

Computer Aided Geometric Design

Shape Approximation

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based on a book by Cohen, Riesenfeld, & Elber

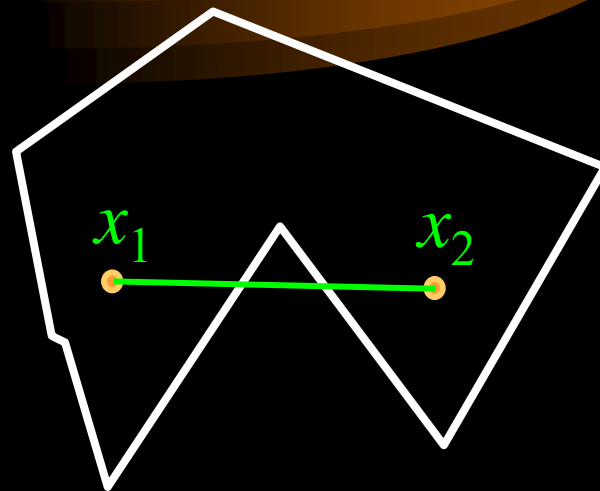
Shape Approximation



- **Point Precision** - Should we require that certain points be interpolated?
- **Continuity** - What level of continuity should we desire (smoothness?)
- **Closeness** - How could we measure the fidelity of a curve “close” to points?
- **Good Fit** - How does one decide what is a good fit?

Definition 1.35

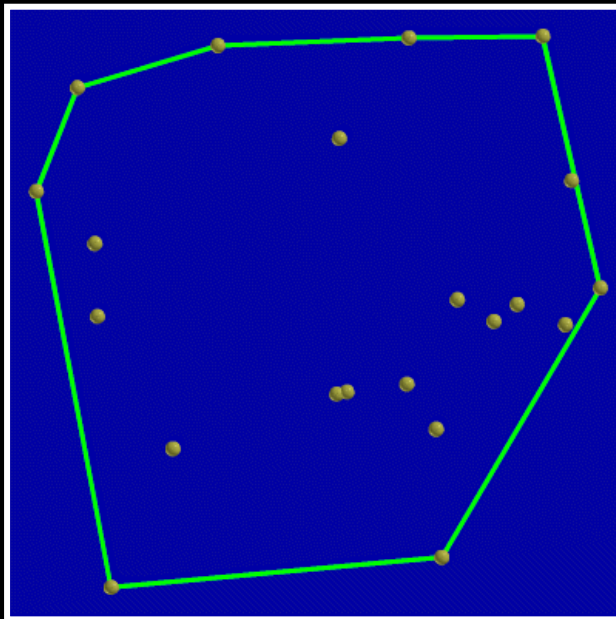
A subset of \mathbf{R}^3 , X is called convex if for all $x_1, x_2 \in X$,
 $(1-t)x_1 + tx_2 \in X$, $t \in [0, 1]$.



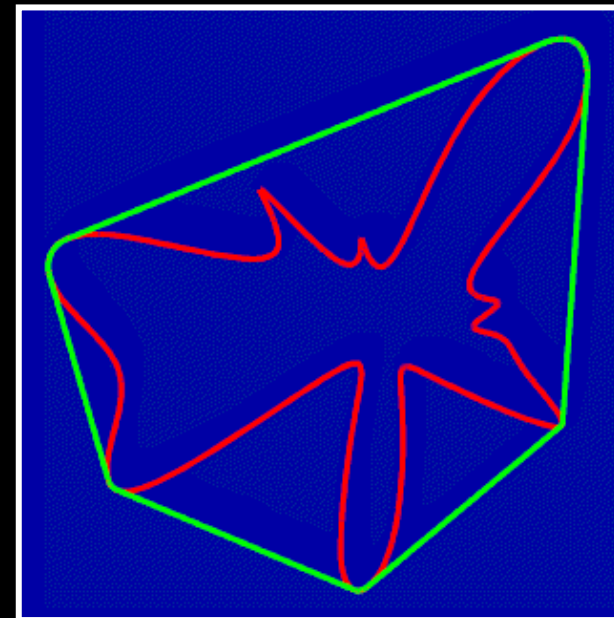
That is, the line segment connecting x_1 and x_2 lies entirely within the set X .

Definition 1.36

The convex hull of a set X is the smallest convex set containing X .



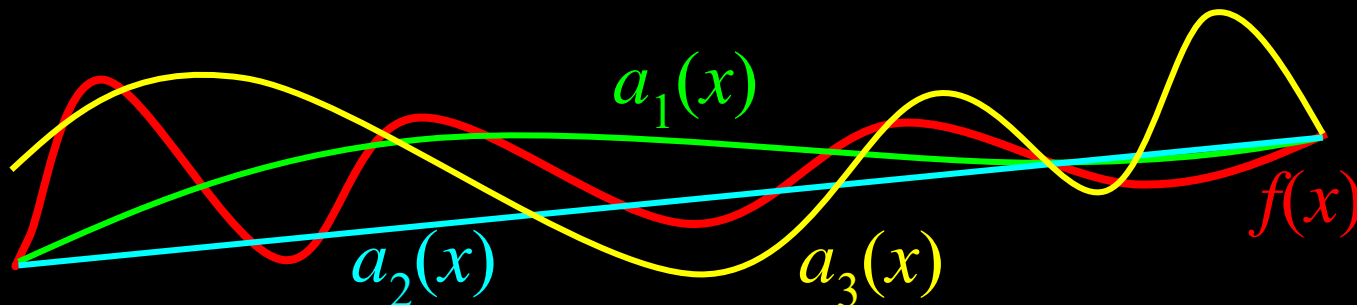
Convex
Hull of
Points
(left)



**Closed
Curve**
(right)

Definition 2.19

We introduce the notion of **variation diminishing**.
If $a(x)$ is an approximation to a function $f(x)$, a is called **variation diminishing** if a straight line can intersect $a(x)$ no more often than it intersects $f(x)$.



$\pi = 2?$

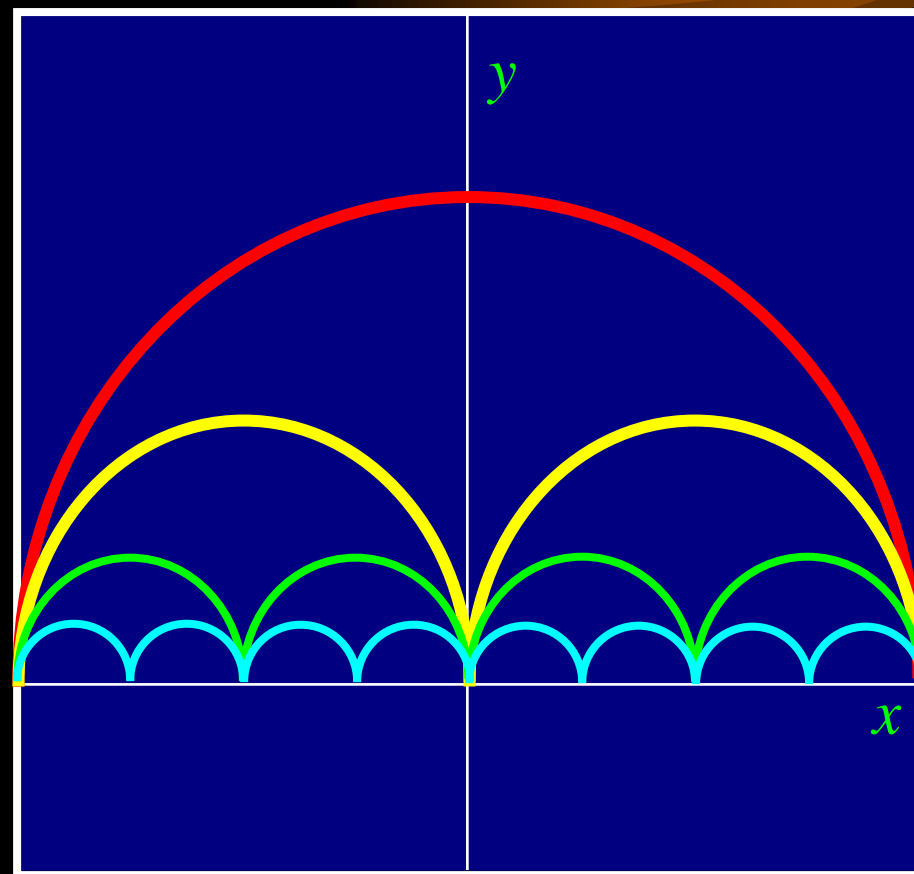
Assume an arc
of radius $r = 1$.

$$\text{Length} = \pi r = \pi$$

$$\text{Length} = 2 \pi(r/2) = \pi$$

$$\text{Length} = 4 \pi(r/4) = \pi$$

$$\text{Length} = 8 \pi(r/8) = \pi$$



Definition 1.29



A function $f(x)$ is called **increasing** on an interval (c,d) if for all $u, v \in (c,d)$, $u \leq v$ implies $f(u) \leq f(v)$.

Definition 1.30

A function $f(x)$ is called **decreasing** on an interval (c,d) if for all $u, v \in (c,d)$, $u \leq v$ implies $f(u) \geq f(v)$.

Theorem 1.31

Suppose $f(x) \in C^{(1)}(c, d)$.

If $f'(x) > 0$ for $x \in (c, d)$, then $f(x)$ is increasing on (c, d) .

If $f'(x) < 0$ for $x \in (c, d)$, then $f(x)$ is decreasing on (c, d) .

Definition 1.32

A **(local) maximum** for a function $f \in C^0$ occurs at a point x_0 if there exists $\varepsilon > 0$ so that $f(x_0) \geq f(x)$ for all $x \neq x_0$ such that $|x - x_0| < \varepsilon$.

A **(local) minimum** to f is defined analogously.

Definition 1.33

The **external points** of a function are the points of the graph at which **maxima** or **minima** occur.

Lemma 1.34

Suppose a function f is piecewise $C^{(1)}$. The extremal points of a function f might occur for only the following values of x :

- $x = a$ and $x = b$, that is the interval endpoints,
- values of x for which $f'(x) = 0$,
- values of x for which $f'(x)$ does not exist.

Definition 1.37

A function $f(x)$ is called **convex** on (c,d) if for all

$$u, v \in [c, d], \quad f\left(\frac{u+v}{2}\right) \leq \frac{f(u) + f(v)}{2}.$$

Definition 1.38

A function $f(x)$ is called **concave** on (c,d) if for all

$$u, v \in (c, d), \quad f\left(\frac{u+v}{2}\right) \geq \frac{f(u) + f(v)}{2}.$$

Theorem 1.39



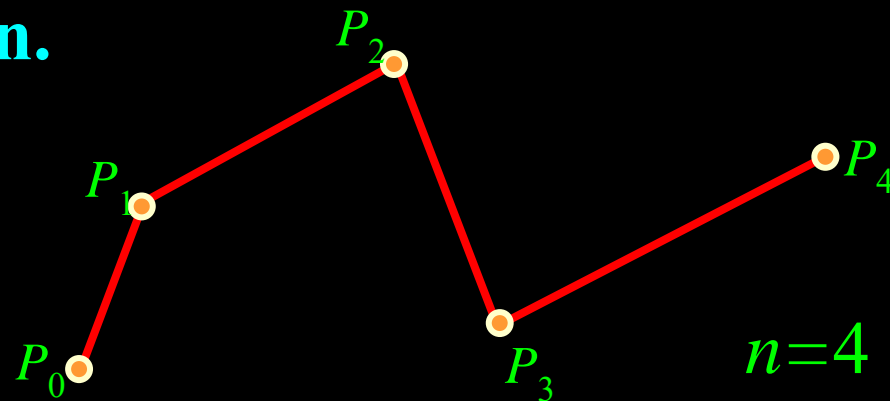
Suppose $f \in C^2$.

If $f''(x) > 0$ for $x \in (c, d)$, then $f(x)$ is **convex** on (c, d) .

If $f''(x) < 0$ for $x \in (c, d)$, then $f(x)$ is **concave** on (c, d) .

Constructive Bezier Curve Algorithm

Consider the $n+1$ points P_0, \dots, P_n and connect the points into a polyline that we will denote hereafter as the **control polygon**.



Given points $P_i, i = 0, \dots, n$, our goal is to determine a curve $\gamma(t)$, for all values $t \in [0, 1]$.

Algorithm 5.1

(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

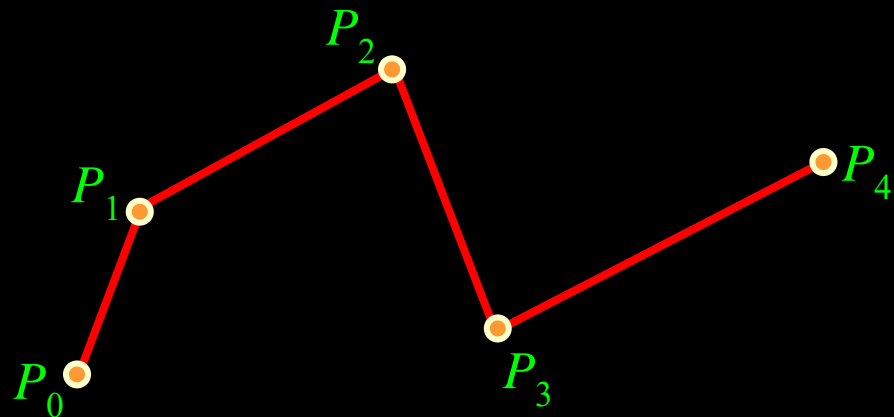
Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 0, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$, for $i = j, \dots, n$.

Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = \frac{1}{2}$$



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(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

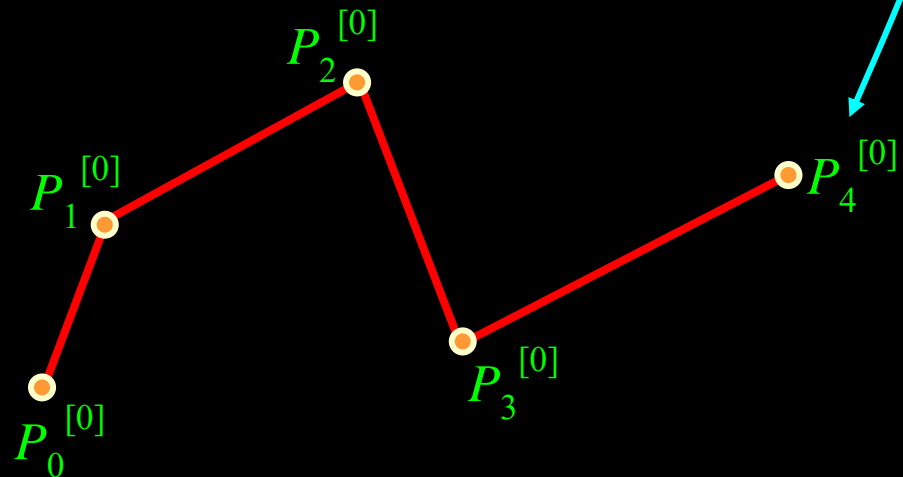
Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 1, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$,
for $i = j, \dots, n$.

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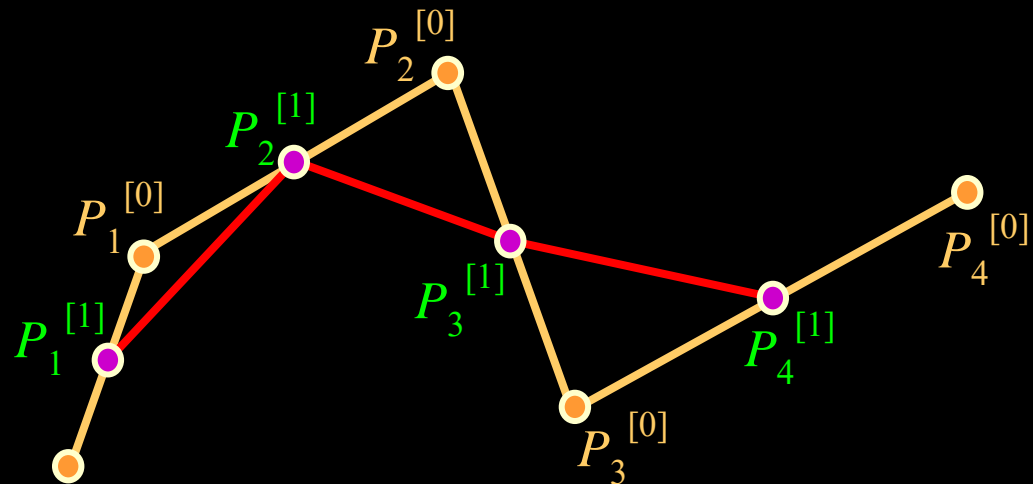
Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = \frac{1}{2}$$

$$j = 1$$

$$P_i^{[1]} = \frac{1}{2}P_{i-1}^{[0]} + \frac{1}{2}P_i^{[0]}, i = 1, \dots, 4$$



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(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 1, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$,
for $i = j, \dots, n$.

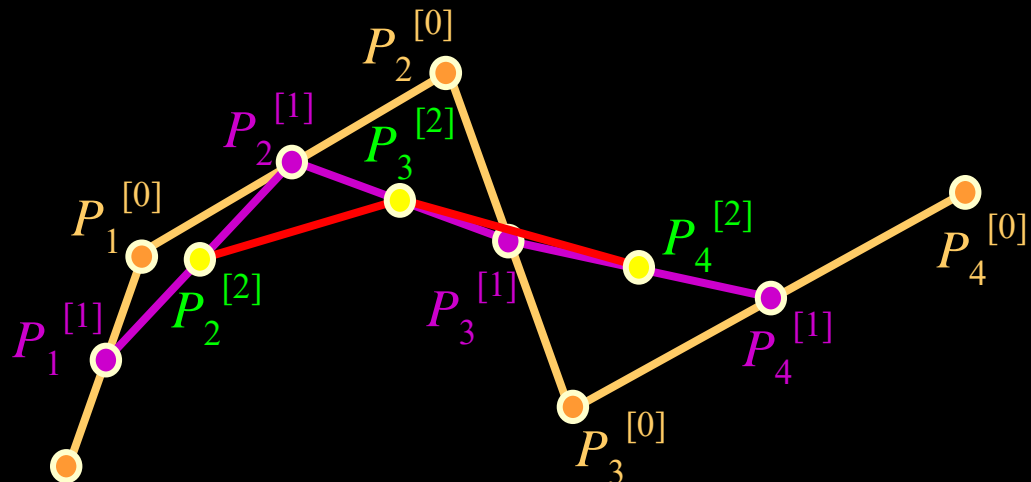
Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = \frac{1}{2}$$

$$j = 2$$

$$P_i^{[2]} = \frac{1}{2}P_{i-1}^{[1]} + \frac{1}{2}P_i^{[1]}, i = 2, \dots, 4$$



Algorithm 5.1

(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 1, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$,
for $i = j, \dots, n$.

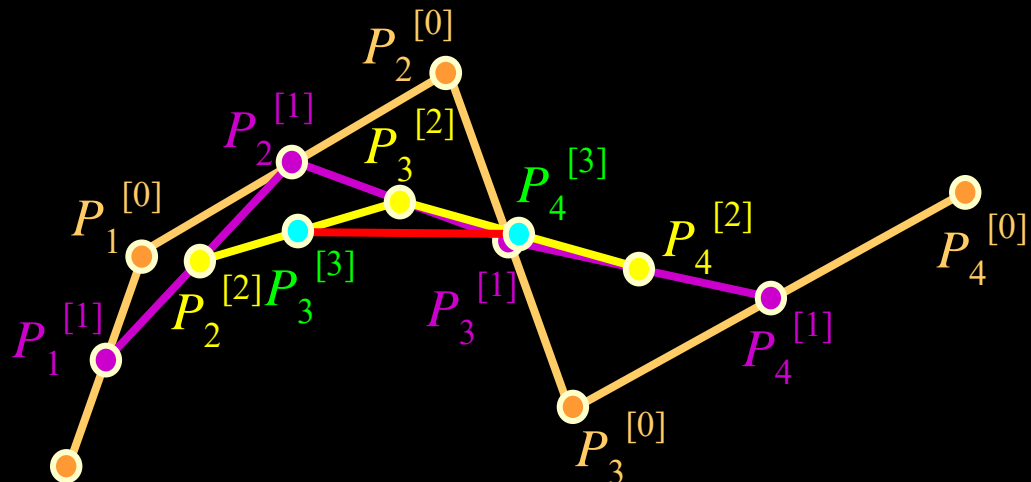
Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = \frac{1}{2}$$

$$j = 3$$

$$P_i^{[3]} = \frac{1}{2}P_{i-1}^{[2]} + \frac{1}{2}P_i^{[2]}, \quad i = 3, \dots, 4$$



Algorithm 5.1

(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 1, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$,
for $i = j, \dots, n$.

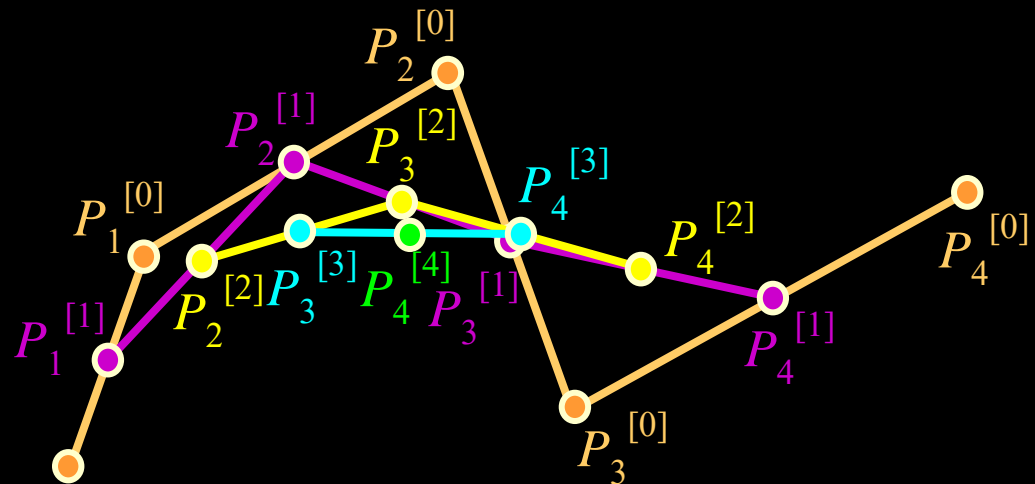
Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = \frac{1}{2}$$

$$j = 4$$

$$P_i^{[4]} = \frac{1}{2}P_{i-1}^{[3]} + \frac{1}{2}P_i^{[3]}, \quad i = 4, \dots, 4$$



Algorithm 5.1

(Constructive Bezier Curve Algorithm)

Step 1: Select a value $t \in [0,1]$. This value remains constant for the rest of the steps.

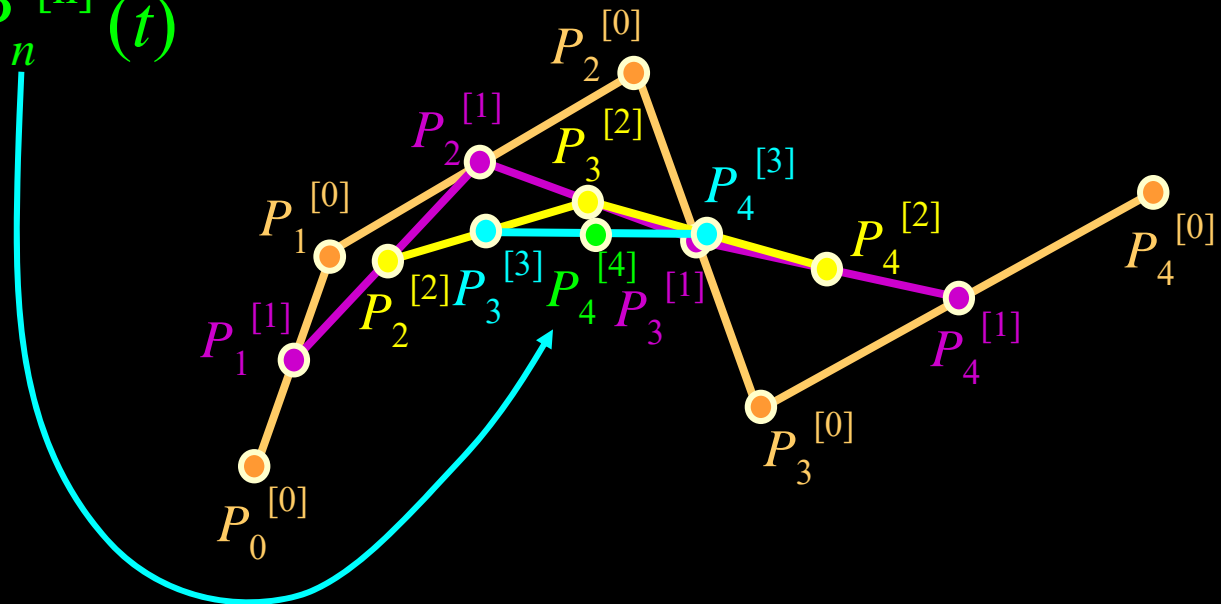
Step 2: Set $P_i^{[0]}(t) = P_i$, for $i = 0, \dots, n$.

Step 3: For $j = 1, \dots, n$, set $P_i^{[j]}(t) = (1-t)P_{i-1}^{[j-1]}(t) + tP_i^{[j-1]}(t)$, for $i = j, \dots, n$.

Step 4: $\gamma(t) = P_n^{[n]}(t)$

$$n = 4$$

$$t = 1/2$$



Properties of Bezier Curve $\gamma(t)$

- $\gamma(t)$ is in the convex hull of $P_i, i = 0, \dots, n$.
- $\gamma(0) = P_0$ and $\gamma(1) = P_n$.
- $\gamma(t)$ looks tangent to the first segment at $t = 0$ and tangent to the last segment at $t = 1$.
- γ is a variation diminishing curve of the control polygon.
- γ seems to follow the general shape of the control polygon.
- (Re)moving P_i , the new shape is easily anticipated.

Different View of Algorithm 5.1

$$\begin{bmatrix} (1-t) & t \end{bmatrix} \begin{bmatrix} P_0 & P_1 & \dots & P_{n-1} \\ P_1 & P_2 & \dots & P_n \end{bmatrix} = \begin{bmatrix} P_1^{[1]} & P_2^{[1]} & \dots & P_n^{[1]} \end{bmatrix}$$

$$\begin{bmatrix} (1-t) & t \end{bmatrix} \begin{bmatrix} P_1^{[1]} & P_2^{[1]} & \dots & P_{n-1}^{[1]} \\ P_2^{[1]} & P_3^{[1]} & \dots & P_n^{[1]} \end{bmatrix} = \begin{bmatrix} P_2^{[2]} & P_3^{[2]} & \dots & P_n^{[2]} \end{bmatrix}$$

⋮

$$\begin{bmatrix} (1-t) & t \end{bmatrix} \begin{bmatrix} P_{n-1}^{[n-1]} \\ P_n^{[n-1]} \end{bmatrix} = P_n^{[n]}.$$

Theorem 5.3

If $\{P_i\}$ is a control polygon, then

$$\gamma(t) = \sum_{i=0}^n P_i \frac{n!}{i!(n-i)!} (1-t)^{n-i} t^i,$$

where, as before, $\gamma(t)$ denotes the point on the Bezier curve constructed with Algorithm 5.1 with parameter value t .

Some Conventions

We use the following conventions:

$$\binom{n}{i} = \frac{n!}{i!(n-i)!} \quad \text{and} \quad \theta_{i,n}(t) = \binom{n}{i} (1-t)^{n-i} t^i,$$

or Theorem 5.3 states that $\gamma(t) = \sum_{i=0}^n P_i \theta_{i,n}(t)$.

Theorem 5.4

Before proving Theorem 5.3, we need to prove

$$\theta_{k,n}(t) = \begin{cases} 1, & \text{for } n = 0, \\ t\theta_{k-1,n-1}(t) + (1-t)\theta_{k,n-1}(t), & \text{for } 0 < k < n, \\ (1-t)\theta_{0,n-1}(t), & \text{for } n > 0, k = 0, \\ t\theta_{n-1,n-1}(t), & \text{for } n = k > 0. \end{cases}$$

Theorem 5.4 (Proof)

We prove only the general case, while the end conditions are left as an exercise.

Because $\binom{n}{k} = \frac{n!}{k!(n-k)!}$, we have

$$\begin{aligned}\binom{n}{k} &= \frac{n}{k} \frac{(n-1)!}{(k-1)!((n-1)-(k-1))!} = \frac{n}{k} \binom{n-1}{k-1} \\ &= \frac{n}{n-k} \frac{(n-1)!}{k!((n-1)-k)!} = \frac{n}{n-k} \binom{n-1}{k}.\end{aligned}$$

Theorem 5.4 (Proof Cont.)

For $0 < k < n$, we have

$$\begin{aligned}\theta_{k,n}(t) &= \binom{n}{k} t^k (1-t)^{n-k} \\ &= \frac{n}{k} \binom{n-1}{k-1} t^k (1-t)^{n-k} \\ &= \frac{n}{k} t \binom{n-1}{k-1} t^{k-1} (1-t)^{(n-1)-(k-1)} \\ &= \frac{n}{k} t \theta_{k-1,n-1}(t) \quad \text{or} \quad \frac{k}{n} \theta_{k,n}(t) = t \theta_{k-1,n-1}(t).\end{aligned}$$

Theorem 5.4 (Proof Cont.)

Similarly, for $0 < k < n$, we have

$$\begin{aligned}\theta_{k,n}(t) &= \binom{n}{k} t^k (1-t)^{n-k} \\ &= \frac{n}{n-k} \binom{n-1}{k} t^k (1-t)^{n-k} \\ &= \frac{n}{n-k} (1-t) \binom{n-1}{k} t^k (1-t)^{(n-1)-k} \\ &= \frac{n}{n-k} (1-t) \theta_{k,n-1}(t) \quad \text{or} \quad \frac{n-k}{n} \theta_{k,n}(t) = (1-t) \theta_{k,n-1}(t).\end{aligned}$$

Theorem 5.4 (Proof Cont.)

From $\frac{k}{n}\theta_{k,n}(t) = t\theta_{k-1,n-1}(t)$

and from $\frac{n-k}{n}\theta_{k,n}(t) = (1-t)\theta_{k,n-1}(t)$

we have:

$$\begin{aligned}\theta_{k,n}(t) &= \frac{k}{n}\theta_{k,n}(t) + \frac{n-k}{n}\theta_{k,n}(t) \\ &= t\theta_{k-1,n-1}(t) + (1-t)\theta_{k,n-1}(t).\end{aligned}$$

Theorem 5.3 (Proof)

Going back to Theorem 5.3, we denote by $\alpha(t)$ the right side of the result we seek,

$$\alpha(t) \stackrel{\text{def}}{=} \sum_{i=0}^n P_i \theta_{i,n}(t).$$

By Theorem 5.4 we can write,

$$\alpha(t) = \sum_{i=0}^n P_i \theta_{i,n}(t)$$

$$= P_0 \theta_{0,n}(t) + \sum_{i=1}^{n-1} P_i \theta_{i,n}(t) + P_n \theta_{n,n}(t)$$

$$= P_0 (1-t) \theta_{0,n-1}(t) + \sum_{i=1}^{n-1} P_i [t \theta_{i-1,n-1}(t) + (1-t) \theta_{i,n-1}(t)] + P_n t \theta_{n-1,n-1}(t)$$

$$\alpha(t) = P_0(1-t)\theta_{0,n-1}(t) + \sum_{i=1}^{n-1} P_i [t\theta_{i-1,n-1}(t) + (1-t)\theta_{i,n-1}(t)] + P_n t\theta_{n-1,n-1}(t)$$

Theorem 5.3 (Proof Cont.)

Regrouping,

$$\begin{aligned} \alpha(t) &= (1-t) \sum_{i=0}^{n-1} P_i \theta_{i,n-1}(t) + t \sum_{i=1}^n P_i \theta_{i-1,n-1}(t) \\ &= (1-t) \sum_{i=0}^{n-1} P_i \theta_{i,n-1}(t) + t \sum_{i=0}^{n-1} P_{i+1} \theta_{i,n-1}(t) \\ &= \sum_{i=0}^{n-1} [(1-t)P_i + tP_{i+1}] \theta_{i,n-1}(t) \\ &= \sum_{i=0}^{n-1} P_{i+1}^{[1]} \theta_{i,n-1}(t) \end{aligned}$$

$$\alpha(t) = \sum_{i=0}^{n-1} P_{i+1}^{[1]} \theta_{i,n-1}(t)$$

Theorem 5.3 (Proof Cont.)

Repeating the application of this procedure j times,

$$\alpha(t) = \sum_{i=0}^{n-j} P_{j+i}^{[j]} \theta_{i,n-j}(t),$$

and after n iterations, we will end up with

$$\alpha(t) = P_n^{[n]} \theta_{0,0}(t) = P_n^{[n]} = \gamma(t).$$

Thereby concluding the proof.

Definition 5.6



The functions

$$\theta_{i,n}(t) = \binom{n}{i} (1-t)^{n-i} t^i,$$

for $i = 0, \dots, n$, for any nonnegative integer, n , are called the **Bernstein basis functions** or the **Bernstein blending functions**.

Definition 5.7

For $t \in [a, b]$, the functions

$$\begin{aligned}\theta_{i,n}[t; a, b] &= \theta_{i,n}\left(\frac{t-a}{b-a}\right) \\ &= \binom{n}{i} \frac{(b-t)^{n-i} (t-a)^i}{(b-a)^n}\end{aligned}$$

for $i = 0, \dots, n$, for any nonnegative integer, n , are called the generalized Bernstein basis functions or the generalized Bernstein blending functions.

Properties of Bernstein Basis Functions



Theorem 5.8

$\theta_{i,n}(t) > 0$ for $t \in (0,1)$, for all n and $i = 0, \dots, n$.

Proof

For $t \in (0,1)$, $t > 0$ and $(1-t) > 0$. $\theta_{i,n}(t)$ is then just the product of n positive factors, and is hence positive.

Theorem 5.9

$$\sum_{i=0}^n \theta_{i,n}(t) \equiv 1, \text{ for } t \in [0,1].$$

Proof

This follows from the binomial theorem in which

$$(a+b)^n = \sum_{i=0}^n \binom{n}{i} a^{n-i} b^i.$$

Letting $a = (1-t)$ and $b = t$ we have

$$((1-t)+t)^n = 1^n = 1 = \sum_{i=0}^n \binom{n}{i} (1-t)^{n-i} t^i = \sum_{i=0}^n \theta_{i,n}(t).$$

Corollary

The Bezier curve lies in the convex hull of the points

$$\{P_i\}_{i=0}^n.$$

Proof

This follows immediately from Theorems 5.8 and 5.9 (convex combination).

Theorem



The Bezier curve, that is considered as an approximation to its control polygon, is **variation diminishing**.

Theorem 5.10



The Bernstein blending functions are **unimodal** with the maximum for the i th blending function of degree n at $t = i/n$.

Comment: A function is considered **unimodal** if it has one extremal point over the domain.

Theorem 5.10 (proof)



Consider $\theta_{0,n}(t) = (1-t)^n$ and $\theta_{n,n}(t) = t^n$. These functions are monotone on the interval $(0,1)$ and hence have one maximum point, at the endpoints.

Now consider that $\theta_{i,n}(t)$ for $i = 1, \dots, n-1$. $\theta_{i,n}(t)$ is continuous and the maximum of $\theta_{i,n}(t)$ can occur at $t = 0$, $t = 1$ or at $\theta'_{i,n}(t) = 0$.

Theorem 5.10 (Proof Cont.)

Consider the derivatives, $\theta'_{i,n}(t)$,

$$\begin{aligned}\theta'_{i,n}(t) &= \binom{n}{i} \left[it^{i-1} (1-t)^{n-i} - (n-i)t^i (1-t)^{n-1-i} \right] \\ &= \binom{n}{i} t^{i-1} (1-t)^{n-1-i} \left[i(1-t) - (n-i)t \right] \\ &= \binom{n}{i} t^{i-1} (1-t)^{n-1-i} \left[i - nt \right]\end{aligned}$$

Theorem 5.10 (Proof Cont.)

The zeros of the derivatives, $\theta_{i,n}'(t) = 0$, in the domain of $(0, 1)$ equal,

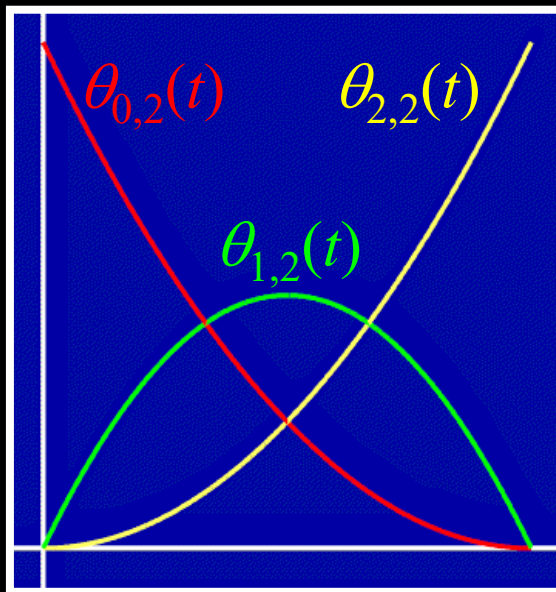
$$\theta_{i,n}'(t) = 0 = \binom{n}{i} t^{i-1} (1-t)^{n-1-i} [i - nt]$$

or

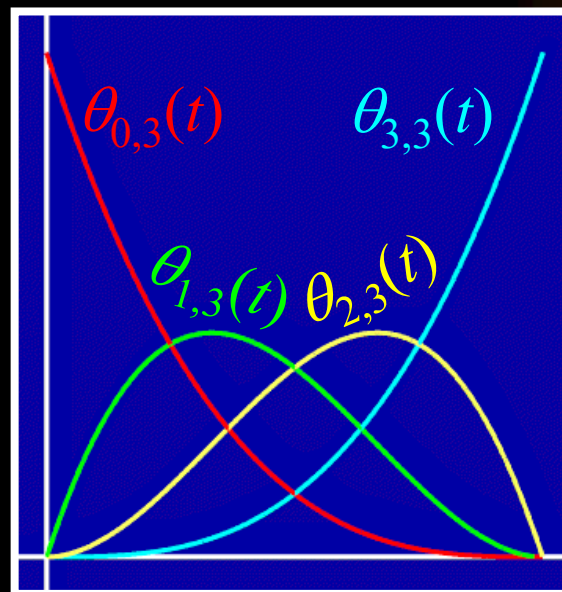
$$t_e = 0 \quad \text{or} \quad t_e = 1 \quad \text{or} \quad t_e = \frac{i}{n}.$$

Because $\theta_{i,n}(0) = \theta_{i,n}(1) = 0$ while being positive in $(0, 1)$, only one maximum can exist at $t = i/n$.

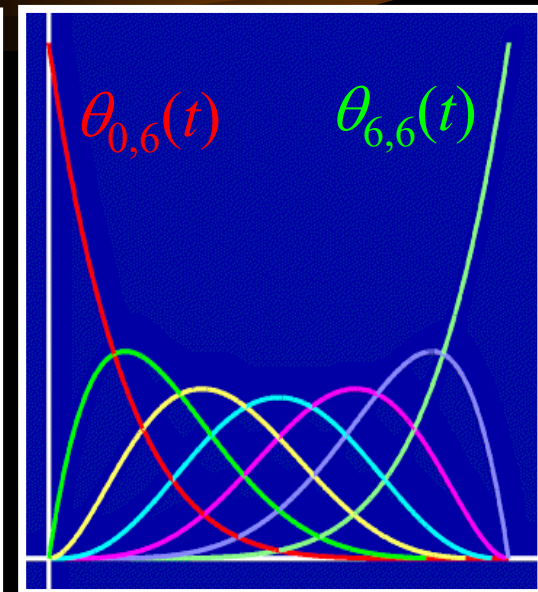
Examples of Basis Functions



Quadratic



Cubic



Degree 6

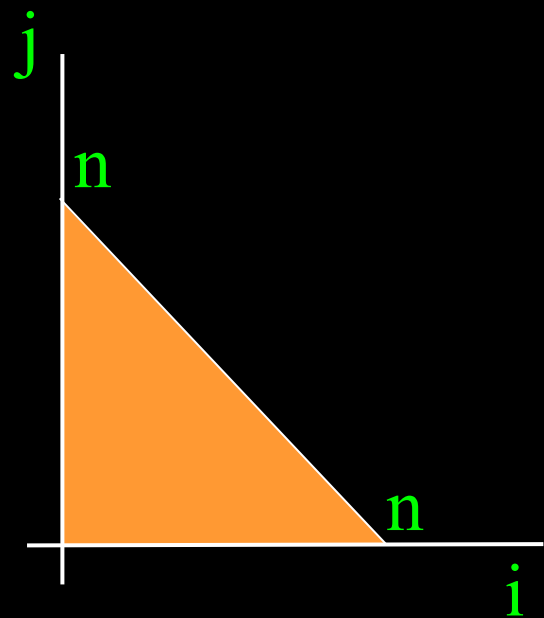
Theorem 5.12

The Bernstein blending functions of degree n form a basis for the polynomials of degree n .

Proof:
$$0 = \sum_{i=0}^n c_i \theta_{i,n}(t)$$
$$= \sum_{i=0}^n c_i \binom{n}{i} t^i (1-t)^{n-i}$$
$$= \sum_{i=0}^n c_i \binom{n}{i} t^i \sum_{j=0}^{n-i} \binom{n-i}{j} (-t)^{n-i-j}$$

Theorem 5.12 (Cont.)

Continuing,



$$\begin{aligned}
 0 &= \sum_{i=0}^n c_i \binom{n}{i} t^i \sum_{j=0}^{n-i} \binom{n-i}{j} (-t)^{n-i-j} \\
 &= \sum_{i=0}^n c_i \binom{n}{i} \sum_{j=0}^{n-i} (-1)^{n-i-j} \binom{n-i}{j} t^{n-j} \\
 &= \sum_{j=0}^n t^{n-j} \underbrace{\sum_{i=0}^{n-j} c_i (-1)^{n-i-j} \binom{n}{i} \binom{n-i}{j}}_{d_i \neq 0} \\
 &= \sum_{j=0}^n t^{n-j} \sum_{i=0}^{n-j} c_i d_i.
 \end{aligned}$$

Theorem 5.12 (Cont.)

Consider $0 = \sum_{j=0}^n t^{n-j} \sum_{i=0}^{n-j} c_i d_i, \quad d_i \neq 0.$

For $j=n$, it falls immediately that c_0 equals zero.

For $j=n-1$, it falls immediately that c_1 equals zero.

In a similar fashion, it is simple to show that all $c_i, i=0, \dots, n$, are zero.

Derivative Evaluation

Differentiating $\theta_{i,n}(t)$, one gets:

$$\theta'_{i,n}(t) = \binom{n}{i} \left[i t^{i-1} (1-t)^{n-i} - (n-i) t^i (1-t)^{n-i-1} \right]$$

Now
$$\binom{n}{i} i = \frac{n! i}{i! (n-i)!} = n \frac{(n-1)!}{(i-1)! (n-i)!} = n \binom{n-1}{i-1},$$

and
$$\binom{n}{i} (n-i) = \frac{n! (n-i)}{i! (n-i)!} = n \frac{(n-1)!}{i! (n-1-i)!} = n \binom{n-1}{i},$$

Derivative Evaluation (Cont.)

$$\begin{aligned} \text{or } \theta'_{i,n}(t) &= n \binom{n-1}{i-1} t^{i-1} (1-t)^{n-i} - n \binom{n-1}{i} t^i (1-t)^{n-i-1} \\ &= n(\theta_{i-1,n-1}(t) - \theta_{i,n-1}(t)). \end{aligned}$$

For $i=0$, $\theta_{0,n}(t)=(1-t)^n$ and we have

$$\theta'_{0,n}(t) = \binom{n}{0} (-n)(1-t)^{n-1} = -n \binom{n-1}{0} (1-t)^{n-1} = -n \theta_{0,n-1}(t),$$

and similarly for $i=n$.

Theorem 5.15



When $\theta_{i,n}(t)$ denotes the i th Bernstein blending function of degree n , then

$$\theta'_{i,n}(t) = n(\theta_{i-1,n-1}(t) - \theta_{i,n-1}(t)) \quad \text{for } i = 0, \dots, n,$$

where

$$\theta_{-1,n-1}(t) \equiv 0 \quad \text{and} \quad \theta_{n,n-1}(t) \equiv 0.$$

Corollary 5.16

If $\gamma(t) = \sum_{i=0}^n P_i \theta_{i,n}(t)$, then

$$\gamma'(t) = \sum_{i=0}^{n-1} Q_i \theta_{i,n-1}(t),$$

where

$$Q_i = n(P_{i+1} - P_i).$$

Corollary 5.17

If $\gamma(t) = \sum_{i=0}^n P_i \theta_{i,n}(t)$, then

$$\gamma^{(j)}(t) = \sum_{i=0}^{n-j} Q_i^{[j]} \theta_{i,n-j}(t),$$

where

$$Q_i^{[j+1]} = \begin{cases} P_i, & j = -1, \\ (n-j)[Q_{i+1}^{[j]} - Q_i^{[j]}], & j = 0, \dots, n-1. \end{cases}$$

The Hodograph

Given a **curve**, the graph of its derivative curve is also denoted the **Hodograph curve**.



Interpolation Using the Bernstein Blending Functions

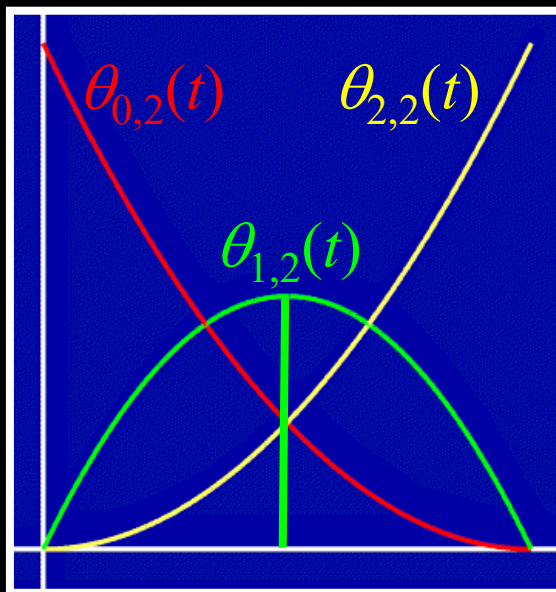
Question: Can we employ the Bezier/Bernstein polynomials for interpolation?

Question: Can we coerce Bezier curve $\gamma(t) = \sum P_i \theta_i(t)$ to interpolate certain locations? For example, $\gamma(i/n) = V_i$, $i=0, \dots, n$?

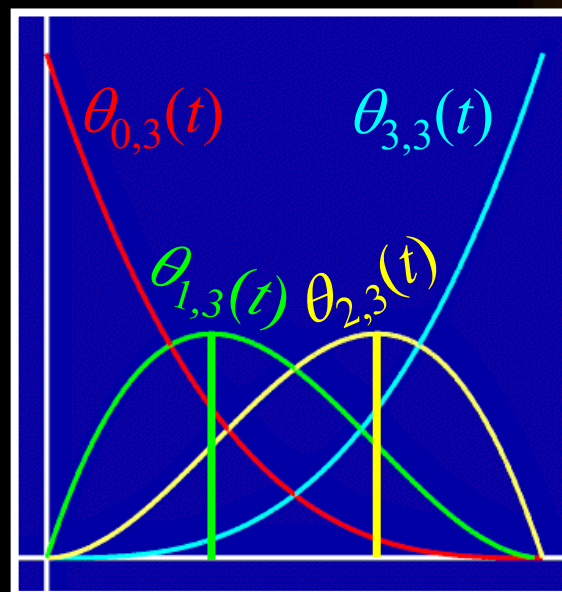
Points i/n , $i=0, \dots, n$ are known as node points.

Question: What is unique about the node points?

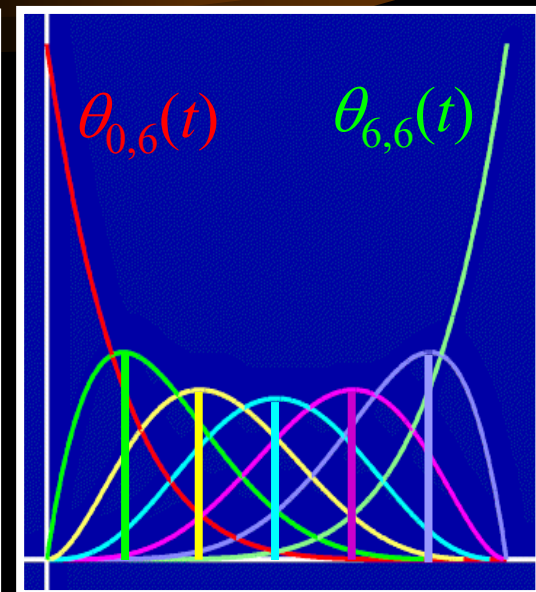
Nodal Points



Quadratic



Cubic



Degree 6

Interpolation Using the Bernstein Blending Functions

Impose the interpolation problem as a linear system of equations, $\Theta P = V$:

$$\left[\theta_{j,n}(i/n) \right] \left[P_0 \quad P_1 \quad \dots \quad P_n \right]^T = \left[V_0 \quad V_1 \quad \dots \quad V_n \right]^T$$

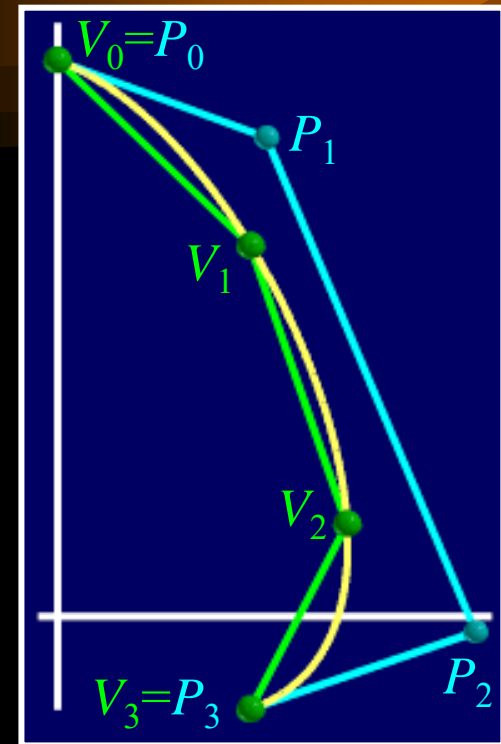
or,

$$\begin{bmatrix} \theta_{0,n}\left(\frac{0}{n}\right) & \theta_{1,n}\left(\frac{0}{n}\right) & \theta_{2,n}\left(\frac{0}{n}\right) & \dots & \theta_{n,n}\left(\frac{0}{n}\right) \\ \theta_{0,n}\left(\frac{1}{n}\right) & \theta_{1,n}\left(\frac{1}{n}\right) & \theta_{2,n}\left(\frac{1}{n}\right) & \dots & \theta_{n,n}\left(\frac{1}{n}\right) \\ \theta_{0,n}\left(\frac{2}{n}\right) & \theta_{1,n}\left(\frac{2}{n}\right) & \theta_{2,n}\left(\frac{2}{n}\right) & \dots & \theta_{n,n}\left(\frac{2}{n}\right) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \theta_{0,n}\left(\frac{n}{n}\right) & \theta_{1,n}\left(\frac{n}{n}\right) & \theta_{2,n}\left(\frac{n}{n}\right) & \dots & \theta_{n,n}\left(\frac{n}{n}\right) \end{bmatrix} \begin{bmatrix} P_0 \\ P_1 \\ P_2 \\ \vdots \\ P_n \end{bmatrix} = \begin{bmatrix} V_0 \\ V_1 \\ V_2 \\ \vdots \\ V_n \end{bmatrix}.$$

Interpolation Using the Bernstein Blending Functions (Example 5.30)

Consider a cubic Bezier interpolating the following four points, $V_0=(0,6)$, $V_1=(2,4)$, $V_2=(3,1)$, $V_3=(2,-1)$:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 8/27 & 4/9 & 2/9 & 1/27 \\ 1/27 & 2/9 & 4/9 & 8/27 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} P_0 \\ P_1 \\ P_2 \\ P_3 \end{bmatrix} = \begin{bmatrix} (0,6) \\ (2,4) \\ (3,1) \\ (2,-1) \end{bmatrix}$$



Question: What can one say about the solution's existence? About the stability (condition number) of Θ ?

Bernstein Approximation



Question: By sampling curve $f(x)$, $x \in [0,1]$ at the node points $f(i/n)=V_i$, $i=0,\dots,n$, can a Bezier curve reproduce

- Constant $f(x)$ functions?
- Linear $f(x)$ functions?
- Quadratic $f(x)$ function?
- Arbitrary order $f(x)$ functions?

Definition 5.19



For a function $f \in C^{(m)}[0,1]$, the n th Bernstein polynomial approximation is

$$B_n(f; x) = \sum_{k=0}^n f\left(\frac{k}{n}\right) \theta_{k,n}(x).$$

Definition 5.20



A sequence of functions $\{s_j(x)\}_j$ is said to converge uniformly to a function $s(x)$ on an interval I if for every $\varepsilon > 0$ there exists a natural number N such that, for all $j > N$, $|s_j(x) - s(x)| < \varepsilon$.

Theorem 5.21

If $f(x)$ is bounded on $[0,1]$, then

$$\lim_{n \rightarrow \infty} B_n(f; x) = f(x)$$

at any point x where f is continuous. If

$f \in C^{(m)}[0,1]$, then it converges uniformly. Further, if

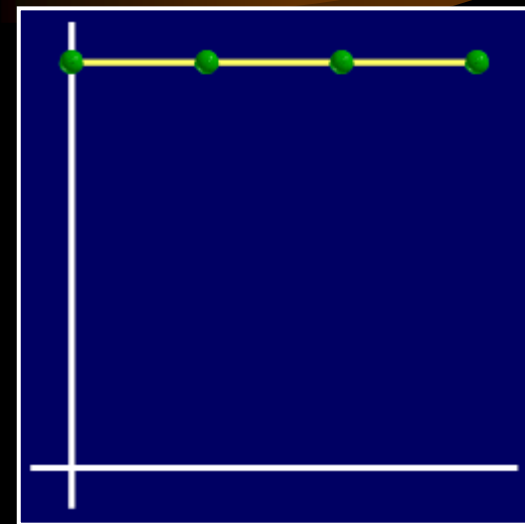
$$f \in C^{(m)}[0,1], \text{ then } \lim_{n \rightarrow \infty} B_n^{(m)}(f; x) = f^{(m)}(x)$$

uniformly on $[0,1]$.

Example 5.22 ($f(x) = 1$ Function)

If $f(x)$ is constant one, we have,

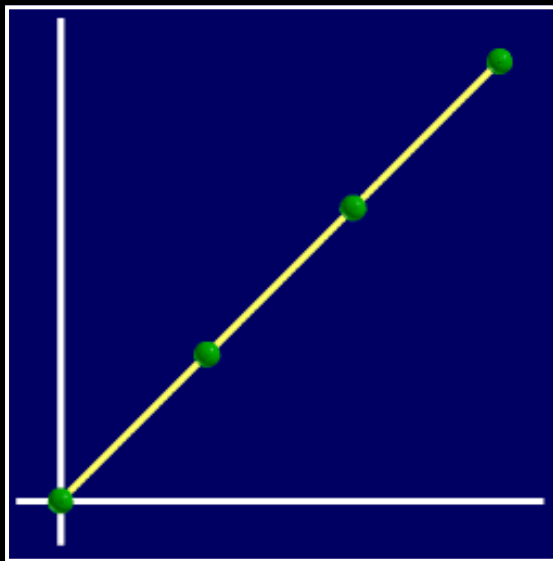
$$\begin{aligned} B_n(1; x) &= \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k} \\ &= (x + (1-x))^n \\ &= 1. \end{aligned}$$



In other words, the Bernstein approximation reproduces constants.

Example 5.23 (Linear $f(x)$)

If $f(x) = x$, we have
 $f(k/n) = k/n$ and,



reproducing linear functions.

$$\begin{aligned}
 B_n(x; x) &= \sum_{k=0}^n \frac{k}{n} \binom{n}{k} x^k (1-x)^{n-k} \\
 &= \sum_{k=1}^n \frac{k}{n} \binom{n}{k} x^k (1-x)^{n-k} \\
 &= \sum_{k=1}^n \binom{n-1}{k-1} x^k (1-x)^{n-k} \\
 &= x \sum_{k=1}^n \binom{n-1}{k-1} x^{k-1} (1-x)^{(n-1)-(k-1)} \\
 &= x \sum_{k=0}^{n-1} \binom{n-1}{k} x^k (1-x)^{(n-1)-k} = x.
 \end{aligned}$$

Example 5.24 (Quadratic $f(x)$)

If $f(x)$ is quadratic, assume $f(x) = x^2$ and we have,

$$B_n(x^2; x) = \sum_{k=0}^n \left(\frac{k}{n}\right)^2 \binom{n}{k} x^k (1-x)^{n-k}$$

In contrast, assume $n \geq 2$,

$$x^2 = x^2 \sum_{r=0}^{n-2} \binom{n-2}{r} x^r (1-x)^{n-2-r} = \sum_{r=0}^{n-2} \binom{n-2}{r} x^{r+2} (1-x)^{n-2-r}$$

which yields,

Quadratic $f(x)$ (Cont.)

$$\begin{aligned}x^2 &= \sum_{r=0}^{n-2} \binom{n-2}{r} x^{r+2} (1-x)^{n-2-r} \\&= \sum_{r=0}^{n-2} \frac{(r+2)(r+1)}{n(n-1)} \binom{n}{r+2} x^{r+2} (1-x)^{n-(r+2)} \\&= \sum_{k=2}^n \frac{k(k-1)}{n(n-1)} \binom{n}{k} x^k (1-x)^{n-k} \\&= \sum_{k=0}^n \frac{k(k-1)}{n(n-1)} \binom{n}{k} x^k (1-x)^{n-k}\end{aligned}$$

Quadratic $f(x)$ (Cont.)

So we have,

$$B_n(x^2; x) = \sum_{k=0}^n \left(\frac{k}{n}\right)^2 \binom{n}{k} x^k (1-x)^{n-k},$$

$$\text{and } x^2 = \sum_{k=0}^n \frac{k(k-1)}{n(n-1)} \binom{n}{k} x^k (1-x)^{n-k},$$

and it is clear that $B_n(x^2; x) \neq x^2$.

Quadratic $f(x)$ (Cont.)

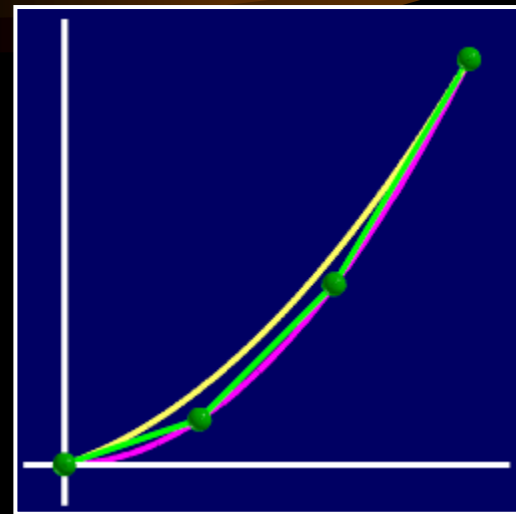
The error (difference) equals,

$$\left| B_n(x^2; x) - x^2 \right| = \sum_{k=0}^n \left| \frac{k^2}{n^2} - \frac{k(k-1)}{n(n-1)} \right| \theta_{k,n}(t)$$

$$= \sum_{k=0}^n \left| \frac{k}{n} \left(\frac{k}{n} - \frac{(k-1)}{(n-1)} \right) \right| \theta_{k,n}(t)$$

$$= \sum_{k=0}^n \left| \frac{k}{n^2} \left(\frac{n-k}{n-1} \right) \right| \theta_{k,n}(t)$$

$$\leq \sum_{k=0}^n \frac{1}{n} \theta_{k,n}(t), \quad \text{or} \quad \left| B_n(x^2; x) - x^2 \right| \leq \frac{1}{n}.$$



Lemma 5.26



Suppose $s(x) = ax + b$ is an arbitrary straight line and $B_n(f;x)$ is the n th degree Bernstein approximation. Then, the number of times $B_n(f;x)$ crosses s is exactly the same as the number of sign changes of the Bernstein approximation to $f - s$.

Proof (of Lemma 5.26)

$$\begin{aligned} B_n(f - s; x) &= \sum_{i=0}^n (f(i/n) - s(i/n)) \theta_{i,n}(x) \\ &= \sum_{i=0}^n f(i/n) \theta_{i,n}(x) - \sum_{i=0}^n s(i/n) \theta_{i,n}(x) \\ &= \sum_{i=0}^n f(i/n) \theta_{i,n}(x) - s(x), \end{aligned}$$

because $B_n(x; x) = x$, and $B_n(1; x) = 1$.

Counting zeros is equivalent to counting intersections.

Descartes' Rule of Signs

If

$$P(x) = \sum_{r=0}^n p_r x^r$$

has all real coefficients p_r , then the number N_p of positive zeros ($x > 0$) does not exceed the number of sign changes in the sequence p_n, p_{n-1}, \dots, p_0 .

Variation Diminishing



Denote by $Z(a_0, a_1, \dots, a_n)$ the number of sign changes of the sequence a_0, a_1, \dots, a_n .

Denote by $Z[f]$ the number of sign changes of an arbitrary continuous function f .

Variation Diminishing (Cont.)



Theorem 5.27

For an arbitrary continuous function, f , over the interval $[0,1]$,

$$Z[B_n(f; x)] \leq Z(f(0), f(1/n), \dots, f(n/n)).$$

Question: Why is it that Theorem 5.27 also proves the variation diminishing property of the Bezier curves?

Variation Diminishing (Cont.)

Proof (of Theorem 5.27)

For $x \in (0, 1)$,

$$\begin{aligned} Z[B_n(f; x)] &= Z\left[\frac{B_n(f; x)}{(1-x)^n}\right] \\ &= Z\left[\sum_{i=0}^n f(i/n) \binom{n}{i} \left(\frac{x}{1-x}\right)^i\right] = Z\left[\sum_{i=0}^n f(i/n) \binom{n}{i} z^i\right], \end{aligned}$$

where $z = x/(1-x)$, taking all positive real values.

Variation Diminishing (Cont.)

Proof (of Theorem 5.27)

However,

$$\begin{aligned} Z \left[\sum_{i=0}^n f(i/n) \binom{n}{i} z^i \right] &\leq Z \left(f\left(\frac{0}{n}\right) \binom{n}{0}, f\left(\frac{1}{n}\right) \binom{n}{1}, \dots, f\left(\frac{n}{n}\right) \binom{n}{n} \right) \\ &= Z \left(f\left(\frac{0}{n}\right), f\left(\frac{1}{n}\right), \dots, f\left(\frac{n}{n}\right) \right) \end{aligned}$$

by the Descartes rule of signs.

Variation Diminishing in R^3

Definition 5.29

If $f(t)$ is a parametric curve in R^3 and $V[f](t)$ is a parametric variation diminishing approximation of $f(t)$, then an arbitrary plane can intersect $V[f](t)$ no more often than it intersects $f(t)$.

Piecing Together Bezier Curves

We are interested in the question of connecting different Bezier curves with some prescribed continuity G^k

Let $\gamma_1(t) = \sum_{i=0}^N P_i \theta_{i,N}(t)$ and $\gamma_2(t) = \sum_{i=0}^N Q_i \theta_{i,N}(t)$ be two

Bezier curves and denote by $\gamma(t)$ their compound curve,

$$\gamma(t) = \begin{cases} \gamma_1(t), & t \in [0,1) \\ \gamma_2(t-1), & t \in [1,2]. \end{cases}$$

Question: What should P_i and Q_i be for G^k continuity?

Piecing Together Bezier Curves (Cont.)

From Chapter 4, we know that the condition for G^1 continuity equals $\gamma'_1(1) = k_1 \gamma'_2(0)$.

For two n 'th degree Bezier curves, we have,

$$\gamma'_1(1) = n(P_n - P_{n-1}) \quad \text{and} \quad \gamma'_2(0) = n(Q_1 - Q_0),$$

or $(P_n - P_{n-1}) = k_1(Q_1 - Q_0)$ and recalling $P_n = Q_0$.

Question: What are the conditions for C^1 continuity?

Piecing Together Bezier Curves (Cont.)

From Chapter 4, we know that the condition for G^2 continuity equals $\gamma''_1(1) = c_3 \gamma'_2(0) + (k_1)^2 \gamma''_2(0)$.

For two n 'th degree Bezier curves, we have,

$$\begin{aligned} \gamma''_1(1) &= n(n-1)(P_n - 2P_{n-1} + P_{n-2}) \quad \text{and} \\ \gamma''_2(0) &= n(n-1)(Q_0 - 2Q_1 + Q_2). \end{aligned}$$

Having C^0 continuity that yields $P_n = Q_0$, and assuming G^1 continuity or $(P_n - P_{n-1}) = k_1(Q_1 - Q_0)$,

$$\gamma''_1(1) = c_3 \gamma'_2(0) + (k_1)^2 \gamma''_2(0).$$

Piecing Together Bezier Curves (Cont.)

$$n(n-1)(P_n - 2P_{n-1} + P_{n-2}) = c_3 n(Q_1 - Q_0) + (k_1)^2 n(n-1)(Q_0 - 2Q_1 + Q_2)$$

or

$$(n-1)(P_n - 2P_{n-1} + P_{n-2}) = c_3(Q_1 - Q_0) + (k_1)^2(n-1)(Q_0 - 2Q_1 + Q_2)$$

or

$$(P_{n-2} - 2P_{n-1} + P_n) = (k_1)^2(Q_0 - 2Q_1 + Q_2) + \frac{c_3}{n-1}(Q_1 - Q_0)$$

By the G^1 continuity, $(P_n - P_{n-1}) = k_1(Q_1 - Q_0)$, and,

$$P_{n-2} - P_{n-1} = \left(k_1 - \frac{c_3}{n-1} + (k_1)^2 \right) (Q_0 - Q_1) - (k_1)^2 (Q_2 - Q_1)$$

$$P_{n-2} - P_{n-1} = \left(k_1 - \frac{c_3}{n-1} + (k_1)^2 \right) (Q_0 - Q_1) - (k_1)^2 (Q_2 - Q_1)$$


Piecing Together Bezier Curves (Cont.)

By G^1 continuity, $P_{n-1} = P_n - k_1(Q_1 - Q_0)$, and since $P_n = Q_0$ from C^0 continuity,

$$\begin{aligned} P_{n-2} &= P_{n-1} + \left(k_1 - \frac{c_3}{n-1} + (k_1)^2 \right) (Q_0 - Q_1) - (k_1)^2 (Q_2 - Q_1) \\ &= P_n - k_1(Q_1 - Q_0) + \left(k_1 - \frac{c_3}{n-1} + (k_1)^2 \right) (Q_0 - Q_1) - (k_1)^2 (Q_2 - Q_1) \\ &= Q_0 + \left(2k_1 - \frac{c_3}{n-1} + (k_1)^2 \right) (Q_0 - Q_1) - (k_1)^2 (Q_2 - Q_1). \end{aligned}$$

Question: What if $\gamma_1(t)$ and $\gamma_2(t)$ have different orders?

Degree Raising



Question: Given a Bezier curve, $\gamma(t)$, of degree n , can one represent $\gamma(t)$ as a Bezier curve of degree m , $m > n$?
 $m < n$?

Question: Why would one need a Bezier curve in a different degree of the exact same shape?

Degree Raising (Cont.)

We recall the following combinatorial identities,

$$\frac{k}{n+1} \theta_{k,n+1} = t \theta_{k-1,n}(t) \quad \text{and} \quad \frac{n+1-k}{n+1} \theta_{k,n+1}(t) = (1-t) \theta_{k,n}(t).$$

Rewriting the first identity as $\frac{k+1}{n+1} \theta_{k+1,n+1}(t) = t \theta_{k,n}(t)$,

we get the sum of

$$\theta_{k,n}(t) = \frac{k+1}{n+1} \theta_{k+1,n+1}(t) + \frac{n+1-k}{n+1} \theta_{k,n+1}(t).$$

$$\theta_{k,n}(t) = \frac{k+1}{n+1} \theta_{k+1,n+1}(t) + \frac{n+1-k}{n+1} \theta_{k,n+1}(t).$$

Degree Raising (Cont.)

Given a curve, $\gamma_p(t)$, the degree raised curve equals,

$$\begin{aligned} \gamma_p(t) &= \sum_{i=0}^n P_i \theta_{i,n}(t) \\ &= \sum_{i=0}^n P_i \left[\frac{i+1}{n+1} \theta_{i+1,n+1}(t) + \frac{n+1-i}{n+1} \theta_{i,n+1}(t) \right] \\ &= \sum_{i=0}^{n+1} \left[\frac{iP_{i-1} + (n+1-i)P_i}{n+1} \right] \theta_{i,n+1}(t), \end{aligned}$$

with the assumption that P_{-1} and P_{n+1} are “don’t care”.

Theorem 5.31

A Bezier curve $\gamma(t) = \sum_{i=0}^n P_i \theta_{i,n}(t)$ can be represented as

$$\gamma(t) = \sum_{i=0}^{n+1} Q_i \theta_{i,n+1}(t)$$

where

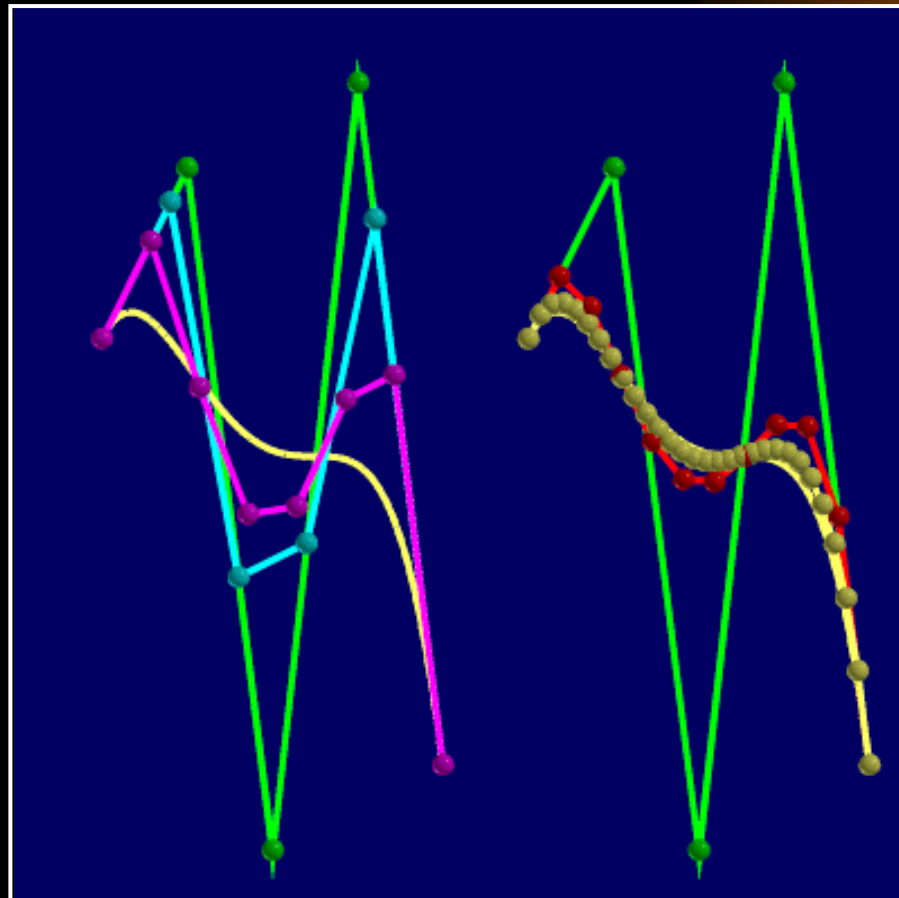
$$Q_i = \frac{iP_{i-1} + (n+1-i)P_i}{n+1}.$$

Theorem 5.31 (Cont.)

degree 4

degree 5

degree 7



degree 4

degree 11

degree 19

Rational Bezier Curves

Let $\theta_{0,n}(t), \dots, \theta_{n,n}(t)$ be the $n + 1$ Bezier basis function of degree n . For a sequence of coefficients $\{P_0, \dots, P_n\}$, which can be scalar or vector and a sequence of scalars, $\{w_0, \dots, w_n\}$, the curve

$$\gamma(t) = \frac{\sum_{i=0}^n w_i P_i \theta_{i,n}(t)}{\sum_{i=0}^n w_i \theta_{i,n}(t)},$$

is called a **rational Bezier curve** of degree n .

Rational Bezier Curves

The w_i 's are sometimes called **homogeneous coordinates** for the P_i 's since there is one w_i associated with each P_i . The determining parameters of the rational Bezier curve are the **homogeneous point**

$$H_i = (h_{x,i}, h_{y,i}, h_{z,i}, h_{w,i}) \quad \text{where if } P_i = (x_i, y_i, z_i), \\ h_{x,i} = w_i x_i, \quad h_{y,i} = w_i y_i, \quad h_{z,i} = w_i z_i \quad \text{and} \quad h_{w,i} = w_i.$$

Warning: Some systems use $P_i = (w_i x_i, w_i y_i, w_i z_i)$!

Integrals of Bezier Basis Functions



Question: How can we compute the integral of a Bezier basis function?

Recall the derivative of a Bezier basis function,

$$\theta'_{i,n}(t) = n(\theta_{i-1,n-1}(t) - \theta_{i,n-1}(t)).$$

Then,

$$\int \theta_{i-1,n-1}(t) dt = \frac{1}{n} \theta_{i,n}(t) + \int \theta_{i,n-1}(t) dt.$$

$$\int \theta_{i-1,n-1}(t) dt = \frac{1}{n} \theta_{i,n}(t) + \int \theta_{i,n-1}(t) dt$$

Integrals of Bezier Basis Functions (Cont.)

$$\begin{aligned} \text{Or, } \int \theta_{i-1,n-1}(t) dt &= \frac{1}{n} \theta_{i,n}(t) + \frac{1}{n} \theta_{i+1,n}(t) + \int \theta_{i+1,n-1}(t) dt \\ &= \frac{1}{n} \sum_{j=i}^{n-1} \theta_{j,n}(t) + \int \theta_{n-1,n-1}(t) dt \\ &= \frac{1}{n} \sum_{j=i}^{n-1} \theta_{j,n}(t) + \frac{1}{n} \theta_{n,n}(t) = \frac{1}{n} \sum_{j=i}^n \theta_{j,n}(t). \end{aligned}$$

Question: What is the integral of a Bezier function,

$$\int c(t) dt = \int \sum_i P_i \theta_{i,n}(t) dt, \text{ equal to?}$$

Products of Bezier (Basis) Functions

Products (and integrals) of geometry are crucial analysis and/or synthesis operators:

- Distances between geometric entities,
- Area/Volume/Moments of geometric entities,
- Curvature analysis, recalling $\kappa(t) = \frac{\|\beta' \times \beta''\|}{\|\beta'\|^3}$.

Question: What about arc-length computation?

Products of Bezier Basis Functions

Recall that $\theta_{i,n}(t) = \binom{n}{i} (1-t)^{n-i} t^i$,

Then,
$$\begin{aligned} \theta_{i,n}(t)\theta_{j,m}(t) &= \binom{n}{i} (1-t)^{n-i} t^i \binom{m}{j} (1-t)^{m-j} t^j \\ &= \binom{n}{i} \binom{m}{j} (1-t)^{n+m-i-j} t^{i+j}. \end{aligned}$$

$$\theta_{i,n}(t)\theta_{j,m}(t) = \binom{n}{i} \binom{m}{j} (1-t)^{n+m-i-j} t^{i+j}.$$

Products of Bezier Basis Functions.

But also by definition $\theta_{i+j,n+m}(t) = \binom{n+m}{i+j} (1-t)^{n+m-i-j} t^{i+j}$,

$$\text{or, } \theta_{i,n}(t)\theta_{j,m}(t) = \frac{\binom{n}{i} \binom{m}{j}}{\binom{n+m}{i+j}} \theta_{i+j,n+m}(t).$$

Question: To what is the product of two Bezier functions, $c_1(t) c_2(t)$, equal?

Bezier Demo

