditions**

Knots u_2 and u_{K-3} are the first and last domain knots

Bessel tangents method: extract tangents from interpolating parabola through first and last three data points

– See The Essentials of CAGD for equation details

Cubic Spline Interpolation

Interpolation conditions:

$$\mathbf{p}_0 = \mathbf{x}(u_2)$$

$$\mathbf{t}_s = \dot{\mathbf{x}}(u_2)$$

$$\mathbf{p}_1 = \mathbf{x}(u_3)$$

$$\vdots$$

$$\mathbf{t}_e = \dot{\mathbf{x}}(u_{K-3})$$

$$\mathbf{p}_{P-1} = \mathbf{x}(u_{K-3})$$

Triple end knots result in

$$\mathbf{d}_0 = \mathbf{p}_0 \quad \text{and} \quad \mathbf{d}_{D-1} = \mathbf{p}_{P-1}$$

⇒ eliminates two unknowns

⇒ eliminates two equations

Simplified interpolation conditions:

$$\mathbf{t}_s = \dot{\mathbf{x}}(u_2)$$

$$\mathbf{p}_1 = \mathbf{x}(u_3)$$

$$\vdots$$

$$\mathbf{p}_{P-2} = \mathbf{x}(u_{K-4})$$

$$\mathbf{t}_e = \dot{\mathbf{x}}(u_{K-3})$$

For unknowns $\mathbf{d}_1, \dots, \mathbf{d}_{D-2}$

Cubic Spline Interpolation

Example

Revisit previous example

$$\mathbf{p}_0 = \mathbf{x}(0)$$

$$\mathbf{t}_s = \dot{\mathbf{x}}(0)$$

$$\mathbf{p}_1 = \mathbf{x}(1)$$

$$\mathbf{p}_2 = \mathbf{x}(2)$$

$$\mathbf{p}_3 = \mathbf{x}(3)$$

$$\mathbf{t}_e = \dot{\mathbf{x}}(4)$$

$$\mathbf{p}_4 = \mathbf{x}(4)$$

Assigning $\mathbf{d}_0 = \mathbf{p}_0$ and $\mathbf{d}_6 = \mathbf{p}_4$
System becomes

$$\mathbf{t}_s = \dot{\mathbf{x}}(0)$$

$$\mathbf{p}_1 = \mathbf{x}(1)$$

$$\mathbf{p}_2 = \mathbf{x}(2)$$

$$\mathbf{p}_3 = \mathbf{x}(3)$$

$$\mathbf{t}_e = \dot{\mathbf{x}}(4)$$

\Rightarrow five equations for the unknowns
 $\mathbf{d}_1, \dots, \mathbf{d}_5$

Cubic Spline Interpolation

Each data point yields an equation of the form

$$\mathbf{p}_i = \mathbf{d}_0 N_0^3(u_{2+i}) + \dots + \mathbf{d}_{D-1} N_{D-1}^3(u_{2+i})$$

Due to the local support property of B-spline curves

$$\mathbf{p}_i = \mathbf{d}_i N_i^3(u_{2+i}) + \mathbf{d}_{i+1} N_{i+1}^3(u_{2+i}) + \mathbf{d}_{i+2} N_{i+2}^3(u_{2+i})$$

⇒ **tridiagonal** structure

End conditions: first and last equation in the system

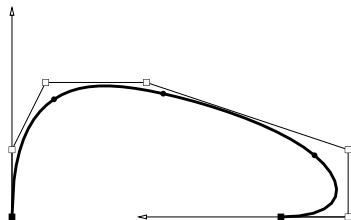
– For tridiagonal structure, must involve only the first and last unknowns

For the special case of equally spaced interior knots

$$6\mathbf{p}_i = \mathbf{d}_i + 4\mathbf{d}_{i+1} + \mathbf{d}_{i+2}$$

for each equation involving a data point

Cubic Spline Interpolation



Example

- Equally spaced knots
- End tangent equations: $\mathbf{t}_s = 3(\mathbf{d}_1 - \mathbf{d}_0)$ and $\mathbf{t}_e = 3(\mathbf{d}_6 - \mathbf{d}_5)$

$$\begin{bmatrix} 1 & & & & & & \\ 3/2 & 7/2 & 1 & & & & \\ & 1 & 4 & 1 & & & \\ & & & 1 & 7/2 & 3/2 & \\ & & & & & 1 & \\ & & & & & & 1 \end{bmatrix} \begin{bmatrix} \mathbf{d}_1 \\ \mathbf{d}_2 \\ \mathbf{d}_3 \\ \mathbf{d}_4 \\ \mathbf{d}_5 \end{bmatrix} = \begin{bmatrix} \mathbf{d}_0 + \frac{1}{3}\mathbf{t}_s \\ 6\mathbf{p}_1 \\ 6\mathbf{p}_2 \\ 6\mathbf{p}_3 \\ \mathbf{d}_6 - \frac{1}{3}\mathbf{t}_e \end{bmatrix}$$

Cubic Spline Interpolation in a Nutshell

Input:

- Data points $\mathbf{p}_0, \dots, \mathbf{p}_{P-1}$
- A cubic B-spline knot sequence

$$u_0 = u_1 = u_2, \quad u_3, \dots, u_{K-4}, \quad u_{K-3} = u_{K-2} = u_{K-1}$$

$$K = P + 4 \quad \Rightarrow \quad P - 1 \text{ curve segments}$$

Find: cubic B-spline interpolant

- Control points $\mathbf{d}_0, \dots, \mathbf{d}_{D-1}$ where $D = P + 2$
- Each data point \mathbf{p}_i is associated with parameter u_{2+i}

Compute:

- Set $\mathbf{d}_0 = \mathbf{p}_0$ and $\mathbf{d}_{D-1} = \mathbf{p}_{P-1}$
- Create tangents \mathbf{t}_s and \mathbf{t}_e using Bessel tangent equations
- Set up the tridiagonal linear system of equations
- Solve the $(D - 2) \times (D - 2)$ linear system for $\mathbf{d}_1, \dots, \mathbf{d}_{D-2}$