

## Subdivision curves

1

$$\mathbf{M}\mathbf{V} = \mathbf{V}\Lambda$$

2

## Subdivision curves

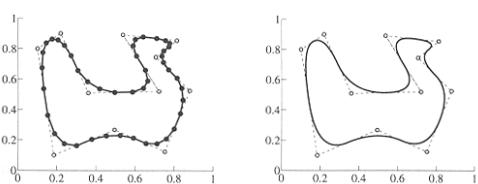
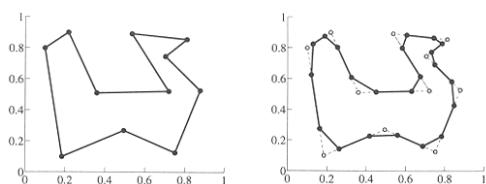
Idea:

- repeatedly refine the control polygon

$$P^1 \rightarrow P^2 \rightarrow P^3 \rightarrow \dots$$

- curve is the limit of an infinite process

$$Q = \lim_{j \rightarrow \infty} P^j$$



3

## Reading

Recommended:

- Stollnitz, DeRose, and Salesin. *Wavelets for Computer Graphics: Theory and Applications*, 1996, section 6.1-6.3, A.5.

Note: there is an error in Stollnitz, et al., section A.5.  
Equation A.3 should read:

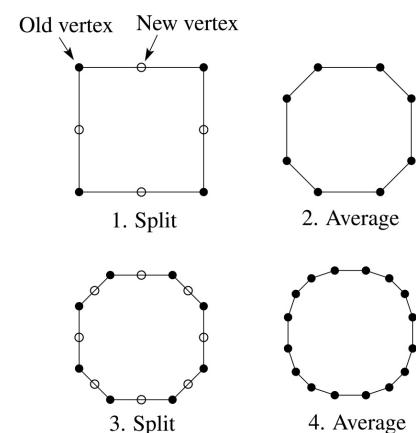
$$\mathbf{M}\mathbf{V} = \mathbf{V}\Lambda$$

4

## Chaikin's algorithm

Chakin introduced the following "corner-cutting" scheme in 1974:

- Start with a piecewise linear curve
- Insert new vertices at the midpoints (the **splitting step**)
- Average each vertex with the "next" (clockwise) neighbor (the **averaging step**)
- Go to the splitting step



## Averaging masks

The limit curve is a quadratic B-spline!

Instead of averaging with the nearest neighbor, we can generalize by applying an **averaging mask** during the averaging step:

$$r = (\dots, r_{-1}, r_0, r_1, \dots)$$

In the case of Chaikin's algorithm:

$$r =$$

5

## Lane-Riesenfeld algorithm (1980)

Use averaging masks from Pascal's triangle:

$$r = \frac{1}{2^n} \left( \binom{n}{0}, \binom{n}{1}, \dots, \binom{n}{n} \right)$$

Gives B-splines of degree  $n+1$ .

$n=0$ :

$n=1$ :

$n=2$ :

6

## Subdivide ad nauseum?

After each split-average step, we are closer to the **limit curve**.

How many steps until we reach the final (limit) position?

Can we push a vertex to its limit position without infinite subdivision? Yes!

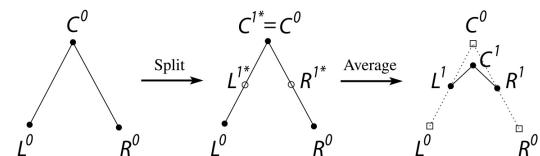
7

## Local subdivision matrix

Consider the cubic B-spline subdivision mask:

$$\begin{pmatrix} \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{pmatrix}$$

Now consider what happens during splitting and averaging in a small neighborhood:



We can write equations that relate points at one subdivision level to points at the previous:

8

## Local subdivision matrix

We can write this as a recurrence relation in matrix form:

$$\begin{pmatrix} L^j \\ C^j \\ R^j \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/8 & 3/4 & 1/8 \\ 0 & 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} L^{j-1} \\ C^{j-1} \\ R^{j-1} \end{pmatrix}$$

$$Q^j = S Q^{j-1}$$

Where the  $L, R, C$ 's are (for convenience) *row vectors* and  $S$  is the **local subdivision matrix**.

We can expand these row vectors :

$$\begin{pmatrix} L_x^j & L_y^j \\ C_x^j & C_y^j \\ R_x^j & R_y^j \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/8 & 3/4 & 1/8 \\ 0 & 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} L_x^{j-1} & L_y^{j-1} \\ C_x^{j-1} & C_y^{j-1} \\ R_x^{j-1} & R_y^{j-1} \end{pmatrix}$$

and now think in terms of the behaviors of the  $x$  and  $y$  components, treating them as column vectors:

$$\begin{bmatrix} X^j & Y^j \end{bmatrix} = S \begin{bmatrix} X^{j-1} & Y^{j-1} \end{bmatrix}$$

9

## Eigenvectors and eigenvalues

To solve this problem, we need to look at the eigenvectors and eigenvalues of  $S$ . First, a review...

Let  $v$  be a vector such that:

$$Sv = \lambda v$$

We say that  $v$  is an eigenvector with eigenvalue  $\lambda$ .

An  $n \times n$  matrix can have  $n$  eigenvalues and eigenvectors:

$$Sv_1 = \lambda_1 v_1$$

$\vdots$

$$Sv_n = \lambda_n v_n$$

If the eigenvectors are linearly independent (which means that  $S$  is *non-defective*), then they form a basis, and we can re-write  $X$  in terms of the eigenvectors:

$$X = \sum_i^n a_i v_i$$

11

## Local subdivision matrix, cont'd

Let's focus on just the behavior of the  $x$  components:

$$\begin{pmatrix} L_x^j \\ C_x^j \\ R_x^j \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/8 & 3/4 & 1/8 \\ 0 & 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} L_x^{j-1} \\ C_x^{j-1} \\ R_x^{j-1} \end{pmatrix}$$

$$X^j = S X^{j-1}$$

(The analysis of the  $y$  component will be the same.)

Tracking the  $x$  components through subdivision:

$$X^j = S X^{j-1} = S \cdot S X^{j-2} = S \cdot S \cdot S X^{j-3} = \dots = S^j X^0$$

The limit position of the  $x$ 's is then:

$$X^\infty = \lim_{j \rightarrow \infty} S^j X^0$$

OK, so how do we apply a matrix an infinite number of times??

10

## To infinity, but not beyond...

Now let's apply the matrix to the vector  $X$ :

$$X^1 = S X^0 = S \sum_i^n a_i v_i = \sum_i^n a_i S v_i = \sum_i^n a_i \lambda_i v_i$$

Applying it  $j$  times:

$$X^j = S^j X = S^j \sum_i^n a_i v_i = \sum_i^n a_i S^j v_i = \sum_i^n a_i \lambda_i^j v_i$$

Let's assume the eigenvalues are non-negative and sorted so that:

$$\lambda_1 > \lambda_2 > \lambda_3 \geq \dots \geq \lambda_n \geq 0$$

(The form of the inequalities is important.) Now let  $j$  go to infinity:

$$X^\infty = \lim_{j \rightarrow \infty} S^j X^0 = \lim_{j \rightarrow \infty} \sum_i^n a_i \lambda_i^j v_i$$

If  $\lambda_1 > 1$ , then:

If  $\lambda_1 < 1$ , then:

If  $\lambda_1 = 1$ , then:

12

## Evaluation masks

What are the eigenvalues and eigenvectors of our cubic B-spline subdivision matrix?

$$\begin{aligned}\lambda_1 &= 1 & \lambda_2 &= \frac{1}{2} & \lambda_3 &= \frac{1}{4} \\ \mathbf{v}_1 &= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} & \mathbf{v}_2 &= \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} & \mathbf{v}_3 &= \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix}\end{aligned}$$

We're OK! (In fact, for proper subdivision matrices, the first eigenvector will always be  $[1 \dots 1]^T$ . Why?)

So, the first thing we do is expand the  $x$  and  $y$  components of a vertex in terms of the eigenbasis:

$$X = \sum_i^n a_i \mathbf{v}_i \quad Y = \sum_i^n b_i \mathbf{v}_i$$

Then, after infinite subdivision, the  $x$  components will end up at...?

What about the  $y$ -coordinates?

13

## Evaluation masks, cont'd

Note that we need not start with the 0<sup>th</sup> level control points and push them to the limit.

If we subdivide and average the control polygon  $j$  times, we can push the vertices of the refined polygon to the limit as well:

$$X^\infty = \mathbf{S}^\infty X^j = \mathbf{u}_1^T X^j$$

The same result obtains for the  $y$ -coordinate:

$$Y^\infty = \mathbf{S}^\infty Y^j = \mathbf{u}_1^T Y^j$$

## Evaluation masks, cont'd

To finish up, we need to compute  $a_1$ . First, we can reorganize the expansion of  $X$  into the eigenbasis:

$$X^0 = a_1 \mathbf{v}_1 + a_2 \mathbf{v}_2 + \dots + a_n \mathbf{v}_n = \begin{bmatrix} \vdots & \vdots & & \vdots \\ \mathbf{v}_1 & \mathbf{v}_2 & \cdots & \mathbf{v}_n \\ \vdots & \vdots & & \vdots \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \mathbf{V} \mathbf{A}$$

We can then solve for the coefficients in this new basis:

$$\mathbf{A} = \mathbf{V}^{-1} X^0$$

$$\begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} \dots & \mathbf{u}_1^T & \dots \\ \dots & \mathbf{u}_2^T & \dots \\ \vdots & & \vdots \\ \dots & \mathbf{u}_n^T & \dots \end{bmatrix} X^0$$

Now we can compute the limit position of the  $x$ -coordinate:

$$x_C^\infty = a_1 = \mathbf{u}_1^T X^0$$

We call  $\mathbf{u}_1$  the **evaluation mask**.

14

## Left eigenvectors

What are these  $u$ -vectors? Consider the eigenvector relation:

$$\mathbf{S} \mathbf{v}_i = \lambda_i \mathbf{v}_i$$

We can re-write this as a matrix (we'll use the 3x3 case for illustration here):

$$\begin{aligned}\mathbf{S} [\mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3] = [\lambda_1 \mathbf{v}_1 & \lambda_2 \mathbf{v}_2 & \lambda_3 \mathbf{v}_3] \\ \mathbf{S} [\mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3] &= [\mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3] \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \\ \mathbf{S} \mathbf{V} &= \mathbf{V} \Lambda\end{aligned}$$

where  $\mathbf{V}$  is the concatenation of the eigenvectors into a matrix and  $\Lambda$  is a diagonal matrix filled with the eigenvalues of  $\mathbf{S}$ .

15

16

## Left eigenvectors (cont'd)

Now let's multiply both sides by  $\mathbf{V}^{-1}$  from the left and right and then simplify:

$$\mathbf{V}^{-1}(\mathbf{S}\mathbf{V})\mathbf{V}^{-1} = \mathbf{V}^{-1}(\mathbf{V}\Lambda)\mathbf{V}^{-1}$$

$$\mathbf{V}^{-1}\mathbf{S} = \Lambda\mathbf{V}^{-1}$$

$$\mathbf{U}\mathbf{S} = \Lambda\mathbf{U}$$

where  $\mathbf{U} \equiv \mathbf{V}^{-1}$ . If we "de-construct" this relation, we get:

$$\begin{aligned} \mathbf{U}\mathbf{S} &= \Lambda\mathbf{U} \\ \begin{bmatrix} \mathbf{u}_1^T \\ \mathbf{u}_2^T \\ \mathbf{u}_3^T \end{bmatrix} \mathbf{S} &= \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \begin{bmatrix} \mathbf{u}_1^T \\ \mathbf{u}_2^T \\ \mathbf{u}_3^T \end{bmatrix} \\ \begin{bmatrix} \mathbf{u}_1^T \\ \mathbf{u}_2^T \\ \mathbf{u}_3^T \end{bmatrix} \mathbf{S} &= \begin{bmatrix} \lambda_1 \mathbf{u}_1^T \\ \lambda_2 \mathbf{u}_2^T \\ \lambda_3 \mathbf{u}_3^T \end{bmatrix} \end{aligned}$$

Thus, we find that the  $u$ -vectors obey the relation:

$$\mathbf{u}_i^T \mathbf{S} = \lambda_i \mathbf{u}_i^T$$

These are the "**left eigenvectors**" of  $\mathbf{S}$ . (Alternatively, they are the eigenvectors of  $\mathbf{S}^T$ .)

17

18

## Tangent analysis

What is the tangent to the cubic B-spline curve?

First, let's consider how we represent the  $x$  and  $y$  coordinate neighborhoods:

$$X^0 = a_1 \mathbf{v}_1 + a_2 \mathbf{v}_2 + a_3 \mathbf{v}_3$$

$$Y^0 = b_1 \mathbf{v}_1 + b_2 \mathbf{v}_2 + b_3 \mathbf{v}_3$$

We can view the point neighborhoods then as:

$$\begin{aligned} \mathbf{Q}^0 &= [X^0 \quad Y^0] = [a_1 \mathbf{v}_1 \quad b_1 \mathbf{v}_1] + [a_2 \mathbf{v}_2 \quad b_2 \mathbf{v}_2] + [a_3 \mathbf{v}_3 \quad b_3 \mathbf{v}_3] \\ &= \mathbf{v}_1 [a_1 \quad b_1] + \mathbf{v}_2 [a_2 \quad b_2] + \mathbf{v}_3 [a_3 \quad b_3] \end{aligned}$$

After  $j$  subdivisions, we would get:

$$\begin{aligned} \mathbf{Q}^j &= S^j \{ \mathbf{v}_1 [a_1 \quad b_1] + \mathbf{v}_2 [a_2 \quad b_2] + \mathbf{v}_3 [a_3 \quad b_3] \} \\ &= \lambda_1^j \mathbf{v}_1 [a_1 \quad b_1] + \lambda_2^j \mathbf{v}_2 [a_2 \quad b_2] + \lambda_3^j \mathbf{v}_3 [a_3 \quad b_3] \end{aligned}$$

We can write this more explicitly as:

$$\begin{bmatrix} L^j \\ C^j \\ R^j \end{bmatrix} = \lambda_1^j \begin{bmatrix} \mathbf{v}_{1,L} \\ \mathbf{v}_{1,C} \\ \mathbf{v}_{1,R} \end{bmatrix} [a_1 \quad b_1] + \lambda_2^j \begin{bmatrix} \mathbf{v}_{2,L} \\ \mathbf{v}_{2,C} \\ \mathbf{v}_{2,R} \end{bmatrix} [a_2 \quad b_2] + \lambda_3^j \begin{bmatrix} \mathbf{v}_{3,L} \\ \mathbf{v}_{3,C} \\ \mathbf{v}_{3,R} \end{bmatrix} [a_3 \quad b_3]$$

## Recipe for subdivision curves

The evaluation mask for the cubic B-spline is:

$$\begin{pmatrix} 1 & 2 & 1 \\ 6 & 3 & 6 \end{pmatrix}$$

Now we can cook up a simple procedure for creating subdivision curves:

- Subdivide (split+average) the control polygon a few times. Use the averaging mask.
- Push the resulting points to the limit positions. Use the evaluation mask.

## Tangent analysis (cont'd)

The tangent to the curve is along the direction:

$$\mathbf{t} = \lim_{j \rightarrow \infty} (R^j - C^j)$$

What's wrong with this definition?

Instead, we'll find the *normalized* tangent direction :

$$\mathbf{t} = \lim_{j \rightarrow \infty} \frac{R^j - C^j}{\|R^j - C^j\|}$$

Now, let's look at the "right" and "center" points in isolation:

$$\begin{aligned} R^j &= \lambda_1^j \mathbf{v}_{1,R} [a_1 \quad b_1] + \lambda_2^j \mathbf{v}_{2,R} [a_2 \quad b_2] + \lambda_3^j \mathbf{v}_{3,R} [a_3 \quad b_3] \\ C^j &= \lambda_1^j \mathbf{v}_{1,C} [a_1 \quad b_1] + \lambda_2^j \mathbf{v}_{2,C} [a_2 \quad b_2] + \lambda_3^j \mathbf{v}_{3,C} [a_3 \quad b_3] \end{aligned}$$

The difference between these is:

$$\begin{aligned} R^j - C^j &= \lambda_1^j (\mathbf{v}_{1,R} - \mathbf{v}_{1,C}) [a_1 \quad b_1] + \\ &\quad \lambda_2^j (\mathbf{v}_{2,R} - \mathbf{v}_{2,C}) [a_2 \quad b_2] + \lambda_3^j (\mathbf{v}_{3,R} - \mathbf{v}_{3,C}) [a_3 \quad b_3] \\ &= \lambda_2^j (\mathbf{v}_{2,R} - \mathbf{v}_{2,C}) [a_2 \quad b_2] + \lambda_3^j (\mathbf{v}_{3,R} - \mathbf{v}_{3,C}) [a_3 \quad b_3] \end{aligned}$$

19

20

## The tangent mask

And now computing the tangent:

$$\begin{aligned}
 \lim_{j \rightarrow \infty} \frac{R^j - C^j}{\|R^j - C^j\|} &= \lim_{j \rightarrow \infty} \frac{\lambda_2^j (v_{2,R} - v_{2,C}) [a_2 \ b_2] + \lambda_3^j (v_{3,R} - v_{3,C}) [a_3 \ b_3]}{\|\lambda_2^j (v_{2,R} - v_{2,C}) [a_2 \ b_2] + \lambda_3^j (v_{3,R} - v_{3,C}) [a_3 \ b_3]\|} \\
 &= \lim_{j \rightarrow \infty} \frac{(v_{2,R} - v_{2,C}) [a_2 \ b_2] + \left(\frac{\lambda_3}{\lambda_2}\right)^j (v_{3,R} - v_{3,C}) [a_3 \ b_3]}{\left\| (v_{2,R} - v_{2,C}) [a_2 \ b_2] + \left(\frac{\lambda_3}{\lambda_2}\right)^j (v_{3,R} - v_{3,C}) [a_3 \ b_3] \right\|} \\
 &= \frac{(v_{2,R} - v_{2,C}) [a_2 \ b_2]}{\|(v_{2,R} - v_{2,C}) [a_2 \ b_2]\|} \\
 &= \frac{[a_2 \ b_2]}{\|[a_2 \ b_2]\|} \\
 &= \frac{[\mathbf{u}_2^T X^0 \ \mathbf{u}_2^T Y^0]}{\|[\mathbf{u}_2^T X^0 \ \mathbf{u}_2^T Y^0]\|} \\
 &= \frac{\mathbf{u}_2^T \mathbf{Q}^0}{\|\mathbf{u}_2^T \mathbf{Q}^0\|}
 \end{aligned}$$

Thus, we can compute the tangent using the *second* left eigenvector! This analysis holds for general subdivision curves and gives us the **tangent mask**.

21

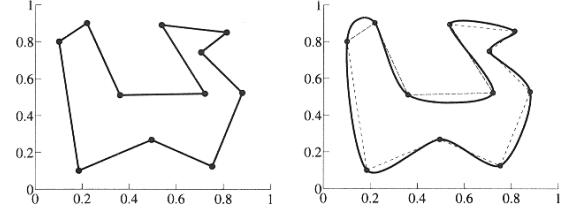
## DLG interpolating scheme (1987)

Slight modification to subdivision algorithm:

- splitting step introduces midpoints
- averaging step *only changes midpoints*

For DLG (Dyn-Levin-Gregory), use:

$$r = \frac{1}{16}(-2, 5, 10, 5, -2)$$



Since we are only changing the midpoints, the points after the averaging step do not move.

22