



Algorithmic Geometry WS 2017/2018

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Bézier Curves



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Affine Geometry

- An *Affine map* Φ can be described as a combination of a linear map A and a translation T :

$$\Phi(x) = Ax + T.$$

- An *Affine combination* is a linear combination

$$x = \sum_{i=0}^n \alpha_i x_i$$

such that the weights of all points sum up to 1: $\sum \alpha_i = 1$.



Properties of affine maps

- An affine map maps a line segment to a line segment.
- Linear interpolation is *affinely invariant*:

$$\Phi((1 - \lambda)P + \lambda Q) = ((1 - \lambda)\Phi(P) + \lambda\Phi(Q)).$$

- Affine maps are ratio preserving.
- Parallel lines stay parallel under affine mapping.



Some examples of affine maps

- The identity. Set $A = I$ the identity matrix, and $T = 0$.
- Translation. It is given by $A = I$
- A scaling. It is given by $T = 0$, and A diagonal matrix.
- A rotation around z axis. $T = 0$, $A = \begin{pmatrix} \cos\alpha & -\sin\alpha & 0 \\ \sin\alpha & \cos\alpha & 0 \\ 0 & 0 & 1 \end{pmatrix}$
- A shear. $T = 0$. $A = \begin{pmatrix} 1 & a & b \\ 0 & 1 & c \\ 0 & 0 & 1 \end{pmatrix}$



Bernstein Polynomials

We will express Bézier Curves in terms of Bernstein Polynomials. First, let's define Bernstein Polynomials.

Definition

Bernstein Polynomials, are defined explicitly by

$$B_i^m(t) := \frac{1}{(b-a)^m} \binom{m}{i} (t-a)^i (b-t)^{m-i}; 0 \leq i \leq m; t \in [a, b],$$

where binomial coefficients are given by

$$\binom{m}{i} = \begin{cases} \frac{m!}{i!(m-i)!} & \text{if } 0 \leq i \leq m, \\ 0 & \text{otherwise.} \end{cases}$$



Bernstein Polynomials - Properties

- 1 The Bernstein polynomials of degree m are linearly independent and span the linear space of degree m . They form a basis of this space.
- 2 $\sum_{i=0}^m B_i^m(t) = 1.$
- 3 $B_i^m(t) \geq 0; t \in [a, b]; 0 \leq i \leq m.$
- 4 $\max B_i^m(t) = \frac{(b-a) \cdot i}{m} + a; t \in [a, b]; 0 < i < m.$
- 5 $B_i^m(t + a) = B_{m-i}^m(b - t).$
- 6 $\frac{d^p}{dt^p} B_i^m(t) = \frac{1}{(b-a)^p} \cdot \frac{m!}{(m-p)!} \sum_{j=0}^p (-1)^{p-j} \cdot \binom{p}{j} \cdot B_{i-j}^{m-j}(t)$



Basic Idea

- approximate a polygon using Bernstein-polynomials \rightarrow curve
- consider the polygon as a "control-structure" for generating curves

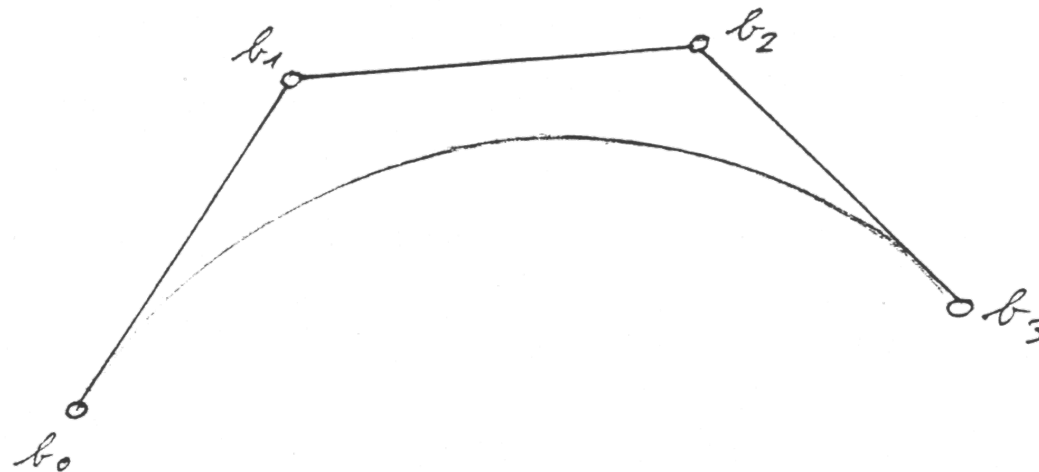


Figure: Bézier Curve



We consider a continuous mapping $F : [0, 1] \rightarrow \mathbb{R}$

$$B_m(F; t) := \sum_{i=0}^m F\left(\frac{i}{m}\right) \cdot B_i^m(t) \quad 0 \leq t \leq 1$$

is the so-called Bernstein approximation of degree m of F .
In the special case of a polygon F_m with vertices $b_0 \dots b_n$,

$$F_m(t) = (m \cdot t - i)b_{i+1} + (i + 1 - m \cdot t)b_i$$

with $0 \leq i \leq m - 1$ and $t \in \left[\frac{i}{m}, \frac{i+1}{m}\right]$,

we get:

$$B_m(F_m; t) = \sum_{i=0}^m b_i \cdot B_i^m(t) \quad t \in [0, 1]$$



This motivates the following

Definition

$$X : [a, b] \rightarrow \mathbb{E}^3$$

$$X(t) := \sum_{i=0}^m b_i \cdot B_i^m(t) = \frac{1}{(b-a)^m} \sum_{i=0}^m b_i \binom{m}{i} (t-a)^i (b-a)^{m-i}$$

is called a Bézier-Segment of degree m ,
with $b_i \in \mathbb{E}^3$, $i = 0, \dots, m$ its Bézier-points.
The polygon is called *Bézier polygon*.



Convex Hull and Variation Diminishing Property

- The Bézier-segment is completely inside the convex hull of the control-polygon.
- Each point $X(t_0)$ is the point of gravity of the Bézier-points b_i with the weights $B_i^m(t_0)$.
- Control polygon and Bézier segment have start point and end point in common.
- Each straight line does not intersect the Bézier segment more often than it intersects the control polygon



The “Convex Hull Property” offers a good bounding-box strategy.
The “Variation Diminishing Property” “guarantees” a smooth curve.
An example of Bézier segment in matrix formulation:
 $m = 3, [a, b] = [0, 1]$

$$\begin{aligned} X(t) &= b_0 \cdot (-t^3 + 3t^2 - 3t + 1) \\ &\quad + b_1 \cdot (3t^3 - 6t^2 + 3t) \\ &\quad + b_2 \cdot (-3t^3 + 3t^2) + b_3(t^3) \\ &= (t^3 t^2 t^1 t^0) \cdot \begin{pmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 3 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} b_0 \\ b_1 \\ b_2 \\ b_3 \end{pmatrix} \end{aligned}$$



The Bernstein polynomials have another interesting property, which we can use for subdivision.

$$\begin{aligned} X(t_0) &= b_0(1-t_0)^3 + 3b_1(1-t_0)^2 \cdot t_0 + 3b_2(1-t_0)t_0^2 + b_3t_0^3 \\ &= b_0(1-t_0)^2(1-t_0) + b_1(1-t_0)^2t_0 + 2b_1(1-t_0)t_0^2 \\ &\quad + 2b_2(1-t_0)t_0^2 + b_2(1-t_0)t_0^2 + b_3t_0^3 \\ &= \underbrace{(b_0(1-t_0) + b_1 \cdot t_0)}_{=:b_1^1} \cdot \underbrace{(1-t_0)^2}_{B_0^2} \\ &\quad + \underbrace{(b_1 \cdot (1-t_0) + b_2 \cdot t_0)}_{=:b_2^1} \cdot \underbrace{(2(1-t_0) \cdot t_0)}_{B_1^2} \\ &\quad + \underbrace{(b_2 \cdot (1-t_0) + b_3 \cdot t_0)}_{=:b_3^1} \cdot \underbrace{t_0^2}_{B_2^2} \end{aligned}$$



$$\begin{aligned} &= \underbrace{(b_1^1 \cdot (1 - t_0) + b_2^1 \cdot t_0)}_{=: b_2^2} \cdot \underbrace{(1 - t_0)}_{B_0^1} \\ &\quad + \underbrace{(b_2^1 \cdot (1 - t_0) + b_3^1 \cdot t_0)}_{=: b_3^2} \cdot \underbrace{t_0}_{B_1^1} \\ &= (b_2^2(1 - t_0) + b_3^2 t_0) = b_3^3 B_0^0 = b_3^3 \end{aligned}$$



General principles

$$\begin{aligned} X(\tilde{t}_0) &= b_0 B_0^m(\tilde{t}_0) + \dots + b_m B_m^m(\tilde{t}_0) \quad \tilde{t}_0 := \frac{t_0 - a}{b - a} \\ &= b_0 B_0^{m-1}(\tilde{t}_0)(1 - \tilde{t}_0) + b_1 B_0^{m-1}(\tilde{t}_0) \cdot \tilde{t}_0 + \dots \\ &\quad + b_{m-1} B_{m-1}^{m-1}(\tilde{t}_0)(1 - \tilde{t}_0) + b_m B_{m-1}^{m-1}(\tilde{t}_0) \cdot \tilde{t}_0 \\ &= \underbrace{(b_0(1 - \tilde{t}_0) + b_1 \tilde{t}_0)}_{=: b_1^1} B_0^{m-1}(\tilde{t}_0) \\ &\quad + \dots + \underbrace{(b_{m-1}(1 - \tilde{t}_0) + b_m \tilde{t}_0)}_{=: b_m^1} \cdot B_{m-1}^{m-1} \end{aligned}$$

linear interpolation $b_i^1 = b_{i-1}(1 - \tilde{t}_0) + b_i \tilde{t}_0$



Iterate:

$$X(\tilde{t}_0) = b_r^r \cdot B_0^{m-r}(\tilde{t}_0) + \dots + b_m^r \cdot B_{m-r}^{m-r}(\tilde{t}_0)$$

$$\text{with } b_i^r = b_{i-1}^{r-1}(1 - \tilde{t}_0) + b_i^{r-1}\tilde{t}_0,$$

$$\text{and } i = r, \dots, m$$

until we reach $r = m$,

$$X(\tilde{t}_0) = b_m^m B_0^0(\tilde{t}_0) = b_m^m$$



De Casteljau Algorithm

The De Casteljau algorithm is used to obtain a point on the curve at parameter value \tilde{t}_0 from the control polygon consisting of the points b_i .

Input: control polygon $P_m = b_0, \dots, b_m$.

Steps:

$$1 \quad b_i^1 := (1 - \tilde{t}_0) \cdot b_{i-1} + \tilde{t}_0 \cdot b_i \quad i = 1, \dots, m$$

$$2 \quad b_i^2 := (1 - \tilde{t}_0) \cdot b_{i-1}^1 + \tilde{t}_0 \cdot b_i^1 \quad i = 2, \dots, m$$

$$m-1 \quad b_i^{m-1} := (1 - \tilde{t}_0) \cdot b_{i-1}^{m-2} + \tilde{t}_0 \cdot b_i^{m-2} \quad i = m - 1, m$$

$$m \quad b_i^m := (1 - \tilde{t}_0) \cdot b_{i-1}^{m-1} + \tilde{t}_0 b_i^{m-1} \quad i = m$$

$$m+1 \quad X(\tilde{t}_0) := b_m^m \quad \tilde{t}_0 := \frac{t_0 - a}{b - a}$$



Visual Scheme for the Algorithm

With the control points in the leftmost column, each pair of consecutive points is used to compute a point of a new control polygon with one fewer point, when arranged in a triangular shape, the target point $X(t_0)$ is then found at the lower right corner.

$$\left. \begin{array}{l} (1 - t_0) \cdot b_i^m \\ + t_0 \cdot b_{i+1}^m \end{array} \right\} b_{i+1}^{m+1}$$

$$\begin{array}{cccc} b_0^0 & & & \\ b_1^0 & b_1^1 & & \\ \vdots & \vdots & \ddots & \\ b_m^0 & b_m^1 & \cdots & b_m^m \end{array}$$

Fact

The De Casteljau Algorithm is affine invariant.



Example

$m = 3, 4$, points on the control polygon are given as

$$P_3 = \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 \\ 4.5 \end{pmatrix}, \begin{pmatrix} 8.5 \\ 6.5 \end{pmatrix}, \begin{pmatrix} 11 \\ 2 \end{pmatrix} \right),$$

compute $X(0.6)$.

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{pmatrix} 2 \\ 4.5 \end{pmatrix} \quad \begin{pmatrix} 1.2 \\ 2.7 \end{pmatrix} \quad \begin{pmatrix} 8.5 \\ 6.5 \end{pmatrix} \quad \begin{pmatrix} 5.9 \\ 5.7 \end{pmatrix} \quad \begin{pmatrix} 4.02 \\ 4.50 \end{pmatrix} \quad \begin{pmatrix} 11 \\ 2 \end{pmatrix} \quad \begin{pmatrix} 10 \\ 3.8 \end{pmatrix} \quad \begin{pmatrix} 8.36 \\ 4.56 \end{pmatrix} \quad \begin{pmatrix} 6.63 \\ 4.54 \end{pmatrix}$$



Graphical example

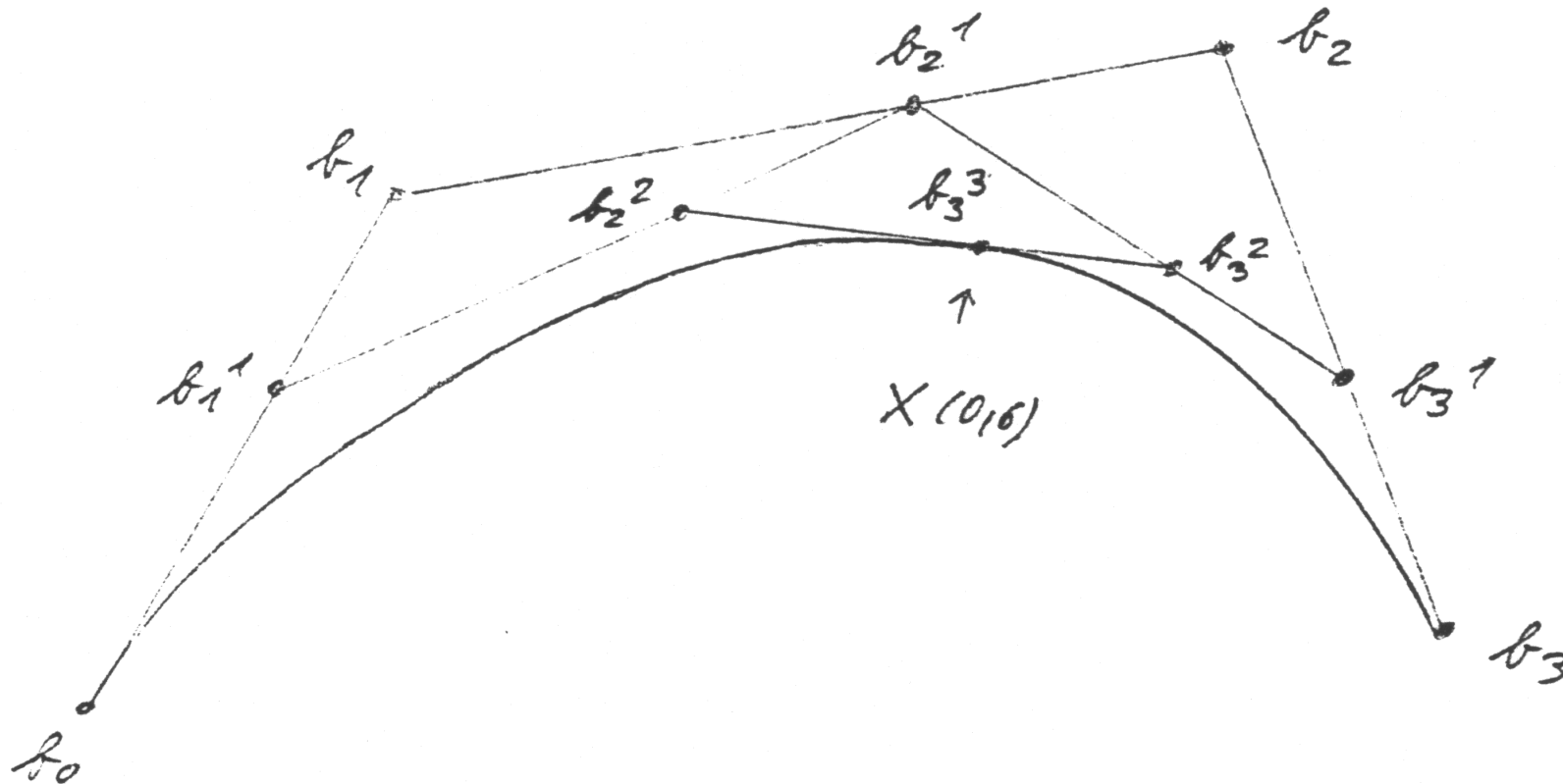


Figure: Visualization of the De Casteljau algorithm



Degree elevation

The Bézier curve generated by the control polygon $P_m = b_0, \dots, b_m$, can also be generated using one more control point by the polygon $P_{m+1} = \widetilde{b}_0, \dots, \widetilde{b}_{m+1}$, with the new control points given by:

$$\widetilde{b}_i = \frac{i}{m+1} \cdot b_{i-1} + \left(1 - \frac{i}{m+1}\right) \cdot b_i \quad i = 1 \dots m$$

Because Bézier curves interpolate the end points, $\widetilde{b}_0 = b_0$ and $\widetilde{b}_{m+1} = b_m$.



For the p th derivative of a Bézier segment

$X(t) := \sum_{i=0}^m b_i B_i^m(t)$, $t \in [a, b]$, one obtains:

$$\frac{d^p}{dt^p} X(t_0) = \frac{m!}{(m-p)!} \cdot \frac{1}{(b-a)^p} \cdot \sum_{i=0}^{m-p} \Delta^p b_i \cdot B_i^{m-p}(t_0)$$

with $\Delta^p b_i := \Delta^{p-1} b_{i+1} - \Delta^{p-1} b_i$,

and $\Delta^0 b_i = b_i$

Fact

The first and last sides of the control polygon are tangential to the Bézier curve.



Some Special Cases

$$X'(t_0) = \frac{m}{b-a} (b_m^{m-1} - b_{m-1}^{m-1})$$
$$X''(t_0) = \frac{m(m-1)}{(b-a)^2} (b_m^{m-2} - 2b_{m-1}^{m-2} + b_{m-2}^{m-2})$$

In our example:

$$X'(0.6) = 3 (b_3^2 - b_2^2) = 3 \left(\begin{pmatrix} 8.36 \\ 4.56 \end{pmatrix} - \begin{pmatrix} 4.02 \\ 4.50 \end{pmatrix} \right) = \begin{pmatrix} 13.02 \\ 0.18 \end{pmatrix}$$

$$X''(0.6) = 6 (b_3^1 - 2b_2^1 + b_1^1)$$
$$= 6 \left(\begin{pmatrix} 10.00 \\ 3.80 \end{pmatrix} - 2 \begin{pmatrix} 5.9 \\ 5.7 \end{pmatrix} + \begin{pmatrix} 1.2 \\ 2.7 \end{pmatrix} \right) = \begin{pmatrix} -3.6 \\ -29.4 \end{pmatrix}$$



Integrating a Bézier segment

With $X(t)$ a Bézier segment on the interval $t \in [a, b]$,

$$\int_a^b X(t)dt = \frac{b-a}{m+1}(b_0 + b_1 + \dots + b_m)$$



Interpolation with Bézier curves

To fit a Bézier curve through a given set of points p_0, \dots, p_r , one needs to solve the linear system of equations:

$$X(t_j) := \sum_{i=0}^r b_i B_i^r(t_j)$$

Once the t_j are chosen, this system has a unique solution b_i , as the basis functions B_i^r are linearly independent.



Example Interpolation

Fitting a Bézier curve through the points

$$p_0 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, p_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, p_2 = \begin{pmatrix} 4 \\ 2 \end{pmatrix}, p_3 = \begin{pmatrix} 6 \\ 6 \end{pmatrix}.$$

With *chordal parametrization*, the t_i are chosen according to the Euclidean distances, i.e. $t_i - t_{i-1} = \|p_i - p_{i-1}\|$ for $i = 1, \dots, m$. For the points in this example, $t_0 = 0$, $t_1 = \sqrt{2}$, $t_2 = t_1 + \sqrt{5}$, $t_3 = t_2 + 2\sqrt{5}$, so $[a, b] = [0, 8.12]$.

$$\begin{aligned} X(t_0) = X(0) &= \begin{pmatrix} 1 \\ 2 \end{pmatrix} & X(t_1) = X(1.41) &= \begin{pmatrix} 2 \\ 1 \end{pmatrix} \\ X(t_2) = X(3.65) &= \begin{pmatrix} 4 \\ 2 \end{pmatrix} & X(t_3) = X(8.12) &= \begin{pmatrix} 6 \\ 6 \end{pmatrix} \end{aligned}$$



The equations then read:

$$\begin{pmatrix} 1 \\ 2 \end{pmatrix} = b_0 \text{ (Because of end-point interpolation)}$$

$$\begin{pmatrix} 2 \\ 1 \end{pmatrix} = \sum_{i=0}^3 b_i B_i^3(1.41)$$

$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} = \sum_{i=0}^3 b_i B_i^3(3.65)$$

$$\begin{pmatrix} 6 \\ 6 \end{pmatrix} = b_3$$



Only b_1 and b_2 need to be solved for:

$$\begin{pmatrix} B_1^3(t_1) & B_2^3(t_1) \\ B_1^3(t_2) & B_2^3(t_2) \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} p_1 - B_0^3(t_1)p_0 - B_3^3(t_1)p_3 \\ p_2 - B_0^3(t_2)p_0 - B_3^3(t_2)p_3 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \begin{pmatrix} 0.36 & 0.08 \\ 0.41 & 0.33 \end{pmatrix}^{(-1)} \begin{pmatrix} p_1 - B_0^3(t_1)p_0 - B_3^3(t_1)p_3 \\ p_2 - B_0^3(t_2)p_0 - B_3^3(t_2)p_3 \end{pmatrix}$$

$$\Rightarrow b_1 = \begin{pmatrix} 2.52 \\ -1.55 \end{pmatrix},$$

$$b_2 = \begin{pmatrix} 6.78 \\ 5.27 \end{pmatrix}$$

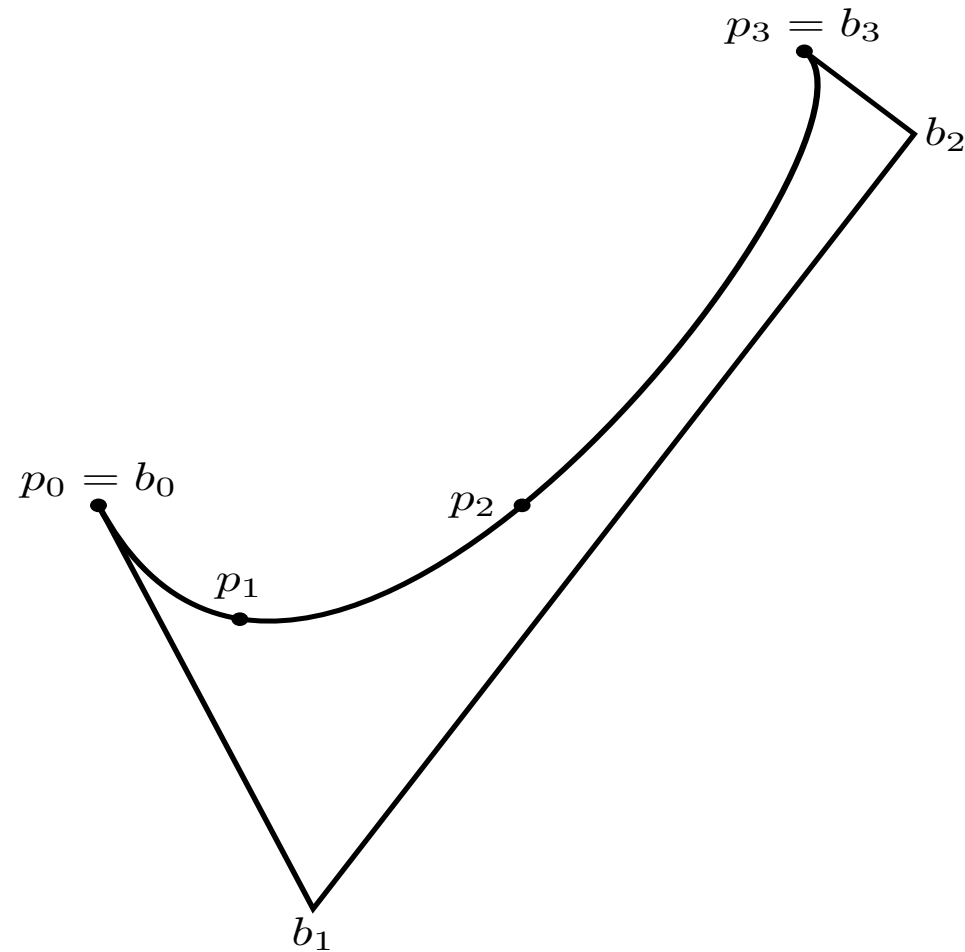


Figure: Interpolating Bézier curve



Further remarks on Interpolation

- previous method offers no additional interaction opportunities like tangent or curvature adjustments.
- another possibility: use the points p_0, \dots, p_m as corner points for combined Bézier segments

Definition

A *Bézier curve* is a spline consisting of Bézier segments X_l , $l = 0, \dots, k$, each of degree m , over the parameter interval $u_l \leq u \leq u_{l+1}$.

$$X_l(u) := \sum_{i=0}^m b_{m_{l+i}} \cdot B_i^m \left(\frac{u - u_l}{u_{l+1} - u_l} \right)$$

The Bernstein polynomials are used as “blending functions”.



Joining the Bézier segments

To obtain a C^n -continuous curve, the function values, as well as derivatives up to the n th, need to be matched.

Function values:

$$X_l(u) := \sum_{i=0}^m b_{m(l+i)} \cdot B_i^m \left(\frac{u - u_l}{u_{l+1} - u_l} \right); \quad u_l \leq u \leq u_{l+1}$$

$$X_{l+1}(u) := \sum_{j=0}^m b_{m(l+i)+j} \cdot B_j^m \left(\frac{u - u_{l+1}}{u_{l+2} - u_{l+1}} \right); \quad u_{l+1} \leq u \leq u_{l+2}$$



Joining the Bézier segments (cont.)

Derivatives in the corner points, with $\lambda_l := u_{l+1} - u_l$:

$$X_l^{(1)}(u_{l+1}) := \frac{m}{\lambda_l} (b_{m(l+1)} - b_{m(l+1)-1})$$

$$X_l^{(2)}(u_{l+1}) := \frac{m(m-1)}{\lambda_l^2} (b_{m(l+1)} - 2b_{m(l+1)-1} + b_{m(l+1)-2})$$

$$X_l^{(3)}(u_{l+1}) := \frac{m(m-1)(m-2)}{\lambda_l^3} \cdot (b_{m(l+1)} - 3b_{m(l+1)-1} + 3b_{m(l+1)-2} - b_{m(l+1)-3})$$



Joining the Bézier segments (cont.)

$$X_{l+1}^{(1)}(u_{l+1}) := \frac{m}{\lambda_{l+1}} (b_{m(l+1)+1} - b_{m(l+1)})$$

$$X_{l+1}^{(2)}(u_{l+1}) := \frac{m(m-1)}{\lambda_{l+1}^2} (b_{m(l+1)+2} - 2b_{m(l+1)+1} + b_{m(l+1)})$$

$$X_{l+1}^{(3)}(u_{l+1}) := \frac{m(m-1)(m-2)}{\lambda_{l+1}^3} \cdot (b_{m(l+1)+3} - 3b_{m(l+1)+2} + 3b_{m(l+1)+1} - b_{m(l+1)})$$



Conditions for Continuity

- C^1 -continuity:

$$b_{m(l+1)} - b_{m(l+1)-1} = \frac{\lambda_l}{\lambda_{l+1}} \cdot (b_{m(l+1)+1} - b_{m(l+1)})$$

- C^2 -continuity: additionally

$$b_{m(l+1)} - 2b_{m(l+1)-1} + b_{m(l+1)-2} = \frac{\lambda_l^2}{\lambda_{l+1}^2} \cdot (b_{m(l+1)+2} - 2b_{m(l+1)+1} + b_{m(l+1)}) \quad (1)$$



Polygon construction

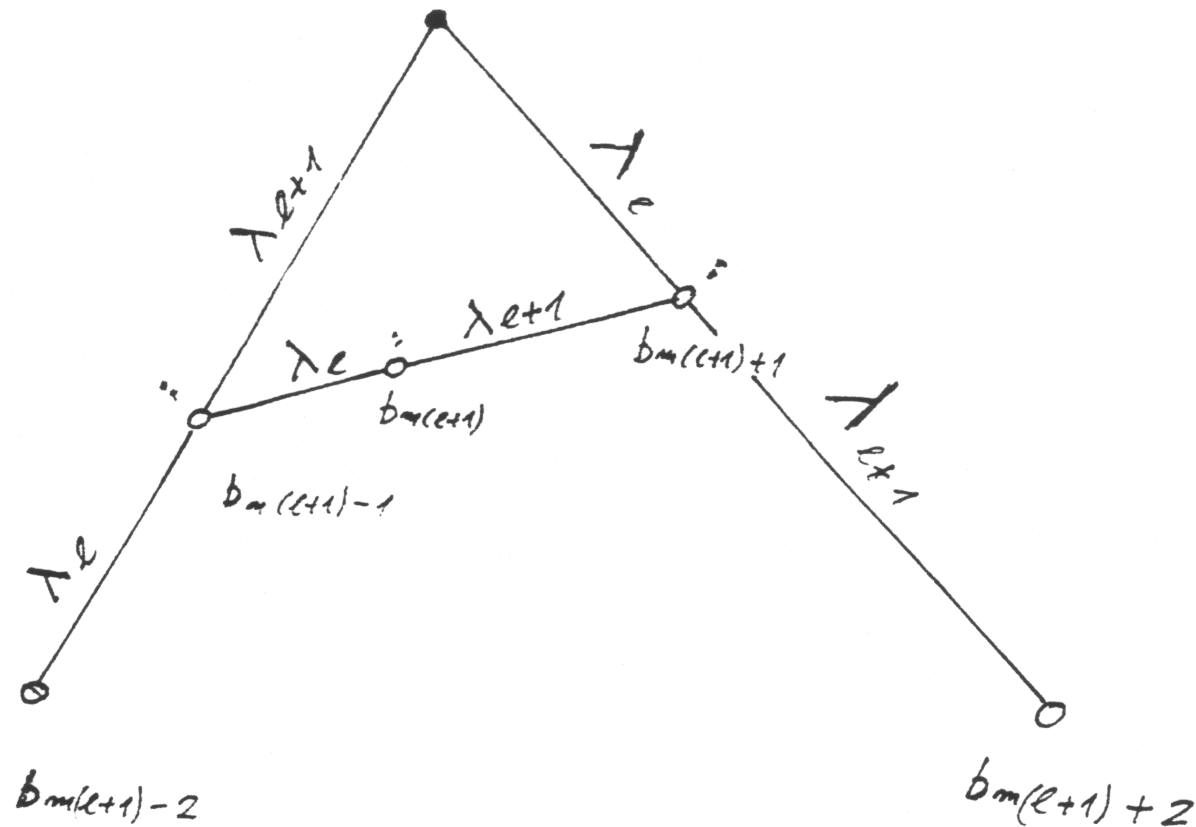


Figure: Two Bézier segments joined C^2 -continuously at $b_{m(l+1)}$



C^3 -continuity

$$\begin{aligned} b_{m(l+1)} - b_{m(l+1)-1} &= \frac{\lambda_l}{\lambda_{l+1}} \cdot (b_{m(l+1)+1} - b_{m(l+1)}) \\ b_{m(l+1)} - 2b_{m(l+1)-1} \\ &\quad + b_{m(l+1)-2} = \frac{\lambda_l^2}{\lambda_{(l+1)}^2} \\ &\quad \cdot (b_{m(l+1)+2} - 2b_{m(l+1)+1} + b_{m(l+1)}) \\ b_{m(l+1)} - 3b_{m(l+1)-1} \\ + 3b_{m(l+1)-2} - b_{m(l+1)-3} &= \frac{\lambda_l^3}{\lambda_{(l+1)}^3} \\ &\quad \cdot (b_{m(l+1)+3} - 3b_{m(l+1)+2} \\ &\quad \quad + 3b_{m(l+1)+1} - b_{m(l+1)}) \end{aligned}$$

Polygon construction

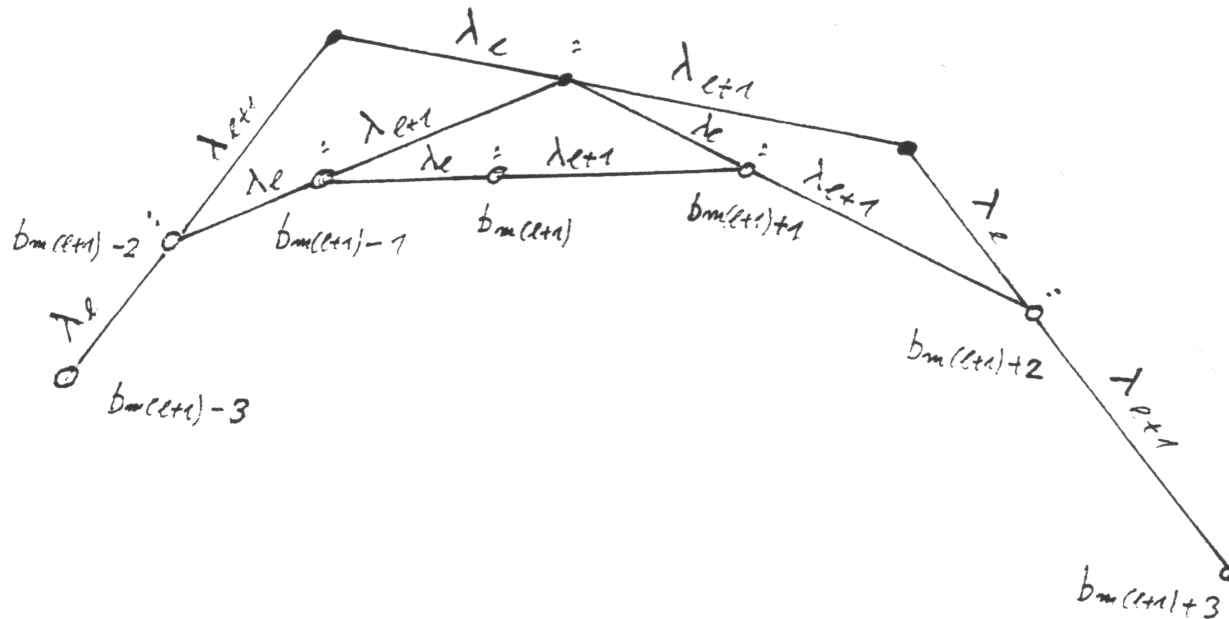


Figure: Two Bézier segments joined with C^3 -continuity

Remark

This is “equivalent” to the De Casteljau recursion



Cubic C^2 -continuous Bézier splines

Fitting the condition for C^1 -continuity,

$$\lambda_{(l+1)} (b_{3(l+1)} - b_{3(l+1)-1}) = \lambda_l (b_{3(l+1)+1} - b_{3(l+1)}),$$

into the C^2 -condition:

$$\begin{aligned} \lambda_{(l+1)}^2 (b_{3(l+1)} - 2b_{3(l+1)-1} + b_{3(l+1)-2}) \\ = \lambda_l^2 (b_{3(l+1)} - 2b_{3(l+1)+1} + b_{3(l+1)}), \end{aligned}$$

we get:

$$\frac{\lambda_{l+1}}{\lambda_l} \left(b_{3(l+1)} - \frac{\lambda_l + \lambda_{(l+1)}}{\lambda_{(l+1)}} b_{3(l+2)} \right) = \frac{\lambda_l}{\lambda_{(l+1)}} \left(b_{3(l+5)} - \frac{\lambda_l + \lambda_{(l+1)}}{\lambda_l} b_{3(l+4)} \right)$$

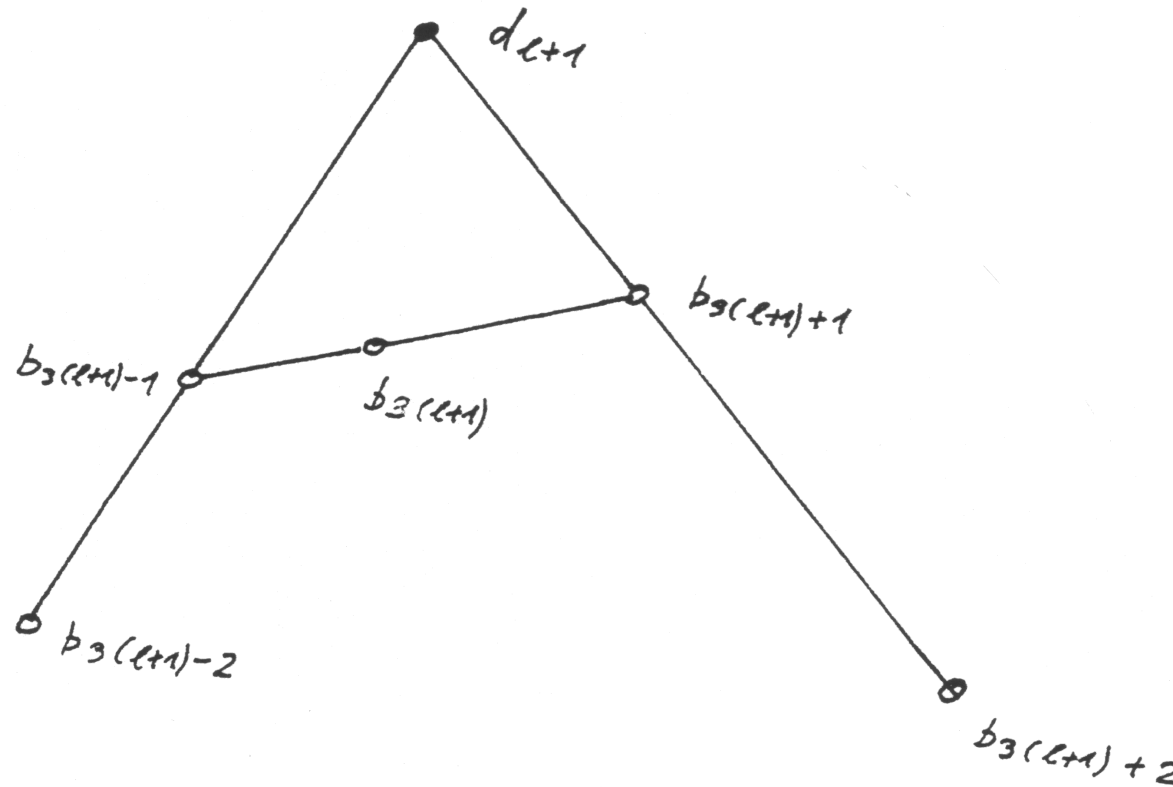


Figure: geometric background for the cubic case. See next slide for the definition of d



Definition

The *weight point* d_{l+1} completes a geometric construction that allows for C^2 -continuous connection of Bézier segments (see previous figure).

$$\begin{aligned}d_{l+1} &:= \frac{\lambda_{l+1}}{\lambda_l} \left(-b_{3(l+1)-2} + \frac{\lambda_l + \lambda_{l+1}}{\lambda_{(l+1)}} b_{3(l+1)-1} \right) \\ &= \frac{\lambda_l}{\lambda_{(l+1)}} \left(-b_{3(l+1)+2} + \frac{\lambda_l + \lambda_{l+1}}{\lambda_{(l+1)}} b_{3(l+1)+1} \right)\end{aligned}$$



Continuity in three steps

To connect a new Bézier segment with C^2 continuity, one can follow three steps:

connect the tangents

$$b_{3(l+1)+1} = b_{3(l+1)} + \frac{\lambda_{(l+1)}}{\lambda_l} (b_{3(l+1)} - b_{3(l+1)-1})$$

find the weight point

$$d_{l+1} = b_{3(l+1)-1} + \frac{\lambda_{(l+1)}}{\lambda_l} (b_{3(l+1)-1} - b_{3(l+1)-2})$$

adjust for C^2 -continuity

$$b_{3(l+1)+2} = b_{3(l+1)+1} + \frac{\lambda_{(l+1)}}{\lambda_l} (-d_{l+1} + b_{3(l+1)+1})$$



Algorithm for interpolating with Bézier segments

$$\begin{aligned}(\lambda_{l-1} + \lambda_l) \cdot d_{l-1} + \lambda_{l-2} d_l &= b_{3l-2} \cdot C_{l-1} & l = 1 \dots k \\ A_l \cdot d_{l-1} + D_l \cdot d_l + A_{l-1} \cdot d_{l+1} &= b_{3l} \cdot B_l & l = 1 \dots k - 1 \\ \lambda_{l+1} \cdot d_l + (\lambda_{l-1} + \lambda_l) \cdot d_{l+1} &= b_{3l+2} \cdot C_l & l = 0 \dots k - 1\end{aligned}$$

with $\lambda_{-1} = \lambda_0$ and $\lambda_k = \lambda_{k-1}$

$$C_l = (\lambda_{l-1} + \lambda_l + \lambda_{l+1})$$

$$A_l = \lambda_l^2 (\lambda_{l-1} + \lambda_l + \lambda_{l+1})$$

$$\begin{aligned}D_l &= \lambda_l (\lambda_{l-2} + \lambda_{l-1}) (\lambda_{l-1} + \lambda_l + \lambda_{l+1}) \\ &\quad + \lambda_{l-1} (\lambda_l + \lambda_{l+1}) (\lambda_{l-2} + \lambda_{l-1} + \lambda_l)\end{aligned}$$

$$B_l = (\lambda_{l-1} + \lambda_l) (\lambda_{l-1} + \lambda_l + \lambda_{l+1}) (\lambda_{l-2} + \lambda_{l-1} + \lambda_l)$$



Special case $\lambda_l = 1$ for all $l = 1 \dots k$.

$$A_l \equiv 3; \quad B_l \equiv 18; \quad C_{l-1} \equiv 3; \quad D_l \equiv 12$$

$$\begin{aligned} 2d_{l-1} + d_l &= 3b_{3l-2} & l = 1, \dots, k \\ d_{l-1} + 4d_l + d_{l+1} &= 6b_{3l} & l = 1, \dots, k-1 \\ d_l + 2d_{l+1} &= 3b_{3l+2} & l = 0, \dots, k-1 \end{aligned}$$

Remarks

- The weight points d_0 and d_k are defined as:

$$d_0 = 2b_1 - b_2$$

$$d_k = -b_{3k-2} + 2b_{3k-1}$$

- Using the previously pictured algorithm allows for calculation of all Bézier points except for b_0 and b_{3k} .



Example for an interpolating cubic Bézier spline

Input: p_0, p_1, p_2 , special parametrization $\lambda_1 = \lambda_2 = 1$.

- set $b_0 = p_0; b_3 = p_1; b_6 = p_2$
- choose the free parameters b_1 and b_5
- solve the linear system:

$$\begin{array}{rclcl} 2d_0 & +d_1 & & = & 3b_1 \\ d_0 & +4d_1 & +d_2 & = & 6b_3 \\ & d_1 & +2d_2 & = & 3b_5 \end{array}$$

- calculate Bézier points b_2, b_4 using the weight point construction

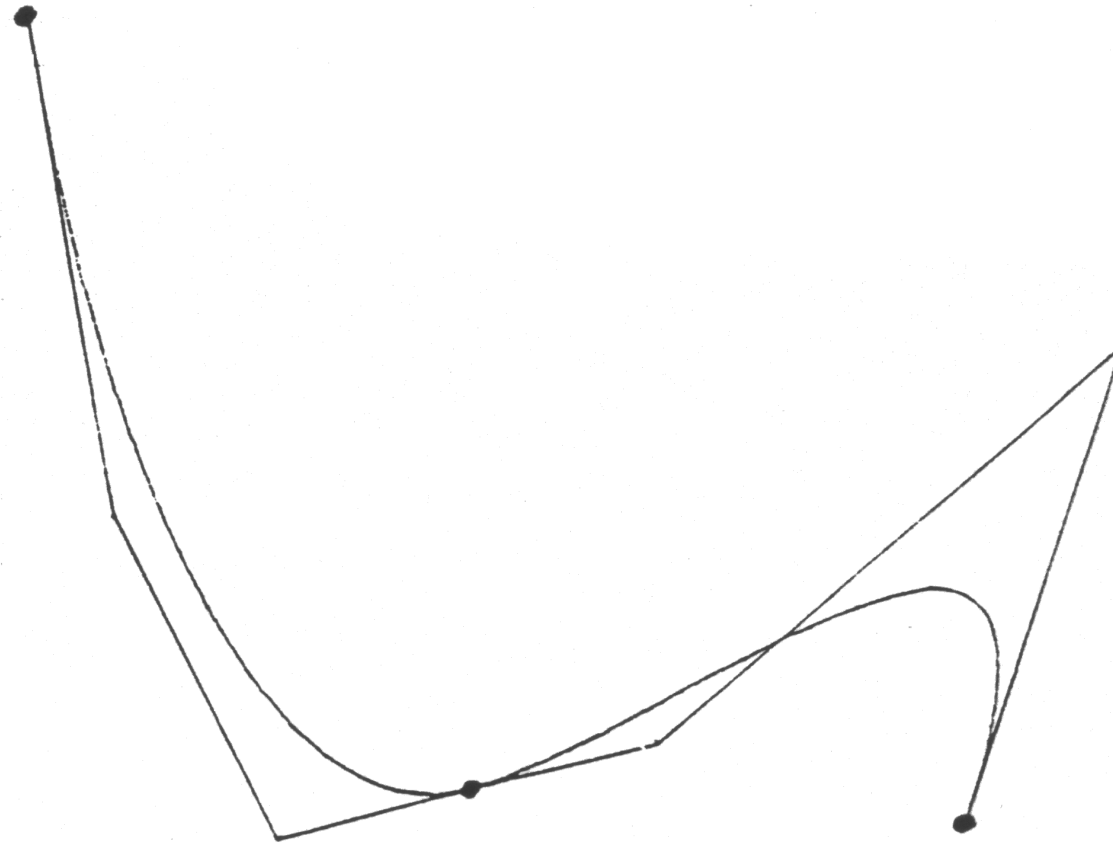


Figure: Interpolating three points with two cubic Bézier curves



Algorithm for Interpolating Bézier Splines

Input: $k + 1$ points $p_i \in \mathbb{E}^3, 0 \leq i \leq k$.

- 1 set $b_0 = p_0, b_3 = p_1, \dots, b_{3k} = p_k$
- 2 choose free parameters b_1 and b_{3k-1} (slope at the end points)
- 3 solve the linear system of equations:

$$\begin{aligned}(\lambda_1 + \lambda_0)d_0 + \lambda_0 \cdot d_1 &= C_0 \cdot b_1 \\ A_l \cdot d_{l-1} + D_l \cdot d_l + A_{l-1} \cdot d_{l+1} &= B_l \cdot b_{3l} \quad l = 1 \dots k - 1 \\ \lambda_{k-1}d_{k-1} + (\lambda_{k-2} + \lambda_{k-1}) \cdot d_k &= C_{k-1}b_{3k-1}\end{aligned}$$

- 4 calculate the Bézier points $b_{3l-2}, b_{3l-1} (l = 1 \dots k)$
- 5 run the De Casteljau algorithm k times:
 $\{b_0, b_1, b_2, b_3\}; \{b_3, b_4, b_5, b_6\};$
 $\dots; \{b_{3k-3}, b_{3k-2}, b_{3k-2}, b_{3k-1}\}$

Example

Interpolate the points

$$\begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} \begin{pmatrix} 4 \\ 4 \end{pmatrix} \begin{pmatrix} 6 \\ 2 \end{pmatrix}$$

by a cubic Bézier Spline (with chordal parametrization),

$$\lambda_0 = \sqrt{2}, \quad \lambda_1 = \sqrt{5}, \quad \lambda_2 = \sqrt{20}$$

$$u_0 = 0, \quad u_1 = u_0 + \lambda_0, \quad u_2 = u_1 + \lambda_1, \quad u_3 = u_2 + \lambda_2$$

$$\text{Step 1: } b_0 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, b_3 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, b_6 = \begin{pmatrix} 4 \\ 4 \end{pmatrix}, b_9 = \begin{pmatrix} 6 \\ 2 \end{pmatrix}$$

Step 2: Choose b_1 and b_8 .



Step 3:

$$\begin{aligned}d_0 + 0.39d_1 &= 1.39b_1 \\0.27d_0 + 0.66d_1 + 0.068d_2 &= \begin{pmatrix} 2 \\ 1 \end{pmatrix} \\0.39d_1 + 0.57d_2 + 0.067d_3 &= \begin{pmatrix} 4 \\ 2 \end{pmatrix} \\0.66d_2 + d_3 &= 1.67b_8\end{aligned}$$

Solution:

$$\begin{aligned}d_0 &= 1.68b_1 - 0.02b_8 - \begin{pmatrix} 1.16 \\ 0.58 \end{pmatrix} & d_1 &= 0.76b_1 + 0.03b_8 + \begin{pmatrix} 3 \\ 1.5 \end{pmatrix} \\d_2 &= 0.25b_8 + \begin{pmatrix} 5.64 \\ 2.82 \end{pmatrix} & d_3 &= -0.35b_1 + 1.84b_8 - \begin{pmatrix} 3.74 \\ 1.87 \end{pmatrix}\end{aligned}$$



Step 4: Run the De Casteljau algorithm three times, for each of the segments $b_{0...3}$, $b_{3...6}$, $b_{6...9}$

$$\begin{aligned} b_2 &= 0.18b_1 + 0.004b_8 + \begin{pmatrix} 0.72 \\ 0.31 \end{pmatrix} & b_4 &= -0.47b_1 - 0.02b_8 + \begin{pmatrix} 2.97 \\ 1.49 \end{pmatrix} \\ b_5 &= -0.03b_1 - 0.1b_8 + \begin{pmatrix} 3.59 \\ 1.79 \end{pmatrix} & b_7 &= 0.35b_1 + 0.17b_8 + \begin{pmatrix} 3.76 \\ 1.88 \end{pmatrix} \end{aligned}$$

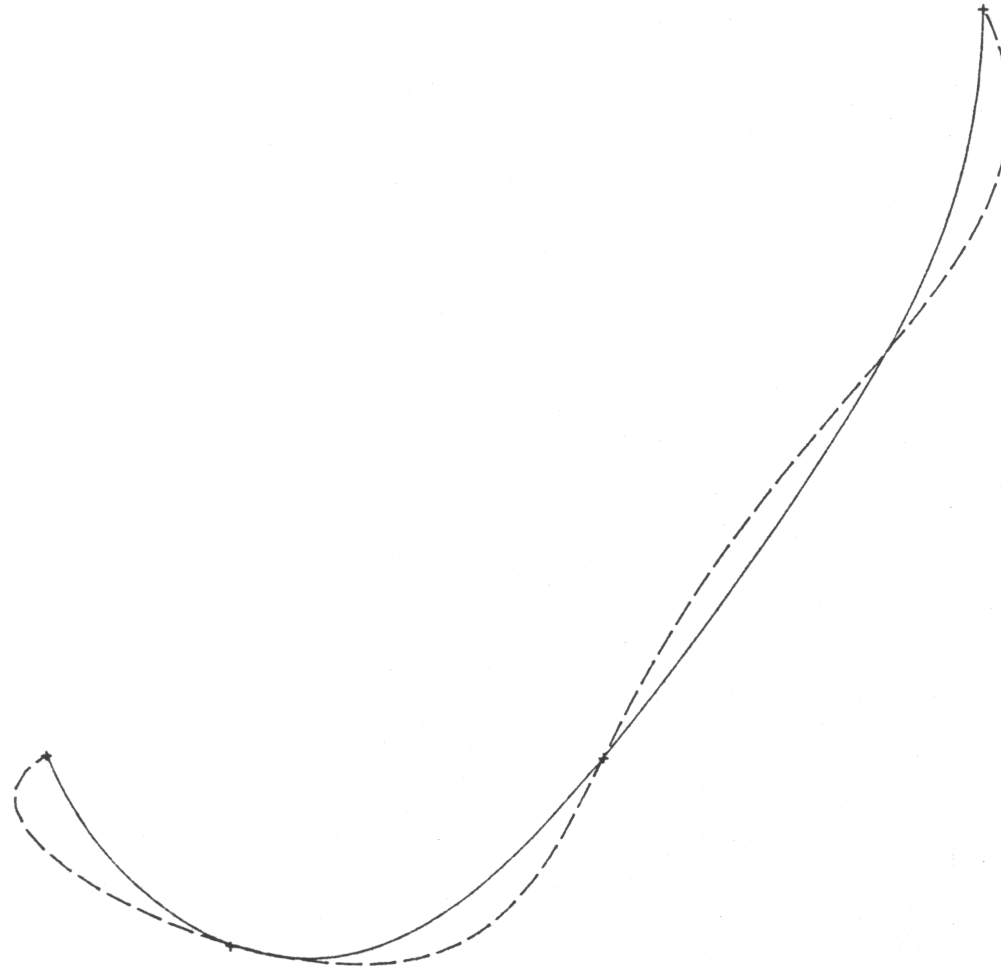


Figure: Different choices for b_1 and b_8



Algorithm for “free modeling” Bézier splines

Input: $k + 1$ weight points d_0, \dots, d_k

Step 1 (a) $b_0 = d_0, b_{3k} = d_k$
(b) $b_0 = b_{3k} = \frac{1}{2}(b_1 + b_{3k-1}); d_0 = d_k$

Step 2 Calculate all inner Bézier points:

$$\begin{aligned} 2d_{l-1} + d_l &= 3b_{3l-2} & l = 1, \dots, k \\ d_{l-1} + 4d_l + d_{l+1} &= 6b_{3l} & l = 1, \dots, k-1 \\ d_l + 2d_{l+1} &= 3b_{3l+2} & l = 0, \dots, k-1 \end{aligned}$$

Step 3 Run De Casteljau algorithm k times, on

$$b_{0\dots3} \dots b_{3k-3\dots3k}$$



Example

Input: weight points

$$d_0 = \begin{pmatrix} -1 \\ 2 \end{pmatrix}, d_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, d_2 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, d_3 = \begin{pmatrix} -1 \\ 2 \end{pmatrix}.$$

$$\text{Step 1: } b_0 = b_9 = \frac{1}{2}(b_1 + b + 8) = \begin{pmatrix} -0.5 \\ 1.5 \end{pmatrix}$$

Step 2:

$$\begin{aligned} b_1 &= 1/3(2d_0 + d_1) &= \begin{pmatrix} -1/3 \\ 5/3 \end{pmatrix} & b_2 &= 1/3(d_0 + 2d_1) &= \begin{pmatrix} 1/3 \\ 4/3 \end{pmatrix} \\ b_3 &= 1/6(d_0 + 4d_1 + d_2) &= \begin{pmatrix} 1/2 \\ 1 \end{pmatrix} & b_4 &= 1/3(2d_1 + d_2) &= \begin{pmatrix} 2/3 \\ 2/3 \end{pmatrix} \\ b_5 &= 1/3(d_1 + 2d_2) &= \begin{pmatrix} 1/3 \\ 1/3 \end{pmatrix} & b_6 &= 1/6(d_1 + 4d_2 + d_3) &= \begin{pmatrix} 0 \\ 1/2 \end{pmatrix} \\ b_7 &= 1/3(2d_2 + d_3) &= \begin{pmatrix} -1/3 \\ 2/3 \end{pmatrix} & b_8 &= 1/3(d_2 + 2d + 3) &= \begin{pmatrix} -2/3 \\ 4/3 \end{pmatrix} \end{aligned}$$

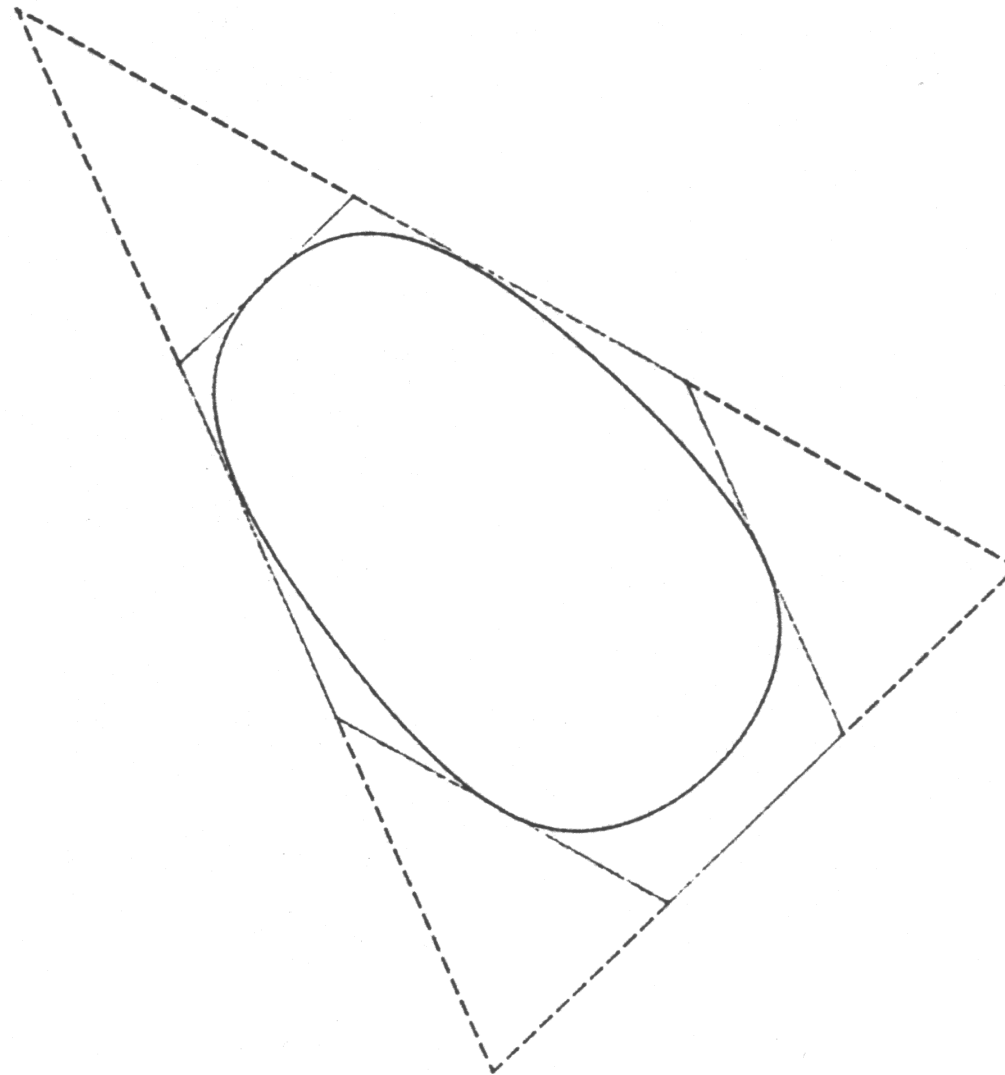


Figure: Result of the previous example calculation



Example

Input: weight points $d_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$, $d_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, $d_2 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$

Step 1: $b_0 = d_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ and $b_6 = d_2 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$

Step 2:

$$\begin{aligned} b_1 &= \frac{1}{3}(2d_0 + d_1) = \begin{pmatrix} 1/3 \\ 1/3 \end{pmatrix} & b_2 &= \frac{1}{3}(d_2 + 2d_1) = \begin{pmatrix} 2/3 \\ 2/3 \end{pmatrix} \\ b_3 &= \frac{1}{6}(d_0 + 4d_1 + d_2) = \begin{pmatrix} 1 \\ 1/2 \end{pmatrix} & b_4 &= \frac{1}{3}(2d_1 + d_2) = \begin{pmatrix} 4/3 \\ 1/3 \end{pmatrix} \\ b_5 &= \frac{1}{3}(d_1 + 2d_2) = \begin{pmatrix} 5/3 \\ -1/3 \end{pmatrix} \end{aligned}$$

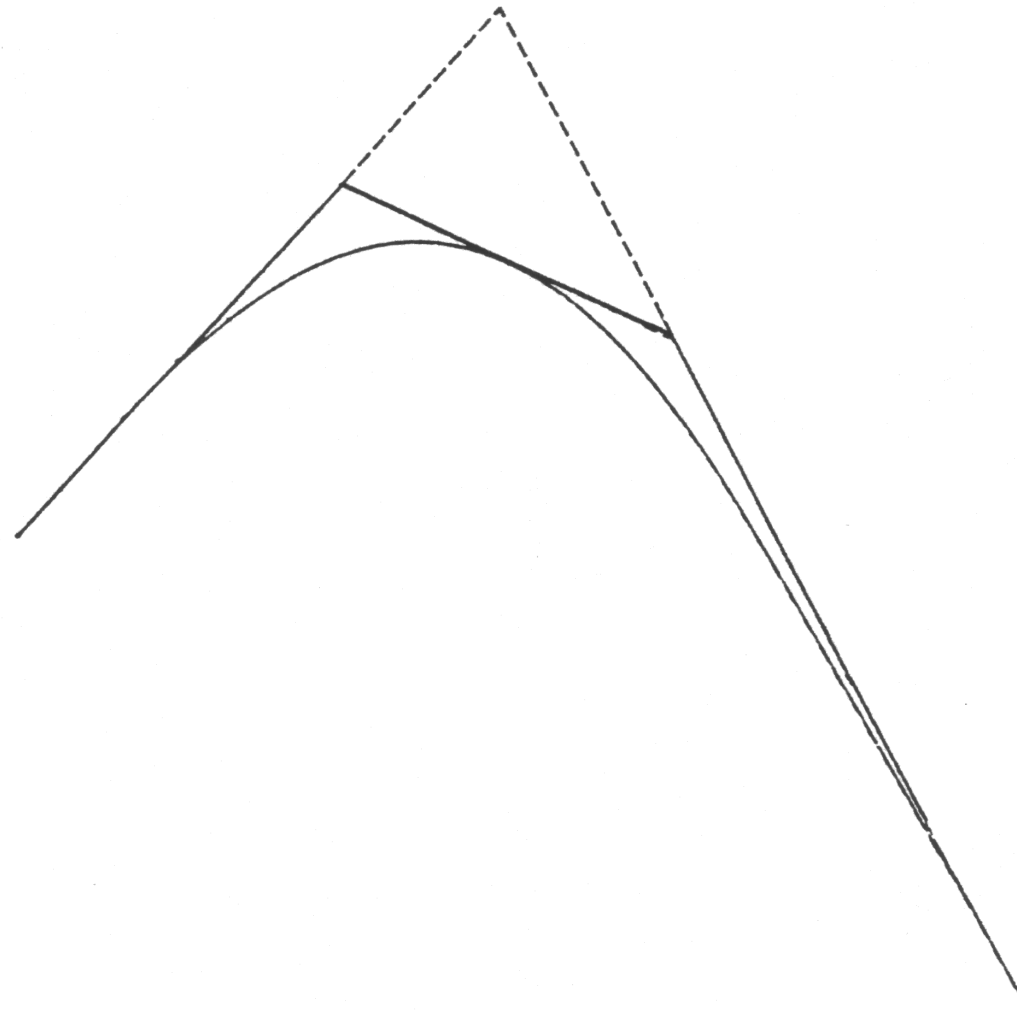


Figure: Result of the previous example



Rational Bézier curves

Blending function methods also work in projective spaces:

$$x(t) = \begin{bmatrix} \sum_{i=0}^n \lambda_i b_{ix} \cdot B_i^n(t) \\ \sum_{i=0}^n \lambda_i b_{iy} \cdot B_i^n(t) \\ \sum_{i=0}^n \lambda_i b_{iz} \cdot B_i^n(t) \\ \sum_{i=0}^n \lambda_i \cdot B_i^n(t) \end{bmatrix}$$

Projective Bézier curve with
homogeneous coordinates



“Going back” to Euclidean space, we get rational Bézier curves:

$$x(t) = \sum_{i=0}^n b_i \cdot \widetilde{B}_i^n(t) = \sum_{i=0}^n b_i \cdot \frac{\lambda_i \cdot B_i^n(t)}{\sum_{i=0}^n \lambda_i B_i^n(t)}$$

Remarks

- 1 For $\lambda_i \equiv 1$ for all $i = 0 \dots n$ we get the “usual” Bézier curves, since $\sum_{i=0}^n B_i^n(t) = 1$
- 2 Under the condition $\lambda_0 > 0$, $\lambda_n > 0$ and $\lambda_i \geq 0$ for $i = 1, \dots, n - 1$, the curve starts in b_0 and ends in b_n .
- 3 All the other properties like the constriction to the convex hull, variation diminishing etc. are still valid.

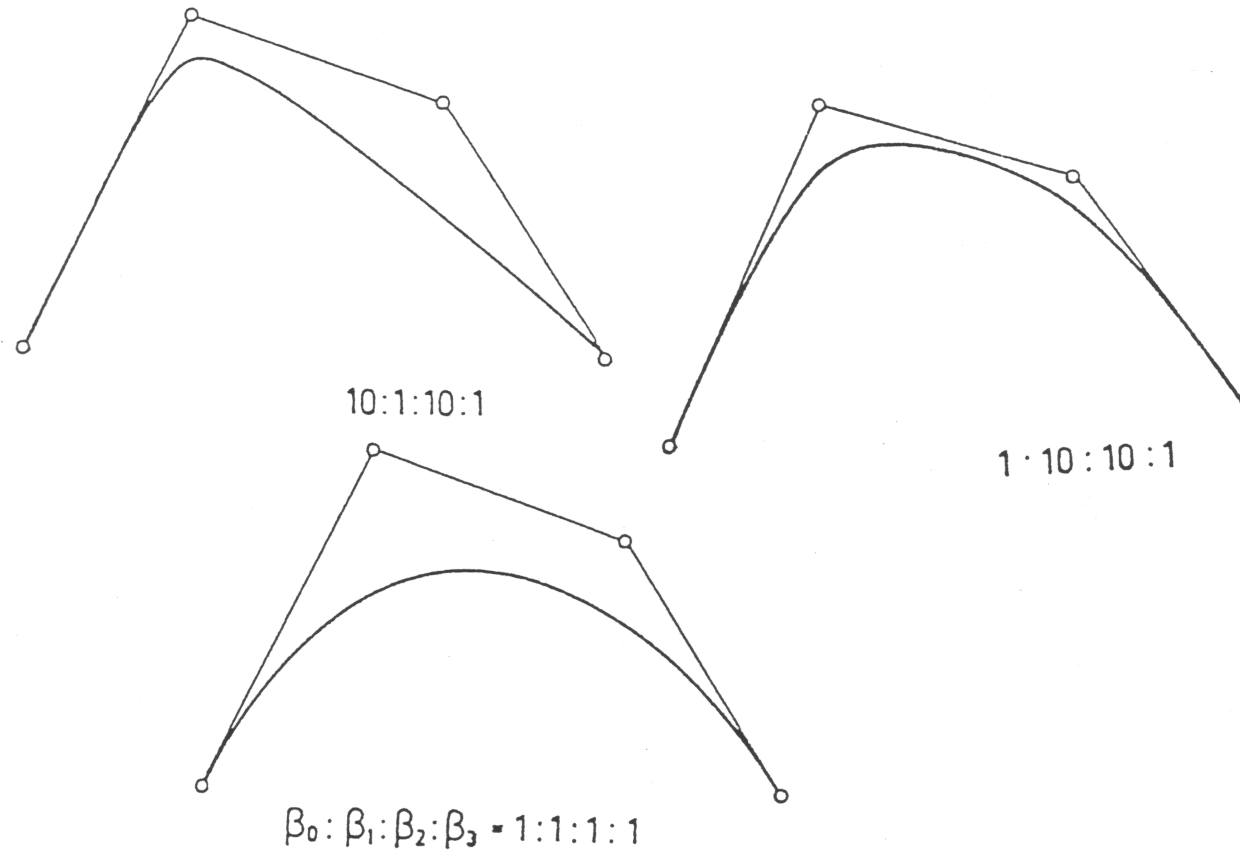


Figure: Influence of weight changes



Main reason for rational Bézier curves

Curves of algebraic order 2 (conics) are special rational Bézier curves of degree 2.

We consider rational, quadratic Bézier curves in this form:

$$x(t) = b_0 \frac{\lambda_0 B_0^2(t)}{\sum_{i=0}^2 \lambda_i B_i^2(t)} + b_1 \frac{\lambda_1 B_1^2(t)}{\sum_{i=0}^2 \lambda_i B_i^2(t)} + b_2 \frac{\lambda_2 B_2^2(t)}{\sum_{i=0}^2 \lambda_i B_i^2(t)},$$

and choose $\lambda_0 = \lambda_2 = 1$.



We now define the *center point* $M = \frac{1}{2}(b_0 + b_2)$ and the *shoulder point* $S = M(1 - s) + b_1$ with $s := \frac{\lambda_1}{1 + \lambda_1}$.

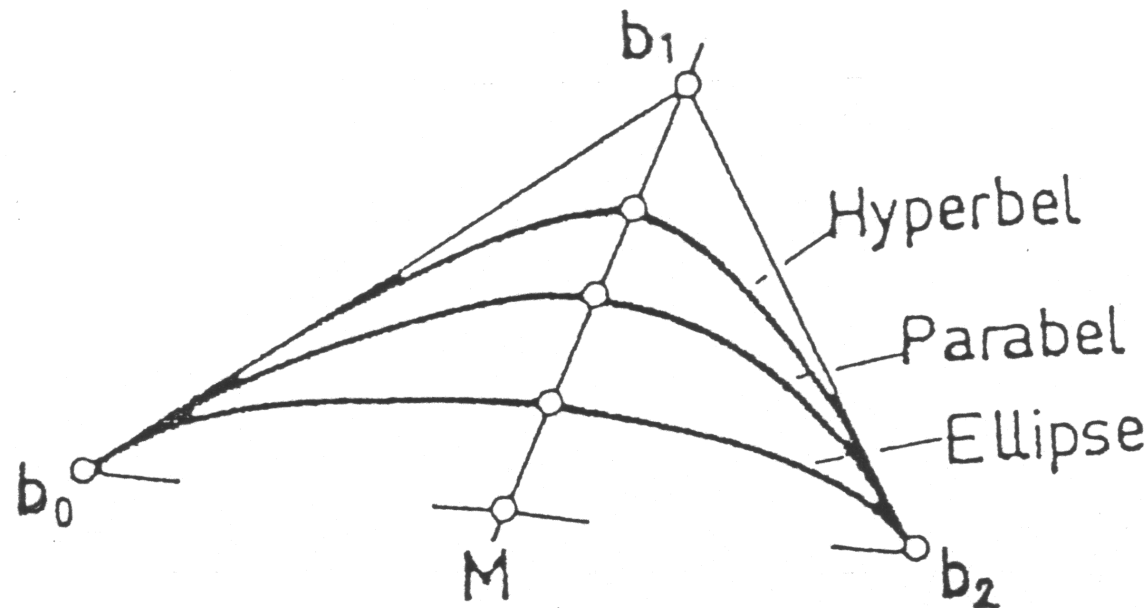


Figure: Shoulder point construction

For $s = 1/2$ we get a parabola, for $s < 1/2$ an ellipse, and for $s > 1/2$ a hyperbola.