

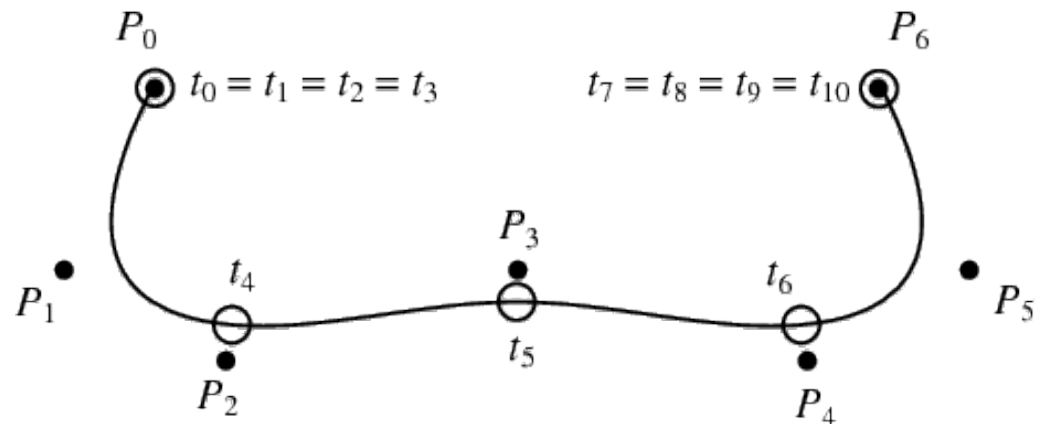
## Outline

- Interpolation and polynomial approximation
- Interpolation
  - Lagrange
  - Cubic Splines

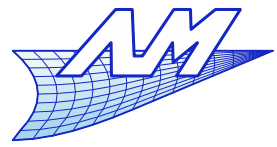
## Approximation

- Bézier curves
- B-Splines

- Some vocabulary (again ;)
  - Control point : Geometric point that serves as support to the curve
  - Knot : a specific value of the parameter  $u$  corresponding to a joint between pieces of a curve
  - Knot sequence : the set of knots values (in an increasing order).



- Interpolation
  - The curve passes through the control points
- Approximation
  - The curve doesn't necessary passes through the control points
    - But these have an influence ...
  - Statistic approaches ?
    - Least squares
    - « Kriging »
    - Not very adapted to geometric modelling



## Interpolation

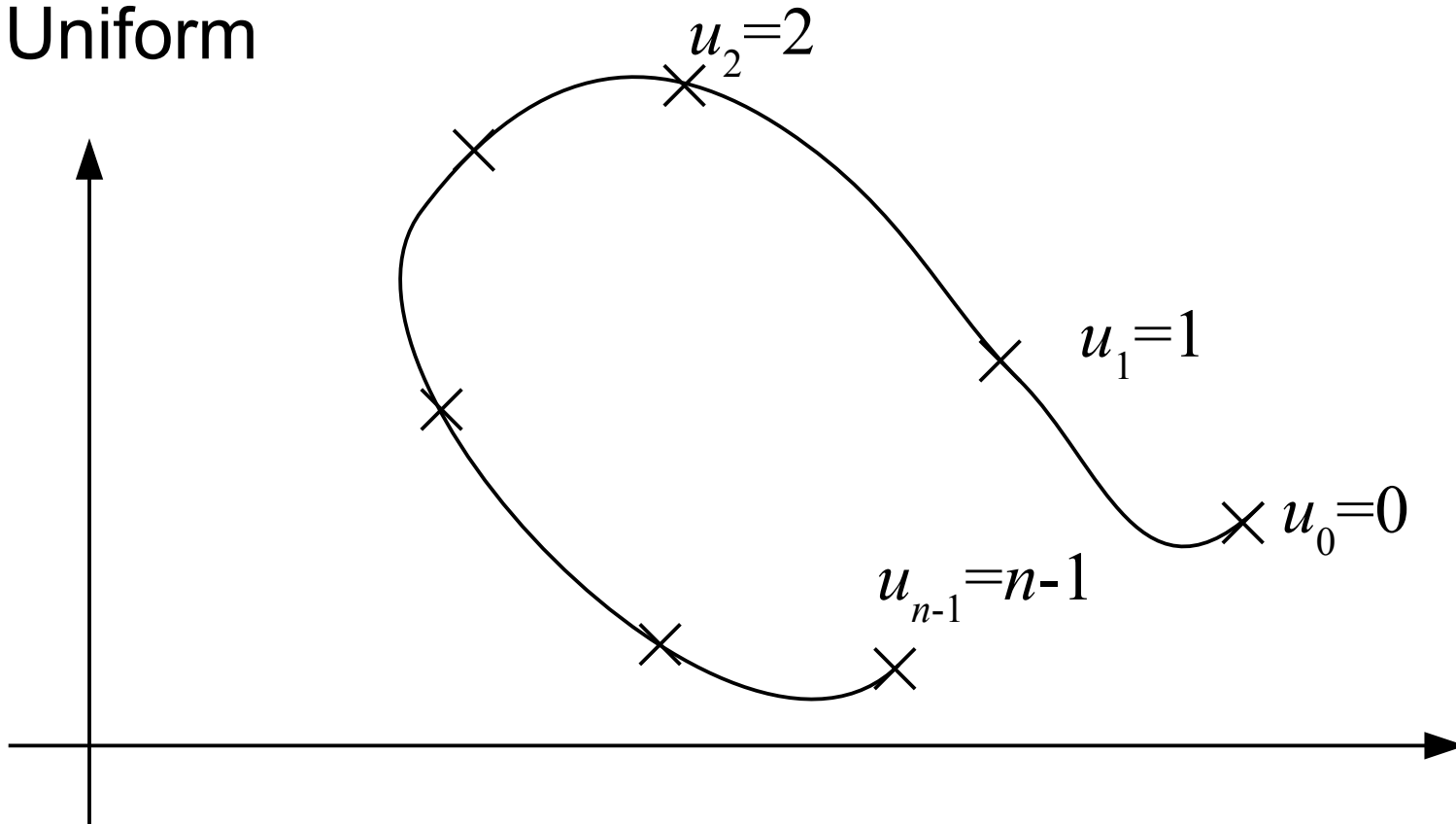
## Interpolation

### Interpolation

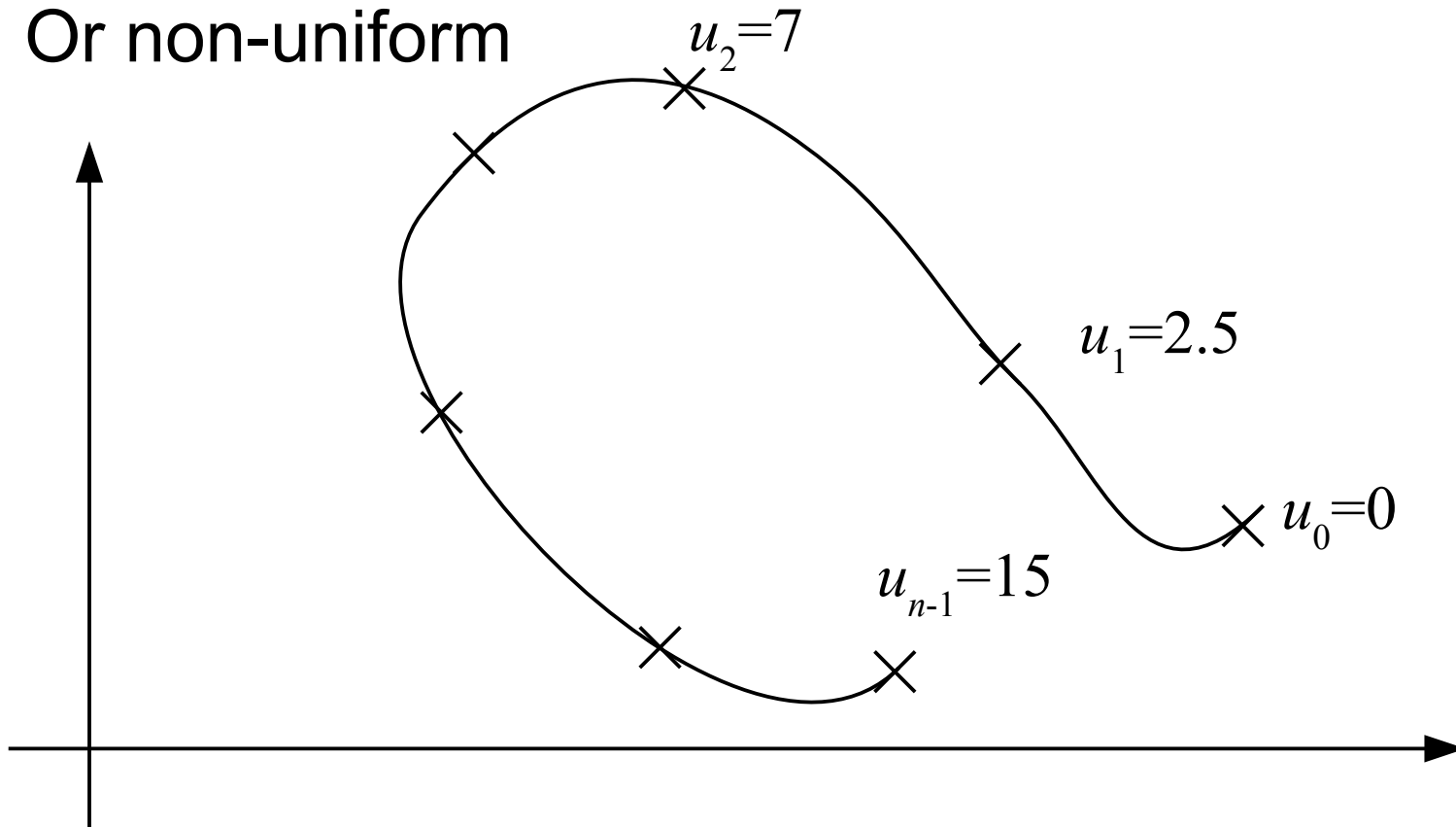
- We want draw a regular and smooth parametric curve through a certain number  $n$  of points  $P_i$ 
  - Several families of base functions are available
  - Most obvious are polynomials
  - There are others -
    - Trigonometric functions (by mean of a Fourier decomposition for instance)
    - Power functions
    - etc...

## Interpolation

- We must choose the parametrization (nodal sequence)
  - Uniform



- Or non-uniform

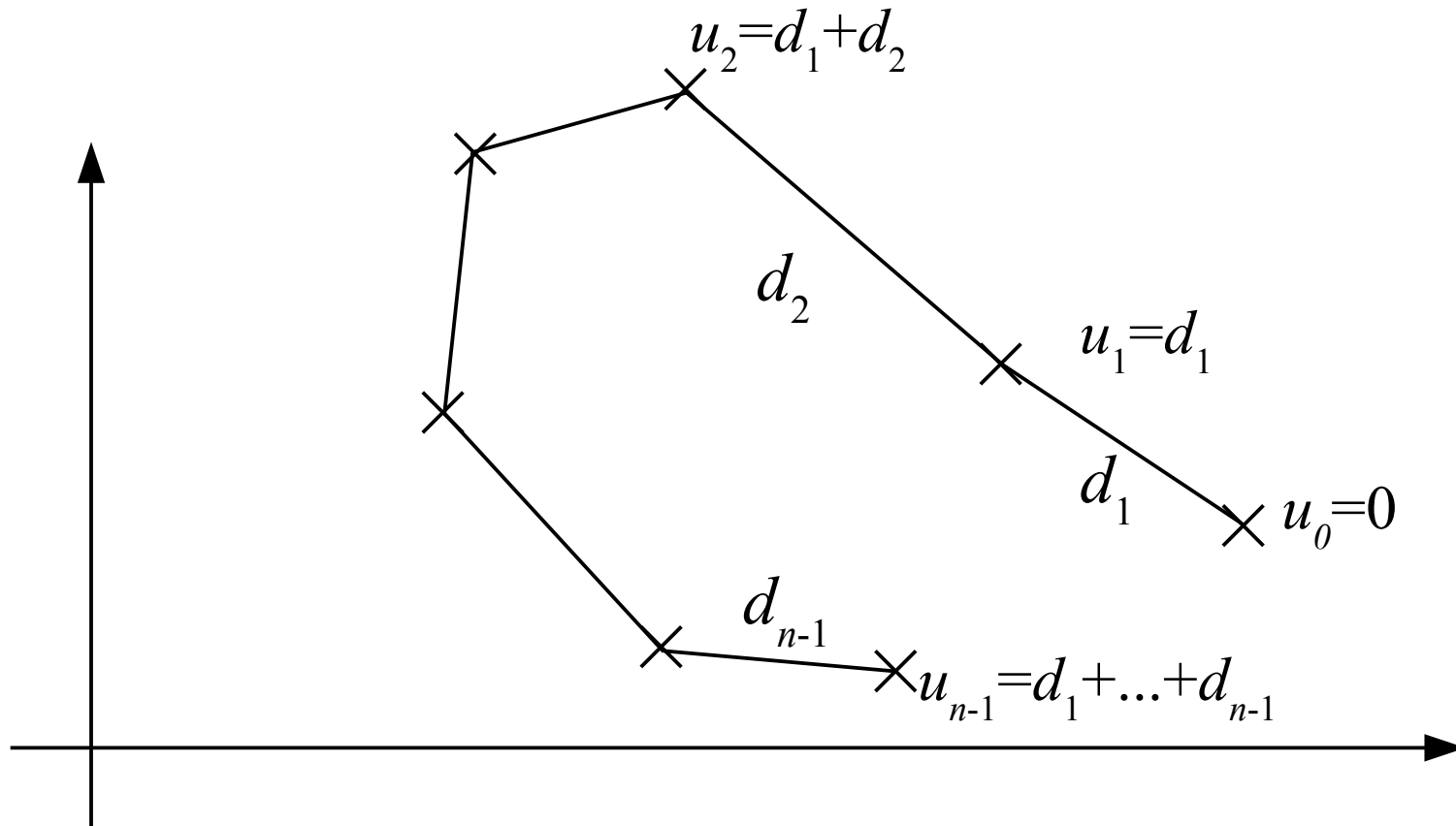


## Interpolation

- Can we choose  $u$  as a curvilinear abscissa ?
  - In principle no since we don't know the final shape of the curve beforehand (with the exception of interpolation points)
  - We will see later that it is often impossible that  $u$  corresponds with  $s$  exactly along the whole curve using analytical functions.
  - But nothing forbids to get close to that – numerically...



- Parametrization as an approximate arc length



## Interpolation

- In all cases, we want to have :

$$P(u_i) = P_i \equiv \begin{cases} x(u_i) = x_i \\ y(u_i) = y_i \end{cases}$$

- We are going to interpolate the functions  $x(u)$  and  $y(u)$  with ONE polynomial with  $n$  parameters
  - This one must be of order  $p=n-1$  :

$$x(u) = a_0 + a_1 u + a_2 u^2 + \cdots + a_{n-1} u^{n-1} = \sum_{j=0}^{n-1} a_j u^j$$

- We set the linear system and solve...

$$\begin{cases} x(u_1) = x_1 \\ \vdots \\ x(u_{n-1}) = x_{n-1} \end{cases}$$

## Interpolation

- Vandermonde matrix

$$\begin{cases} x(u_1) = x_1 \\ \vdots \\ x(u_{n-1}) = x_{n-1} \end{cases} \longrightarrow \begin{pmatrix} 1 & u_0 & u_0^2 & \cdots & u_0^{n-1} \\ 1 & u_1 & u_1^2 & \cdots & u_1^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & u_{n-1} & u_{n-1}^2 & \cdots & u_{n-1}^{n-1} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{pmatrix} = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{pmatrix}$$

Can be solved by classical numerical methods, but ...

- The condition number of this matrix is **VERY** bad
- It must be solved for each RHS member (  $(x_i)$  or  $(y_i)$  ), or have to take the inverse of this matrix, or perform an LU decomposition.

## Interpolation

- Instead of setting the polynomial in  $x$  and in  $y$  and solving, we can put it under the following form :

$$\left\{ \begin{array}{l} x(u) = \sum_0^{n-1} x_i l_i^p(u) \\ y(u) = \sum_0^{n-1} y_i l_i^p(u) \end{array} \right. \Leftrightarrow P(u) = \sum_0^{n-1} P_i l_i^p(u)$$

, where the  $l_i^p(u)$  are a polynomial basis of order  $p=n-1$ .

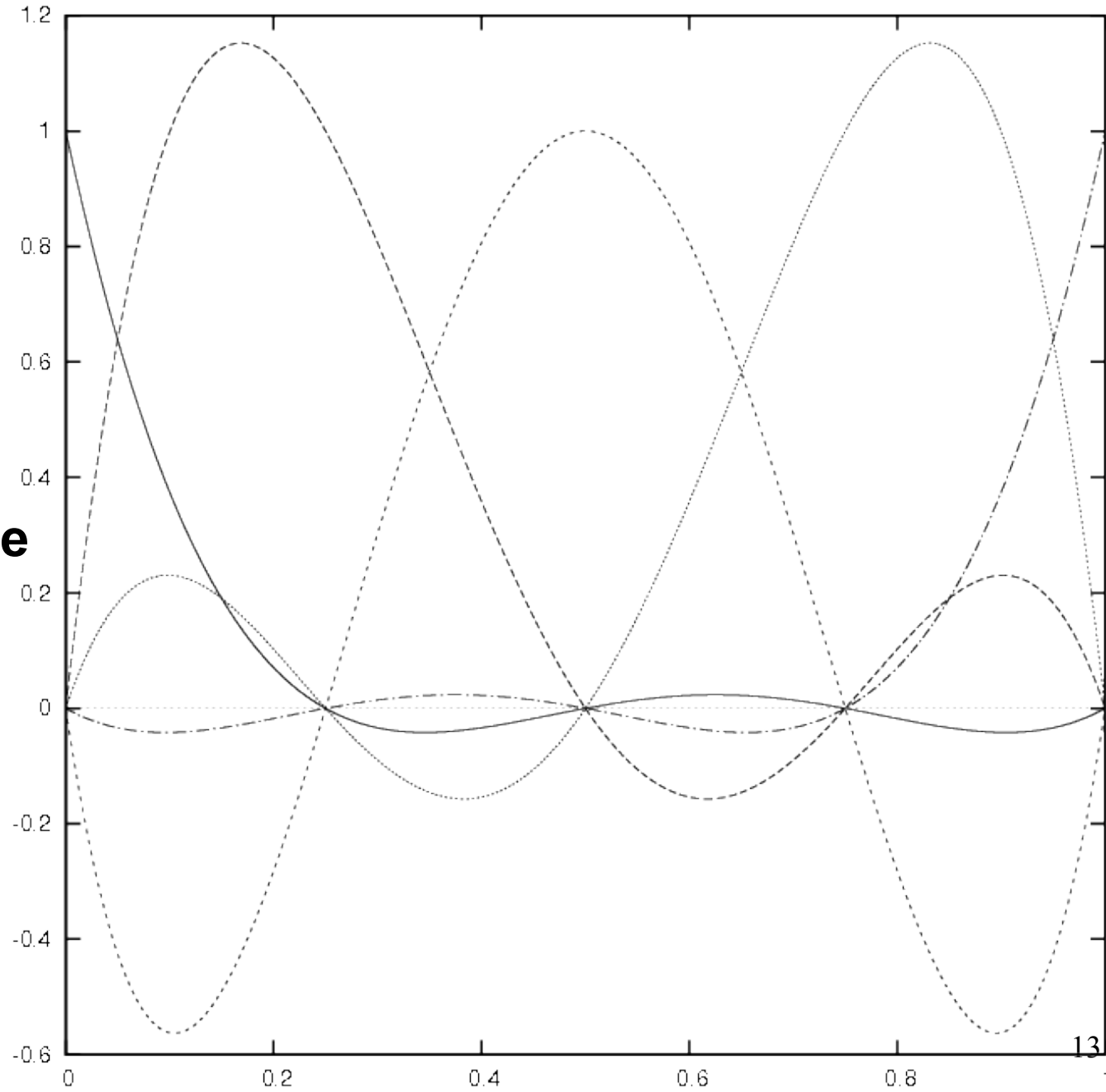
- These polynomials verify, for an interpolation :

$$l_i^p(u_j) = \delta_{ij}$$

- We have only one computation to do for any position of the interpolation points, knowing the  $u_i$  (the parametrization)

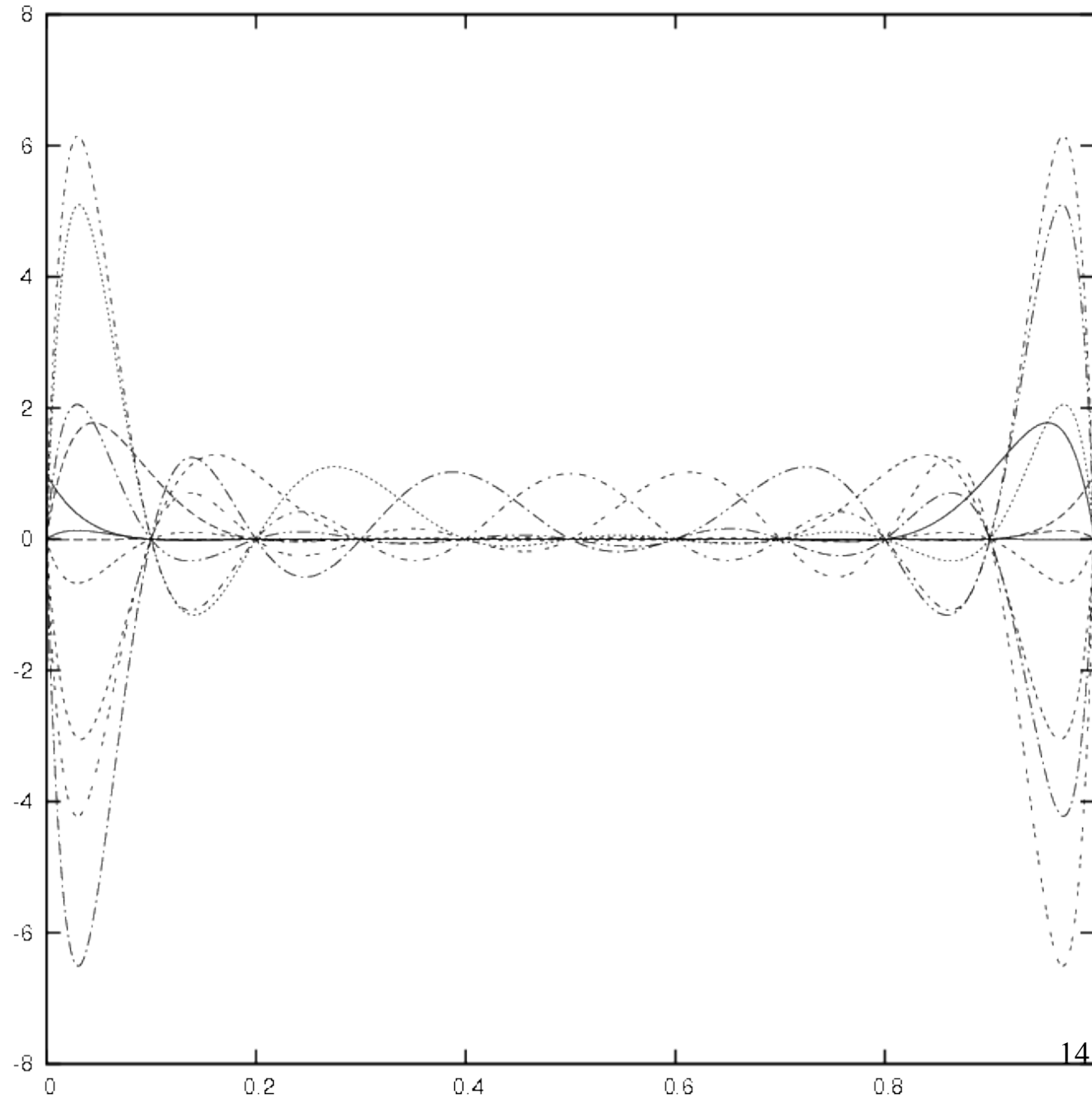
## Lagrange polynomials

- Order 4
- The  $u_i$  are evenly distributed between  $u=0$  and  $u=1$
- The sum is equal to 1 (partition of unity)
- **Presence of negative values**



## Lagrange polynomials

- Order 10
- Presence of huge overshoots near the boundaries



## Lagrange polynomials

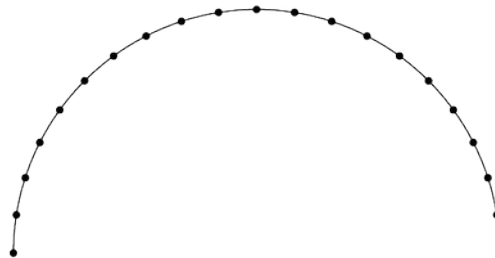
- The interpolation is represented by the following form :

$$P(u) = \sum_{i=0}^{n-1} P_i l_i^p(u) \quad \text{with} \quad l_i^p(u) = \prod_{j=0, i \neq j}^{n-1} \frac{(u - u_j)}{(u_i - u_j)}$$

- Two things worth noting:
  - The curve depends **linearly** on the position of the points
  - It is formed by a weighted sum of **basis functions** that express the influence of each point on the curve

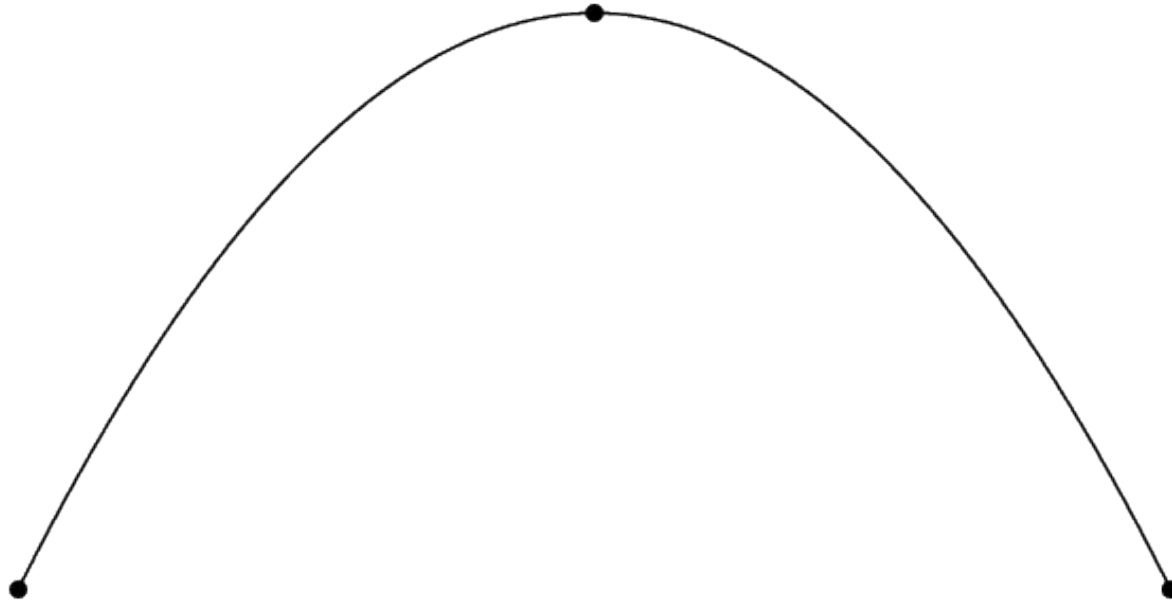
## Lagrange polynomials

- An experiment
  - We approximate a circle by an increasing number of points
  - Simultaneously , approximation order increases
  - In every case, the curve is  $C_\infty$



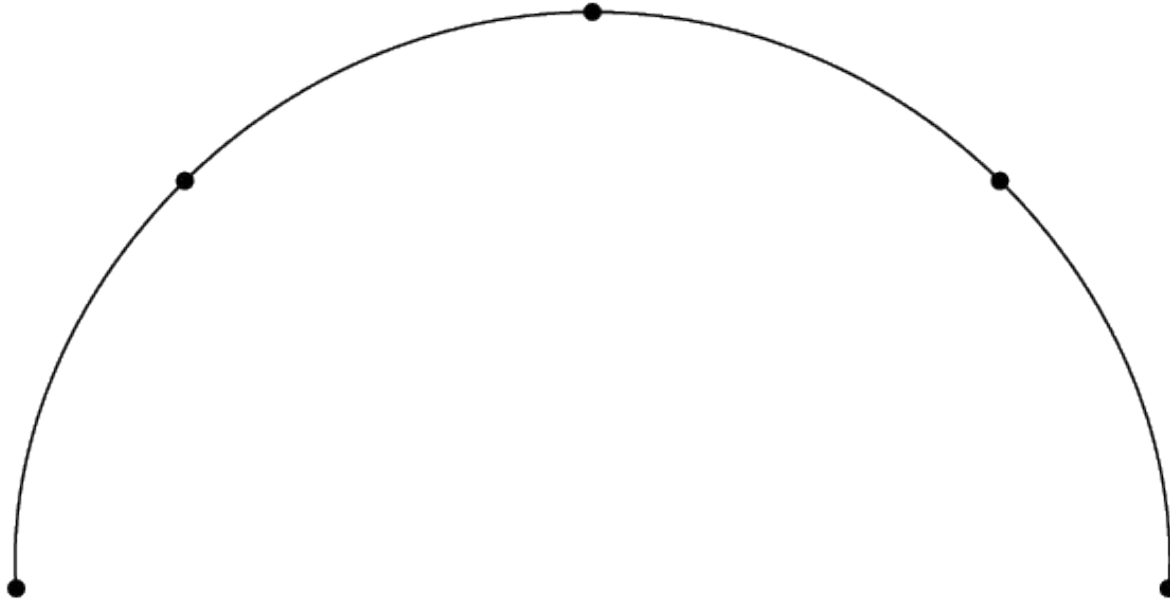


## Lagrange polynomials



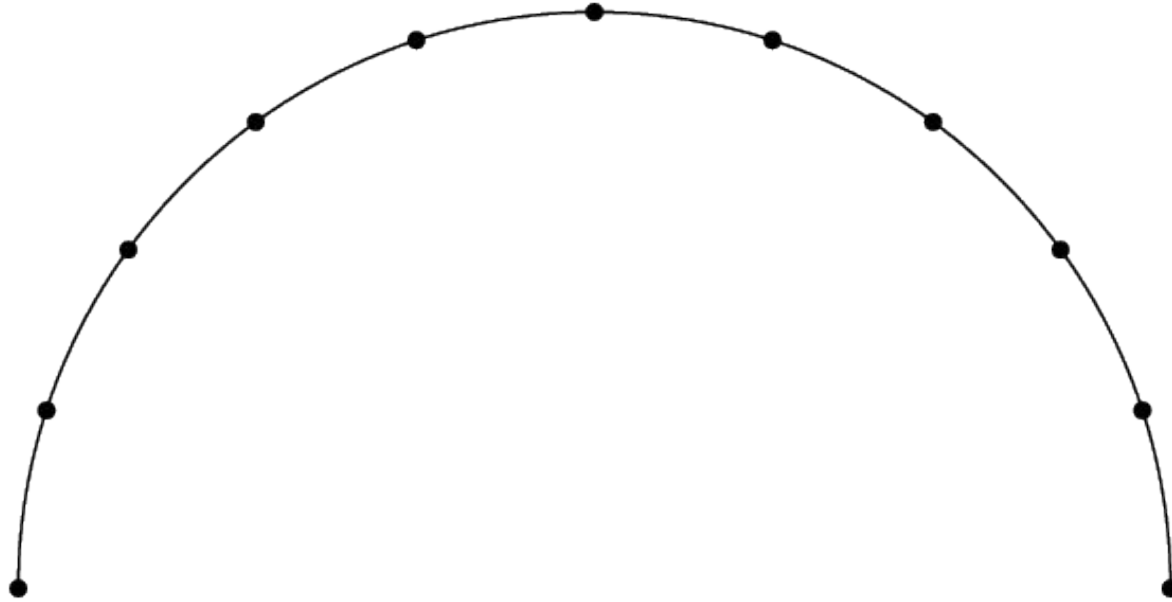
3 points, order 2 (a parabola !)

## Lagrange polynomials



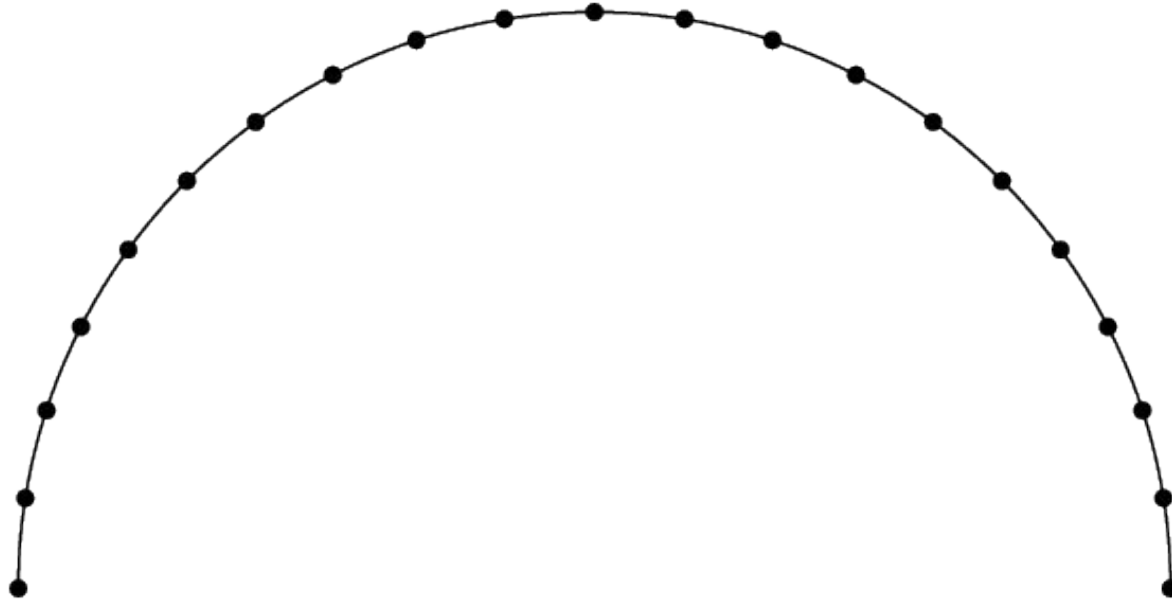
5 points, order 4

## Lagrange polynomials



11 points, order 10

## Lagrange polynomials



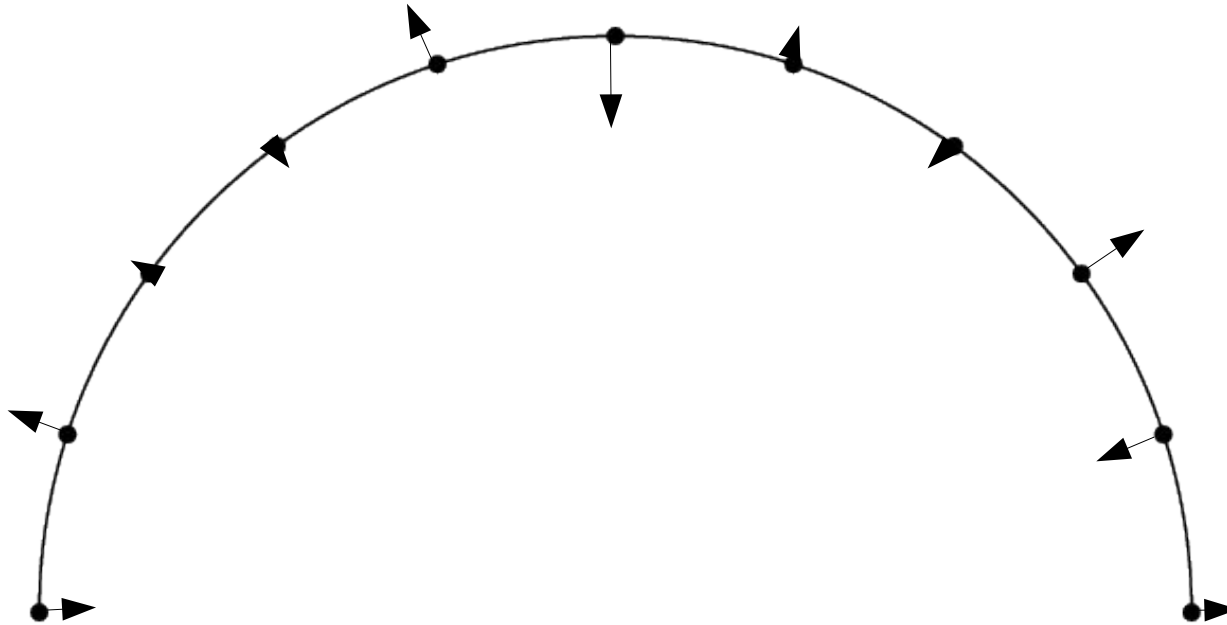
21 points, order 20

## Lagrange polynomials

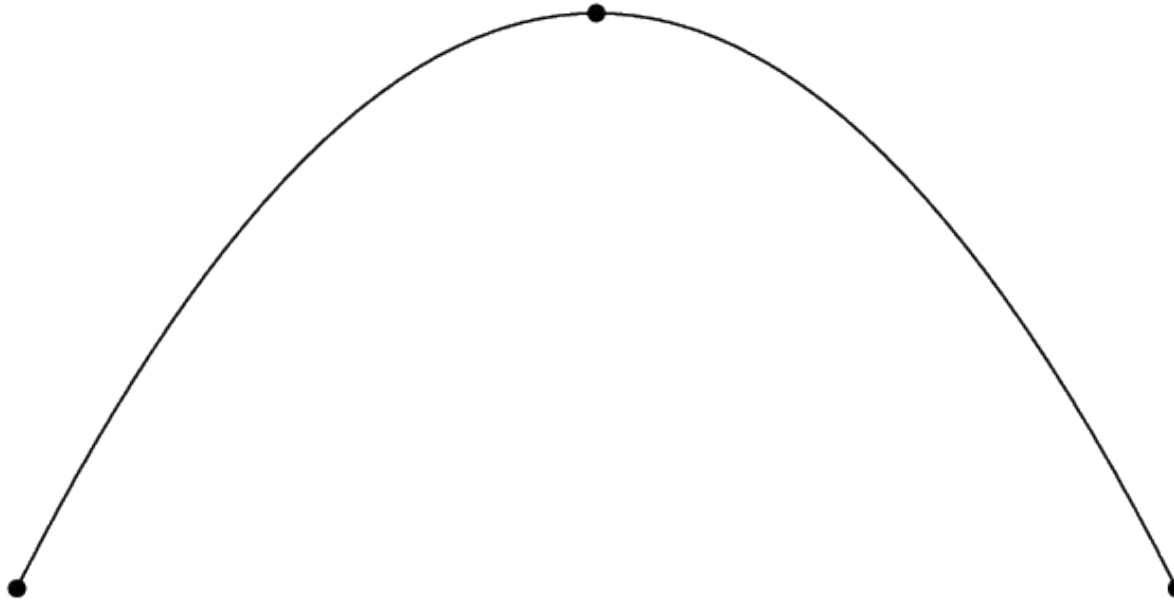
- Does it work ?
- The points were set exactly on the circle
  - What occurs if their position is inaccurate ?
  - Or if the approximated shape is not so simple ?
  - We are going to see two cases
    - The coordinates of the points are perturbed randomly
    - A deterministic increase and decrease of the radius

## Lagrange perturbation

- Random perturbation
  - Each point is moved radially by a value between -0.5 and +0.5 % of the circle's radius

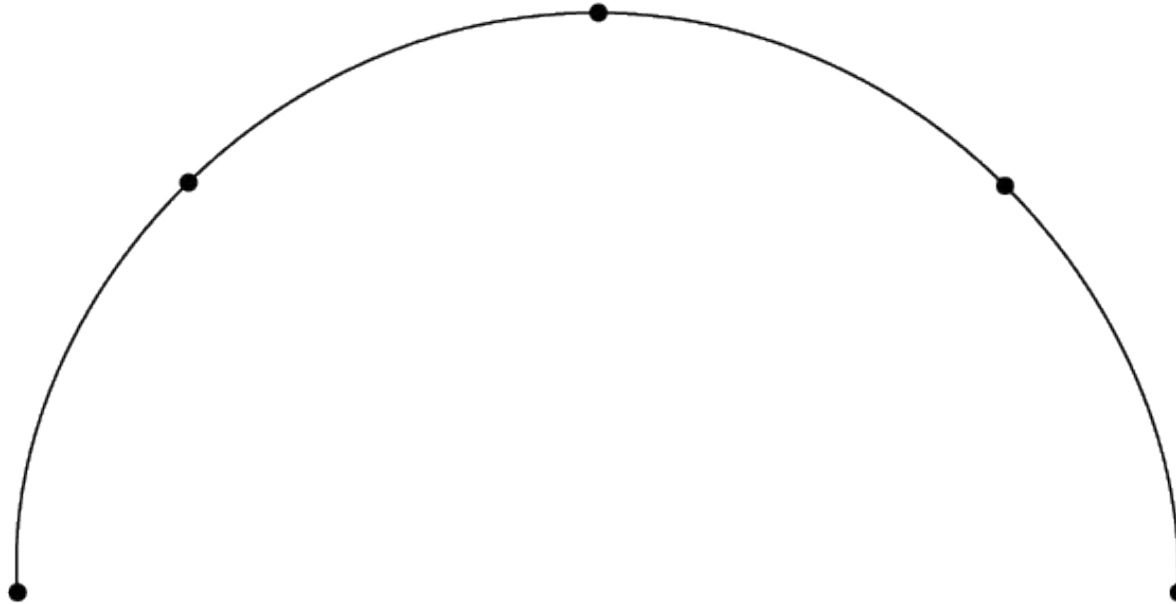


## Lagrange polynomials



3 points, random perturbation 1%

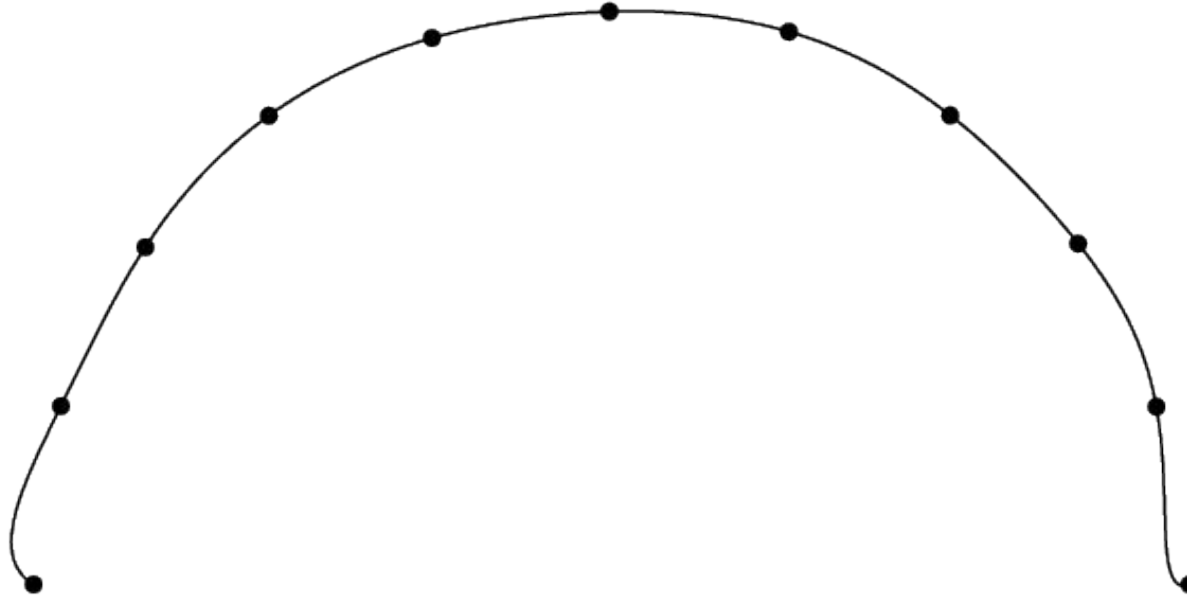
## Lagrange polynomials



5 points, random perturbation 1%



## Lagrange polynomials



11 points, random perturbation 1%

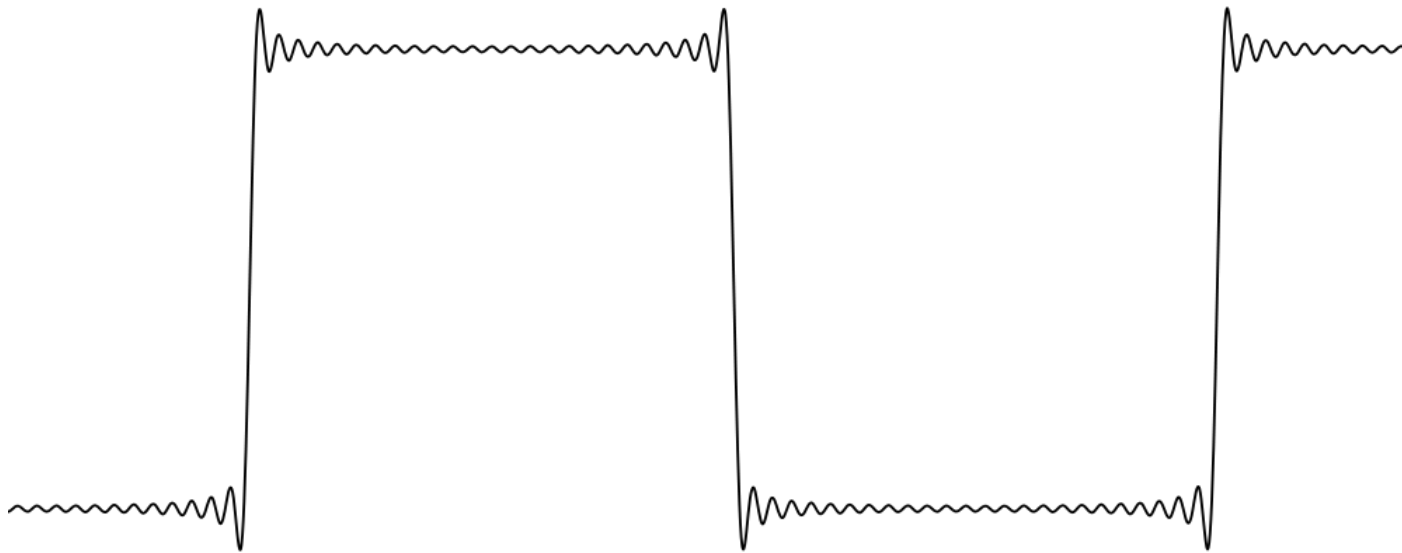
## Lagrange polynomials



21 points, random perturbation 1%

## Lagrange polynomials

- Runge phenomenon
  - Similar to Gibbs phenomenon of the decompositions in harmonic functions



## Lagrange polynomials

- How to minimize Runge's phenomenon ?
  - The problem here is the use of a unique polynomial and regular intervals between knots.
  - Instead, if we concentrate knots at the extremities, the interpolation is less prone to Runge's phenomenon.
  - Make use of Chebyshev knots:

$$u_i^n = \cos\left(\frac{2i-1}{2n}\pi\right) \quad \text{in the interval } [-1,1]$$

$$\text{or } u_i^n = \frac{1}{2} + \frac{1}{2} \cos\left(\frac{2i-1}{2n}\pi\right) \quad \text{in the interval } [0,1]$$

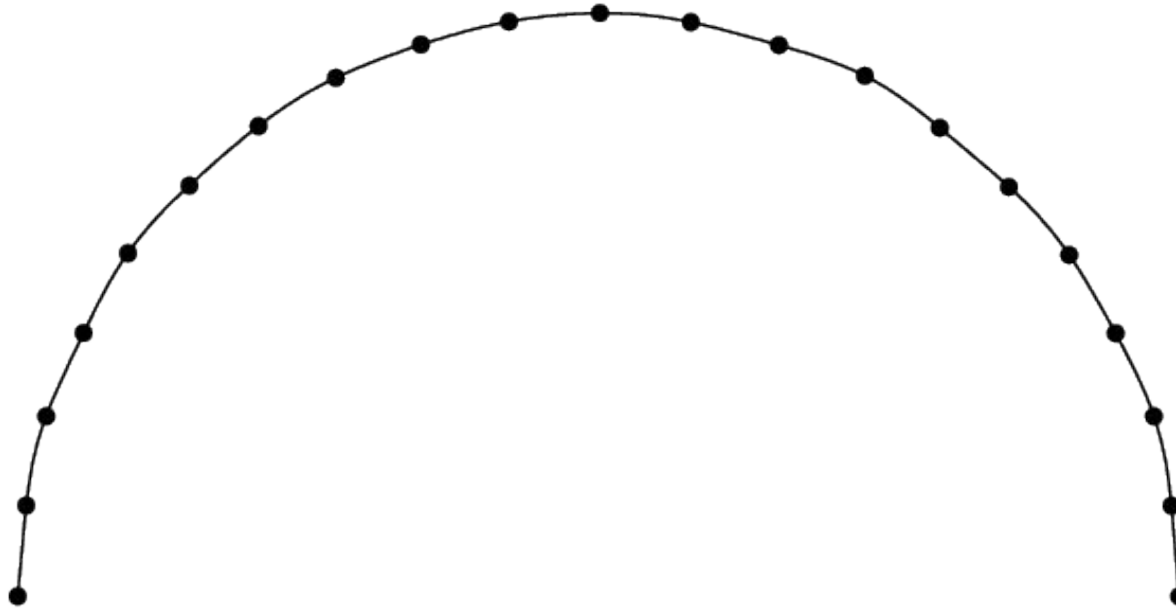
## Lagrange polynomials



21 points, random perturbation 1%

Uniform nodal intervals – lagrange interpolation

## Lagrange polynomials



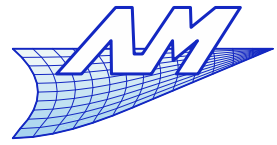
21 points, random perturbation 1%

Using Chebychev knots

## Lagrange polynomials

### Morality :

- Lagrange interpolation is not suited beyond 10's of control points because of the Runge phenomenon
- A modification of the position of a control point leads to global changes of the curve.
- (The evaluation of high order polynomials expressed as monomials leads to numerical problems.)
- No control of the slopes at the boundary of the curve (start and finish).



## Splines

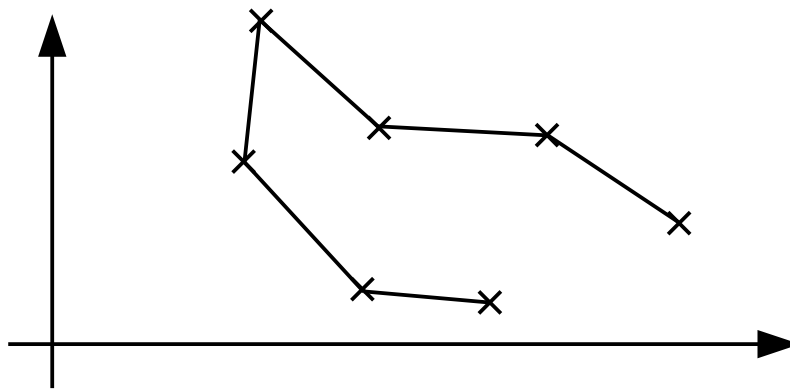


## Splines

- Motivation

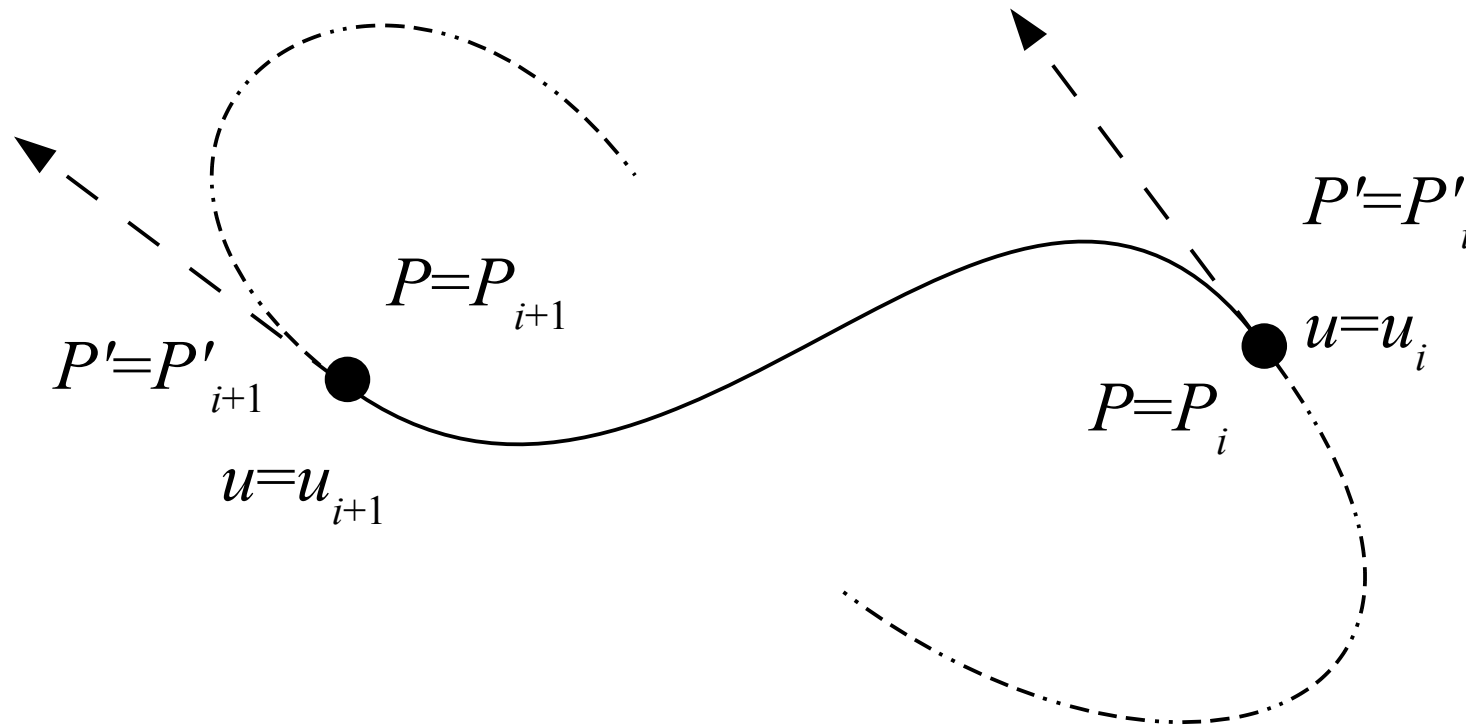
Let us imagine that we have many (100's of) control points

- But we don't want a Lagrange interpolation !
- We should stay with a low order scheme but conserve enough freedom to pass through every point
  - Curve defined by pieces ... and of low order (1)



## Splines

- We are going to build a low order interpolation for each knot interval, such that we can impose slopes at the knots.



## Splines

- In each range  $[i, i+1]$ , we want to have an independent polynomial
- We have 4 parameters : position at each knot and associated tangents.
  - The basis must have 4 degrees of freedom, thus be of order 3 in the case of polynomials.

$$P(u_i) = P_i \equiv \begin{cases} x(u_i) = x_i \\ y(u_i) = y_i \\ \dots \end{cases}$$

$$x_{[i]}(u) = A_{[i]0} + A_{[i]1}u + A_{[i]2}u^2 + A_{[i]3}u^3, \quad u \in [u_i, u_{i+1}]$$

## Splines

- First, every interval has a unit length i.e.

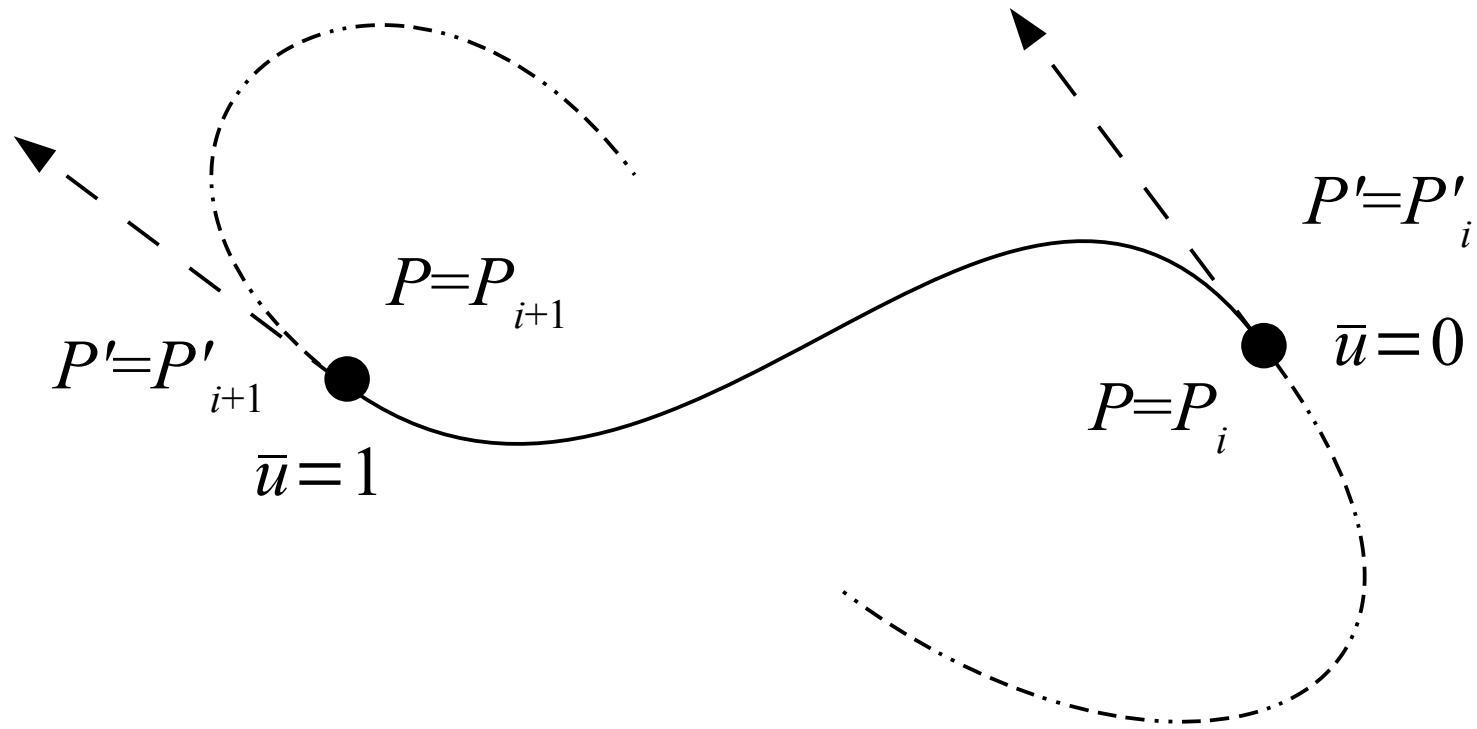
$$u_{i+1} - u_i = 1$$

- Then we ensure identical intervals  $[0..1]$  between each interpolation point :

$$\bar{u} = \frac{u - u_i}{u_{i+1} - u_i} = u - u_i \quad \frac{d\bar{u}}{du} = 1$$

- On each interval  $i$  , we thus have the following relation:

$$x_{[i]}(\bar{u}) = a_{[i]0} + a_{[i]1} \bar{u} + a_{[i]2} \bar{u}^2 + a_{[i]3} \bar{u}^3 \quad , \quad \bar{u} \in [0, 1]$$



## Splines

- We pass through both control points:

$$P(\bar{u}=0) = P_i \Leftrightarrow a_{[i]0} + a_{[i]1} \bar{u} + a_{[i]2} \bar{u}^2 + a_{[i]3} \bar{u}^3 = x_i$$

$$P(\bar{u}=1) = P_{i+1} \Leftrightarrow a_{[i]0} + a_{[i]1} \bar{u} + a_{[i]2} \bar{u}^2 + a_{[i]3} \bar{u}^3 = x_{i+1}$$

- We impose both slopes :

$$P'(\bar{u}_0=0) = P'_i \Leftrightarrow a_{[i]1} + 2 a_{[i]2} \bar{u} + 3 a_{[i]3} \bar{u}^2 = x'_i$$

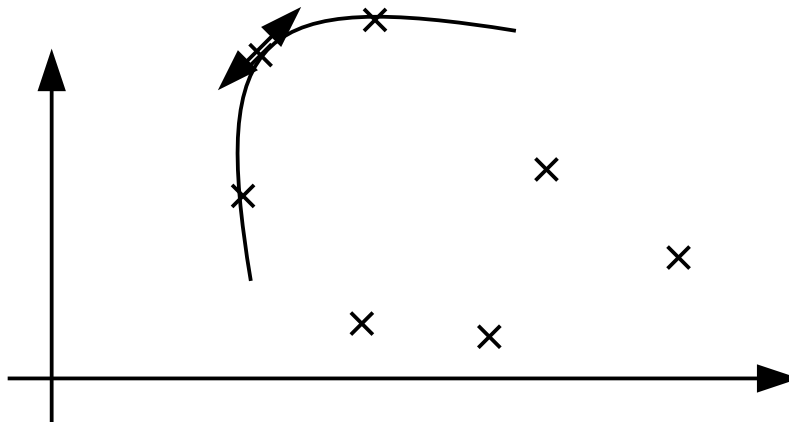
$$P'(\bar{u}=1) = P'_{i+1} \Leftrightarrow a_{[i]1} + 2 a_{[i]2} \bar{u} + 3 a_{[i]3} \bar{u}^2 = x'_{i+1}$$

- At the end :

$$\begin{cases} a_{[i]0} = x_i \\ a_{[i]1} = x'_i \\ a_{[i]2} = 3(x_{i+1} - x_i) - 2x'_i - x'_{i+1} \\ a_{[i]3} = 2(x_i - x_{i+1}) + x'_i + x'_{i+1} \end{cases}$$

## Splines

- We have continuity
- We have continuity of the derivatives
- But how to choose the slopes ?
  - Let the user choose ( “artistic” freedom)
  - Automatically ...



## Splines

- By finite differences with three points :

$$x'_i = \frac{x_{i+1} - x_i}{2(u_{i+1} - u_i)} + \frac{x_i - x_{i-1}}{2(u_i - u_{i-1})}$$

- At the boundaries, we use finite differences (asymmetric)

$$x'_0 = \frac{x_1 - x_0}{u_1 - u_0} \quad x'_{n-1} = \frac{x_{n-1} - x_{n-2}}{u_{n-1} - u_{n-2}}$$

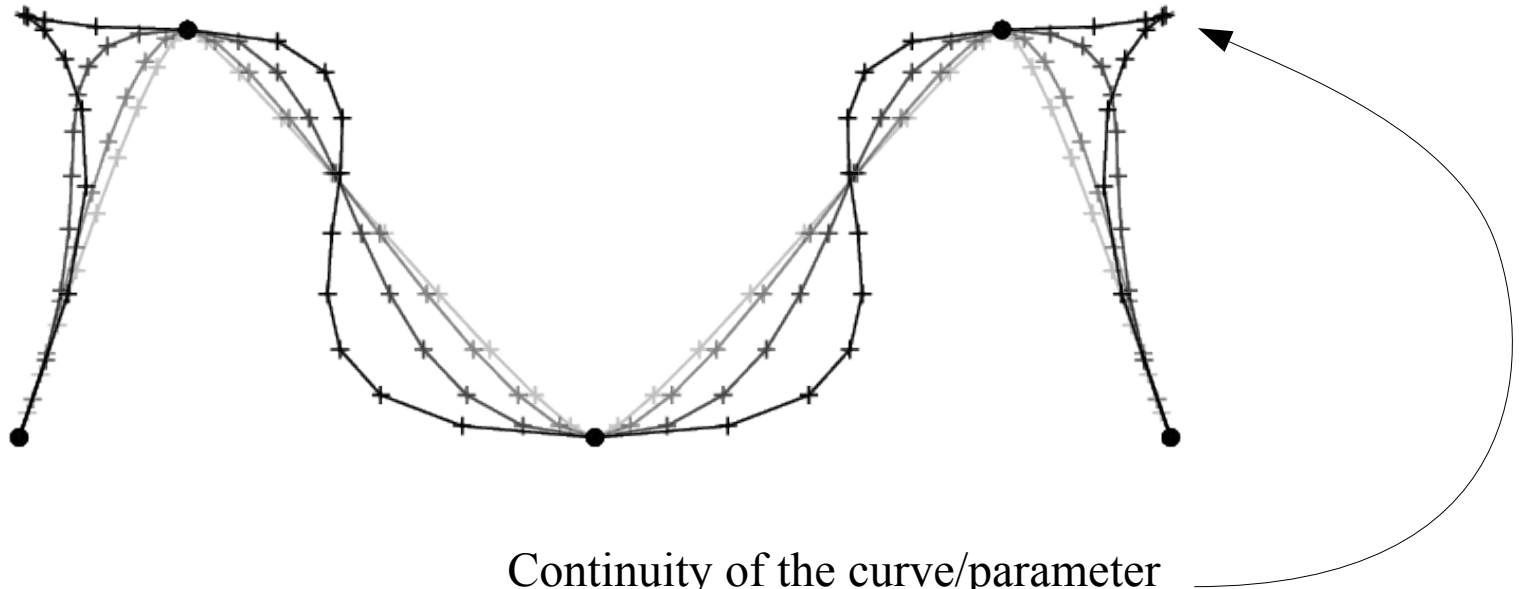
- The result depends on the parametrization !

- Cardinal spline

$$x'_i = (1-c) \frac{x_{i+1} - x_{i-1}}{2}, \quad 0 \leq c \leq 1 \quad x'_0 = (1-c)(x_1 - x_0) \quad x'_{n-1} = (1-c)(x_{n-1} - x_{n-2})$$

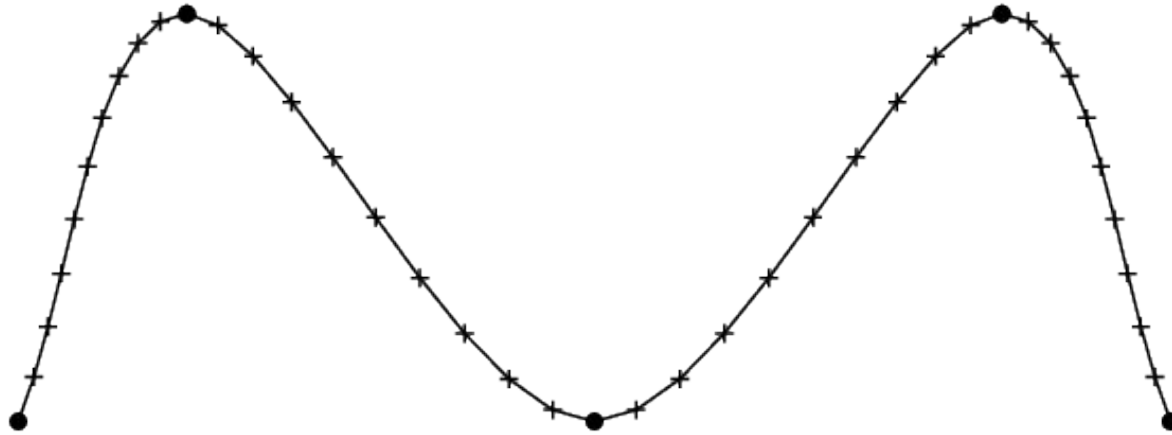
- $c$  is a « tension » parameter.  $c=0$  gives yields the so called “Catmull-Rom” spline,  $c=1$  a zigzagging line.





Continuity of the curve/parameter  
 but loss of regularity (and of geometric continuity  
 in many cases)

5 points, finite differences by varying the parametrization  
 $[0..1]$  ,  $[0..2]$  ,  $[0..5]$  ,  $[0..10]$

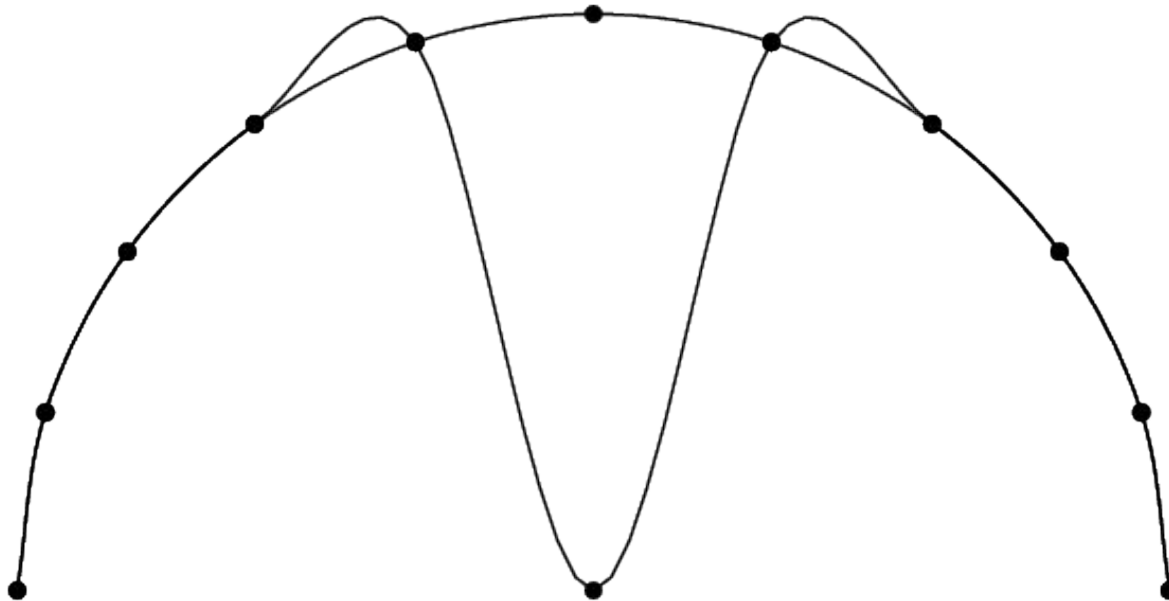


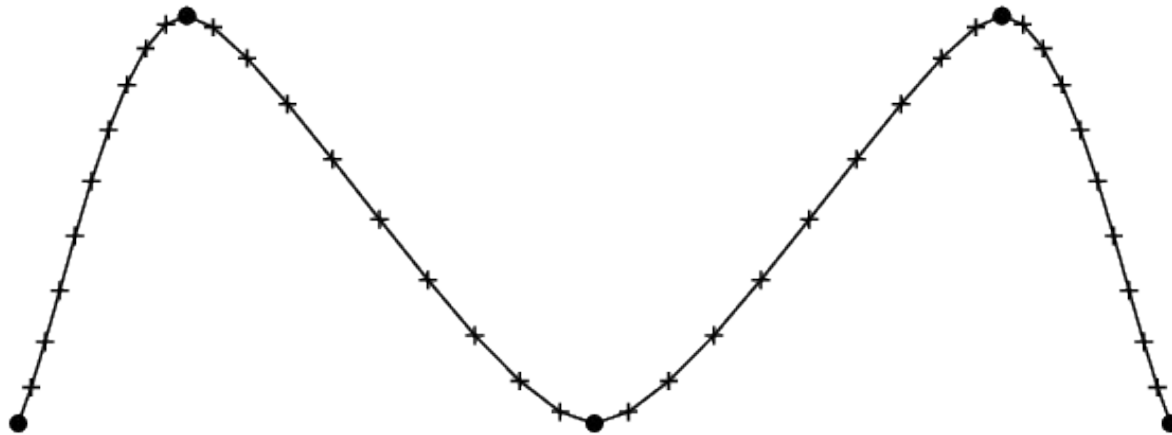
5 points, Cardinal Spline (Catmull-Rom)  $c=0$

# Splines

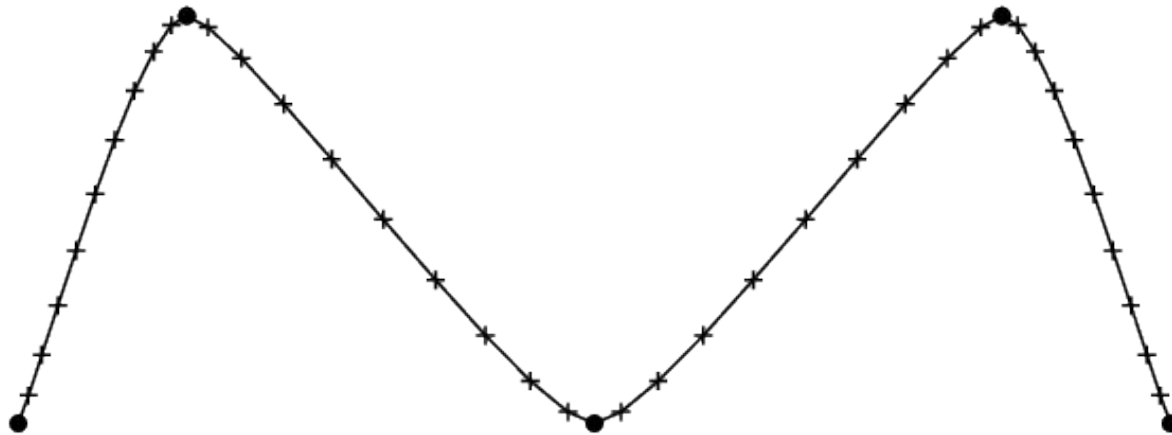
Catmull-Rom Splines are widely used in computer graphics

- Simple to compute, effective
- Local control (price to pay : discontinuous  $\text{sec}^d$  derivative)
- Animations with keyframing
  - Ensures a fluid motion because of the continuity of the slope

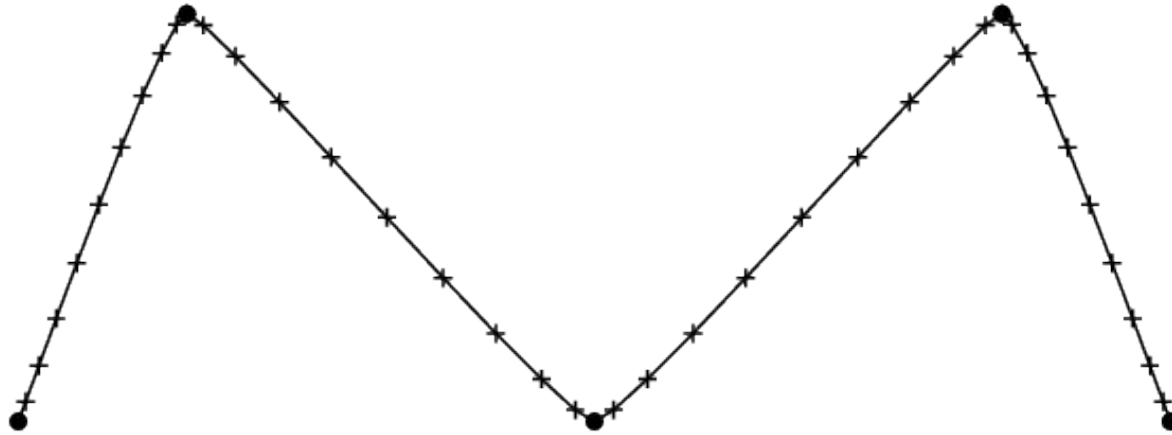




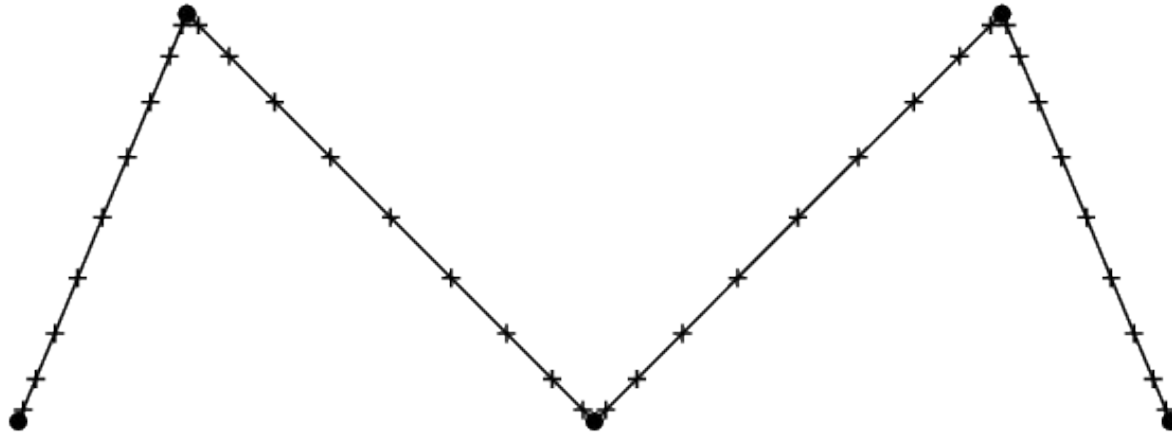
5 points, Cardinal Spline  $c=0.25$



5 points, Cardinal Spline  $c=0.5$



5 points, Cardinal Spline  $c=0.75$



5 points, Cardinal Spline  $c=1.0$

## Splines

- We can impose the continuity of second derivatives...
  - On a curve with  $n$  points, we have  $n$  extra relations to impose
  - We may impose the continuity of the second derivative only on the  $n-2$  interior knots

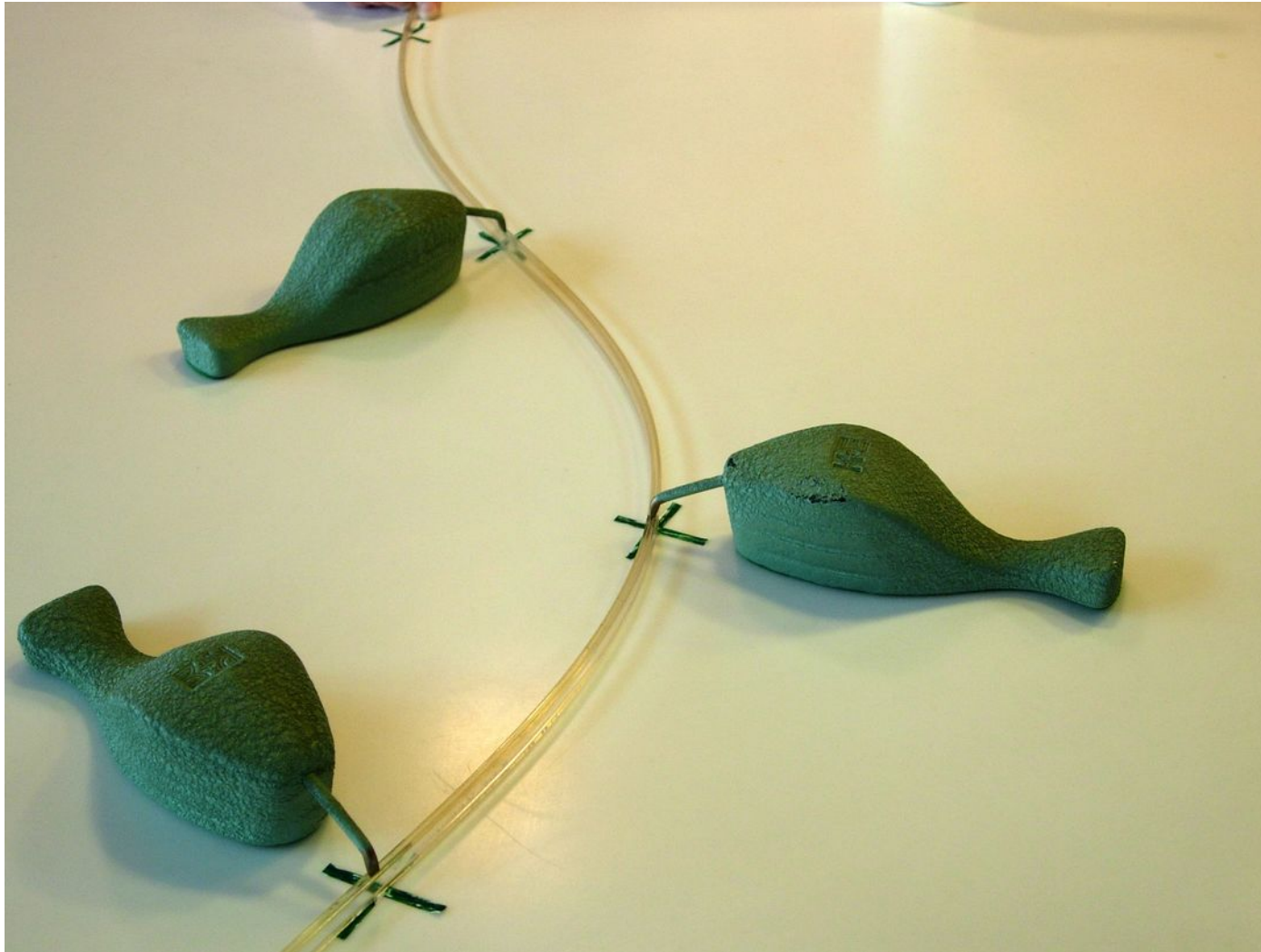
What about the 2 points on the boundary ?

- Impose a vanishing second derivative.  
We obtain what is called « **natural spline** »
- We could also impose the slopes (i.e. only  $n-2$  relations remaining)
- Or , impose that the third derivative is zero on the points 1 and  $n-2$ 
  - That means a single polynomial expression for the first two knot intervals, and the last two.



## Splines

- Natural Spline : mathematical approximation of the spline historically used in naval construction.



## Splines



## Splines

- We impose the continuity of the second derivatives

$$x''_{[i-1]}(1) = x''_{[i]}(0) \Leftrightarrow 2a_{[i-1]2} + 6a_{[i-1]3} = 2a_{[i]2}$$

- We substitute in the “internal” equations

$$\begin{aligned} 2[3(x_i - x_{i-1}) - 2x'_{i-1} - x'_i] + 6[2(x_{i-1} - x_i) + x'_{i-1} + x'_i] \\ = 2[3(x_{i+1} - x_i) - 2x'_i - x'_{i+1}] \end{aligned}$$

- Finally we obtain :

$$x'_{i-1} + 4x'_i + x'_{i+1} = 3(x_{i+1} - x_{i-1})$$

## Splines

- At the boundaries we want

$$x''_{[0]}(0) = 0 \Leftrightarrow 2a_{[0]2} = 0$$

$$2x'_0 + x'_1 = 3(x_1 - x_0)$$

$$x''_{[n-2]}(1) = 0 \Leftrightarrow 2a_{[n-2]2} + 6a_{[n-2]3} = 0$$

$$x'_{n-2} + 2x'_{n-1} = 3(x_{n-1} - x_{n-2})$$

- We have then a linear system with  $n$  unknowns :

$$\begin{pmatrix} 2 & 1 & & & & & \\ 1 & 4 & 1 & & & & \\ & 1 & 4 & 1 & & & \\ & & & \ddots & & & \\ & & & & 1 & 4 & 1 \\ & & & & & 1 & 2 \end{pmatrix} \begin{pmatrix} x'_0 \\ x'_1 \\ x'_2 \\ \vdots \\ x'_{n-2} \\ x'_{n-1} \end{pmatrix} = \begin{pmatrix} 3(x_1 - x_0) \\ 3(x_2 - x_0) \\ 3(x_3 - x_1) \\ \vdots \\ 3(x_{n-1} - x_{n-3}) \\ 3(x_{n-1} - x_{n-2}) \end{pmatrix}$$

## Splines

- By solving the system, we have :

$$\begin{pmatrix} x_0' \\ x_1' \\ x_2' \\ \vdots \\ x_{n-2}' \\ x_{n-1}' \end{pmatrix}, \text{ which is substituted in } \begin{cases} a_{[i]0} = x_i \\ a_{[i]1} = x_i' \\ a_{[i]2} = 3(x_{i+1} - x_i) - 2x_i' - x_{i+1}' \\ a_{[i]3} = 2(x_i - x_{i+1}) + x_i' + x_{i+1}' \end{cases}$$

,to get the polynomial in each portion :

$$x_{[i]}(\bar{u}) = a_{[i]0} + a_{[i]1}\bar{u} + a_{[i]2}\bar{u}^2 + a_{[i]3}\bar{u}^3, \quad 0 \leq \bar{u} < 1$$

From the global parameter  $u$ , we have to find in which portion we are ( the value of  $i$  ), then compute right polynomial...

Catmull-Rom spline

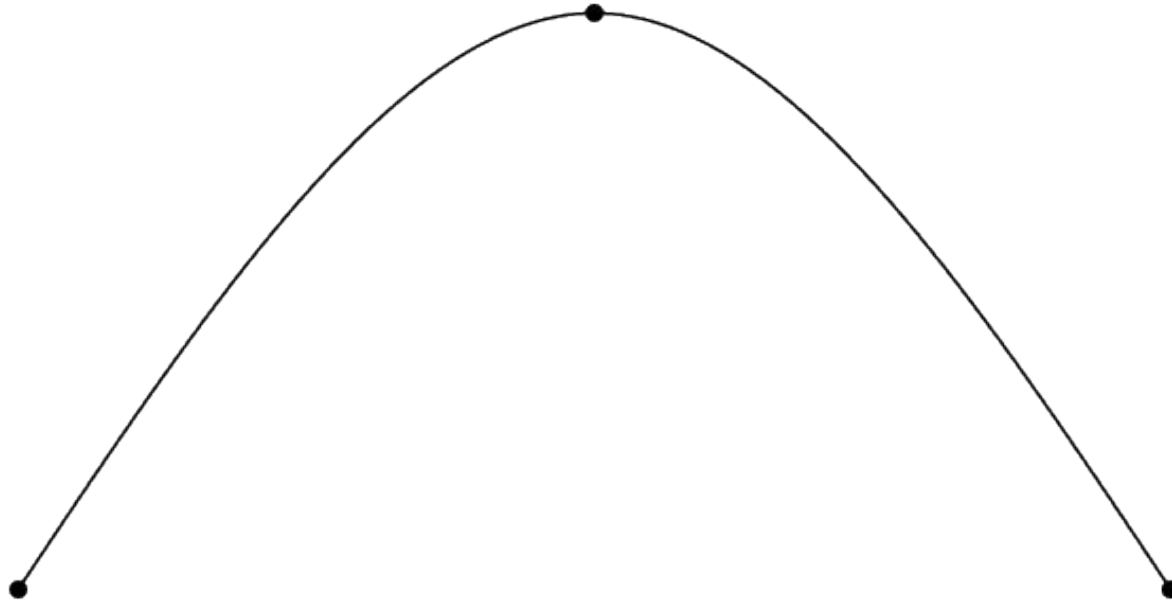


5 control points, natural spline

## Splines

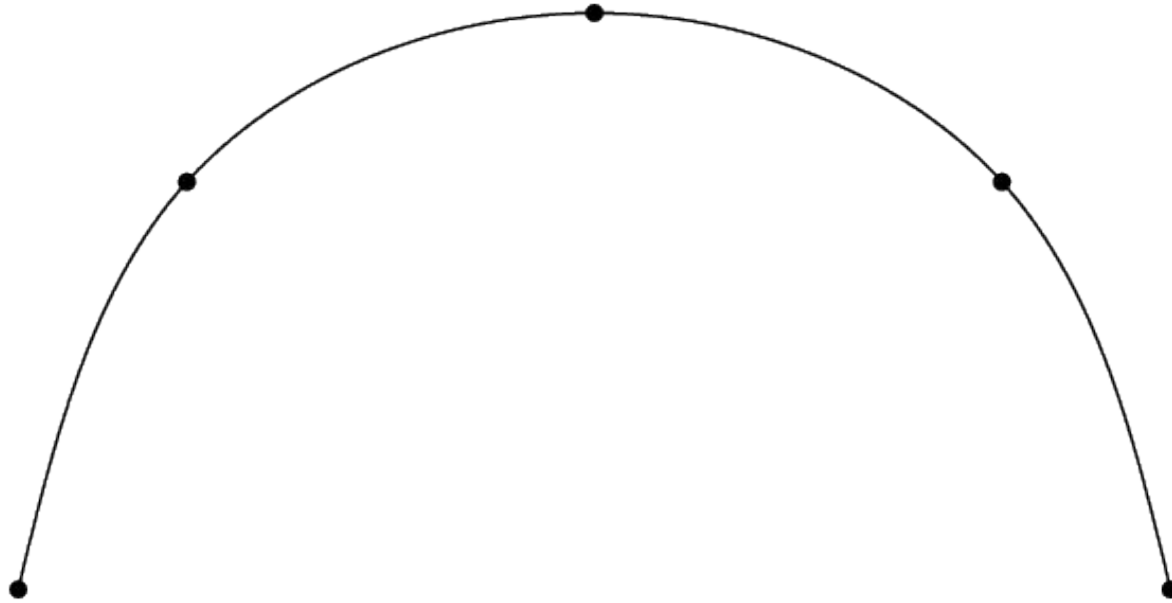
- Another experiment
  - We approximate a circle by a number of increasing points
  - Simultaneously, ~~the order of the approximation~~ the number of pieces increases.
  - In all the cases, the curve is  $C_{\infty} C_2$

## Splines

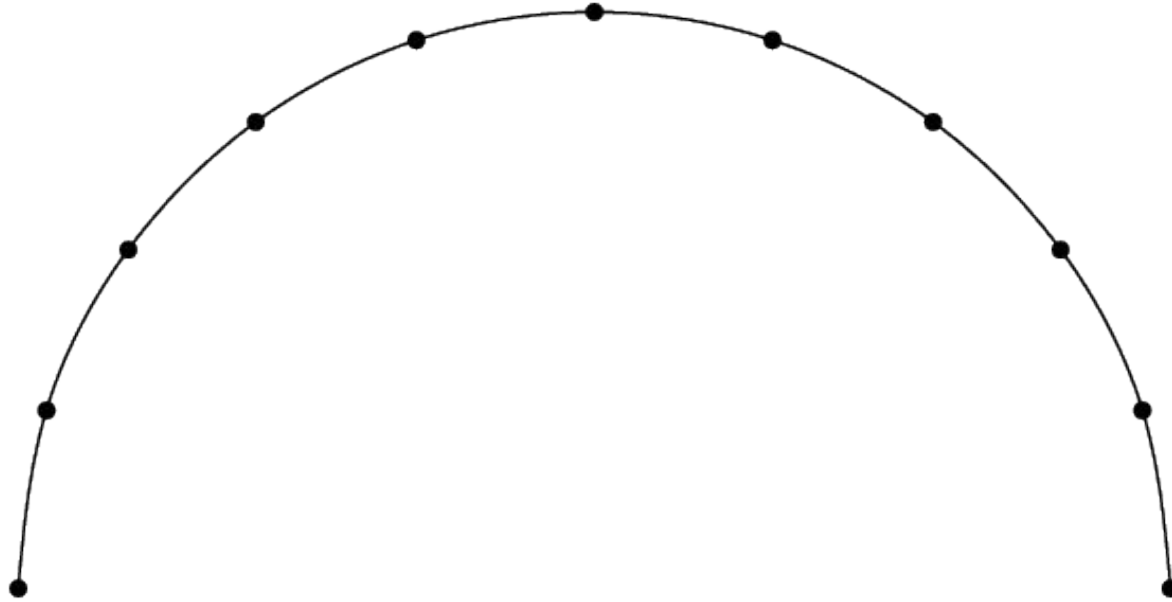


3 points, order 3!

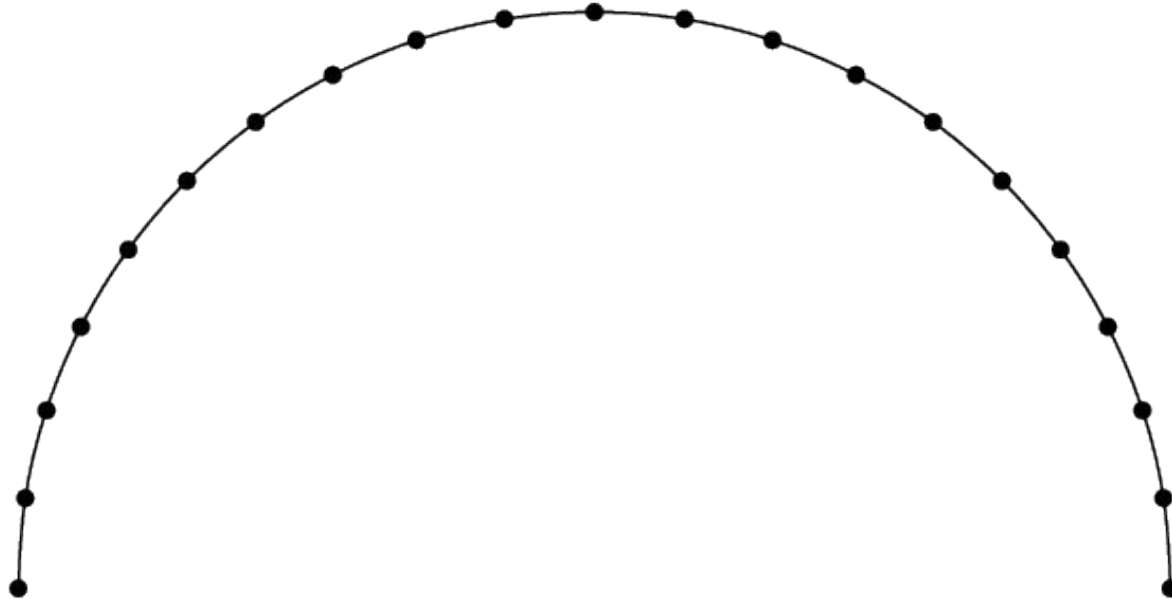




5 points



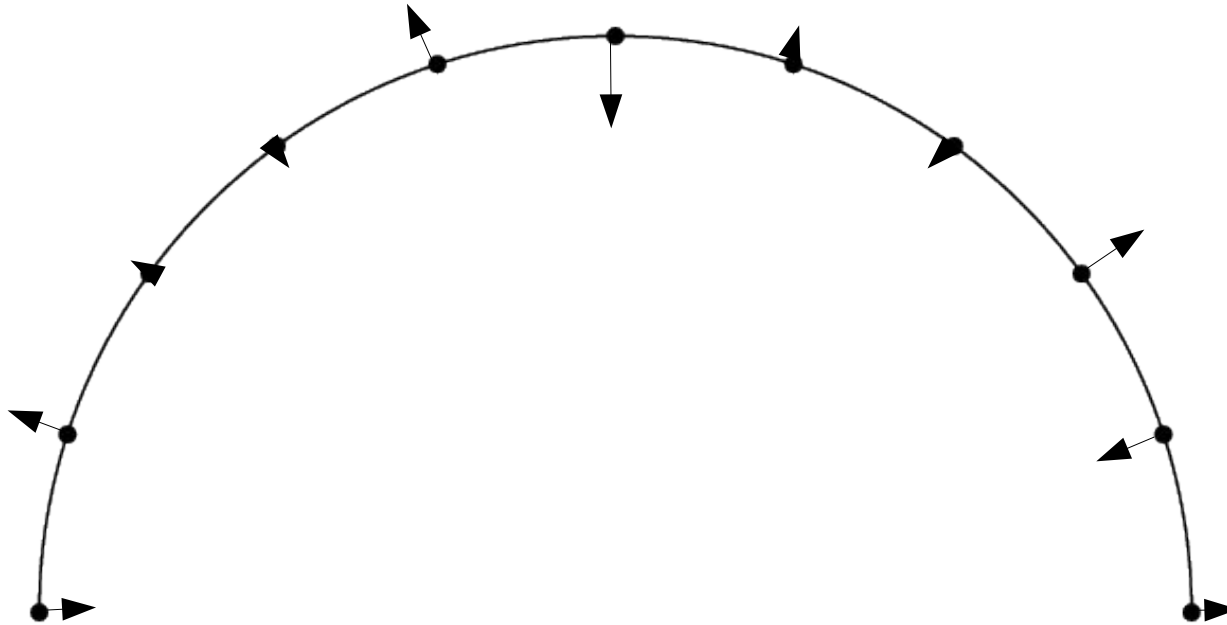
11 points



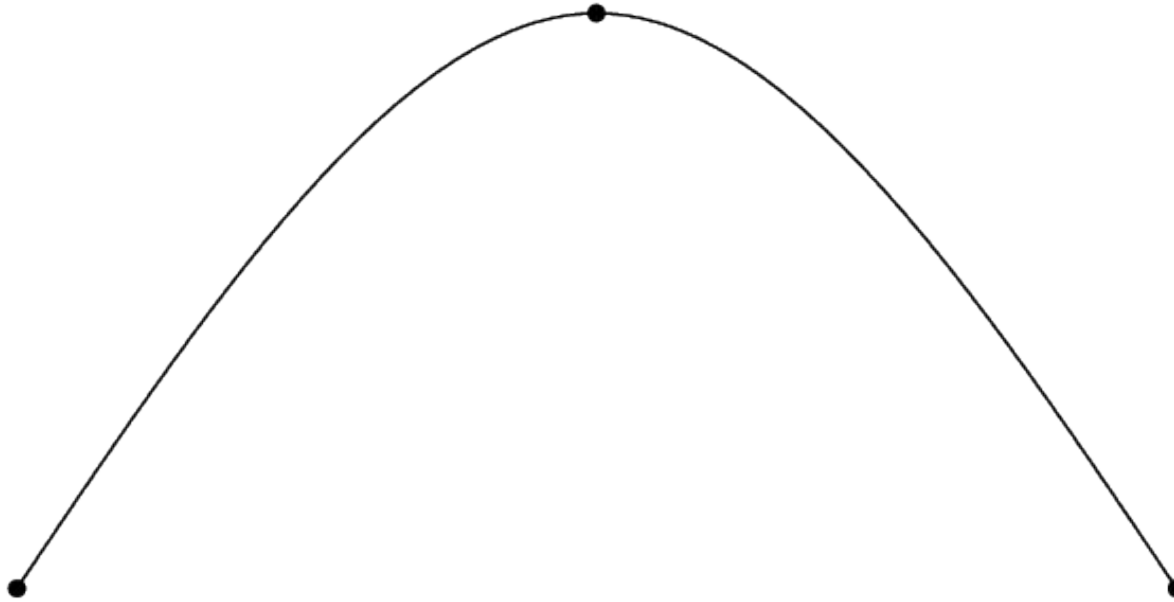
21 points

## Splines

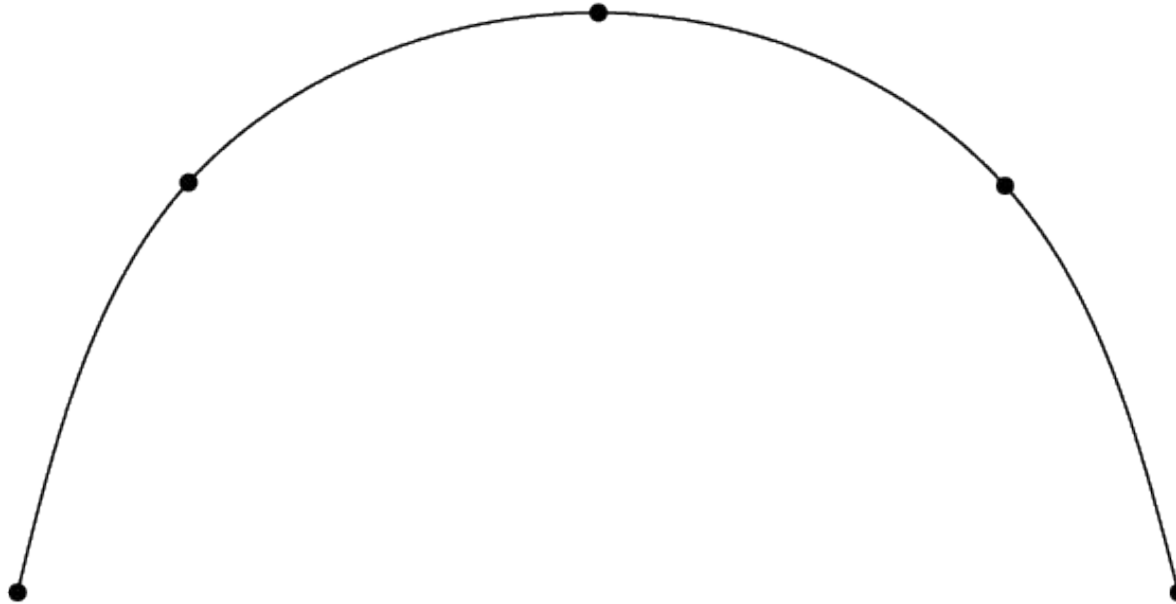
- Random perturbation
  - Each point is moved radially by a value between -0.5 and +0.5 % of the circle's radius



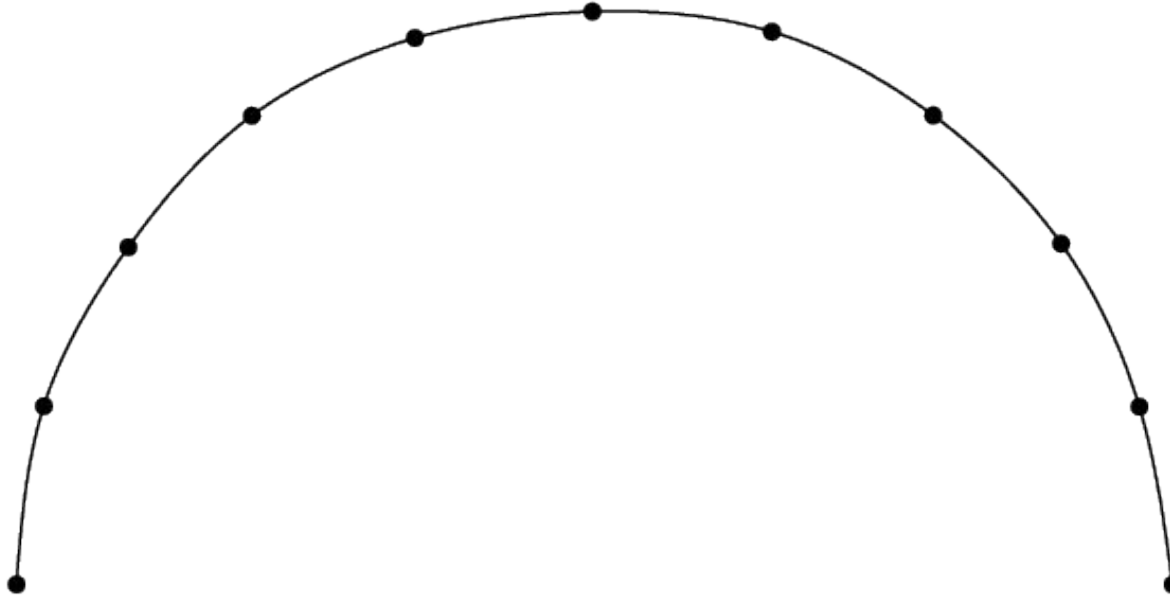
## Splines



3 points, random perturbation 1%

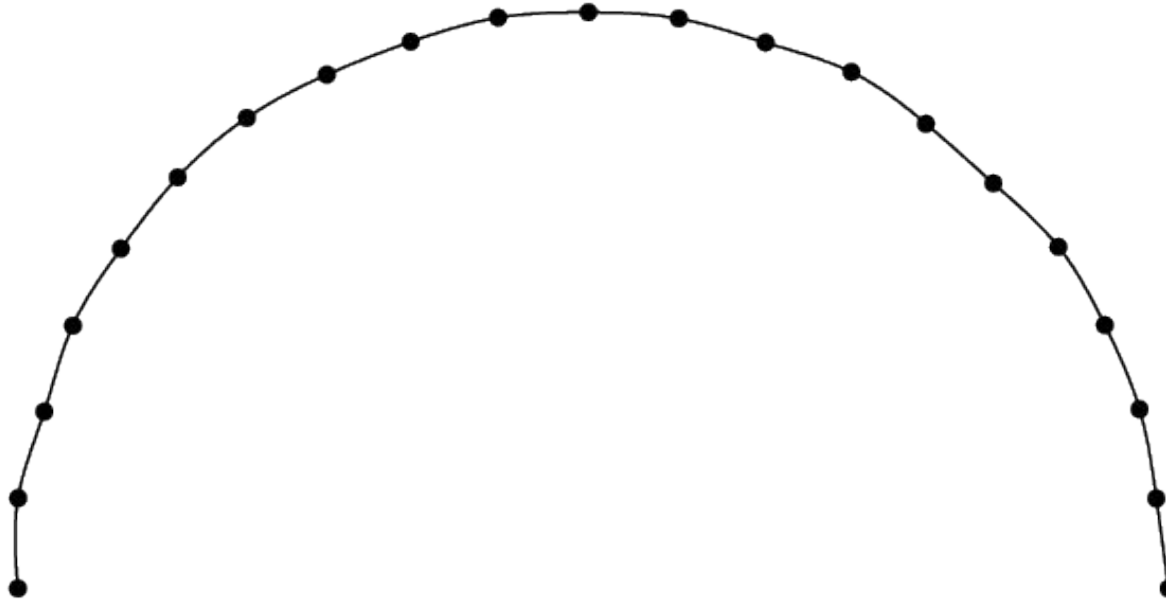


5 points, random perturbation 1%



11 points, random perturbation 1%

## Splines

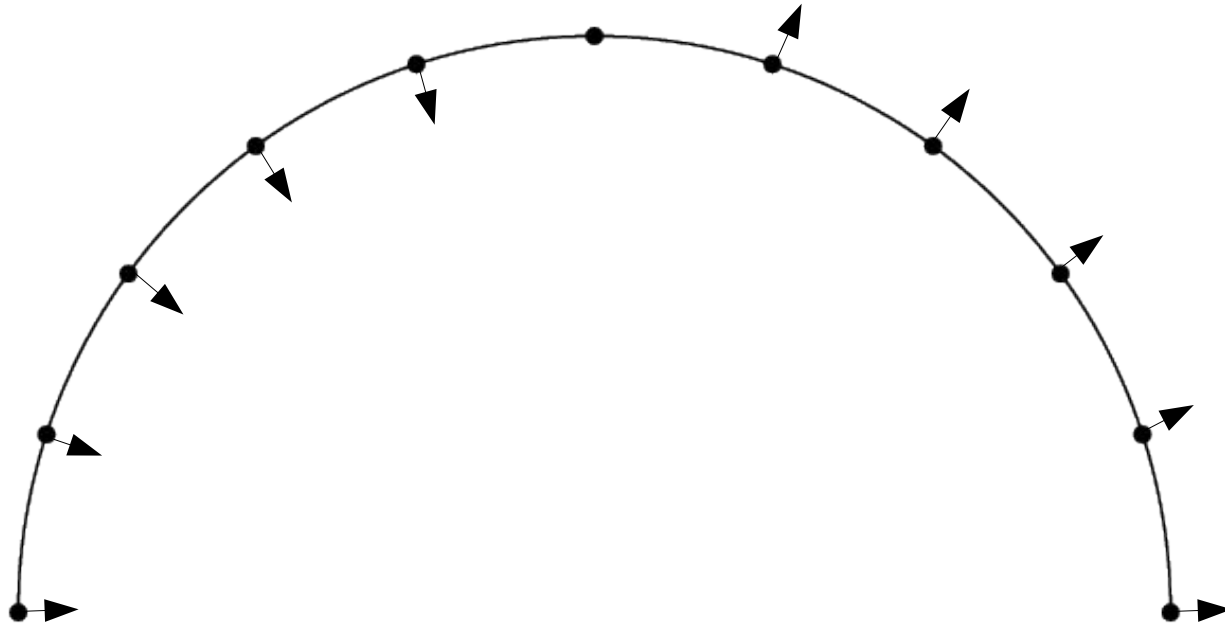


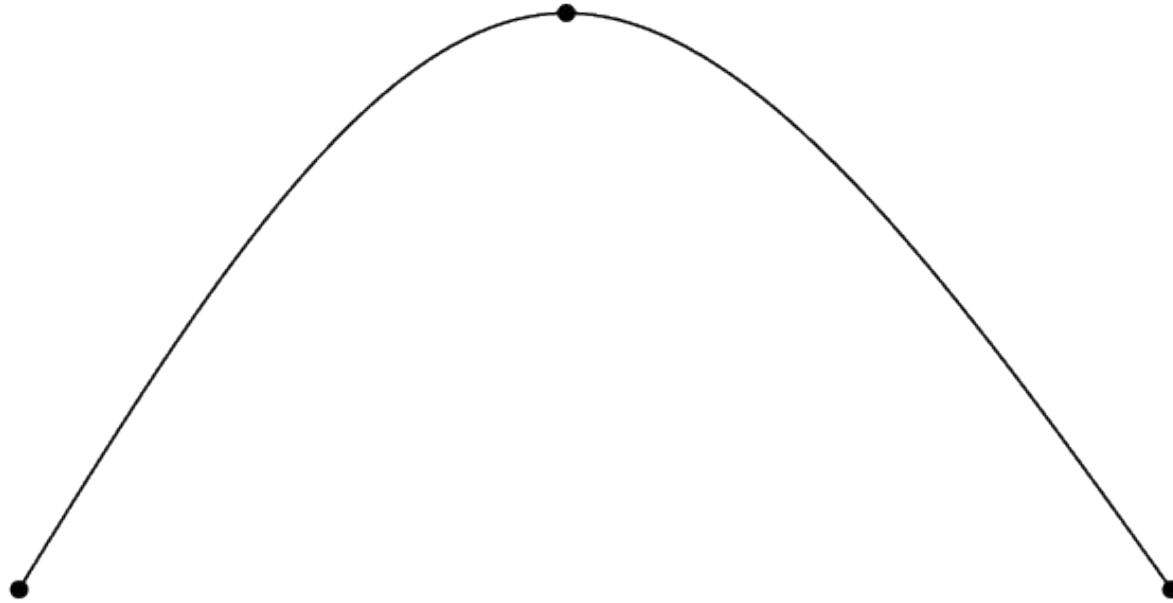
21 points, random perturbation 1%



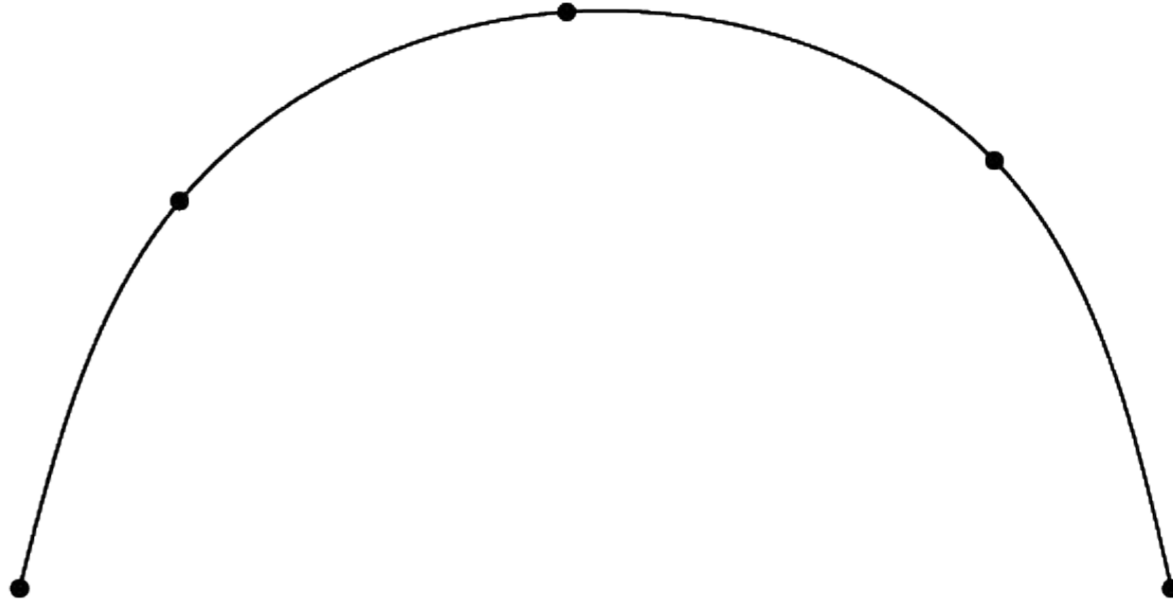
## Splines

- Deterministic perturbation
  - Each point is shifted radially depending on its position by -5 or +5 % of the circle's radius



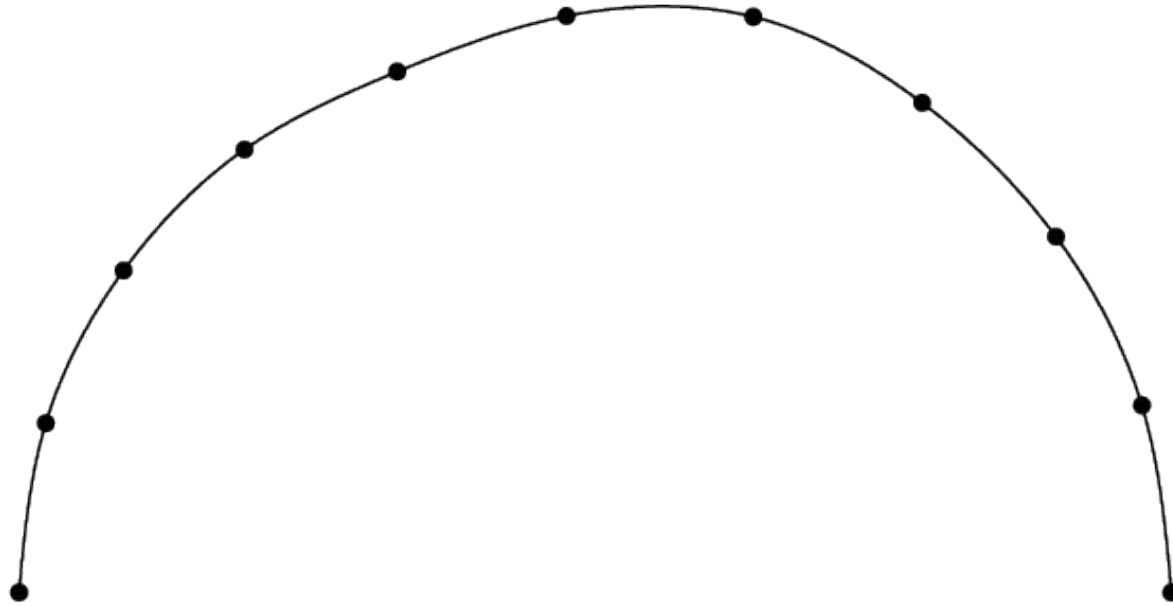


3 points, deterministic perturbation 5%



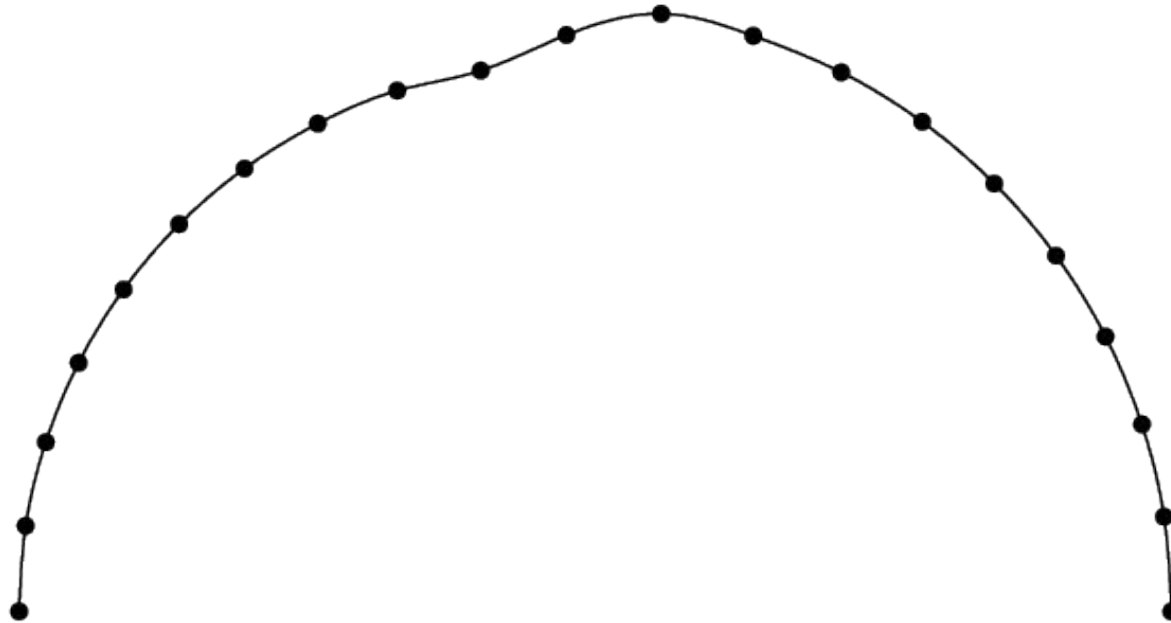
5 points, deterministic perturbation 5%

# Splines



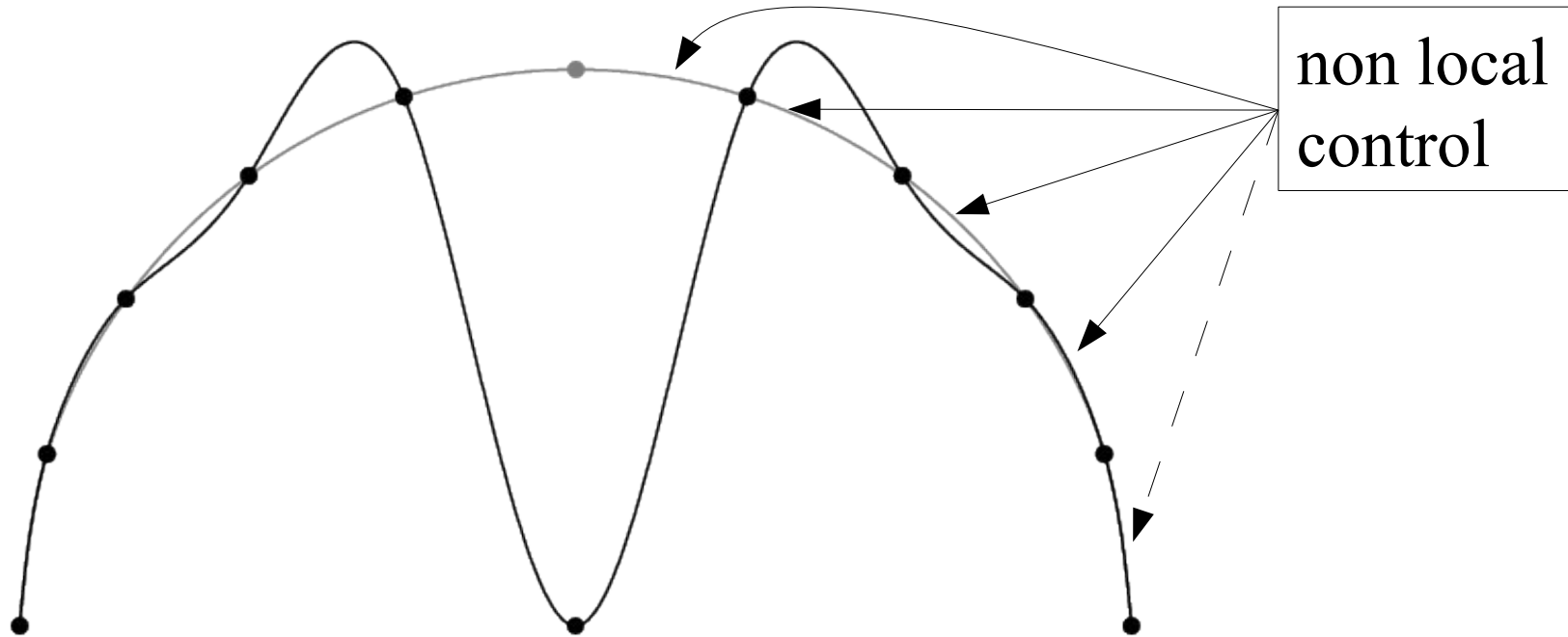
11 points, deterministic perturbation 5%

## Splines



21 points, deterministic perturbation 5%

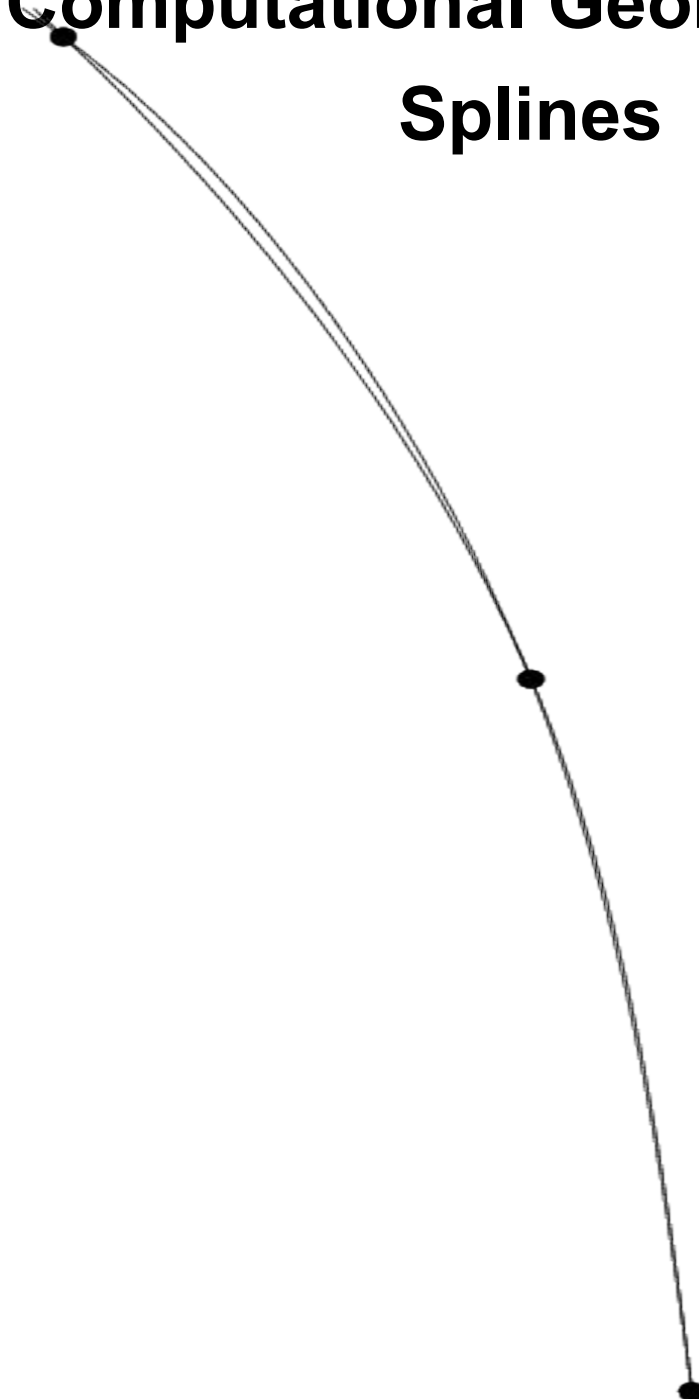
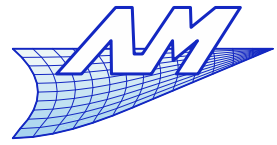
## Splines



11 points

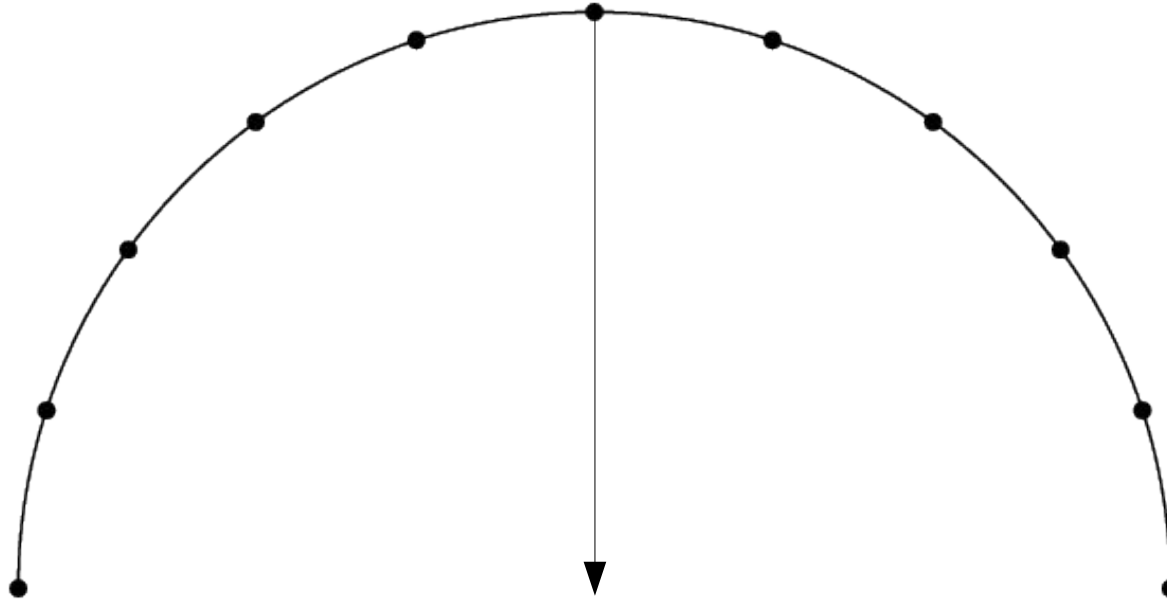
# CAD & Computational Geometry

## Splines

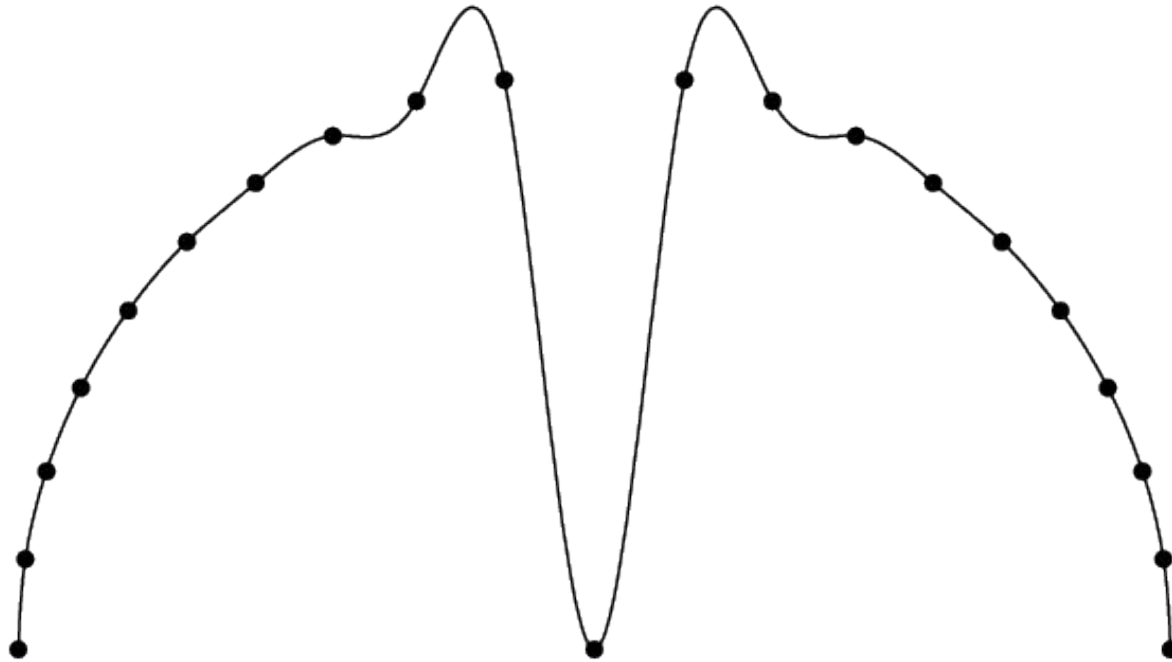


## Splines

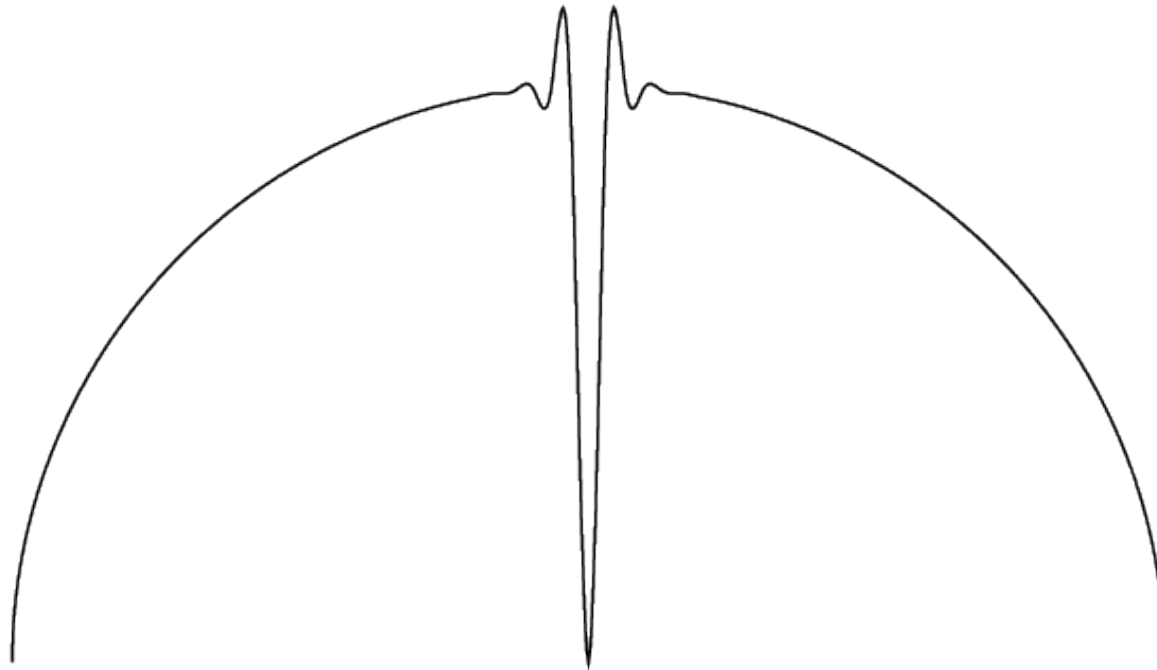
- Perturbation of a point
  - We shift one point by a significant amount



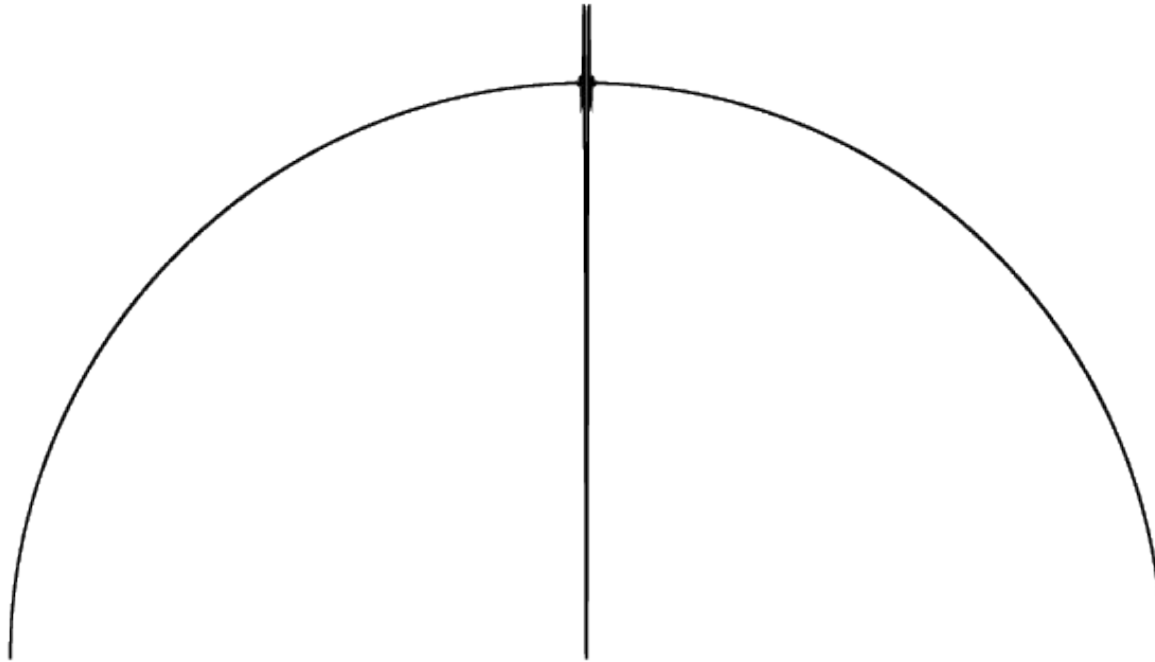




21 points



99 points



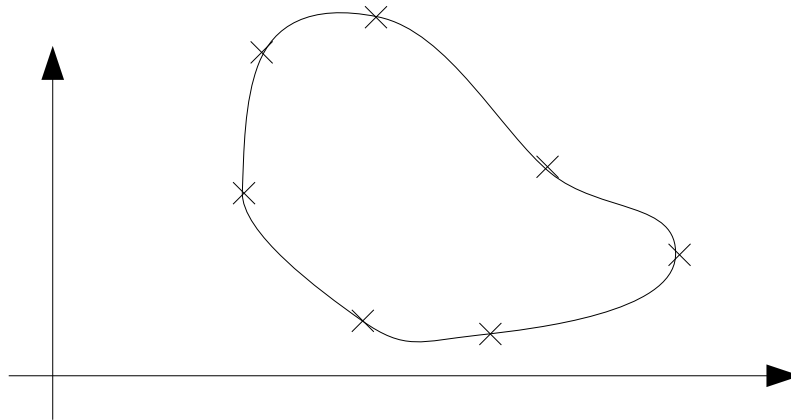
999 points

## Splines

- Stable interpolation scheme
- Weak Runge phenomenon
- The displacement of a point yet affects all the curve
  - Nevertheless, the perturbation fades very quickly further away from the shifted point
  - « Overshoots » are limited.

## Splines

- Closed curve ?
  - The curve can be closed, just impose everywhere that the second derivative is continuous.





## Splines

- General case : arbitrary parametrization

$$P(u_i) = P_i \equiv \begin{cases} x(u_i) = x_i \\ y(u_i) = y_i \end{cases}$$

$$x_{[i]}(u) = A_{[i]0} + A_{[i]1}u + A_{[i]2}u^2 + A_{[i]3}u^3, \quad u \in [u_i, u_{i+1}]$$

- We again change the parametrization...

$$\bar{u} = \frac{u - u_i}{u_{i+1} - u_i} \quad \frac{d\bar{u}}{du} = \frac{1}{u_{i+1} - u_i} = \frac{1}{h_i}$$

$$x_{[i]}(\bar{u}) = a_{[i]0} + a_{[i]1}\bar{u} + a_{[i]2}\bar{u}^2 + a_{[i]3}\bar{u}^3, \quad \bar{u} \in [0, 1]$$

## Splines

- We pass through both points :

$$P(\bar{u}_0=0) = P_i \Leftrightarrow a_{[i]0} + a_{[i]1}\bar{u}_0 + a_{[i]2}\bar{u}_0^2 + a_{[i]3}\bar{u}_0^3 = x_i$$

$$P(\bar{u}_1=1) = P_{i+1} \Leftrightarrow a_{[i]0} + a_{[i]1}\bar{u}_1 + a_{[i]2}\bar{u}_1^2 + a_{[i]3}\bar{u}_1^3 = x_{i+1}$$

- We impose both slopes :

$$P'(\bar{u}_0=0) = \frac{dP}{du}(0) = \frac{dP}{d\bar{u}}(0) \frac{1}{h_i} = P'_i \Leftrightarrow a_{[i]1} + 2a_{[i]2}\bar{u}_0 + 3a_{[i]3}\bar{u}_0^2 = x'_i h_i$$

$$P'(\bar{u}_1=1) = \frac{dP}{du}(1) = \frac{dP}{d\bar{u}}(1) \frac{1}{h_i} = P'_{i+1} \Leftrightarrow a_{[i]1} + 2a_{[i]2}\bar{u}_1 + 3a_{[i]3}\bar{u}_1^2 = x'_{i+1} h_i$$

- Finally :

$$\begin{cases} a_{[i]0} = x_i \\ a_{[i]1} = x'_i h_i \\ a_{[i]2} = 3(x_{i+1} - x_i) - 2x'_i h_i - x'_{i+1} h_i \\ a_{[i]3} = 2(x_i - x_{i+1}) + x'_i h_i + x'_{i+1} h_i \end{cases}$$



## Splines

- We impose the second derivative for a natural spline

$$\frac{d^2 P_{[i]}(\bar{u})}{du^2} = \frac{d^2 P_{[i]} d^2 \bar{u}}{d \bar{u}^2 d u^2} = \frac{d^2 P_{[i]}}{d \bar{u}^2} \frac{1}{h_i^2}$$

$$x_{[i-1]}''(1) = x_{[i]}''(0) \Leftrightarrow \frac{2a_{[i-1]2} + 6a_{[i-1]3}}{h_{i-1}^2} = \frac{2a_{[i]2}}{h_i^2}$$

- We substitute in the internal equations

$$\begin{aligned} & \frac{2[3(x_i - x_{i-1}) - 2x'_{i-1}h_{i-1} - x'_i h_{i-1}] + 6[2(x_{i-1} - x_i) + x'_{i-1}h_{i-1} + x'_i h_{i-1}]}{h_{i-1}^2} \\ & = \frac{2[3(x_{i+1} - x_i) - 2x'_i h_i - x'_{i+1} h_i]}{h_i^2} \end{aligned}$$

## Splines

- We obtain finally :

$$\frac{x'_{i-1} + 2x'_i}{h_{i-1}} + \frac{2x'_i + x'_{i+1}}{h_i} = 3 \frac{(x_i - x_{i-1})}{h_{i-1}^2} + 3 \frac{(x_{i+1} - x_i)}{h_i^2}$$

$$h_i(x'_{i-1} + 2x'_i) + h_{i-1}(2x'_i + x'_{i+1}) = 3 \frac{h_i}{h_{i-1}}(x_i - x_{i-1}) + 3 \frac{h_{i-1}}{h_i}(x_{i+1} - x_i)$$

## Splines

- At the boundaries we want a vanishing second derivative ...

$$x''_{[0]}(0) = 0 \Leftrightarrow \frac{2a_{[0]2}}{(h_0^2)} = 0$$

$$h_0(2x'_0 + x'_1) = 3(x_1 - x_0)$$

$$x''_{[n-2]}(1) = 0 \Leftrightarrow \frac{2a_{[n-2]2} + 6a_{[n-2]3}}{(h_{n-2}^2)} = 0$$

$$h_{n-2}(x'_{n-2} + 2x'_{n-1}) = 3(x_{n-1} - x_{n-2})$$

- We have then a linear system of  $n$  unknowns:  
(next page)

## Splines

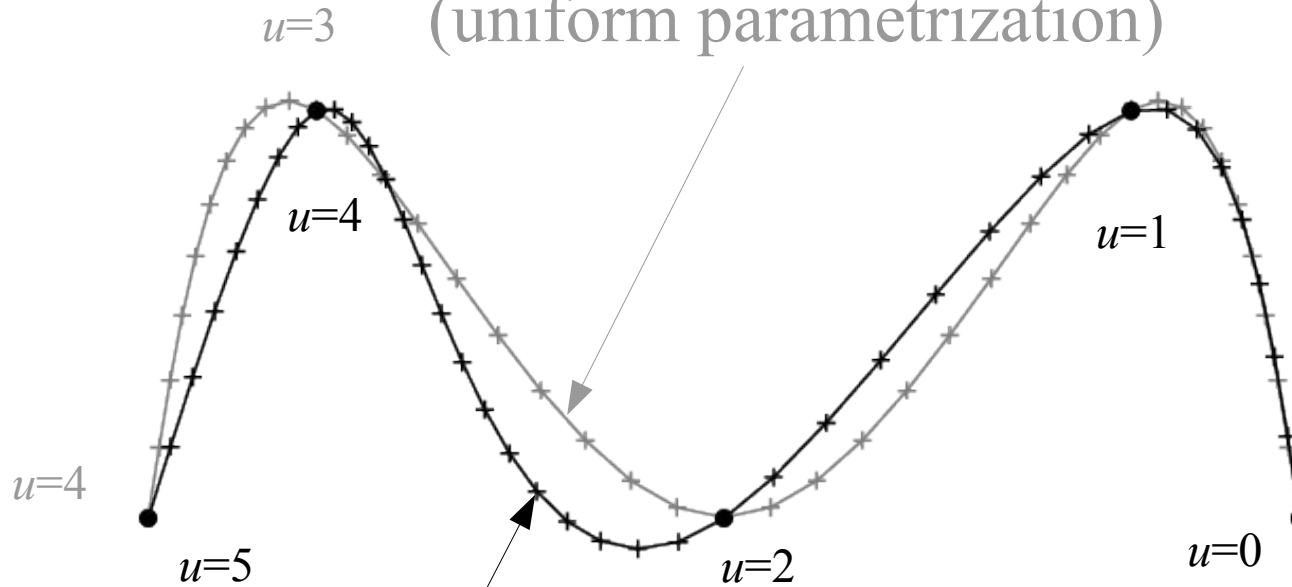
$$\begin{pmatrix} 2h_0 & h_0 & & & & \\ h_1 & 2h_1+2h_0 & h_0 & & & \\ & & \ddots & & & \\ & & & h_{n-2} & 2h_{n-2}+2h_{n-3} & h_{n-3} \\ & & & & h_{n-2} & 2h_{n-2} \end{pmatrix} \begin{pmatrix} x'_0 \\ x'_1 \\ \vdots \\ x'_{n-2} \\ x'_{n-1} \end{pmatrix} =$$

$$\begin{pmatrix} 3(x_1 - x_0) \\ 3\frac{h_1}{h_0}(x_1 - x_0) + 3\frac{h_0}{h_1}(x_2 - x_1) \\ \vdots \\ 3\frac{h_{n-2}}{h_{n-3}}(x_{n-2} - x_{n-3}) + 3\frac{h_{n-3}}{h_{n-2}}(x_{n-1} - x_{n-2}) \\ 3(x_{n-1} - x_{n-2}) \end{pmatrix}$$

## Splines

Natural Spline

(uniform parametrization)



Natural Spline

(non uniform parametrization)

## Splines

- Compact notation

- We would like a compact representation as for the Lagrange interpolation.

$$x_{[i]}(\bar{u}) = a_{[i]0} + a_{[i]1}\bar{u} + a_{[i]2}\bar{u}^2 + a_{[i]3}\bar{u}^3$$

$$0 \leq \bar{u} < 1$$

$$\begin{cases} a_{[i]0} = x_i \\ a_{[i]1} = x'_i \\ a_{[i]2} = 3(x_{i+1} - x_i) - 2x'_i - x'_{i+1} \\ a_{[i]3} = 2(x_i - x_{i+1}) + x'_i + x'_{i+1} \end{cases}$$

$$\begin{cases} x_{[i]}(\bar{u}) = \sum_0^n x_i h_{i0}^p(\bar{u}) + \sum_0^n x'_i h_{i1}^p(\bar{u}) \\ y_{[i]}(\bar{u}) = \sum_0^n y_i h_{i0}^p(\bar{u}) + \sum_0^n y'_i h_{i1}^p(\bar{u}) \Leftrightarrow P_{[i]}(\bar{u}) = \sum_0^n P_i h_{i0}^p(\bar{u}) + \sum_0^n P'_i h_{i1}^p(\bar{u}) \end{cases}$$

...

This on each interval  $[i, i+1]$ .

## Splines

- We have on each interval :

$$\begin{cases} a_{[i]0} = x_i \\ a_{[i]1} = x'_i \\ a_{[i]2} = 3(x_{i+1} - x_i) - 2x'_i - x'_{i+1} \\ a_{[i]3} = 2(x_i - x_{i+1}) + x'_i + x'_{i+1} \end{cases}$$

$$x_{[i]}(\bar{u}) = x_i + x'_i \bar{u} + (3(x_{i+1} - x_i) - 2x'_i - x'_{i+1}) \bar{u}^2 + (2(x_i - x_{i+1}) + x'_i + x'_{i+1}) \bar{u}^3$$

- By rearranging equations

$$\begin{aligned} x_{[i]}(\bar{u}) &= x_i(1 - 3\bar{u}^2 + 2\bar{u}^3) + x'_i(\bar{u} - 2\bar{u}^2 + \bar{u}^3) \\ &\quad + x_{i+1}(3\bar{u}^2 - 2\bar{u}^3) + x'_{i+1}(-\bar{u}^2 + \bar{u}^3) \end{aligned}$$

## Splines

### ■ 3D Curves

- Minimal order so that a curve can have a torsion (non planar curve)

- Let's consider a Lagrange interpolation  $P(u) = \sum_{i=0}^{n-1} P_i l_i^p(u)$
- 2 points  $\rightarrow$  on a straight line (no curvature)
- 3 points  $\rightarrow$  in a plane (no torsion)
- 4 points  $\rightarrow$  torsion becomes possible

- Minimal order to join smoothly two arbitrarily oriented curves = 3