

# 计算机辅助几何设计

## 2023秋学期

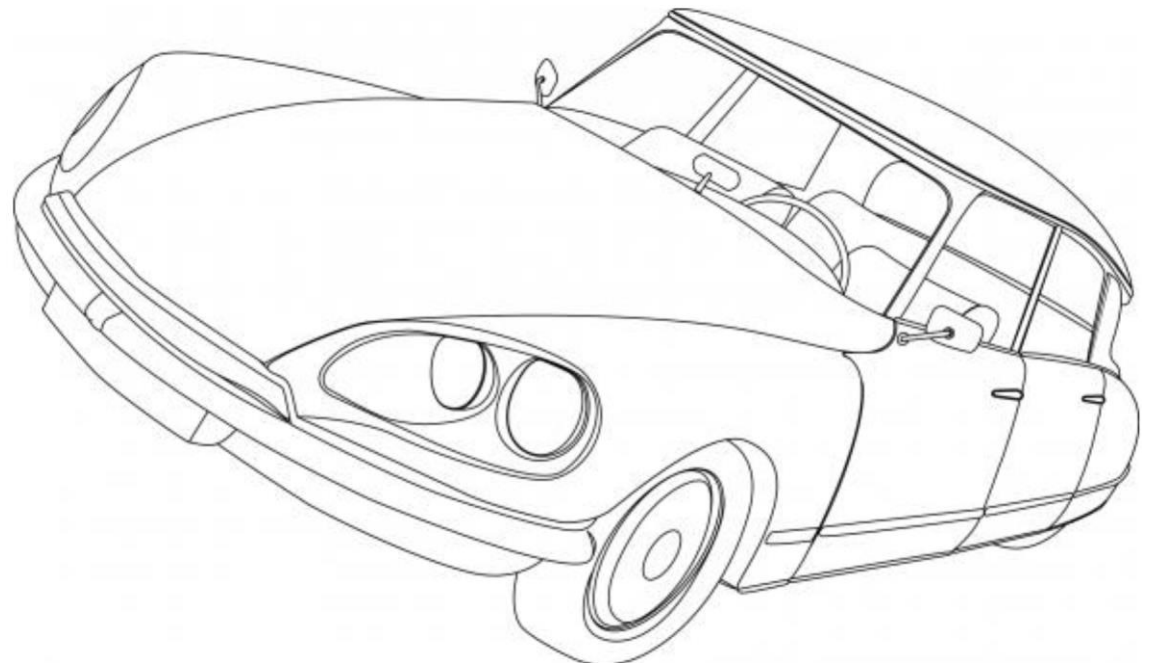
# Bézier Curves

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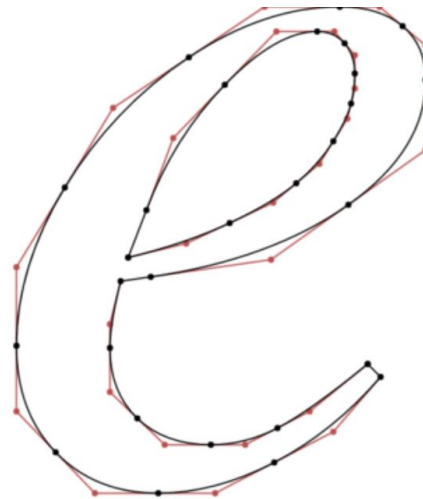
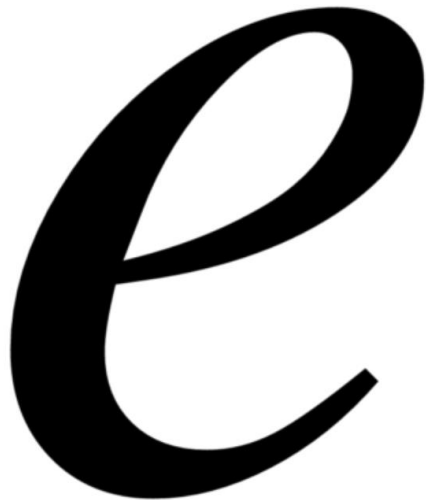
# Bézier curves

- Bézier curves/splines developed by
  - Paul de Casteljaou at Citroen (1959)
  - Pierre Bézier at Renault (1963)for free-form parts in automotive design



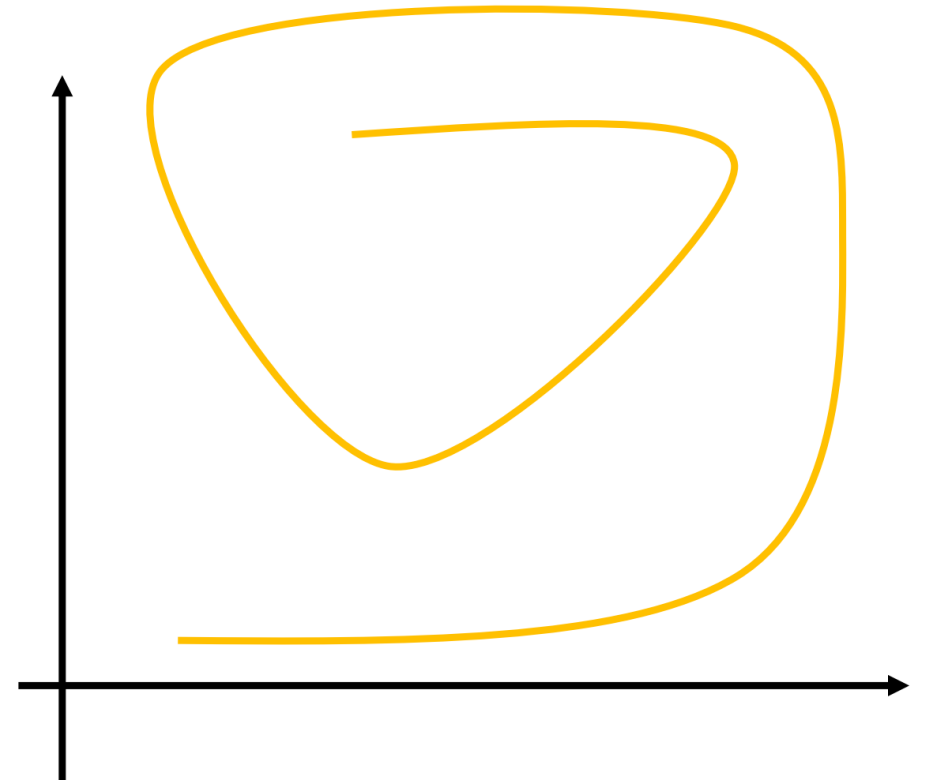
# Bézier curves

- Today: Standard tool for 2D curve editing
- Cubic 2D Bézier curves are everywhere:
  - Inkscape, Corel Draw, Adobe Illustrator, Powerpoint, ...
  - PDF, Truetype (quadratic curves), Windows GDI, ...
- Widely used in 3D curve & surface modeling as well



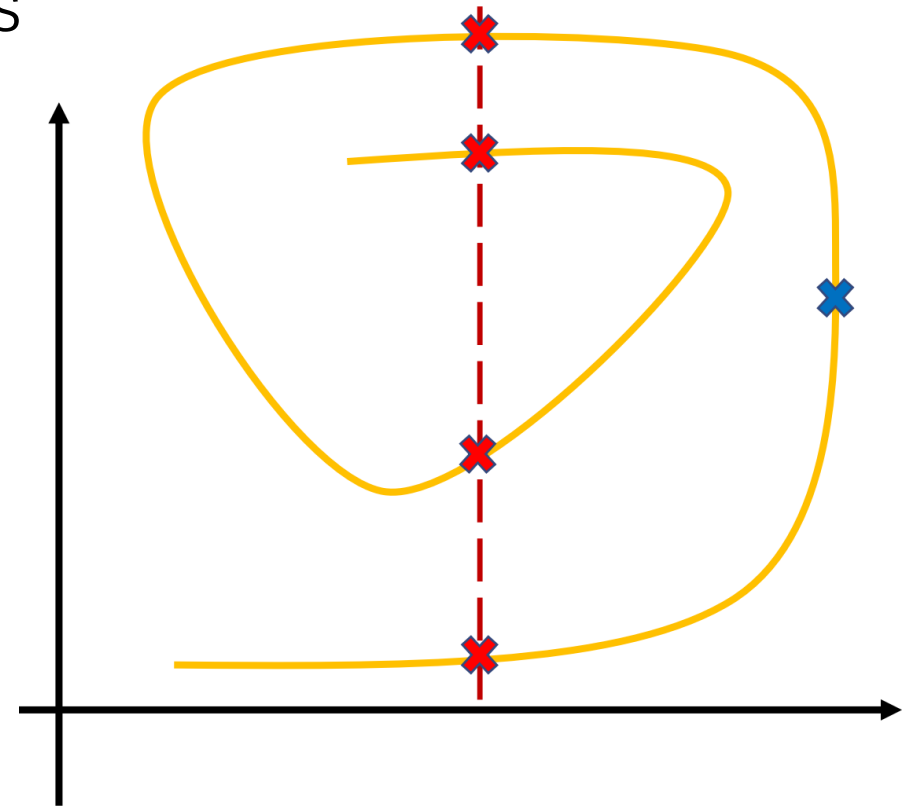
# Curve representation

- The implicit curve form  $f(x, y) = 0$  suffers from several limitations:



# Curve representation

- The implicit curve form  $f(x, y) = 0$  suffers from several limitations:
  - Multiple values for the same  $x$ -coordinates
  - Undefined derivative  $\frac{dy}{dx}$  (see blue cross)
  - Not invariant w.r.t axes transformations

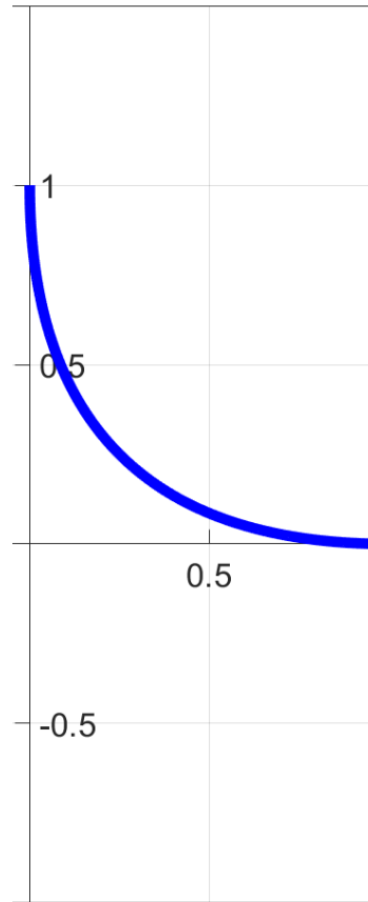


# Parametric representation

- Remedy: parametric representation  $c(t) = (x(t), y(t))$ 
  - Easy evaluations
  - The parameter  $t$  can be interpreted as time
  - The curve can be interpreted as the path traced by a moving particle

# Modeling with the power basis, ...

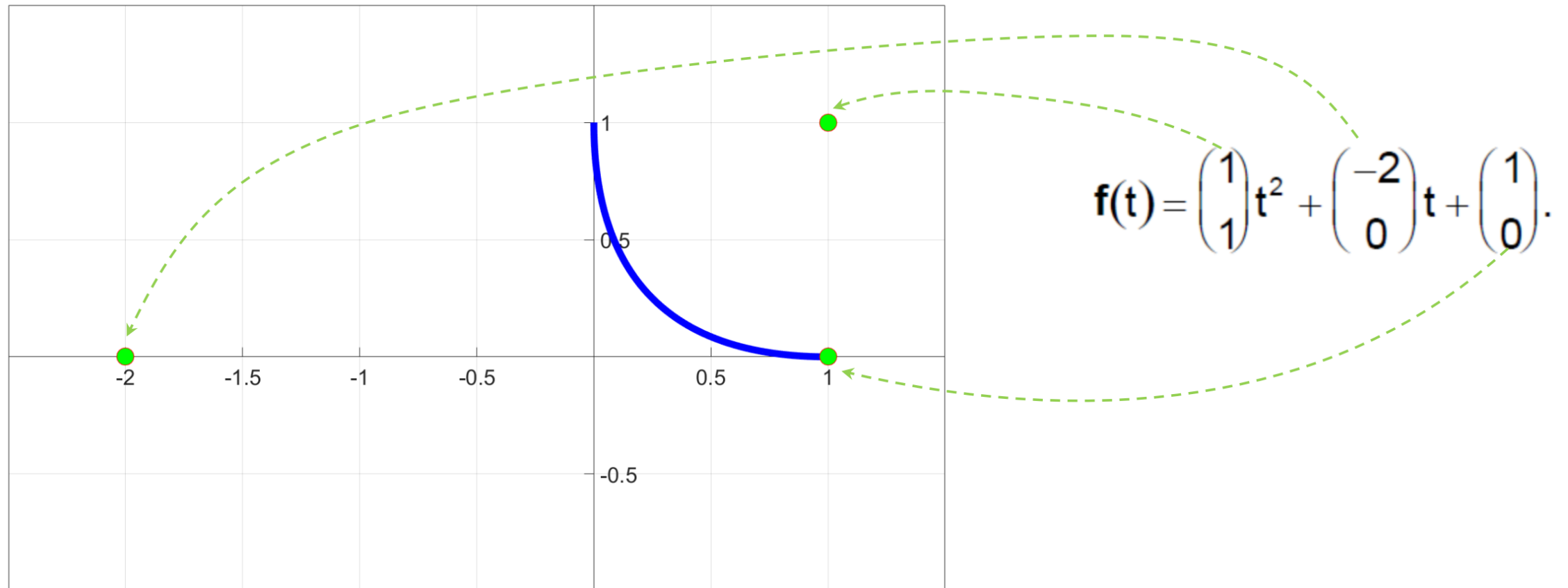
- Example of a parabola:  $f(t) = at^2 + bt + c$



$$f(t) = \begin{pmatrix} 1 \\ 1 \end{pmatrix} t^2 + \begin{pmatrix} -2 \\ 0 \end{pmatrix} t + \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

# Modeling with the power basis, ... no thanks!

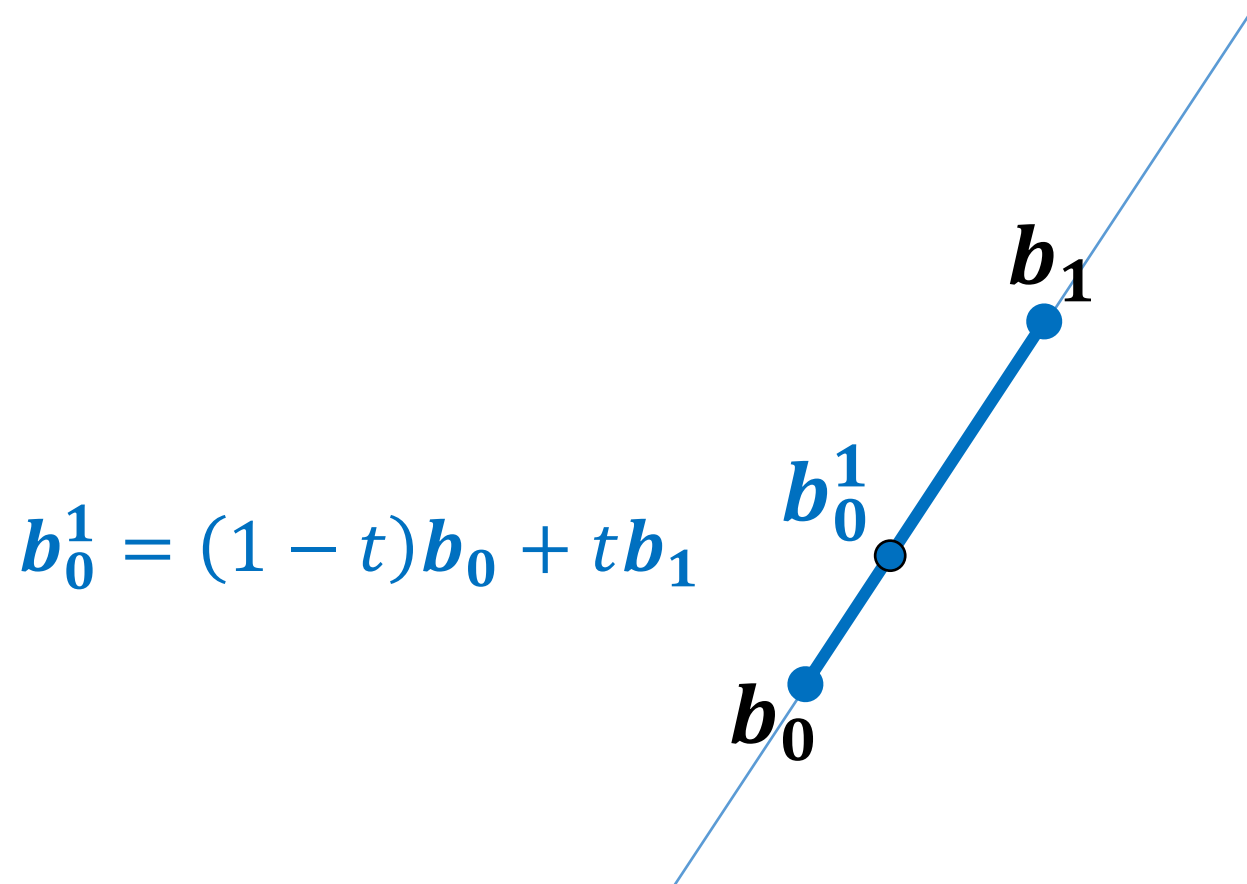
- Examples of a parabola:  $f(t) = at^2 + bt + c$ : the coefficients of the power basis lack intuitive geometric meaning





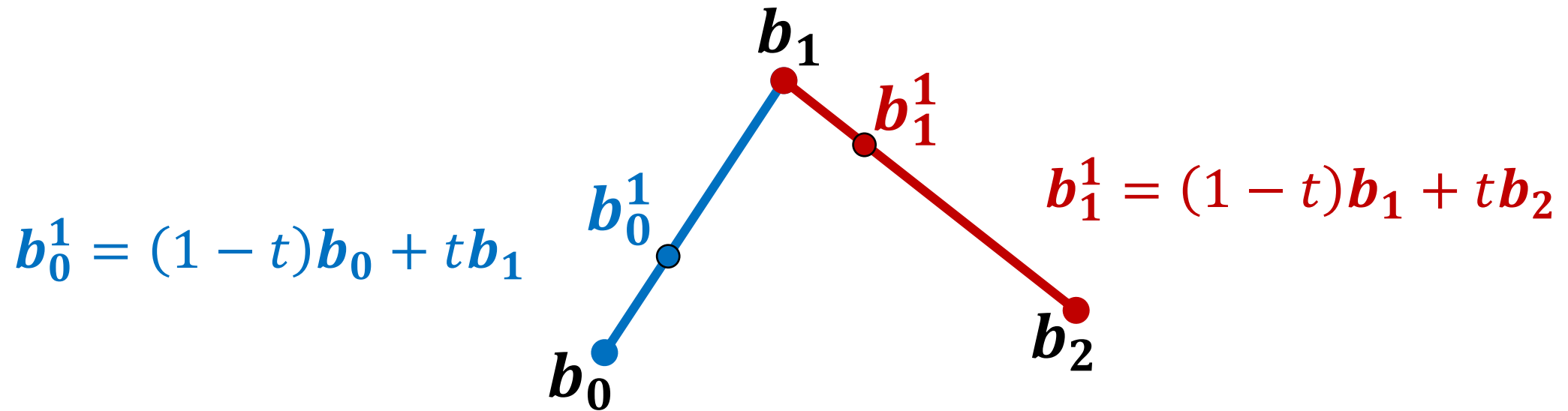
# Back to the drawing board

- A point on a parametric line



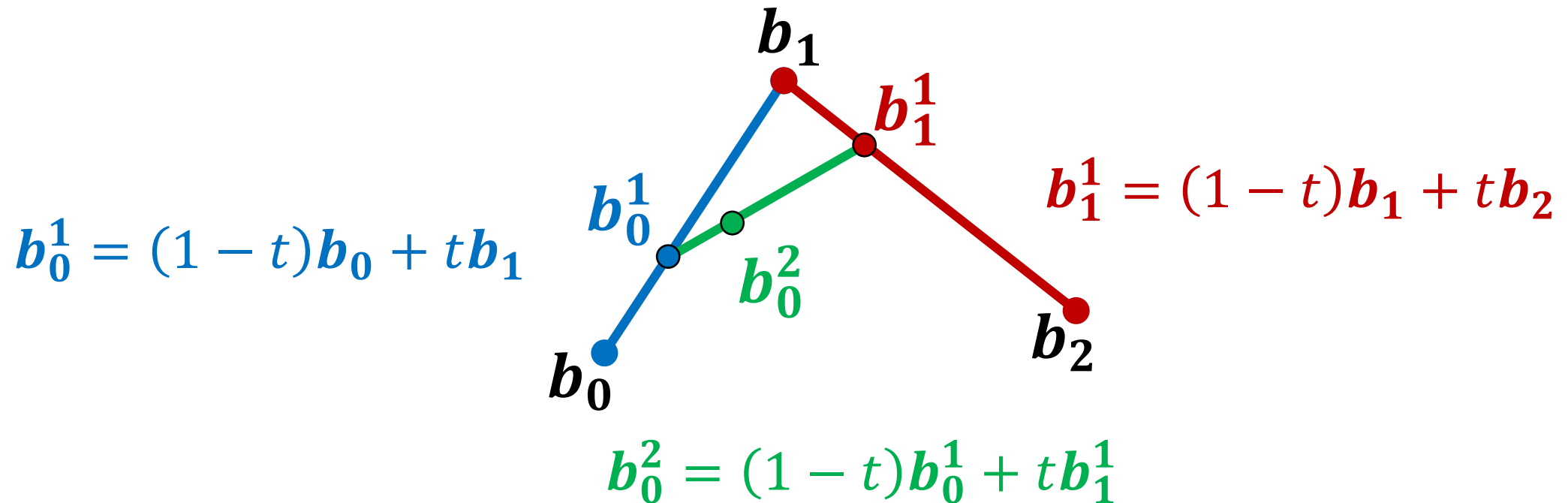
# Back to the drawing board

- Another point on a second parametric line



# Back to the drawing board

- A third point on the line defined by the first two points



# Back to the drawing board

- And then simplify...

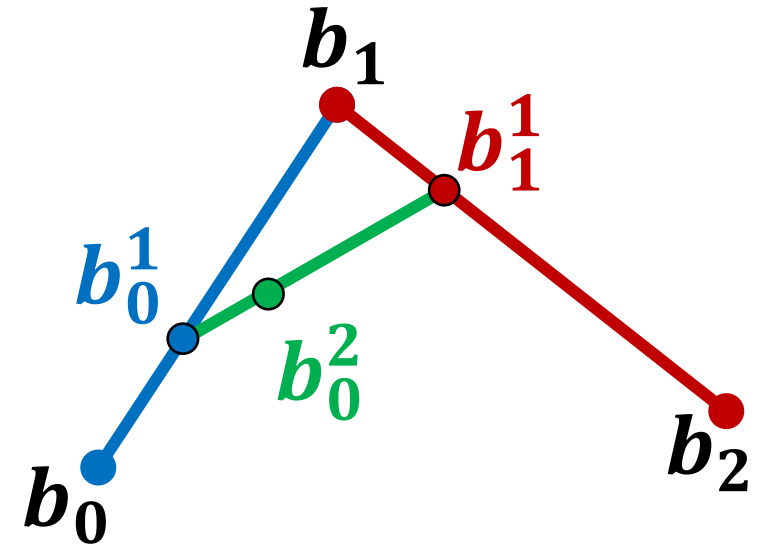
$$b_0^1 = (1 - t)b_0 + tb_1$$

$$b_0^2 = (1 - t)b_0^1 + tb_1^1$$

$$b_1^1 = (1 - t)b_1 + tb_2$$

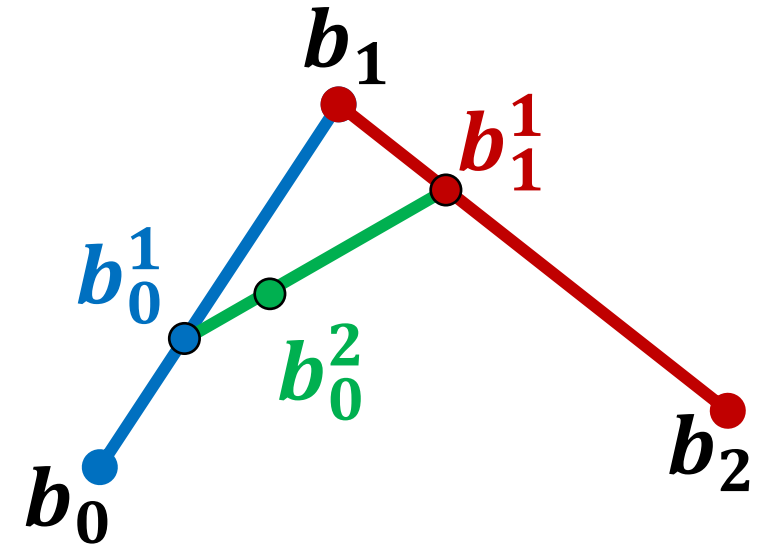
$$b_0^2 = (1 - t)[(1 - t)b_0 + tb_1] + t[(1 - t)b_1 + tb_2]$$

$$b_0^2 = (1 - t)^2 b_0 + 2t(1 - t)b_1 + t^2 b_2$$



# Back to the drawing board

- We obtained another description of parabolic curves
- The coefficients  $b_0, b_1, b_2$  have a geometric meaning



$$b_0^2 = (1 - t)^2 b_0 + 2t(1 - t)b_1 + t^2 b_2$$

# Example re-visited

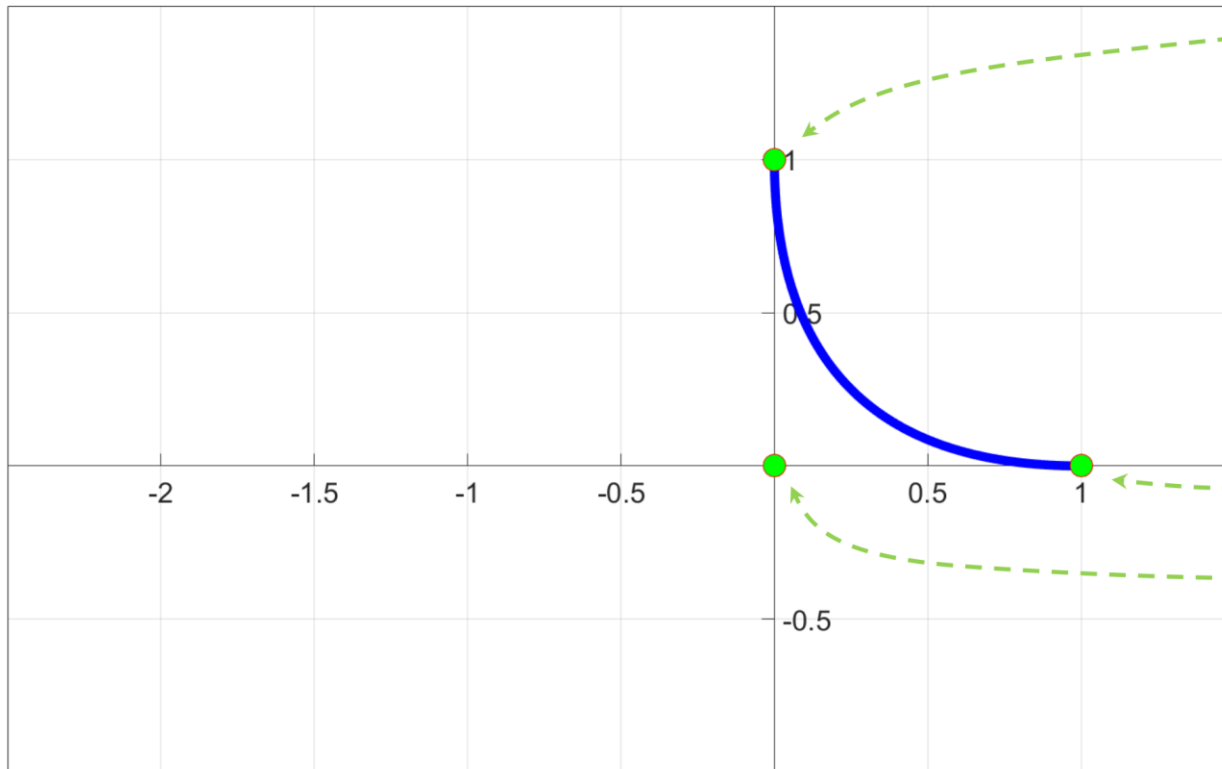
- Let's rewrite our initial parabolic curve example in the new basis

$$\mathbf{f}(t) = \begin{pmatrix} 1 \\ 1 \end{pmatrix} t^2 + \begin{pmatrix} -2 \\ 0 \end{pmatrix} t + \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\mathbf{f}(t) = \begin{pmatrix} 1 \\ 0 \end{pmatrix} (1-t)^2 + \begin{pmatrix} 0 \\ 0 \end{pmatrix} 2t(1-t) + \begin{pmatrix} 0 \\ 1 \end{pmatrix} t^2$$

# Example re-visited

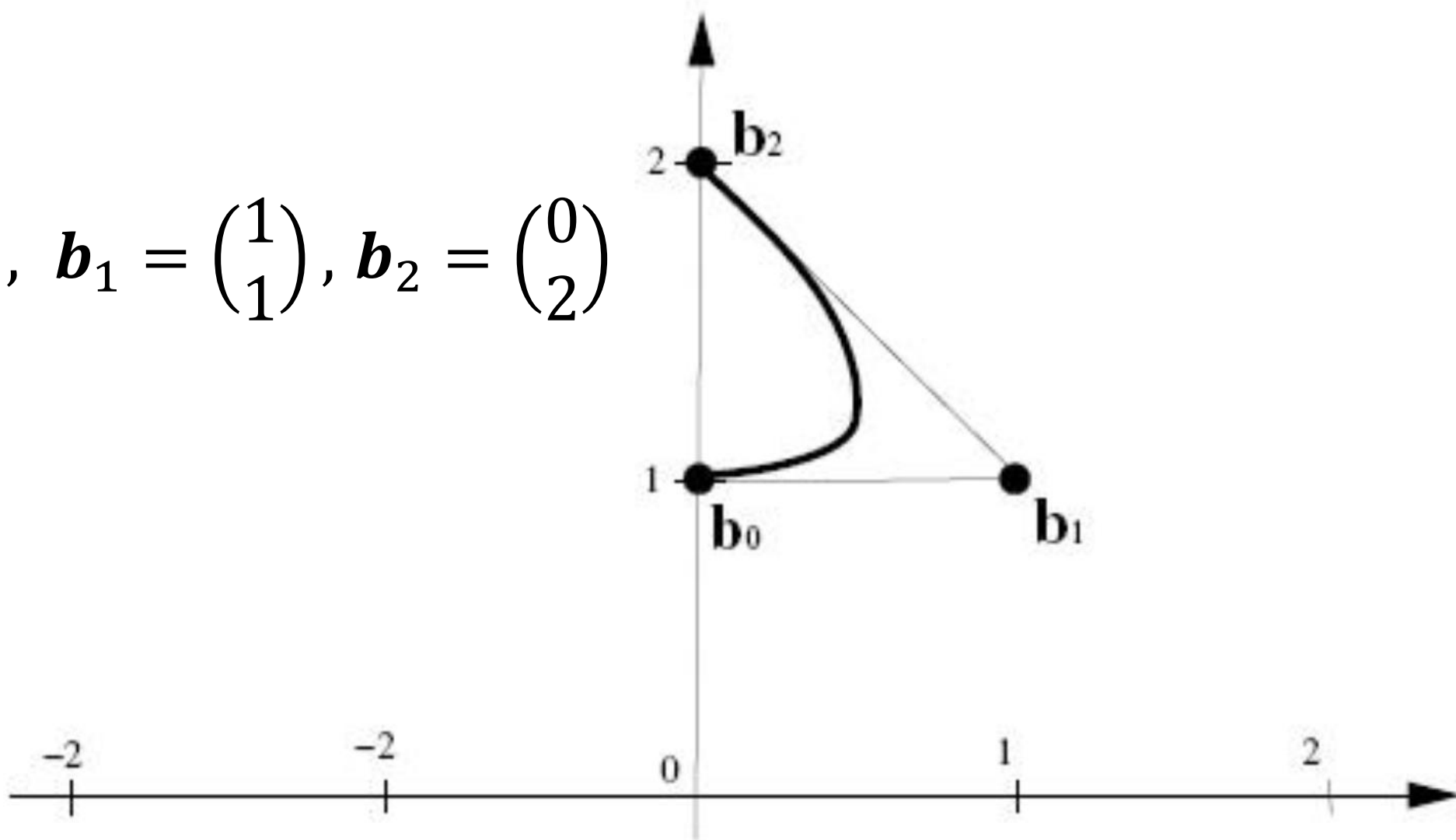
- The coefficients have a geometric meaning
- More intuitive for curve manipulation



$$\mathbf{f}(t) = \begin{pmatrix} 1 \\ 0 \end{pmatrix} (1-t)^2 + \begin{pmatrix} 0 \\ 0 \end{pmatrix} 2t(1-t) + \begin{pmatrix} 0 \\ 1 \end{pmatrix} t^2$$

# Another example

$$\mathbf{b}_0 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \mathbf{b}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \mathbf{b}_2 = \begin{pmatrix} 0 \\ 2 \end{pmatrix}$$





# Going further

- Cubic approximation

- Given 4 points:  $\mathbf{p}_0^0(t) = \mathbf{p}_0$ ,  $\mathbf{p}_1^0(t) = \mathbf{p}_1$ ,  $\mathbf{p}_2^0(t) = \mathbf{p}_2$ ,  $\mathbf{p}_3^0(t) = \mathbf{p}_3$

- First iteration

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

$$\mathbf{p}_1^1 = (1 - t)\mathbf{p}_1 + t\mathbf{p}_2$$

$$\mathbf{p}_2^1 = (1 - t)\mathbf{p}_2 + t\mathbf{p}_3$$

- 2<sup>nd</sup> iteration

$$\mathbf{p}_0^2 = (1 - t)^2\mathbf{p}_0 + 2t(1 - t)\mathbf{p}_1 + t^2\mathbf{p}_2$$

$$\mathbf{p}_1^2 = (1 - t)^2\mathbf{p}_1 + 2t(1 - t)\mathbf{p}_2 + t^2\mathbf{p}_3$$

- Curve

$$\mathbf{c}(t) = (1 - t)^3\mathbf{p}_0 + 3t(1 - t)^2\mathbf{p}_1 + 3t^2(1 - t)\mathbf{p}_2 + t^3\mathbf{p}_3$$

Throughout these examples, we just re-invented a primitive version of the de Casteljau algorithm

Now let's examine it more closely ...

## CAGD杂志将出版专辑，纪念Paul de Casteljau的开创性贡献

原创 ggc 图形学与几何计算 2022-09-18 15:58 发表于北京

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#图形资讯

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2022年3月24日，CAGD的先驱之一，长期在法国雪铁龙公司工作的Paul de Faget de Casteljau先生不幸逝世。为了纪念他的开创性贡献，CAGD杂志准备出版一期专辑怀念他，欢迎投稿！

### de Casteljau先生的历史性贡献

de Casteljau先生于1930年11月19日出生于法国的Besançon，是一位法国的物理学家和数学家，任职于雪铁龙公司，研究汽车外形设计的算法和系统。他和法国另一个汽车公司雷诺公司的工程师Pierre Bézier，分别独立地发展了一套后来被称为Bézier曲线曲面的理论。

de Casteljau先生因为他名字命名的de Casteljau算法闻名，对于一条 $n+1$ 个控制顶点的Bézier曲线，

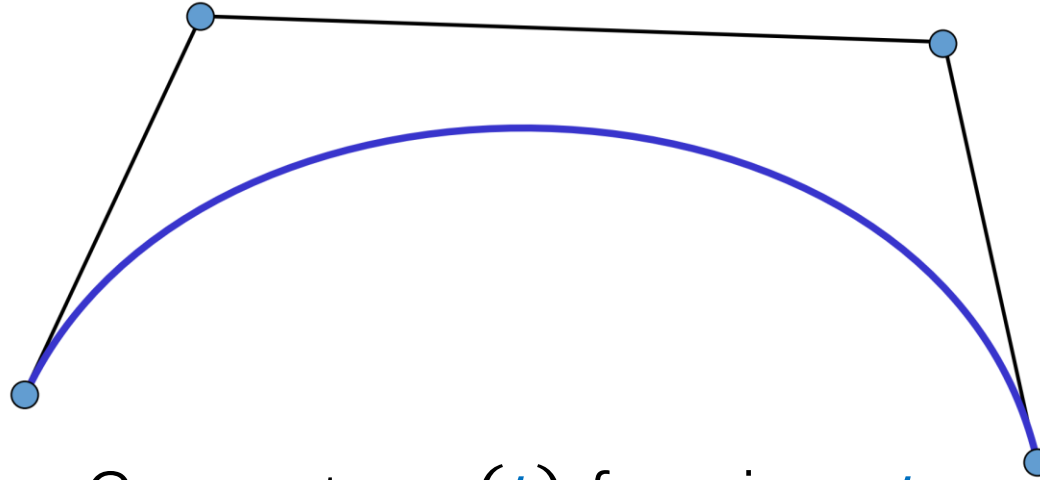
$$P(t) = \sum_{i=0}^n P_i B_{i,n}(t), \quad t \in [0,1]$$

曲线上参数  $t$  对应的型值点可由如下递归算法计算：

$$P_i^k = \begin{cases} P_i & k = 0 \\ (1-t)P_i^{k-1} + tP_{i+1}^{k-1} & k = 1, 2, \dots, n, \\ & i = 0, 1, \dots, n-k \end{cases}$$

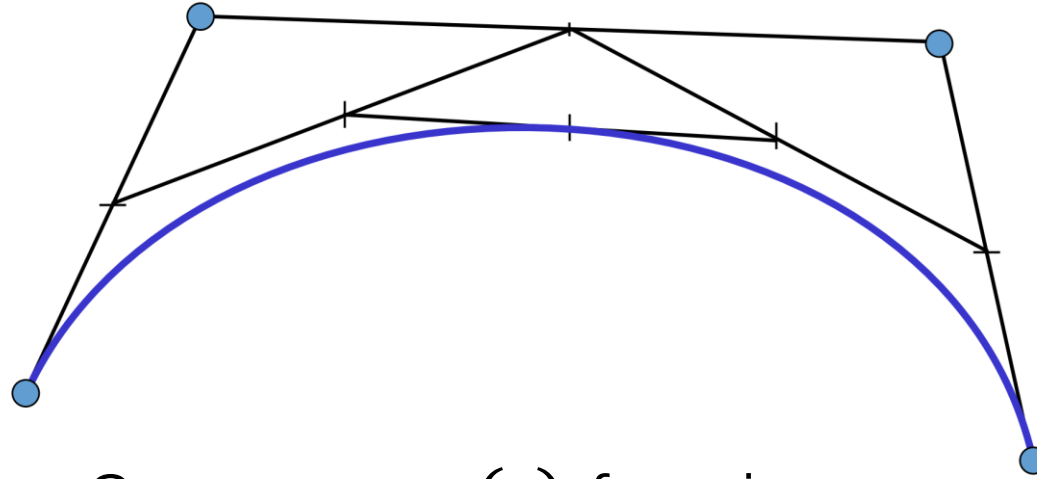
de Casteljau先生于2012年荣获Bézier奖。这是几何造型领域的最高奖，于2007年由Solid Modeling Association (SMA) 设立，并以另一位CAGD先驱Pierre Bézier的名字命名。由Vadim Shairo (主席), Pere Brunet, Christoph Hoffmann, Shi-Min Hu, Kunwoo Lee, Diensh Manocha和Malcolm Sabin组成的Bézier奖委员会在其获奖公告中提到了de Casteljau先生的学术贡献。

# De Casteljau algorithm



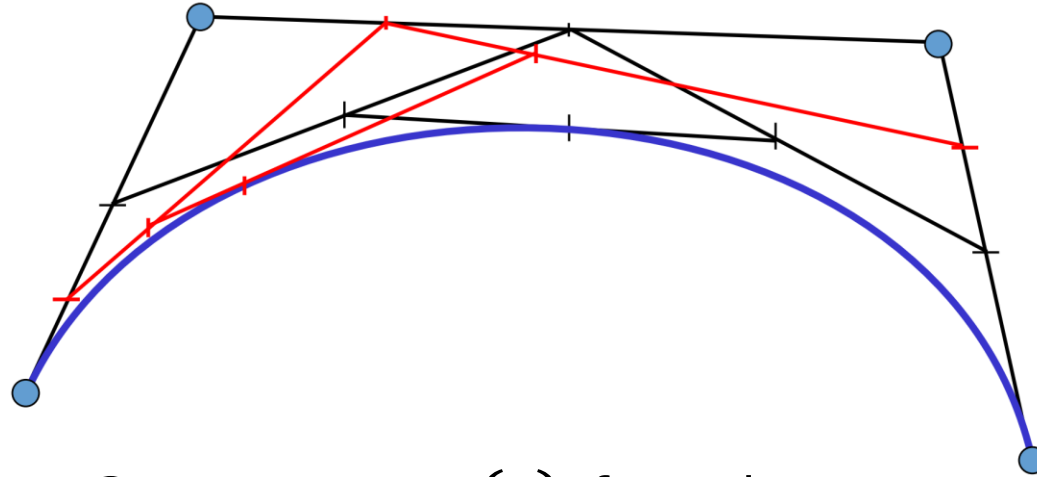
- De Casteljau Algorithm: Computes  $x(t)$  for given  $t$ 
  - Bisect control polygon in ratio  $t: (1 - t)$
  - Connect the new dots with lines (adjacent segments)
  - Interpolate again with the same ratio
  - Iterate, until only one points is left

# De Casteljau algorithm



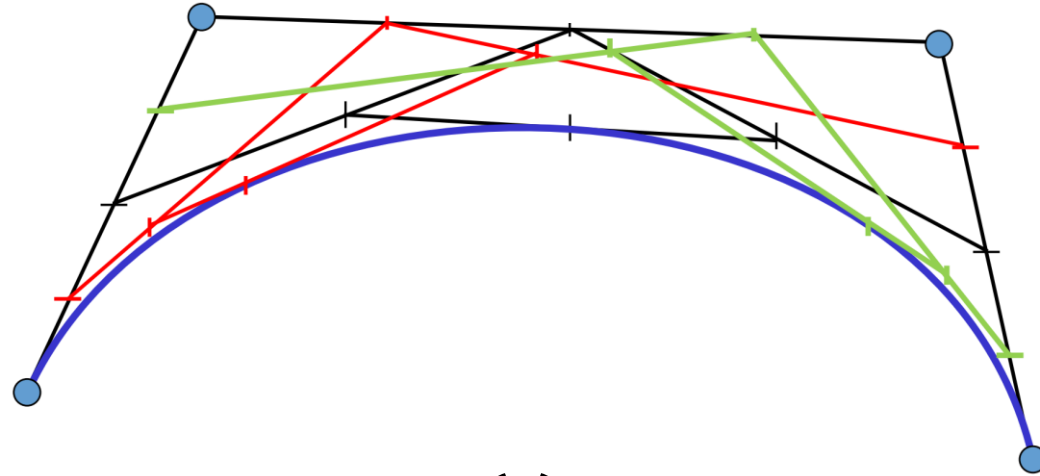
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# De Casteljau algorithm



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# De Casteljau algorithm

- Algorithm description

- Input: points  $\mathbf{b}_0, \mathbf{b}_1, \dots, \mathbf{b}_n \in \mathbb{R}^3$
- Output: curve  $\mathbf{x}(t), t \in [0,1]$

- Geometric construction of the points  $\mathbf{x}(t)$  for given  $t$ :

$$\mathbf{b}_i^0(t) = \mathbf{b}_i, \quad i = 0, \dots, n$$

$$\mathbf{b}_i^r(t) = (1 - t)\mathbf{b}_i^{r-1}(t) + t\mathbf{b}_{i+1}^{r-1}(t)$$

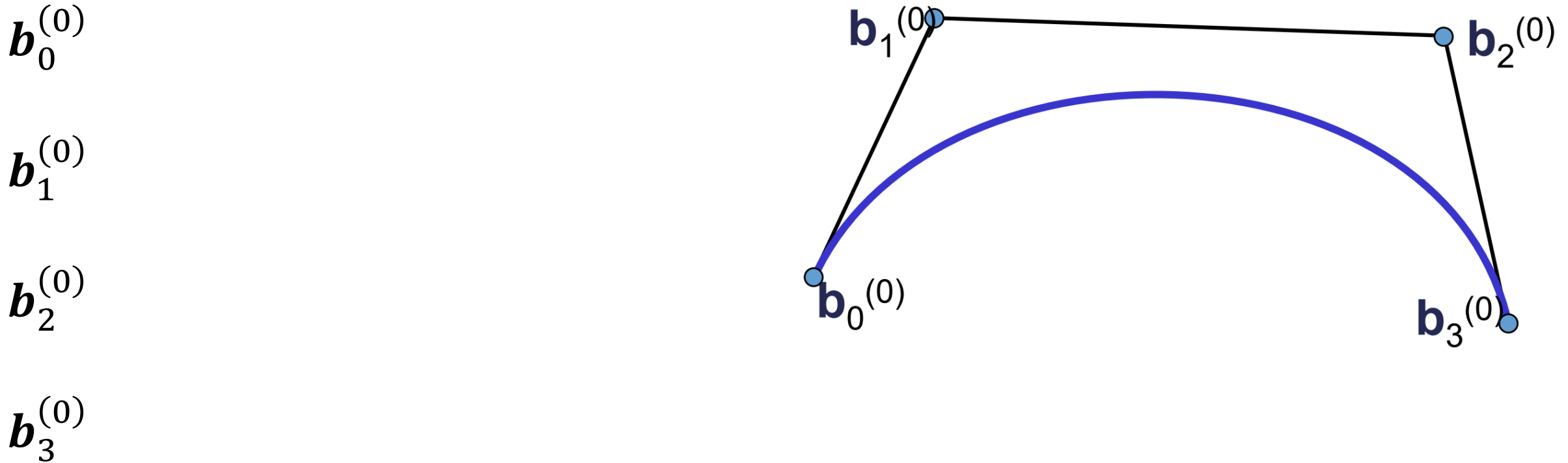
$$r = 1, \dots, n \quad i = 0, \dots, n - r$$

- Then  $\mathbf{b}_0^n(t)$  is the searched curve point  $\mathbf{x}(t)$  at the parameter value  $t$

# De Casteljau algorithm

- Repeated convex combination of control points

$$\mathbf{b}_i^{(r)} = (1 - t)\mathbf{b}_i^{(r-1)} + t\mathbf{b}_{i+1}^{(r-1)}$$

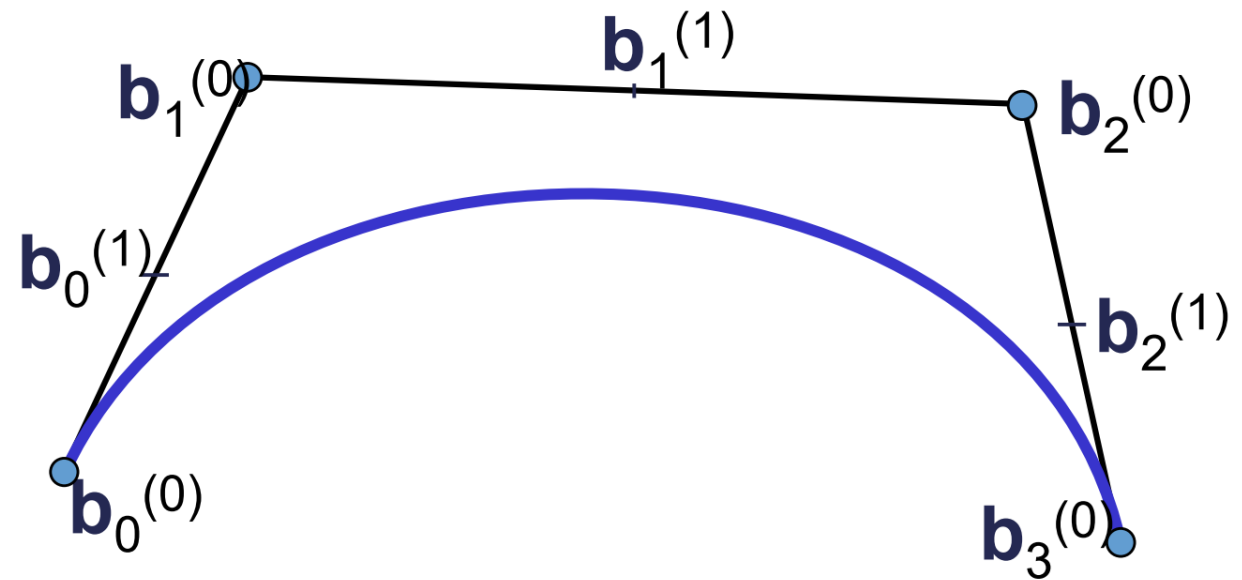
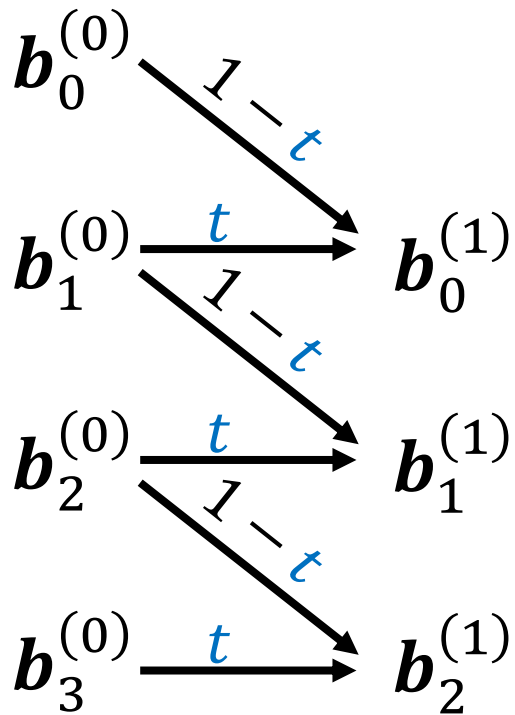




# De Casteljau algorithm

- Repeated convex combination of control points

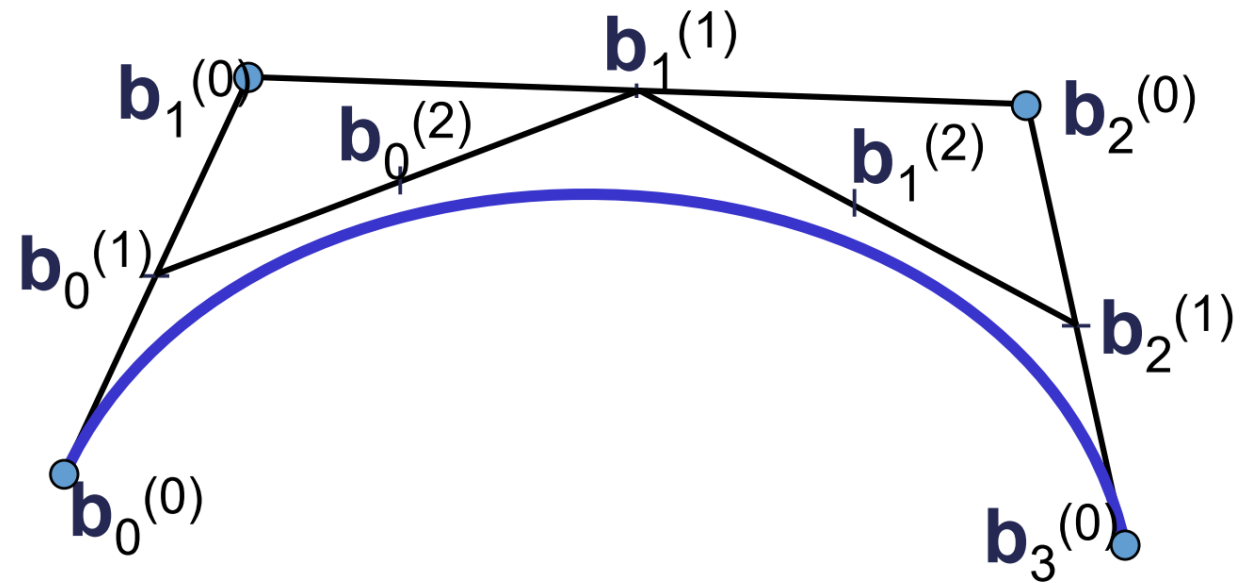
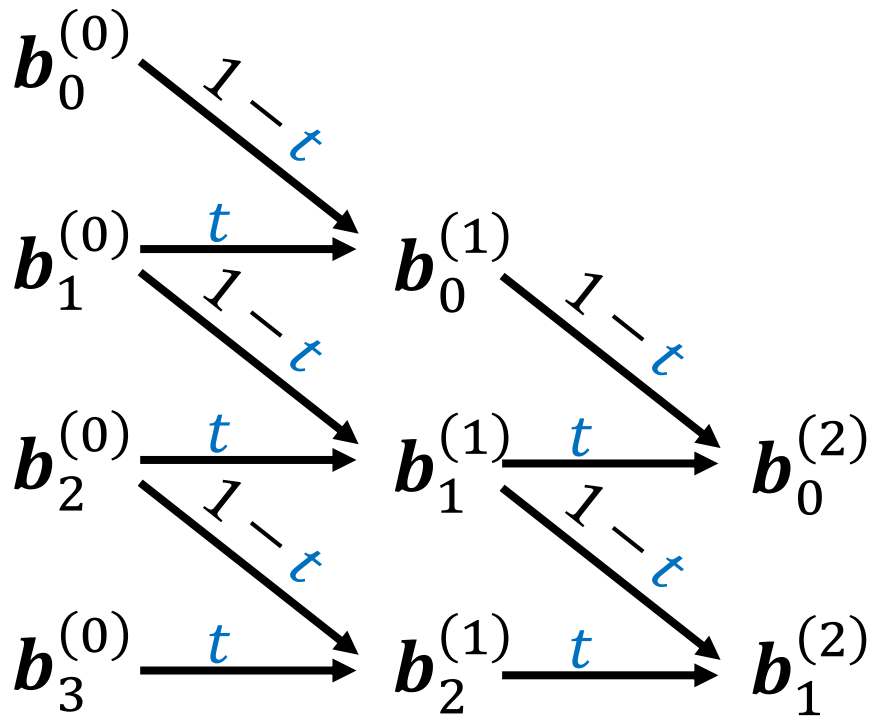
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# De Casteljau algorithm

- Repeated convex combination of control points

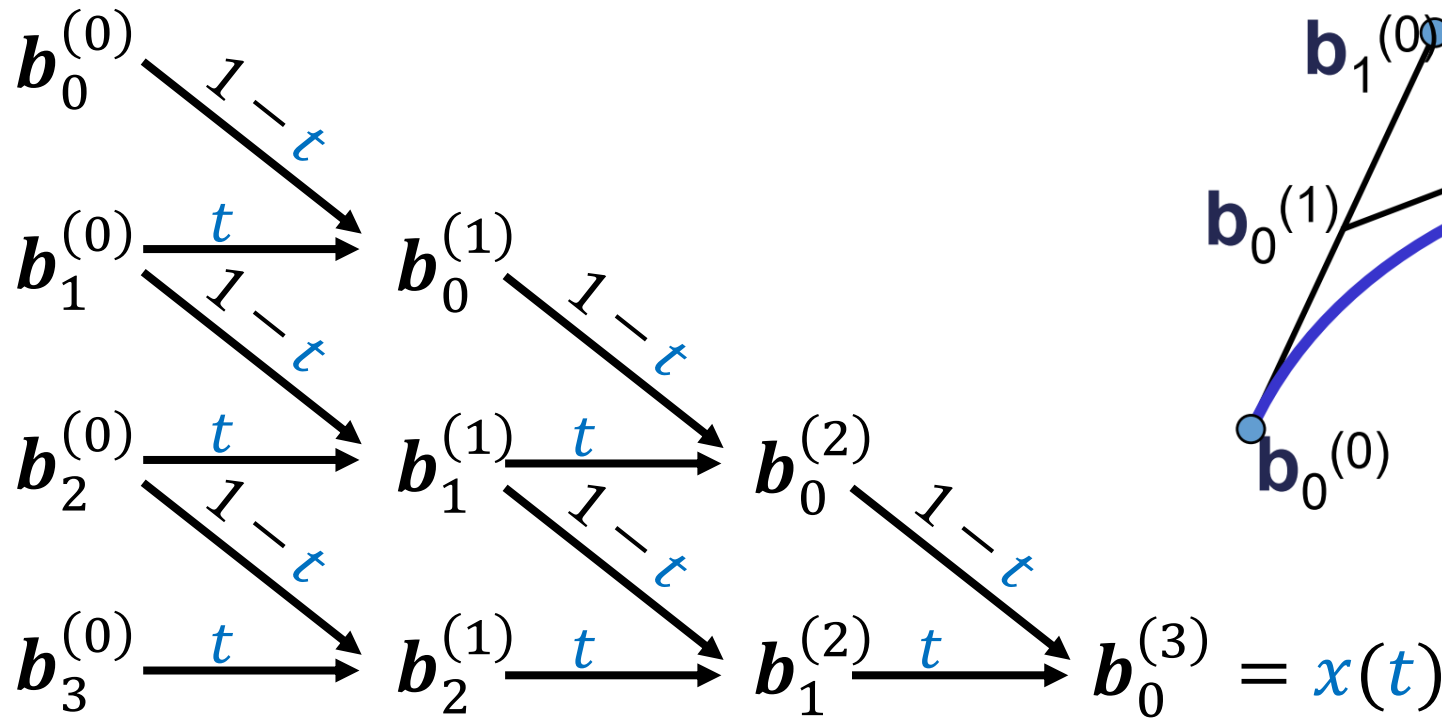
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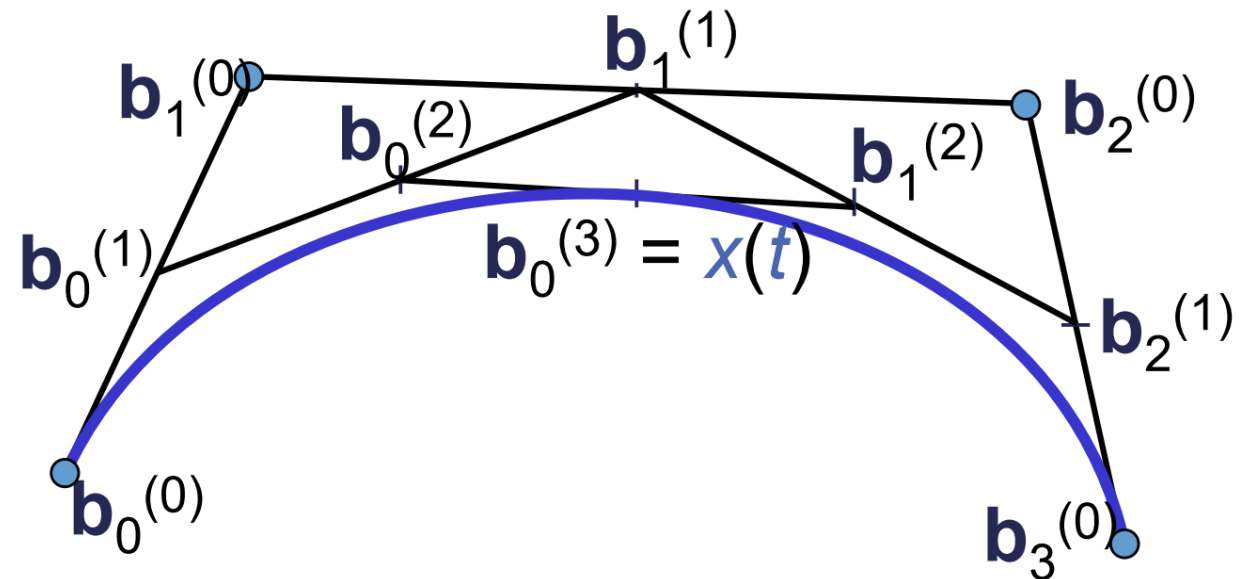
# De Casteljau algorithm

- Repeated convex combination of control points

$$\mathbf{b}_i^{(r)} = (1 - t)\mathbf{b}_i^{(r-1)} + t\mathbf{b}_{i+1}^{(r-1)}$$



De Casteljau scheme



# De Casteljau algorithm

- The intermediate coefficients  $\mathbf{b}_i^r(t)$  can be written in a triangular matrix: the de Casteljau scheme:

$$\mathbf{b}_0 = \mathbf{b}_0^0$$

$$\mathbf{b}_1 = \mathbf{b}_1^0 \quad \mathbf{b}_0^1$$

$$\mathbf{b}_2 = \mathbf{b}_2^0 \quad \mathbf{b}_1^1 \quad \mathbf{b}_0^2$$

$$\mathbf{b}_3 = \mathbf{b}_3^0 \quad \mathbf{b}_2^1 \quad \mathbf{b}_1^2 \quad \mathbf{b}_0^3$$

.....

$$\mathbf{b}_{n-1} = \mathbf{b}_{n-1}^0 \quad \mathbf{b}_{n-2}^1 \quad \dots \quad \mathbf{b}_0^{n-1}$$

$$\mathbf{b}_n = \mathbf{b}_n^0 \quad \mathbf{b}_{n-1}^1 \quad \dots \quad \mathbf{b}_1^{n-1} \quad \mathbf{b}_0^n = x(t)$$

# De Casteljau algorithm

Algorithm:

for  $r=1..n$

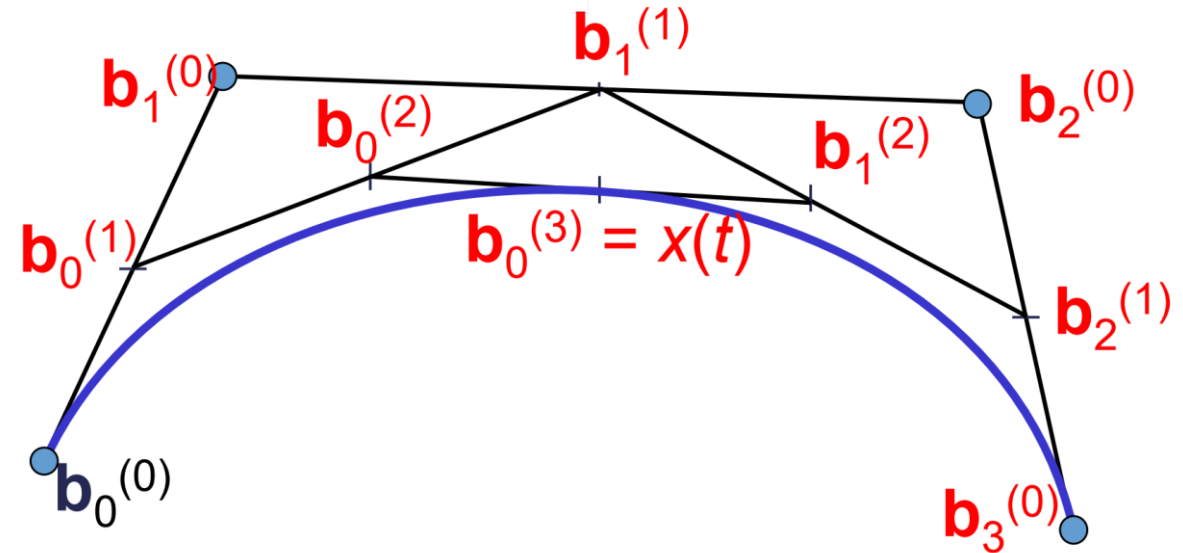
for  $i=0..n-r$

$$\mathbf{b}_i^{(r)} = (1 - t) \mathbf{b}_i^{(r-1)} + t \mathbf{b}_{i+1}^{(r-1)}$$

end

end

return  $\mathbf{b}_0^{(n)}$



The whole algorithm consists only of repeated linear interpolations.

# De Casteljau algorithm: Properties

- The polygon consisting of the points  $\mathbf{b}_0, \dots, \mathbf{b}_n$  is called **Bézier polygon** (control polygon)
- The points  $\mathbf{b}_i$  are called **Bézier points** (control points)
- The curve defined by the Bézier points  $\mathbf{b}_0, \dots, \mathbf{b}_n$  and the de Casteljau algorithm is called **Bézier curve**
- The de Casteljau algorithm is numerically stable, since only convex combinations are applied.
- Complexity of the de Casteljau algorithm
  - $O(n^2)$  time
  - $O(n)$  memory
  - with  $n$  being the number of Bézier points

# De Casteljau algorithm: Properties

- **Properties of Bézier curves:**

- Given: Bézier points  $\mathbf{b}_0, \dots, \mathbf{b}_n$

Bézier curve  $\mathbf{x}(t)$

- Bézier curve is polynomial curve of degree  $n$
- End points interpolation:  $\mathbf{x}(0) = \mathbf{b}_0$ ,  $\mathbf{x}(1) = \mathbf{b}_n$ . The remaining Bézier points are only approximated in general

- **Convex hull property:**

Bézier curve is completely inside the convex hull of its Bézier polygon

# De Casteljau algorithm: Properties

- **Variation diminishing**
  - No line intersects the Bézier curve more often than its Bézier polygon
- **Influence of Bézier points:** global but pseudo-local
  - Global: moving a Bézier points changes the whole curve progression
  - Pseudo-local:  $\mathbf{b}_i$  has its maximal influence on  $x(t)$  at  $t = \frac{i}{n}$
- **Affine invariance:**
  - Bézier curve and Bézier polygon are invariant under affine transformations
- **Invariance under affine parameter transformations**



# De Casteljau algorithm: Properties

- **Symmetry**

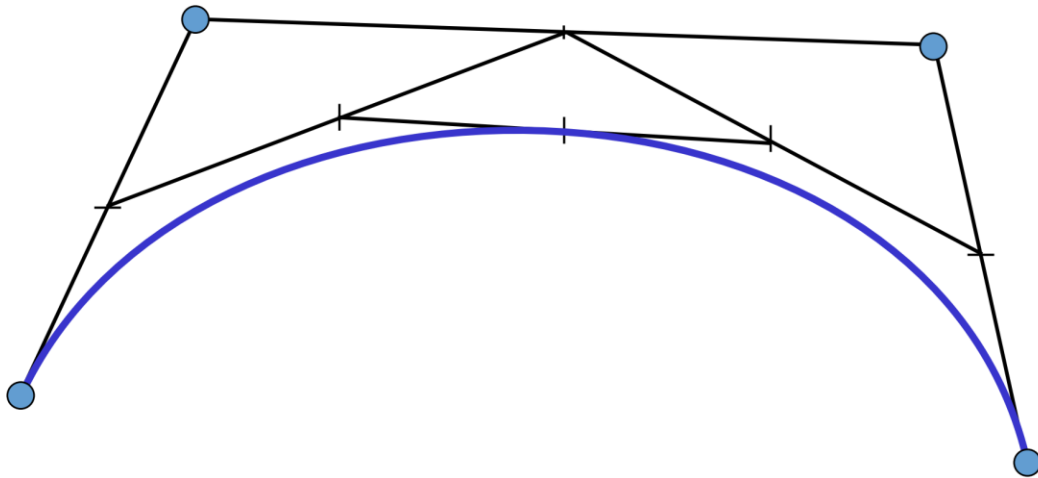
- The following two Bézier curves coincide, they are only traversed in opposite directions:

$$\mathbf{x}(t) = [\mathbf{b}_0, \dots, \mathbf{b}_n] \quad \mathbf{x}'(t) = [\mathbf{b}_n, \dots, \mathbf{b}_0]$$

- **Linear Precision:**

- Bézier curve is line segment, if  $\mathbf{b}_0, \dots, \mathbf{b}_n$  are colinear
- Invariance under barycentric combinations

# Recap



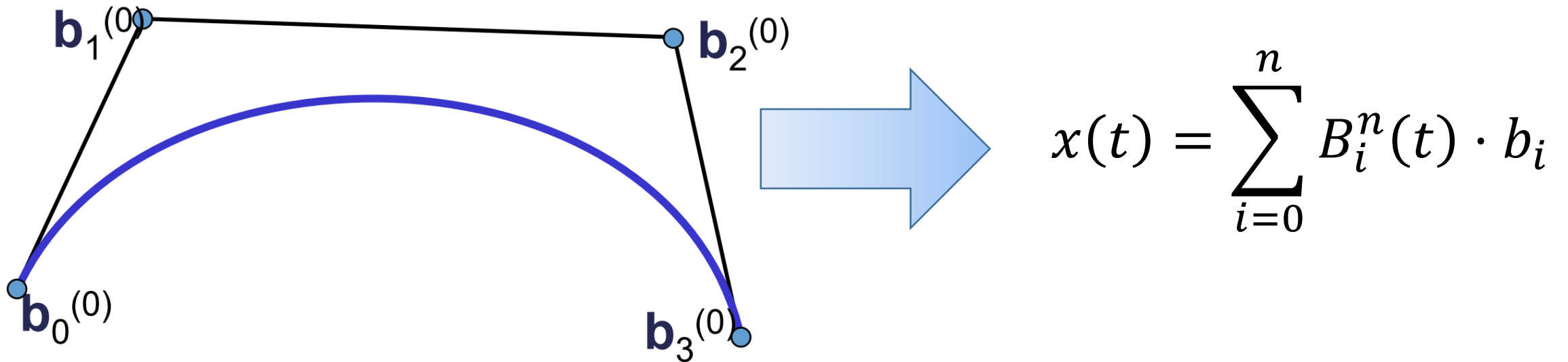
de Casteljau algorithm

# Bézier Curves

Towards a polynomial description

# Bézier Curves

## Towards a polynomial description



# Polynomial description of Bézier curves

- The same problem as before:
  - Given:  $(n + 1)$  control points  $\mathbf{b}_0, \dots, \mathbf{b}_n$
  - Wanted: Bézier curve  $\mathbf{x}(t)$  with  $t \in [0,1]$
- Now with an algebraic approach using basis functions

# Desirable Properties

- Useful requirements for a basis:
  - Well behaved curve
    - Smooth basis functions

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# Desirable Properties

- Useful requirements for a basis:
    - Well behaved curve
      - Smooth basis functions
    - Local control (or at least semi-local)
      - Basis functions with compact support
    - **Affine invariance:**
      - Applying an affine map  $\mathbf{x} \rightarrow A\mathbf{x} + \mathbf{b}$  on
        - Control points
        - Curve
- Should have the same effect**
- In particular: rotation, translation
  - Otherwise: interactive curve editing very difficult



# Desirable Properties

- Useful requirements for a basis:
  - **Convex hull property:**
    - The curve lays within the convex hull of its control points
    - Avoids at least too weird oscillations
  - Advantages
    - Computational advantages (recursive intersection tests)
    - More predictable behavior

# Summary

- Useful properties
  - Smoothness
  - Local control / support
  - **Affine invariance**
  - **Convex hull property**

## Notations

Curve

basis function

control points

$$f(t) = \sum_{i=1}^n b_i(t) \mathbf{p}_i$$

# Affine Invariance

- Affine map:  $\mathbf{x} \rightarrow A\mathbf{x} + \mathbf{b}$
- **Part I:** Linear invariance – we get this automatically

- Linear approach:  $\mathbf{f}(t) = \sum_{i=1}^n b_i(t)\mathbf{p}_i = \sum_{i=1}^n b_i(t) \begin{pmatrix} p_i^{(x)} \\ p_i^{(y)} \\ p_i^{(z)} \end{pmatrix}$

- Therefore:  $A(\mathbf{f}(t)) = A(\sum_{i=1}^n b_i(t)\mathbf{p}_i) = \sum_{i=1}^n b_i(t)(A\mathbf{p}_i)$

# Affine Invariance

- Affine Invariance:
  - Affine map:  $\mathbf{x} \rightarrow A\mathbf{x} + \mathbf{b}$
  - **Part II:** Translational invariance

$$\sum_{i=1}^n b_i(t)(\mathbf{p}_i + \mathbf{b}) = \sum_{i=1}^n b_i(t)\mathbf{p}_i + \sum_{i=1}^n b_i(t)\mathbf{b} = \mathbf{f}(t) + \left( \sum_{i=1}^n b_i(t) \right) \mathbf{b}$$

- For translational invariance, the sum of the basis functions must be one *everywhere* (for all parameter values  $t$  that are used).
- This is called “**partition of unity** property”
- The  $b_i$ ’s form an “**affine combination**” of the control points  $\mathbf{p}_i$
- This is very important for modeling

# Convex Hull Property

- Convex combinations:

- A convex combination of a set of points  $\{\mathbf{p}_1, \dots, \mathbf{p}_n\}$  is any point of the form:

$$\sum_{i=1}^n \lambda_i \mathbf{p}_i \text{ with } \sum_{i=1}^n \lambda_i = 1 \text{ and } \forall i = 1 \dots n: 0 \leq \lambda_i \leq 1$$

- (Remark:  $\lambda_i \leq 1$  is redundant)
- The set of all admissible convex combinations forms the convex hull of the point set
  - Easy to see (exercise): The convex hull is the smallest set that contains all points  $\{\mathbf{p}_1, \dots, \mathbf{p}_n\}$  and every complete straight line between two elements of the set

# Convex Hull Property

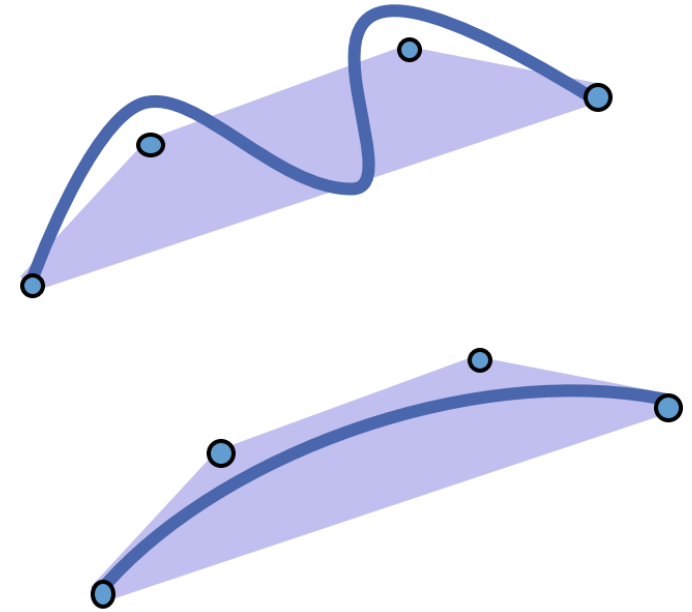
- Accordingly:

- If we have this property

$$\forall t \in \Omega: \sum_{i=1}^n b_i(t) = 1 \text{ and } \forall t \in \Omega, \forall i: b_i(t) \geq 0$$

the constructed curves / surfaces will be:

- Affine invariant (translations, linear maps)
  - Be restricted to the convex hull of the control points
- 
- Corollary: Curves will have *linear precision*
    - All control points lie on a straight line  
⇒ Curve is a straight line segment
  - Surfaces with planar control points will be flat, too



# Convex Hull Property

- Very useful property in practice
  - Avoids at least the worst oscillations
    - no escape from convex hull, unlike polynomial interpolation
  - Linear precision property is intuitive (people expect this)
  - Can be used for fast range checks
    - Test for intersection with convex hull first, then the object
    - Recursive intersection algorithms in conjunctions with subdivision rules (more on this later)



# Polynomial description of Bézier curves

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  - Wanted: Bézier curve  $\mathbf{x}(t)$  with  $t \in [0,1]$
- Now with an algebraic approach using basis functions
- Need to define  $n + 1$  basis functions
  - Such that this describes a Bézier curve:

$$B_0^n(t), \dots, B_n^n(t) \text{ over } [0,1]$$
$$\mathbf{x}(t) = \sum_{i=0}^n B_i^n(t) \cdot \mathbf{b}_i$$



# Bernstein Basis

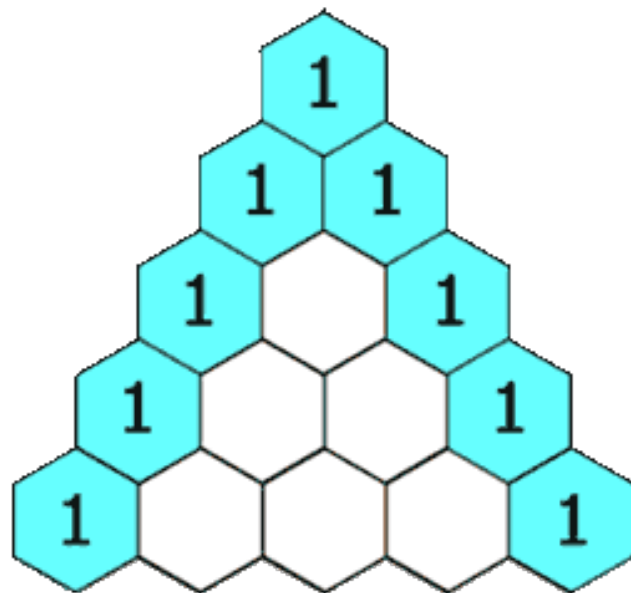
- Let's examine the Bernstein basis:  $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ 
  - Bernstein basis of degree  $n$ :

$$B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i} = B_{i\text{-th basis function}}^{(\text{degree})}$$

where the binomial coefficients are given by:

$$\binom{n}{i} = \begin{cases} \frac{n!}{(n-i)! i!} & \text{for } 0 \leq i \leq n \\ 0 & \text{otherwise} \end{cases}$$

# Binomial Coefficients and Theorem



$$\binom{n}{i} + \binom{n}{i+1} = \binom{n+1}{i+1}$$

					1					
						1		1		
				1		2		1		
			1		3		3		1	
		1		4		6		4		1
	1		5	10		10		5		1

$$(x + y)^n = \sum_{i=0}^n \binom{n}{i} x^i y^{n-i}$$

# Examples: The first few

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

- The first three Bernstein bases:

$$B_0^{(0)} := 1$$

$$B_0^{(1)} := 1 - t \quad B_1^{(1)} := t$$

$$B_0^{(2)} := (1 - t)^2 \quad B_1^{(2)} := 2t(1 - t) \quad B_2^{(2)} := t^2$$

$$B_0^{(3)} := (1 - t)^3 \quad B_1^{(3)} := 3t(1 - t)^2 \quad B_2^{(3)} := 3t^2(1 - t) \quad B_3^{(3)} := t^3$$

# Examples: The first few

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

$$B_0^{(0)} := 1$$

$$B_0^{(1)} := 1 - t$$

$$B_1^{(1)} := t$$

$$B_0^{(2)} := (1-t)^2$$

$$B_1^{(2)} := 2t(1-t)$$

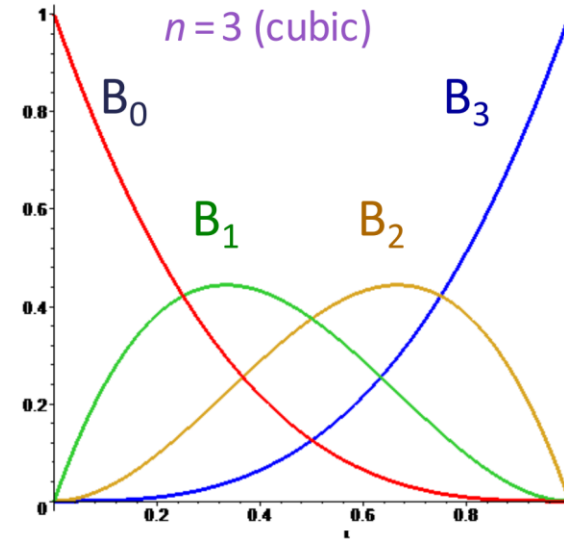
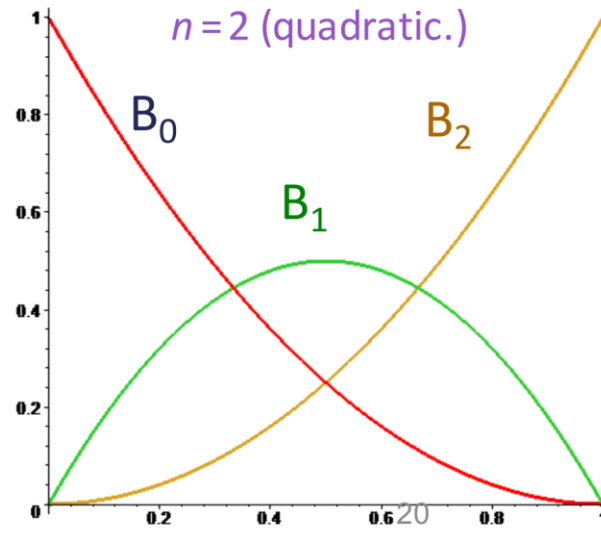
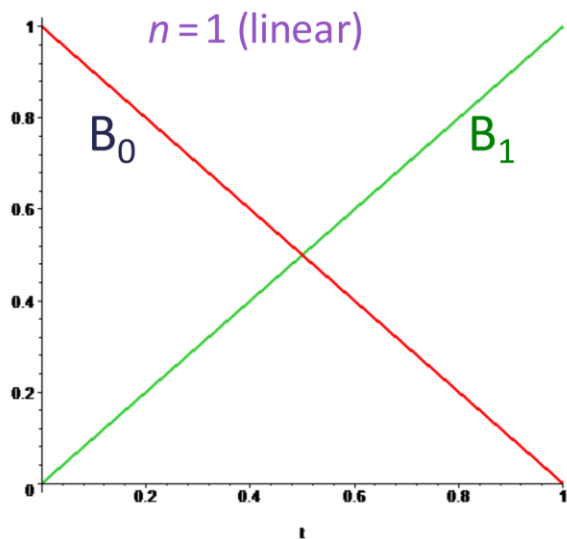
$$B_2^{(2)} := t^2$$

$$B_0^{(3)} := (1-t)^3$$

$$B_1^{(3)} := 3t(1-t)^2$$

$$B_2^{(3)} := 3t^2(1-t)$$

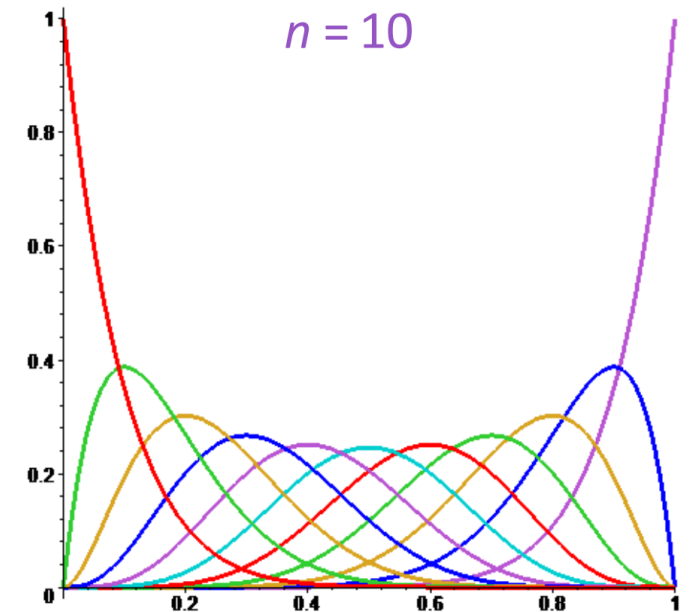
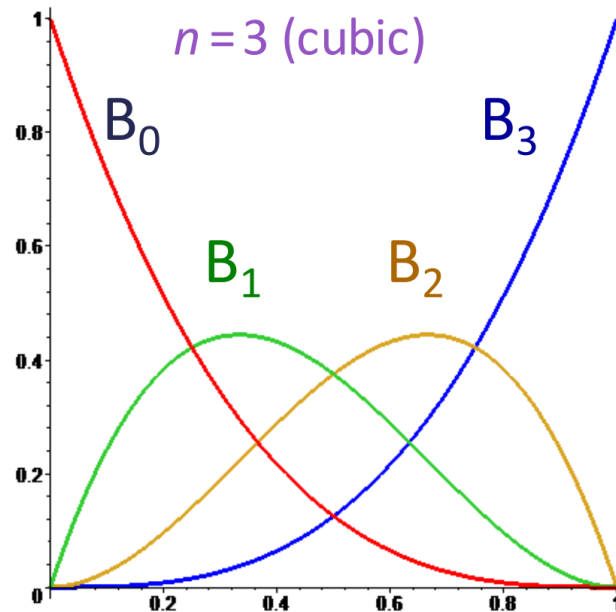
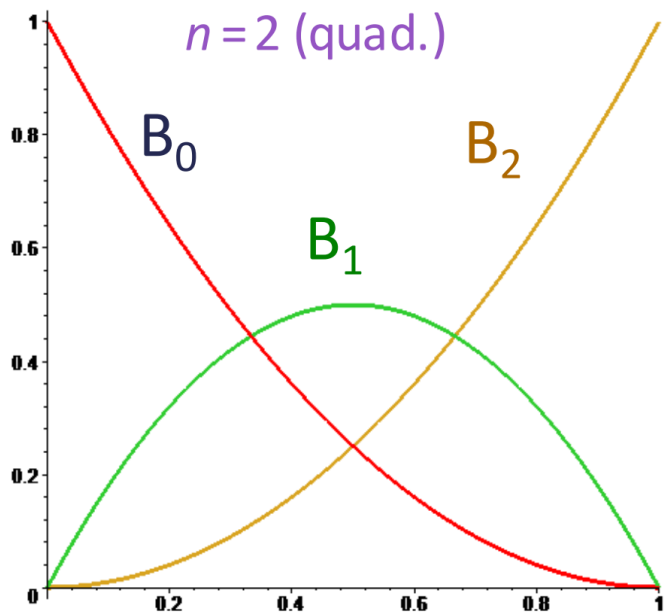
$$B_3^{(3)} := t^3$$



# Bernstein Basis

- Bézier curves use the Bernstein basis:  $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ 
  - Bernstein basis of degree  $n$ :

$$B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i} = B_{i\text{-th basis function}}^{(\text{degree})}$$



# Bernstein Basis

- What about the desired properties?
  - Smoothness
  - Local control / support
  - Affine invariance
  - Convex hull property

# Bernstein Basis: Properties

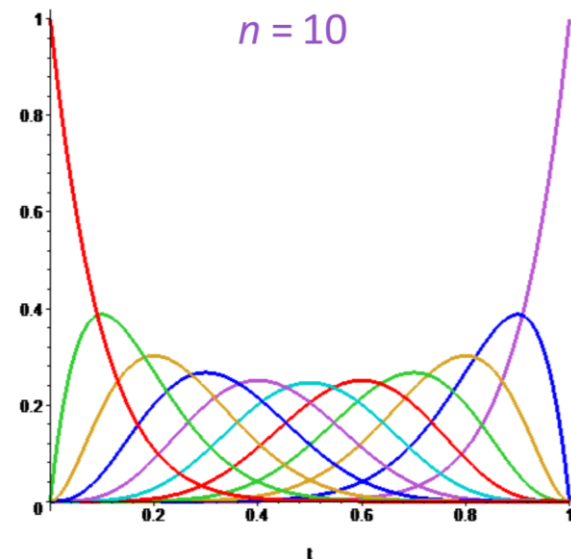
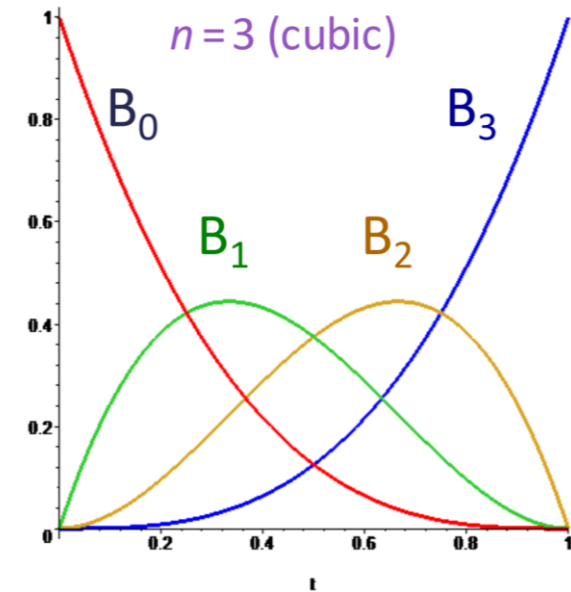
- $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ ,  $B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i}$

- Basis for polynomials of degree  $n$

Smoothness

- Each basis function  $B_i^{(n)}$  has its maximum at  $t = \frac{i}{n}$

Local control (semi-local)



# Bernstein Basis: Properties

- $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ ,  $B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i}$

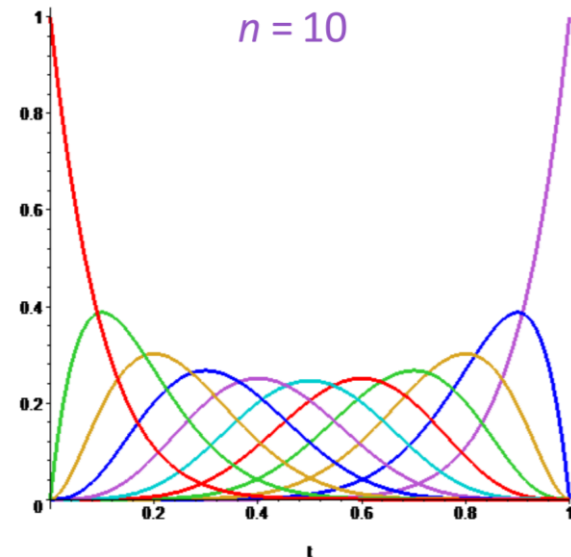
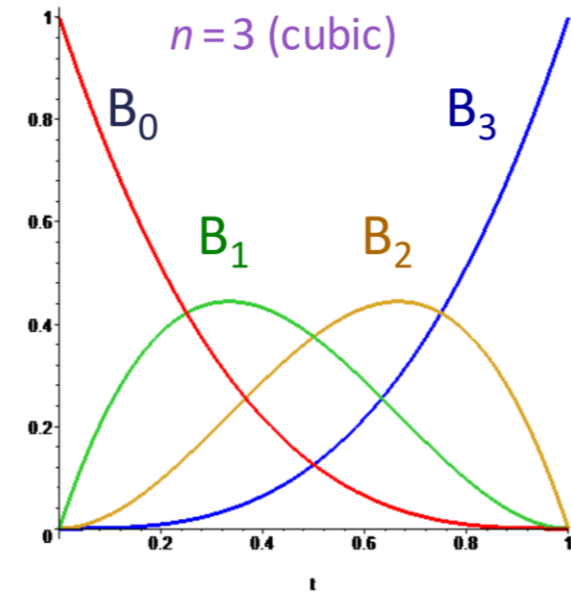
Affine invariance

Convex hull property

- Partition of unity (binomial theorem)

$$1 = (1 - t + t)^n$$

$$\sum_{i=0}^n B_i^{(n)}(t) = (t + (1-t))^n = 1$$





# What about the desired properties?

- Smoothness Yes
- Local control / support To some extent
- Affine invariance Yes
- Convex hull property Yes

$$\binom{n-1}{i} + \binom{n-1}{i-1} = \binom{n}{i}$$

# Bernstein Basis: Properties

- $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ ,  $B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i}$

- Recursive computation

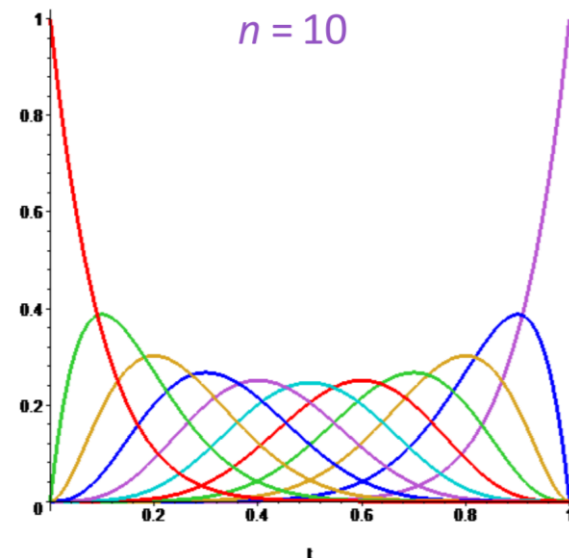
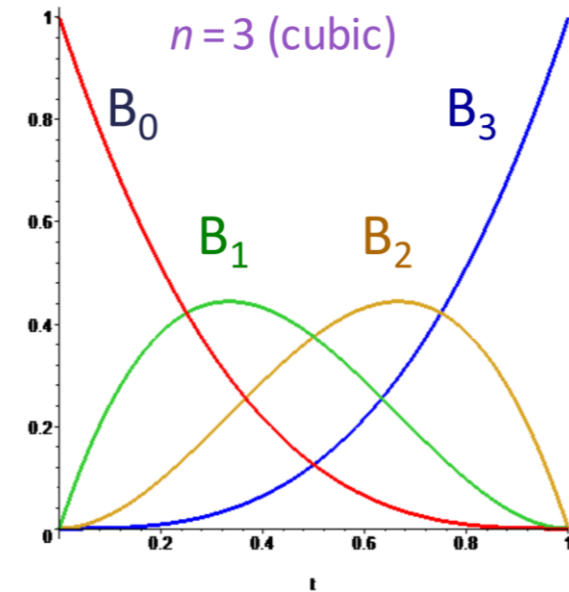
$$B_i^n(t) := (1-t)B_i^{(n-1)}(t) + tB_{i-1}^{(n-1)}(1-t)$$

with  $B_0^0(t) = 1$ ,  $B_i^n(t) = 0$  for  $i \notin \{0 \dots n\}$

- Symmetry

$$B_i^n(t) = B_{n-i}^n(1-t)$$

- Non-negativity:  $B_i^{(n)}(t) \geq 0$  for  $t \in [0..1]$



# Bernstein Basis: Properties

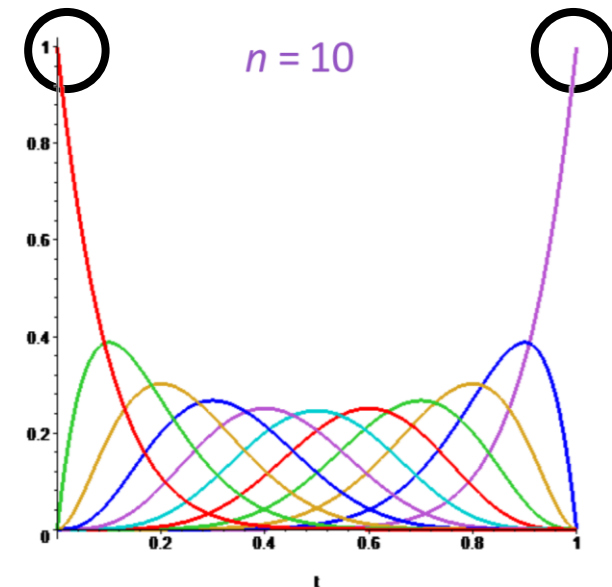
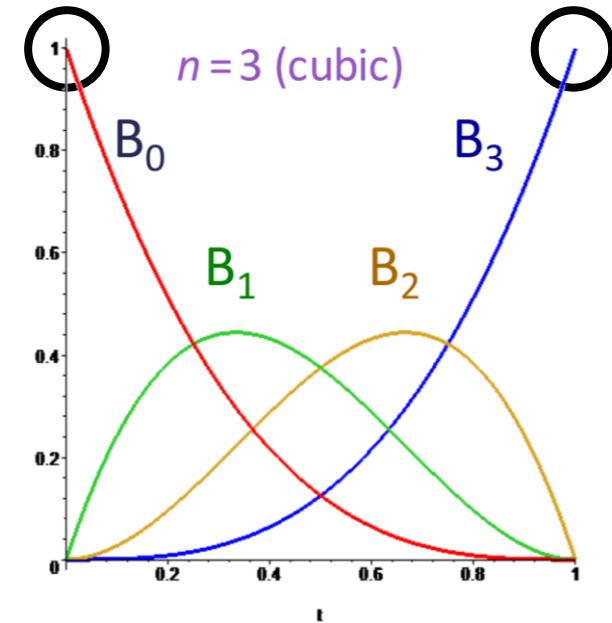
- $B = \{B_0^{(n)}, B_1^{(n)}, \dots, B_n^{(n)}\}$ ,  $B_i^{(n)}(t) = \binom{n}{i} t^i (1-t)^{n-i}$

- Non-negativity II

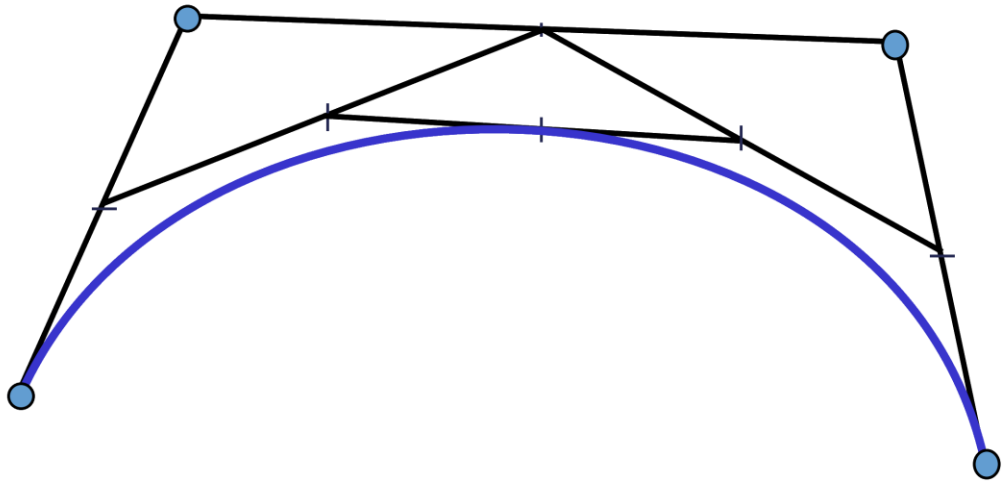
$$B_i^n(t) > 0 \text{ for } 0 < t < 1$$

$$B_0^n(0) = 1, \quad B_1^n(0) = \dots = B_n^n(0) = 0$$

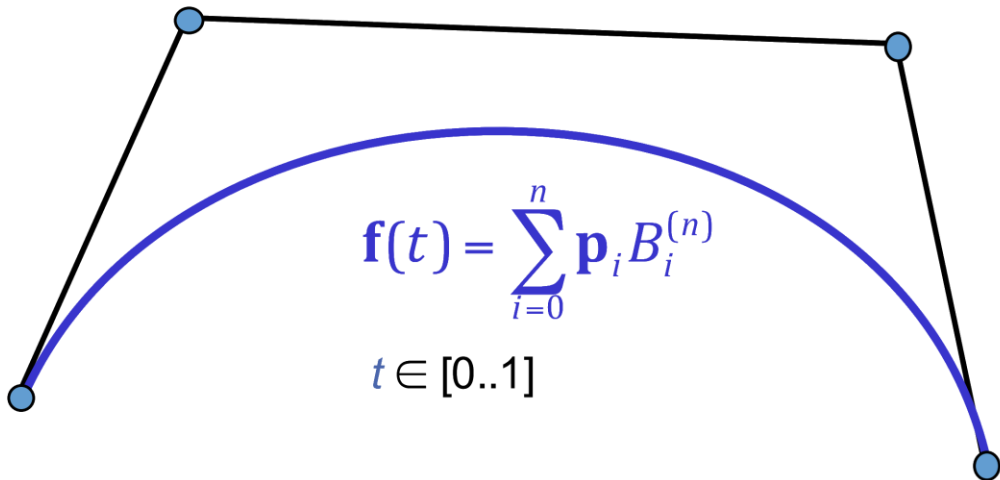
$$B_0^n(1) = \dots = B_{n-1}^n(1) = 0, \quad B_n^n(1) = 1$$



# Recap

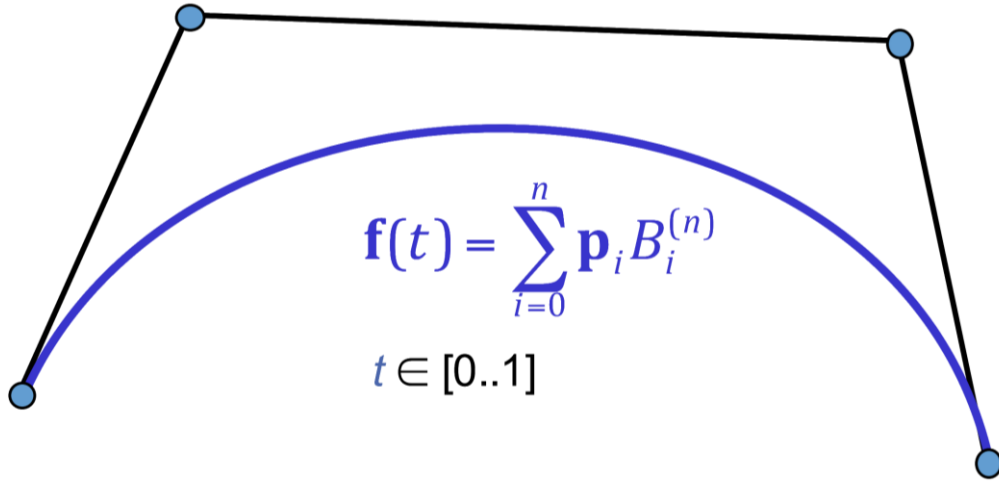


de Casteljau algorithm



**Bernstein form**

# Recap



## Bernstein form

Curve      basis function      control points

$$\mathbf{f}(t) = \sum_{i=1}^n B_i(t) \mathbf{p}_i$$

Dashed arrows point from the labels 'Curve', 'basis function', and 'control points' to the corresponding parts of the equation above.

### Useful properties for basis functions

- Smoothness
- Local control / support
- Affine invariance
- Convex hull property

# Degree elevation

- Given:  $\mathbf{b}_0, \dots, \mathbf{b}_n \rightarrow \mathbf{x}(t)$
- Wanted:  $\bar{\mathbf{b}}_0, \dots, \bar{\mathbf{b}}_n, \bar{\mathbf{b}}_{n+1} \rightarrow \bar{\mathbf{x}}(t)$  with  $\mathbf{x} = \bar{\mathbf{x}}$
- Solution:

# Degree elevation

- Given:  $\mathbf{b}_0, \dots, \mathbf{b}_n \rightarrow \mathbf{x}(t)$
- Wanted:  $\bar{\mathbf{b}}_0, \dots, \bar{\mathbf{b}}_n, \bar{\mathbf{b}}_{n+1} \rightarrow \bar{\mathbf{x}}(t)$  with  $\mathbf{x} = \bar{\mathbf{x}}$
- Solution:

$$\begin{aligned}\bar{\mathbf{b}}_0 &= \mathbf{b}_0 \\ \bar{\mathbf{b}}_{n+1} &= \mathbf{b}_n \\ \bar{\mathbf{b}}_j &= \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j \quad \text{for } j = 1, \dots, n\end{aligned}$$

# Proof

- Let's consider

$$\begin{aligned}(1-t)B_i^n(t) &= (1-t) \binom{n}{i} (1-t)^{n-i} t^i = \binom{n}{i} (1-t)^{n+1-i} t^i \\ &= \frac{n+1-i}{n+1} \binom{n+1}{i} (1-t)^{n+1-i} t^i \\ &= \frac{n+1-i}{n+1} B_i^{n+1}(t)\end{aligned}$$

Similarly

$$tB_i^n(t) = \frac{i+1}{n+1} B_{i+1}^{n+1}(t)$$



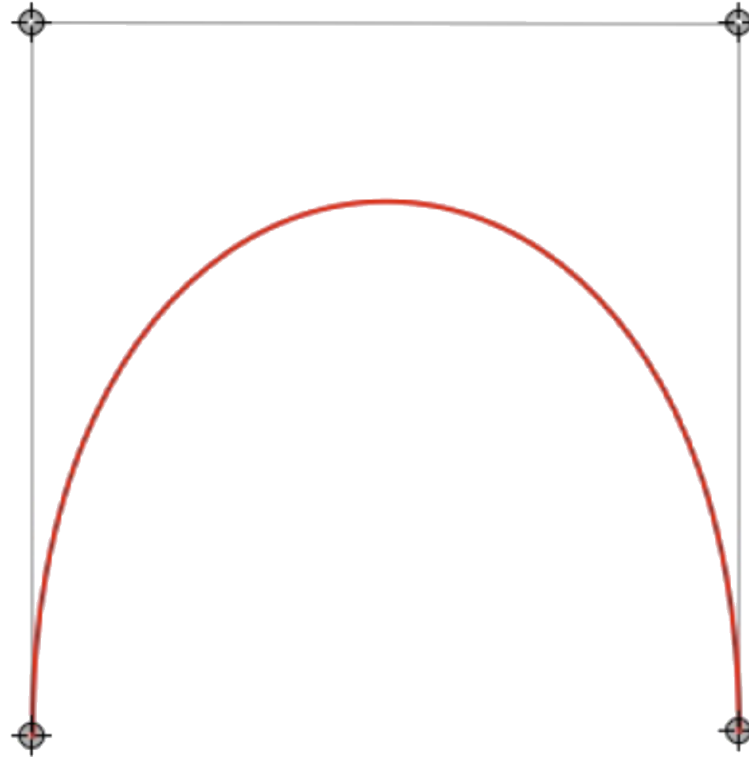
# Proof

$$\begin{aligned}
 f(t) &= [(1-t) + t]f(t) = [(1-t) + t] \sum_{i=0}^n B_i^n(t) \mathbf{P}_i = \sum_{i=0}^n [(1-t)B_i^n(t) + tB_i^n(t)] \mathbf{P}_i \\
 &= \sum_{i=0}^n \left[ \frac{n+1-i}{n+1} B_i^{n+1}(t) + \frac{i+1}{n+1} B_{i+1}^{n+1}(t) \right] \mathbf{P}_i = \sum_{i=0}^n \frac{n+1-i}{n+1} B_i^{n+1}(t) \mathbf{P}_i + \sum_{i=0}^n \frac{i+1}{n+1} B_{i+1}^{n+1}(t) \mathbf{P}_i \\
 &= \sum_{i=0}^n \frac{n+1-i}{n+1} B_i^{n+1}(t) \mathbf{P}_i + \sum_{i=1}^{n+1} \frac{i}{n+1} B_i^{n+1}(t) \mathbf{P}_{i-1} \\
 &= \sum_{i=0}^{n+1} \frac{n+1-i}{n+1} B_i^{n+1}(t) \mathbf{P}_i + \sum_{i=0}^{n+1} \frac{i}{n+1} B_i^{n+1}(t) \mathbf{P}_{i-1} \\
 &= \sum_{i=0}^{n+1} B_i^{n+1}(t) \left[ \frac{n+1-i}{n+1} \mathbf{P}_i + \frac{i}{n+1} \mathbf{P}_{i-1} \right]
 \end{aligned}$$

Using results from last slide

Adding null terms,  $i = n + 1, i = 0$

# Degree elevation: Example



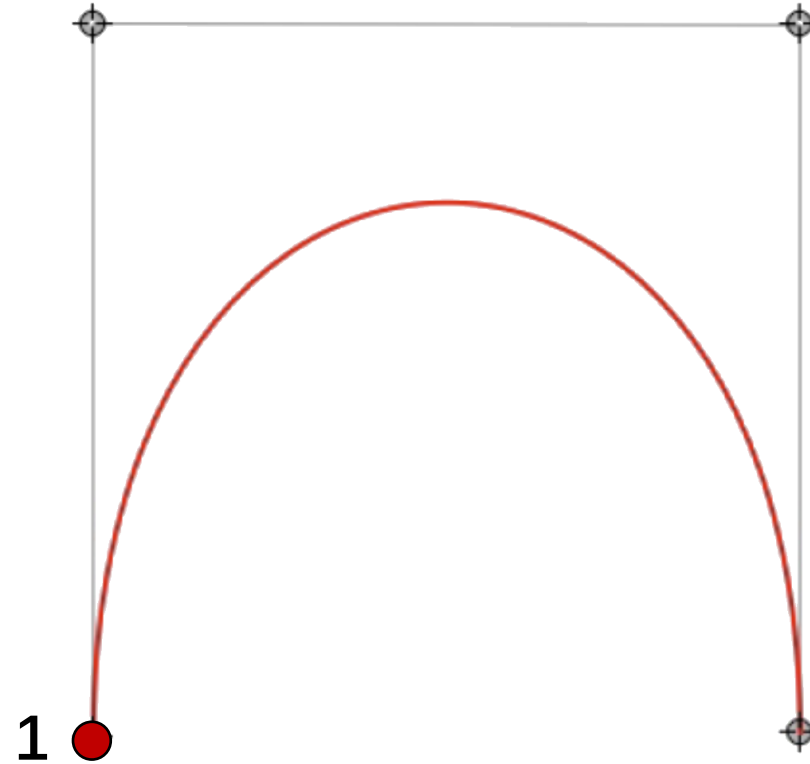
- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

# Degree elevation: Example



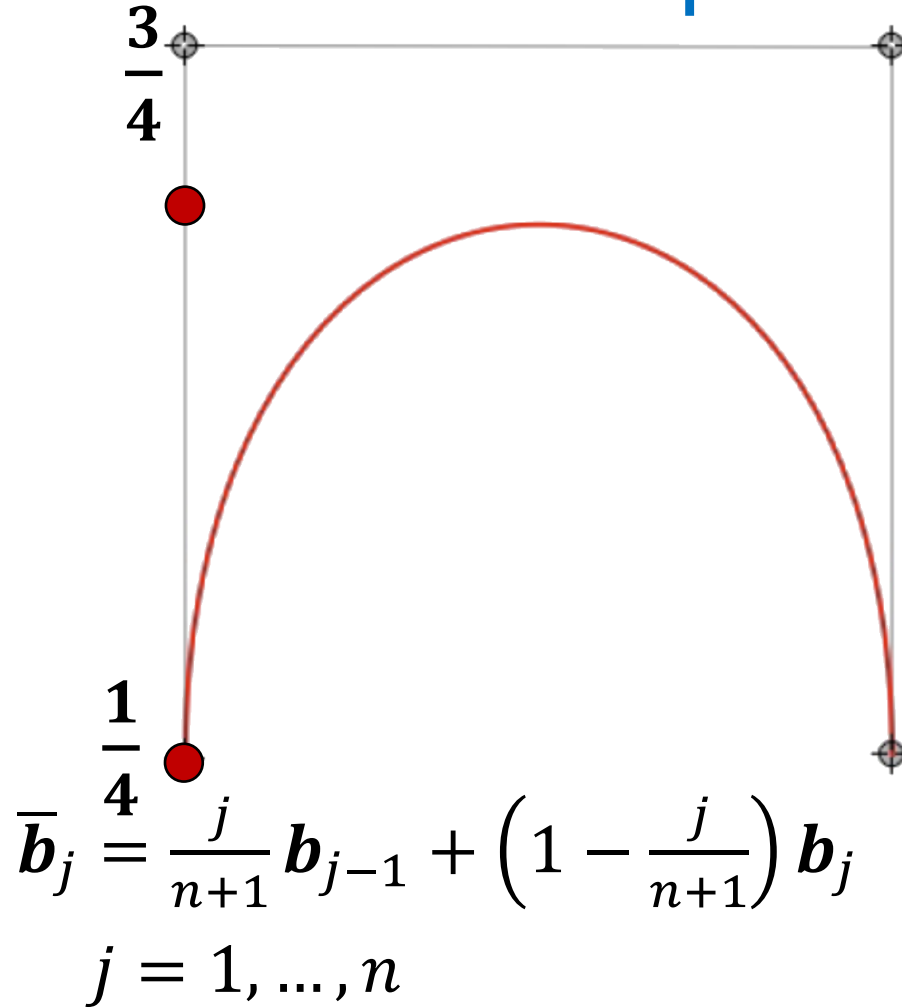
- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

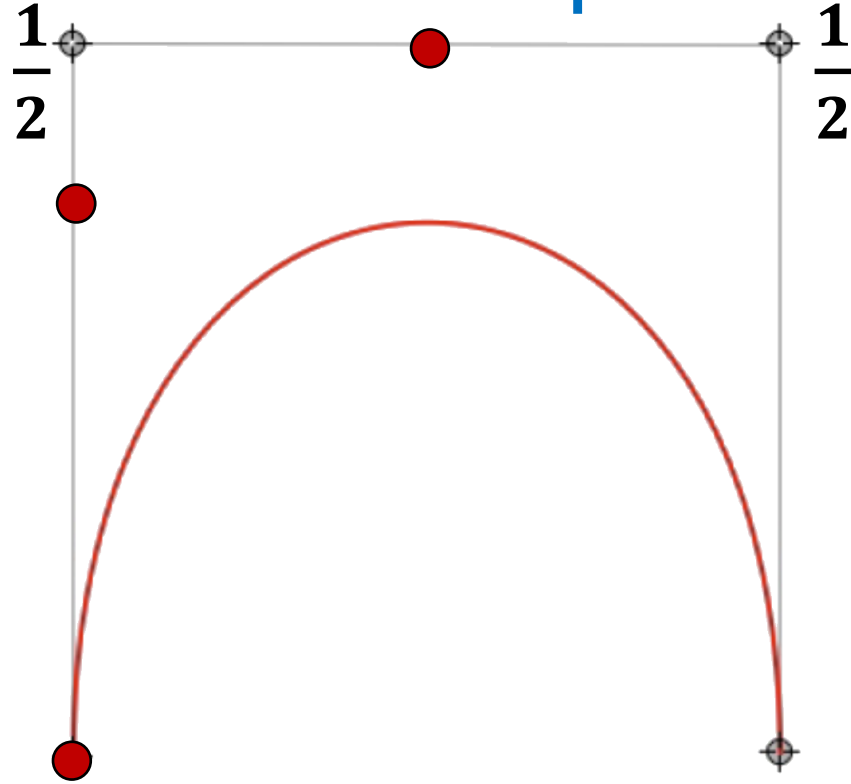
# Degree elevation: Example



- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

# Degree elevation: Example



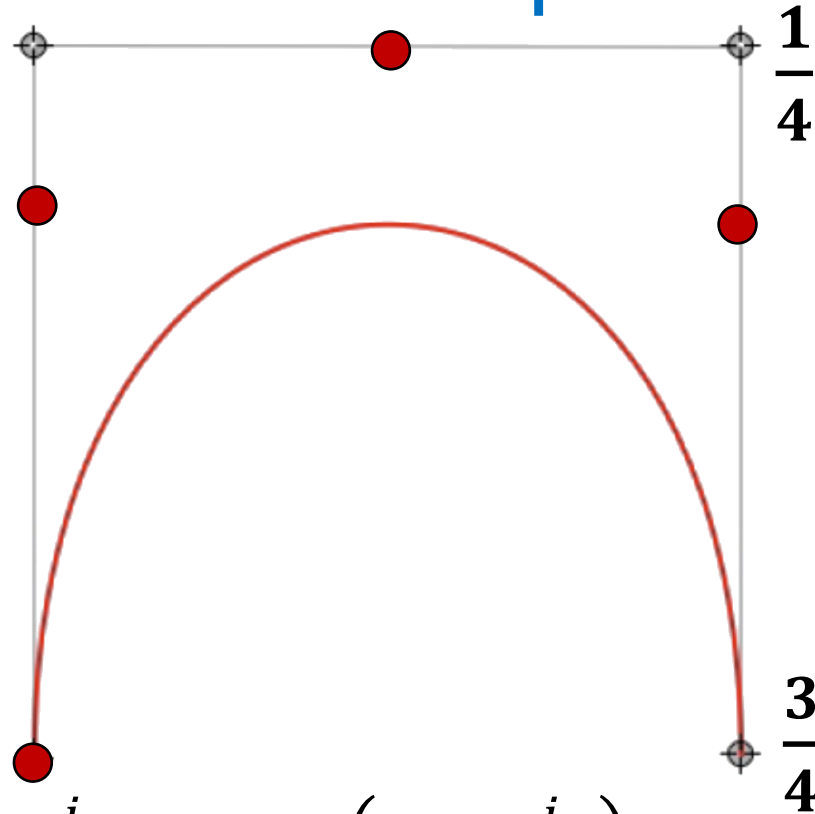
- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

# Degree elevation: Example



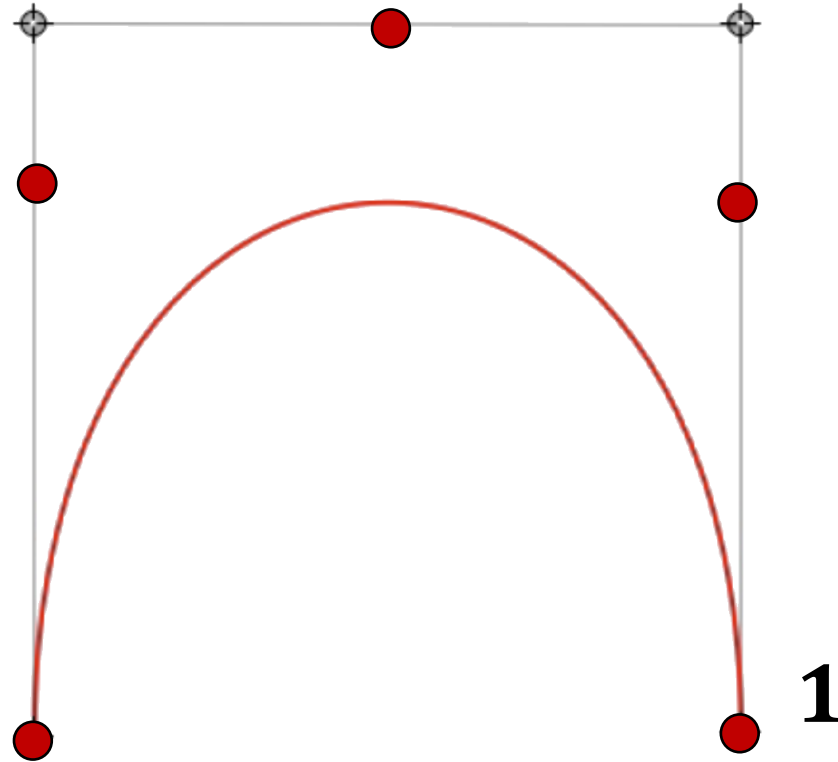
- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

# Degree elevation: Example



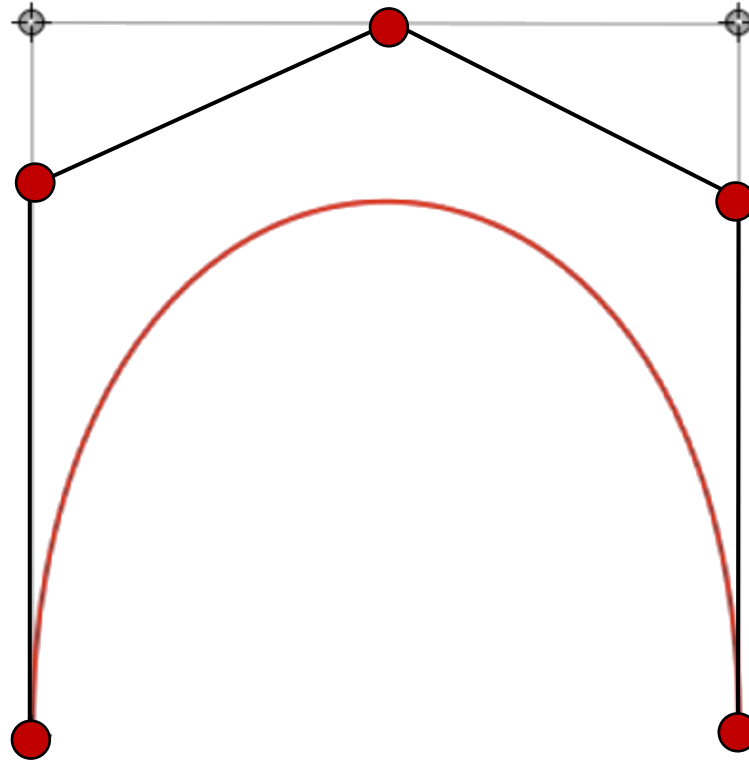
- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

# Degree elevation: Example



- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

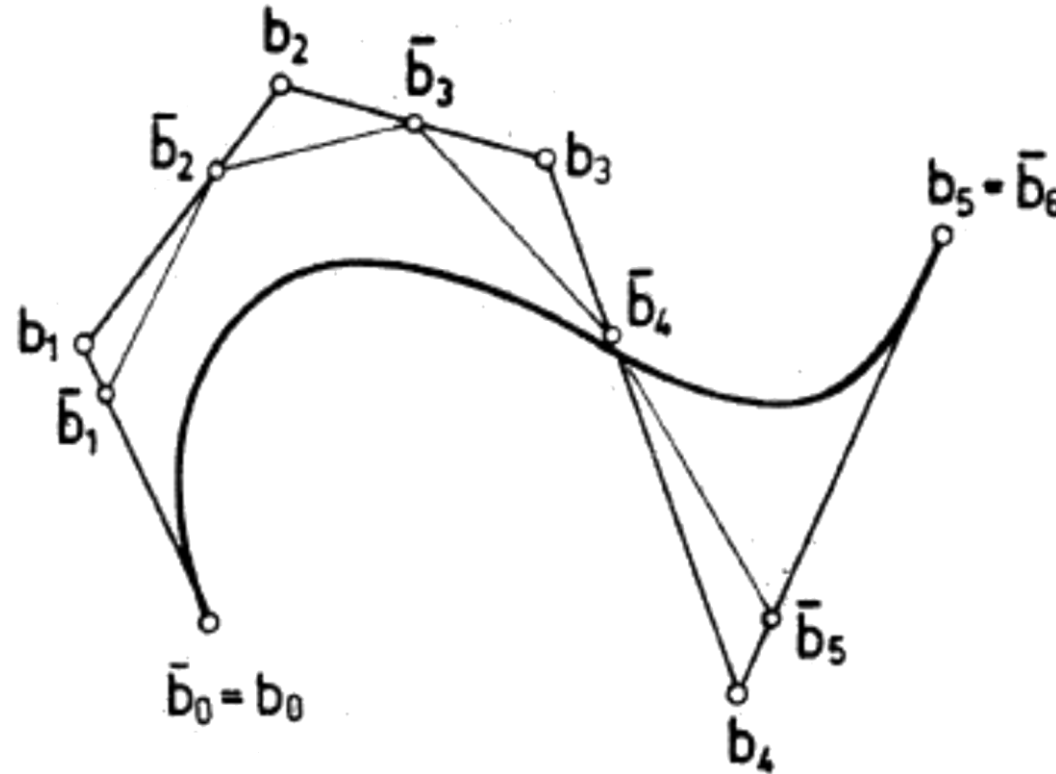
$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

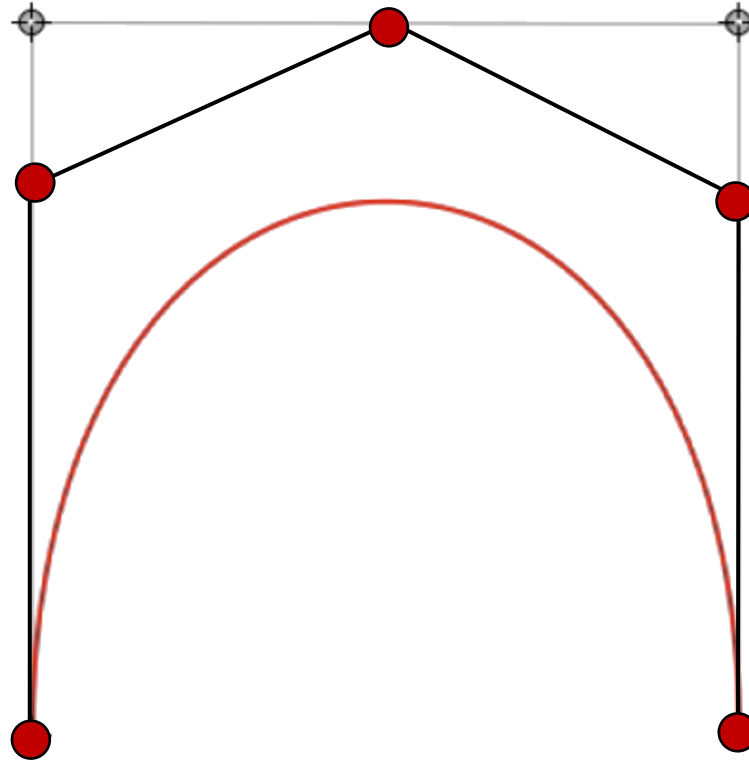


# Degree elevation



For repeated degree elevation, the Bézier polygon converges to the Bézier curve. (slow convergence)

# Degree elevation



- $\bar{\mathbf{b}}_0 = \mathbf{b}_0$

$$\bar{\mathbf{b}}_j = \frac{j}{n+1} \mathbf{b}_{j-1} + \left(1 - \frac{j}{n+1}\right) \mathbf{b}_j$$

- $\bar{\mathbf{b}}_{n+1} = \mathbf{b}_n$

$$j = 1, \dots, n$$

# Bézier Curves

Subdivision

# Subdivision

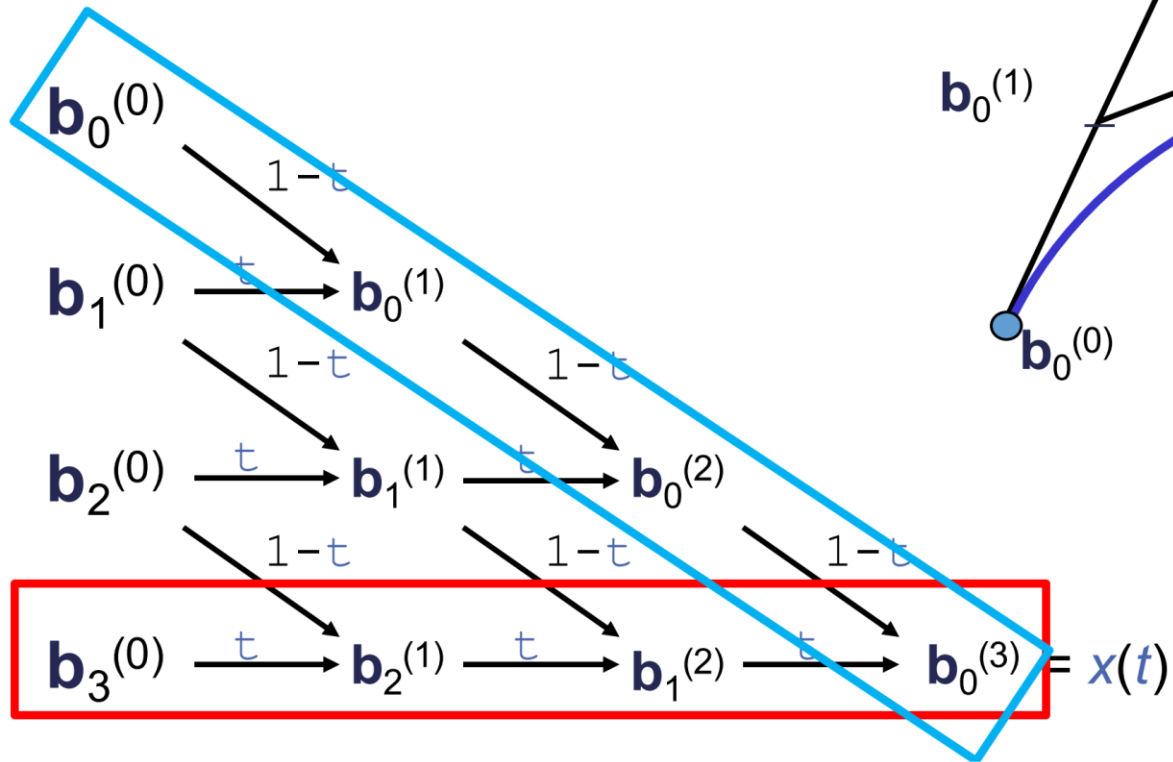
- Given:  $b_0, \dots, b_n \rightarrow x(t), t \in [0,1]$

- Wanted:  $b_0^{(1)}, \dots, b_n^{(1)} \rightarrow x^{(1)}(t),$

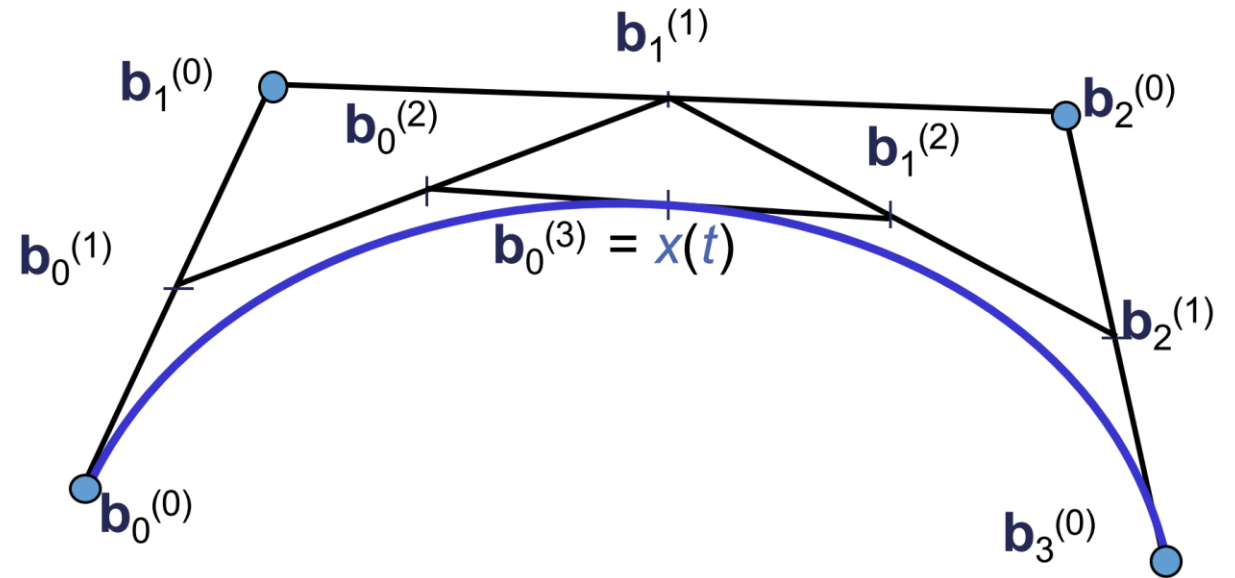
- $b_0^{(2)}, \dots, b_n^{(2)} \rightarrow x^{(2)}(t),$

with  $x = x^{(1)} \cup x^{(2)}$

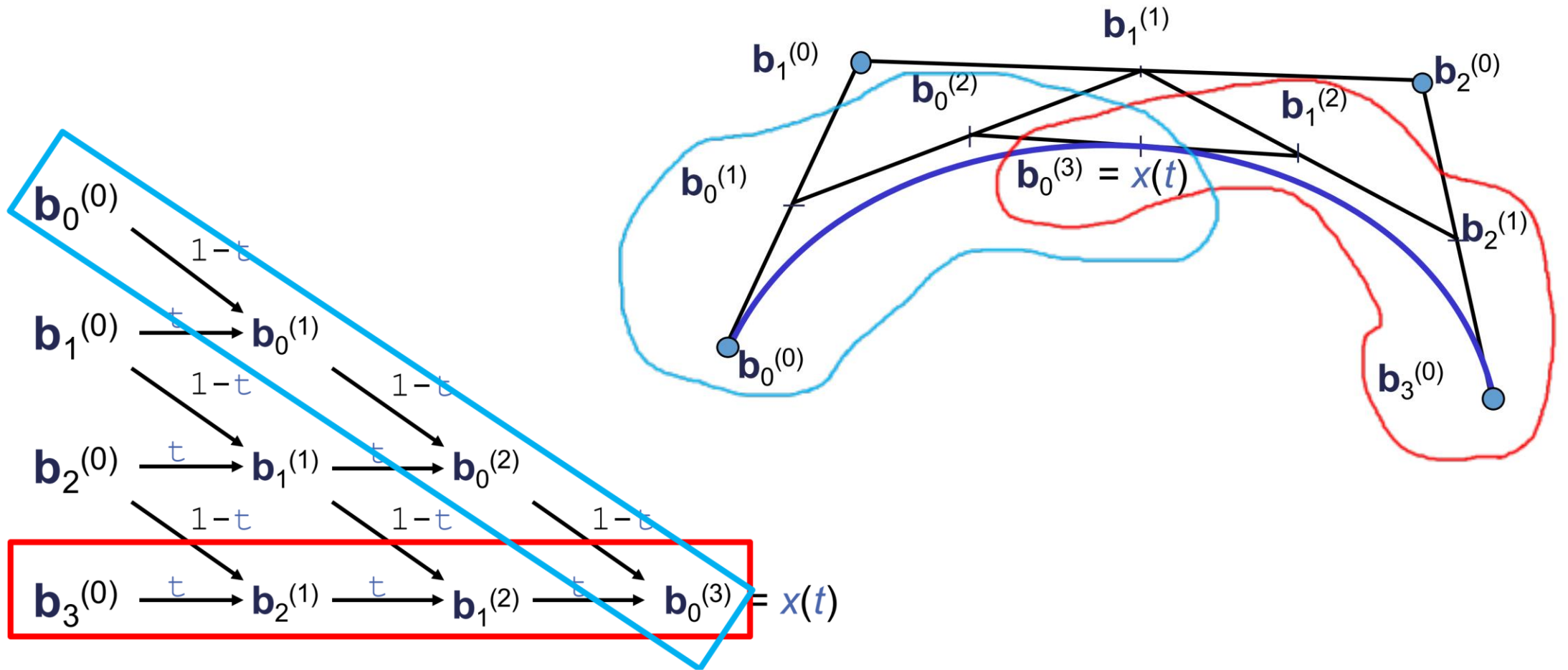
# Subdivision: Example



de Casteljau scheme



# Subdivision: Example

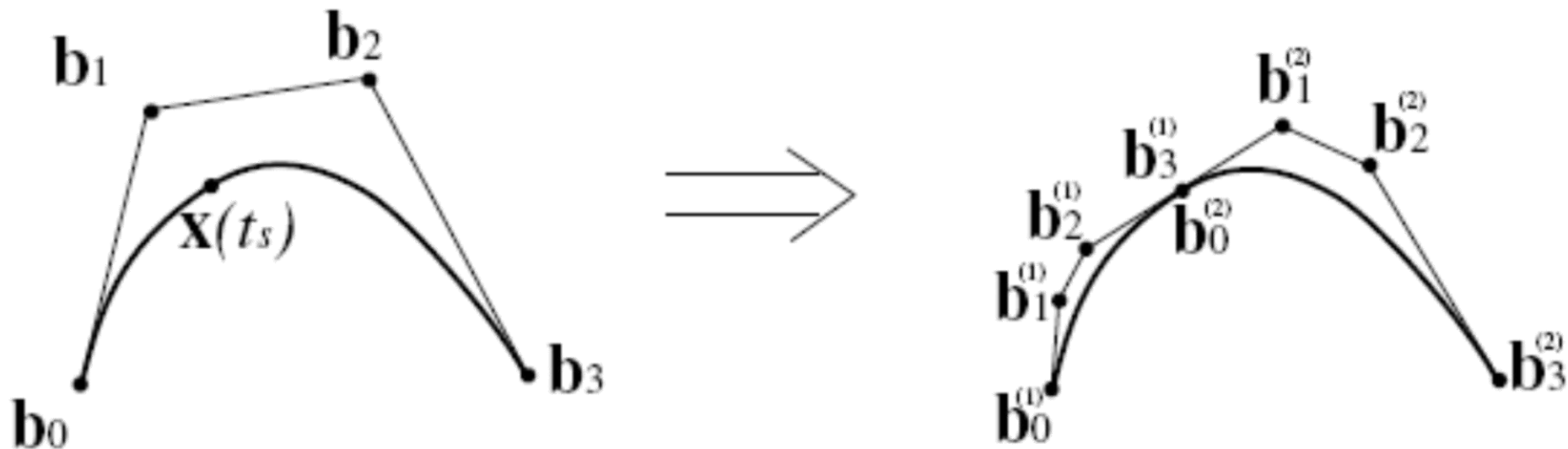


de Casteljau scheme

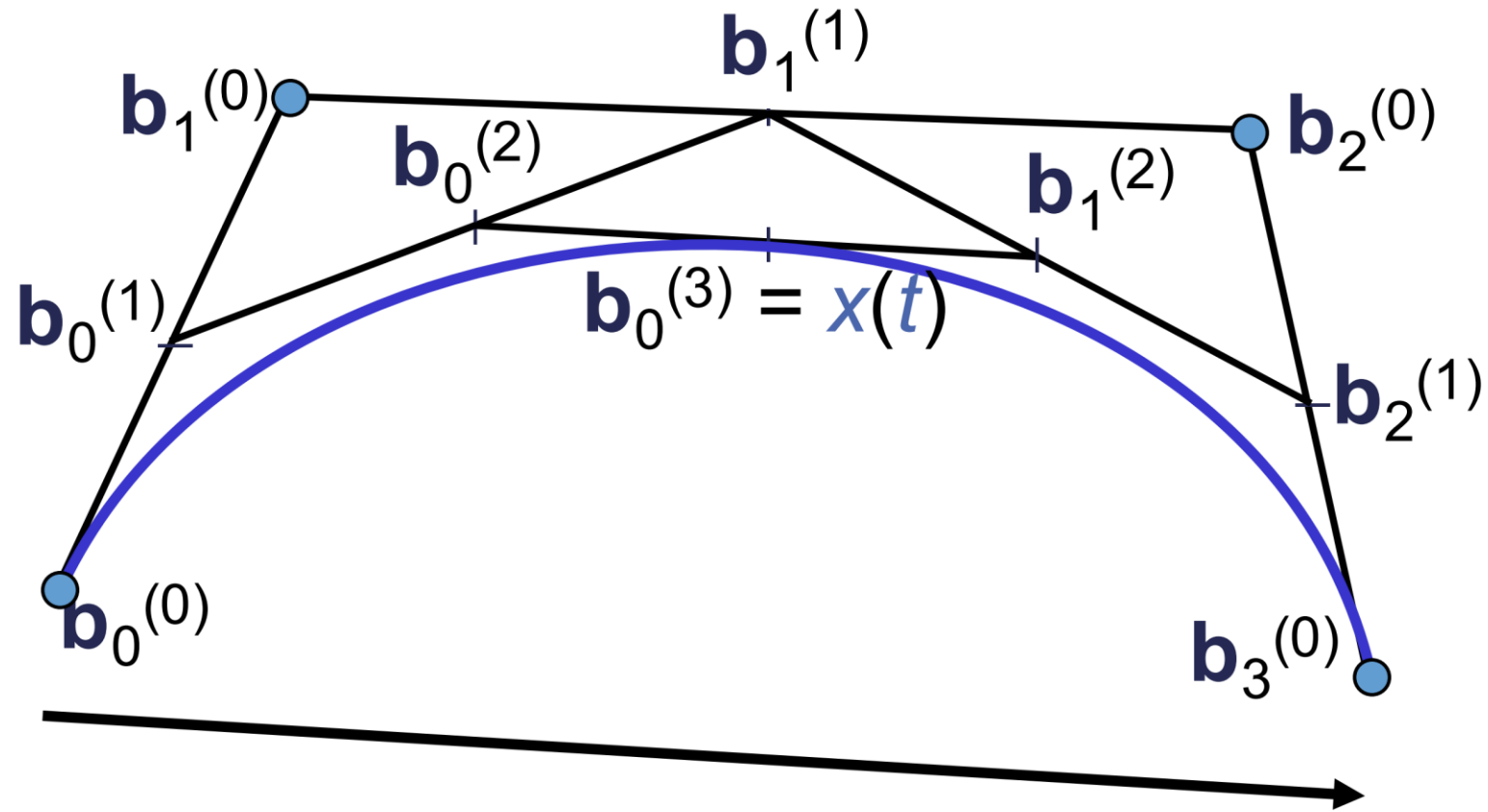
# Subdivision

Solution:  $b_i^{(1)} = b_0^i$ ,  $b_i^{(2)} = b_0^{n-i}$  for  $i = 0, \dots, n$

That means that the new points are intermediate points of the de Casteljau algorithm!



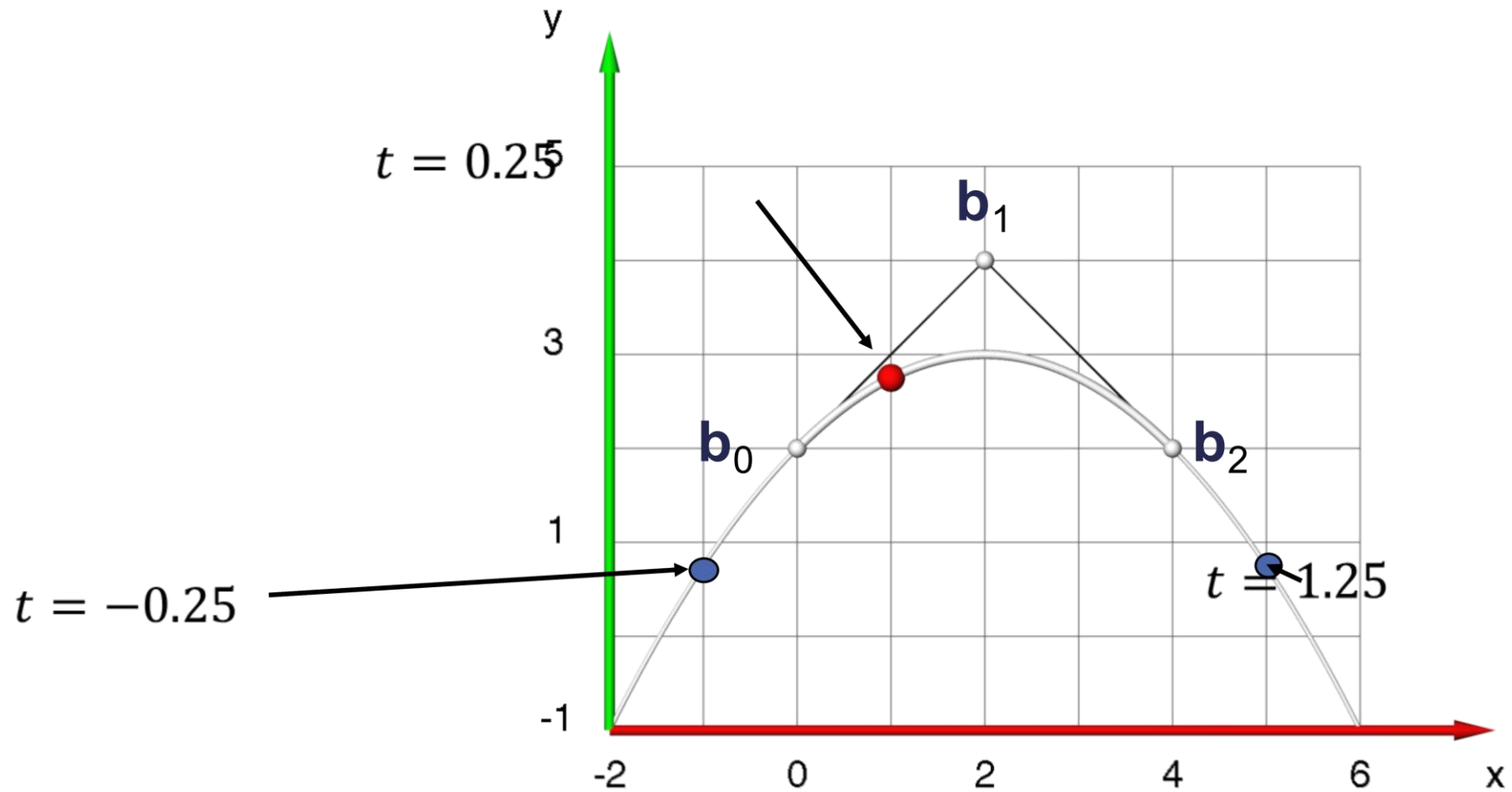
# Curve range



parameterization:  $t \in [0,1]$



# Curve range



# Summary & Outlook

- Bézier curves and curve design
  - The rough form is specified by the position of the control points
  - Results: smooth curve approximating the control points
  - Computation / Representation:
    - de Casteljau algorithm
    - Bernstein form
- Problems:
  - High polynomial degree
  - Moving a control point can change the whole curve
  - Interpolation of points
  - → **Bézier splines**

