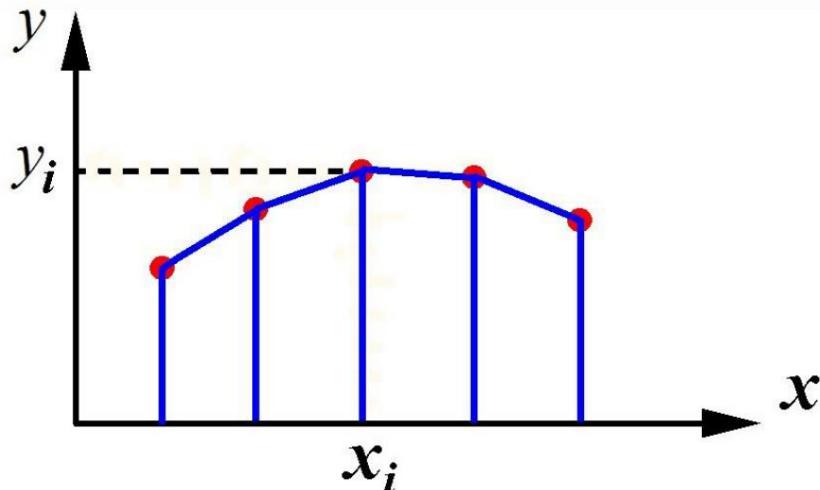


## Chapter 6. Interpolation by splines



recall approximation/interpolation  
by the linear function:

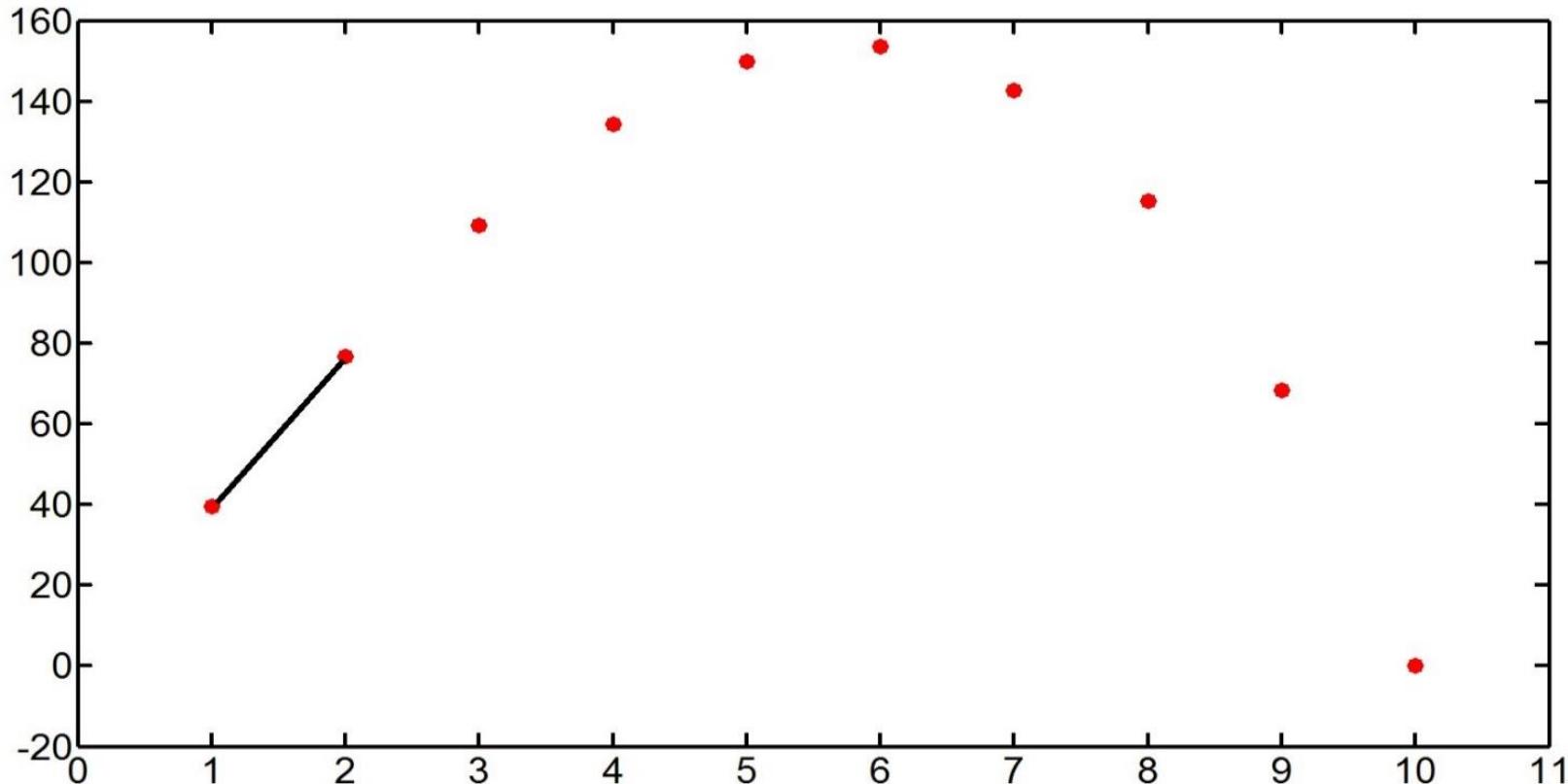
$$f(x) \approx y_i + (y_{i+1} - y_i)(x - x_i)/h$$

No spikes, splashes in this case. However, a drawback:  
**there are jumps, discontinuities of the first-order derivative**

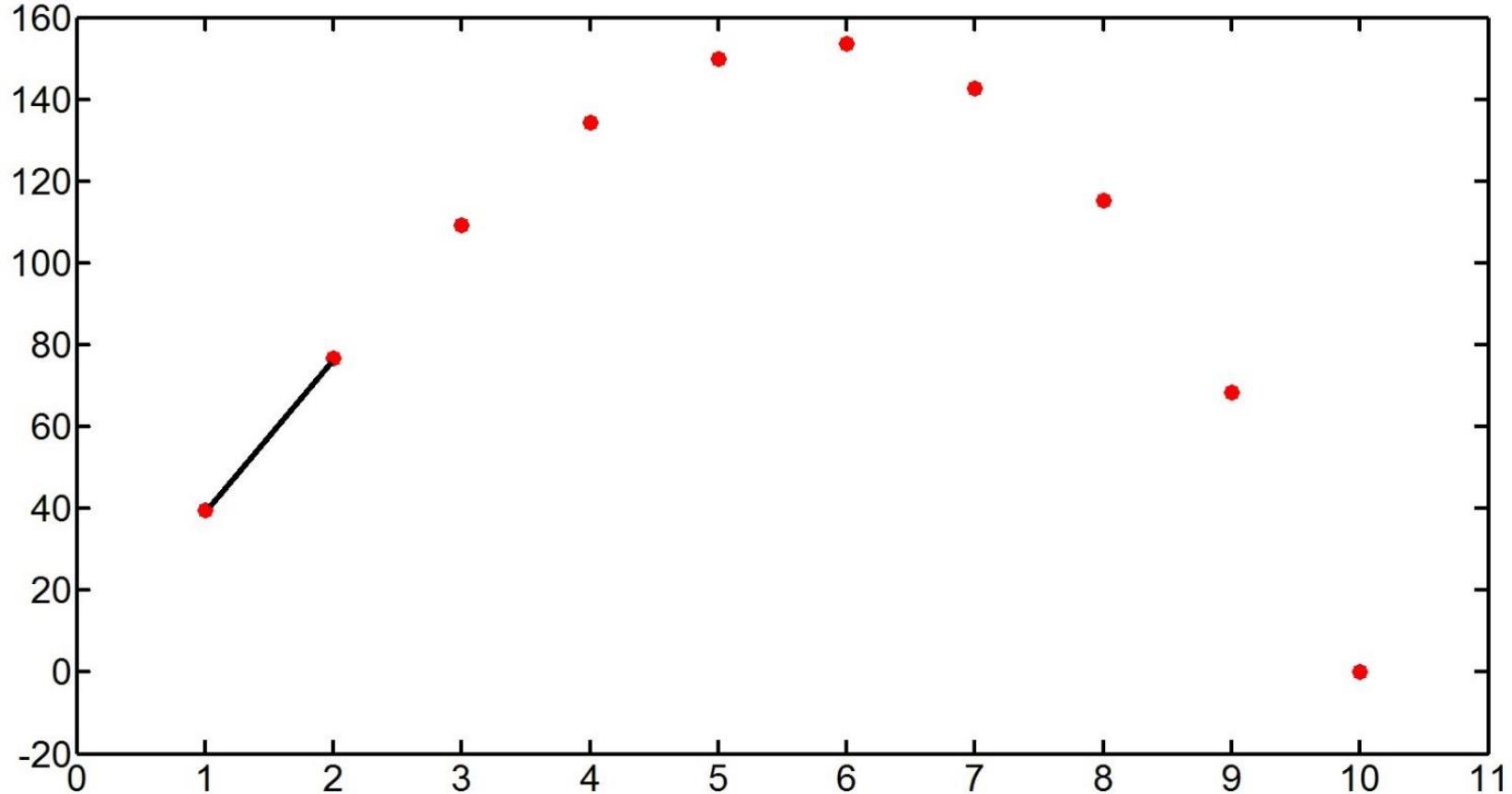
which is  $(y_{i+1} - y_i)/h$     when     $x_i < x < x_{i+1}$ ,

and     $(y_{i+2} - y_{i+1})/h$     when     $x_{i+1} < x < x_{i+2}$ .

That is why, it was suggested to use **splines** for interpolation purposes



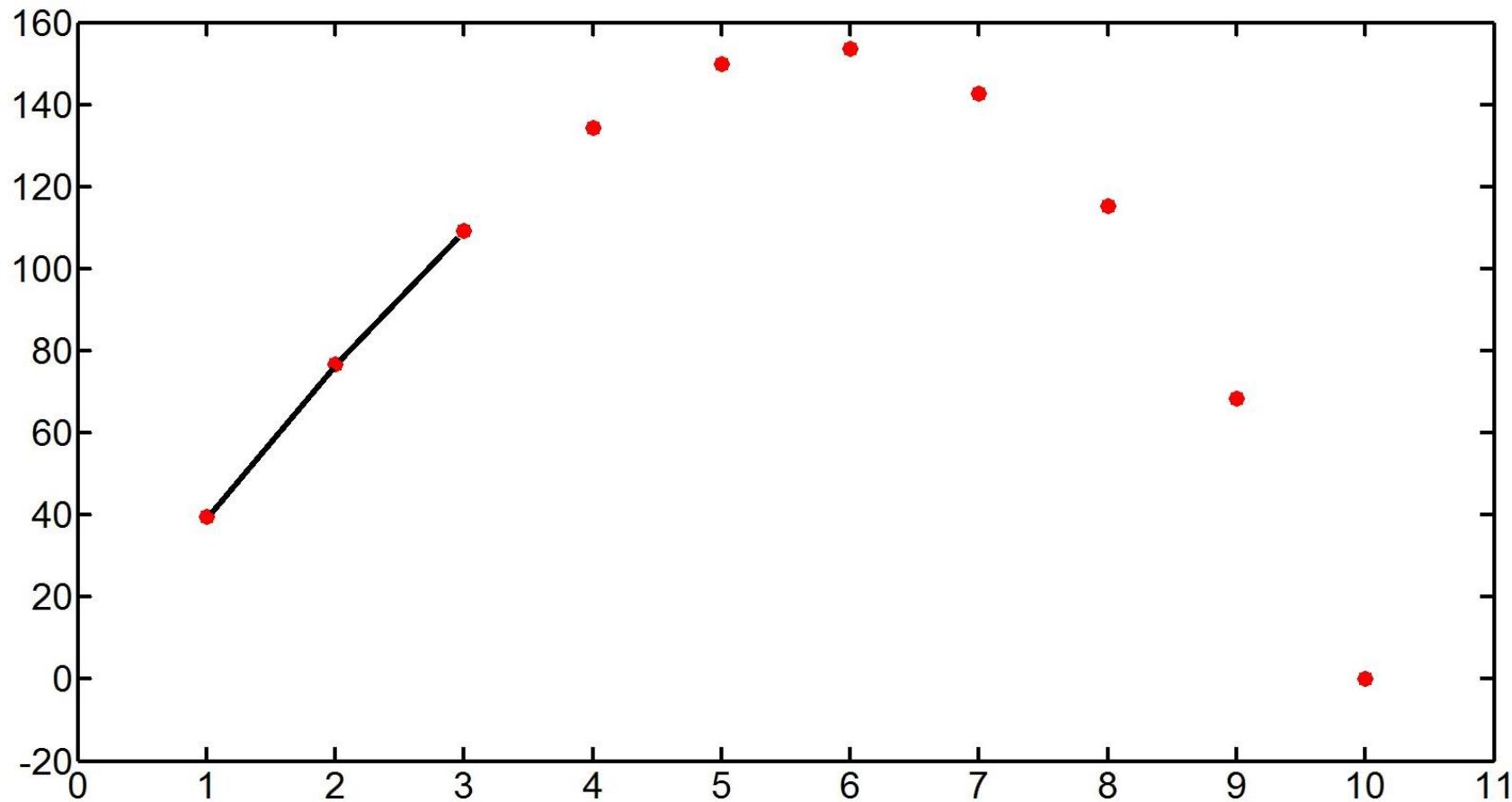
Spline  $S(x)$  is a train of linked parabolas of degree 2 or 3, whose derivative  $dS/dx$  is continuous everywhere, including points  $x_i$  (invented in 1946).

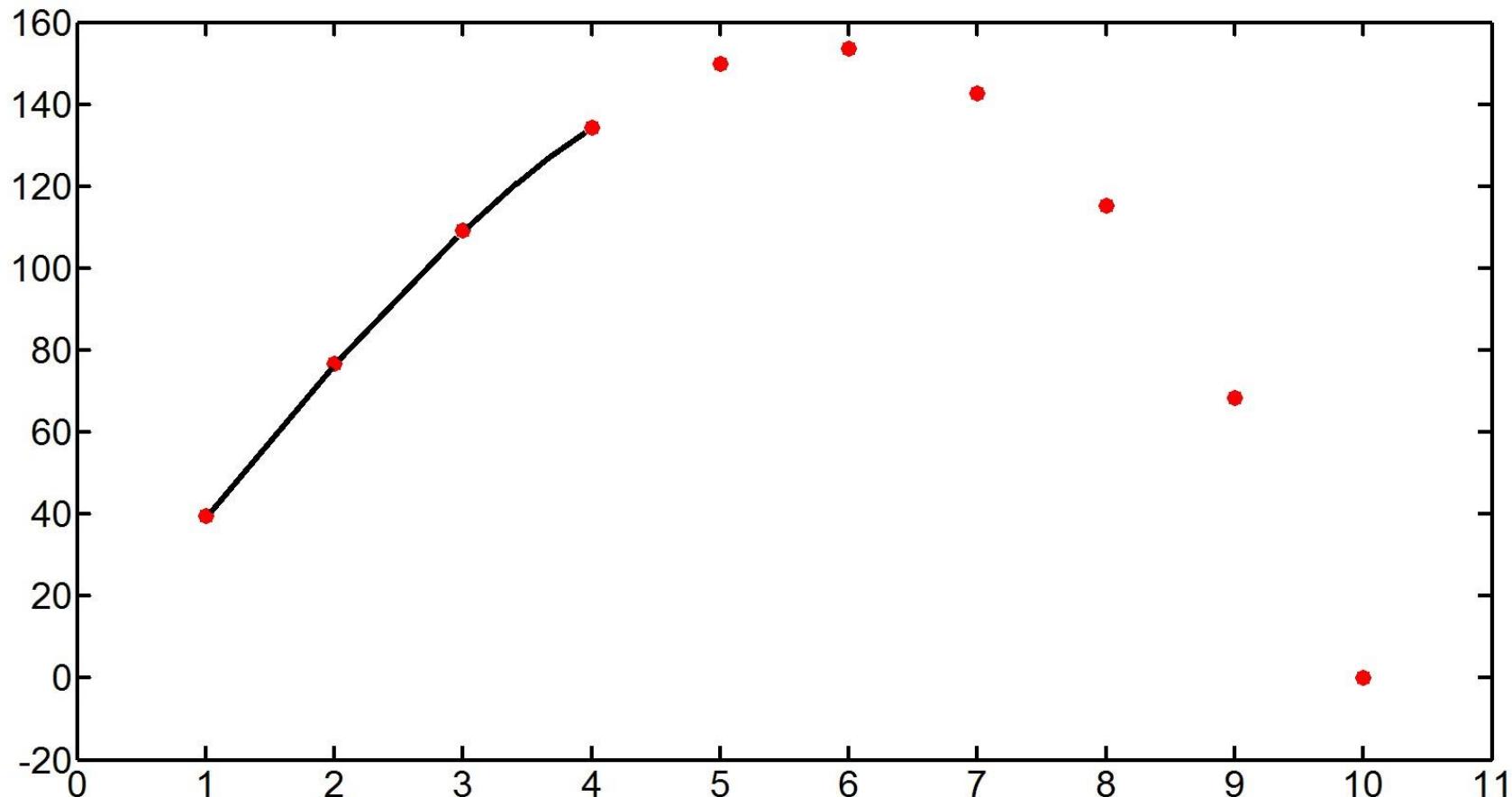


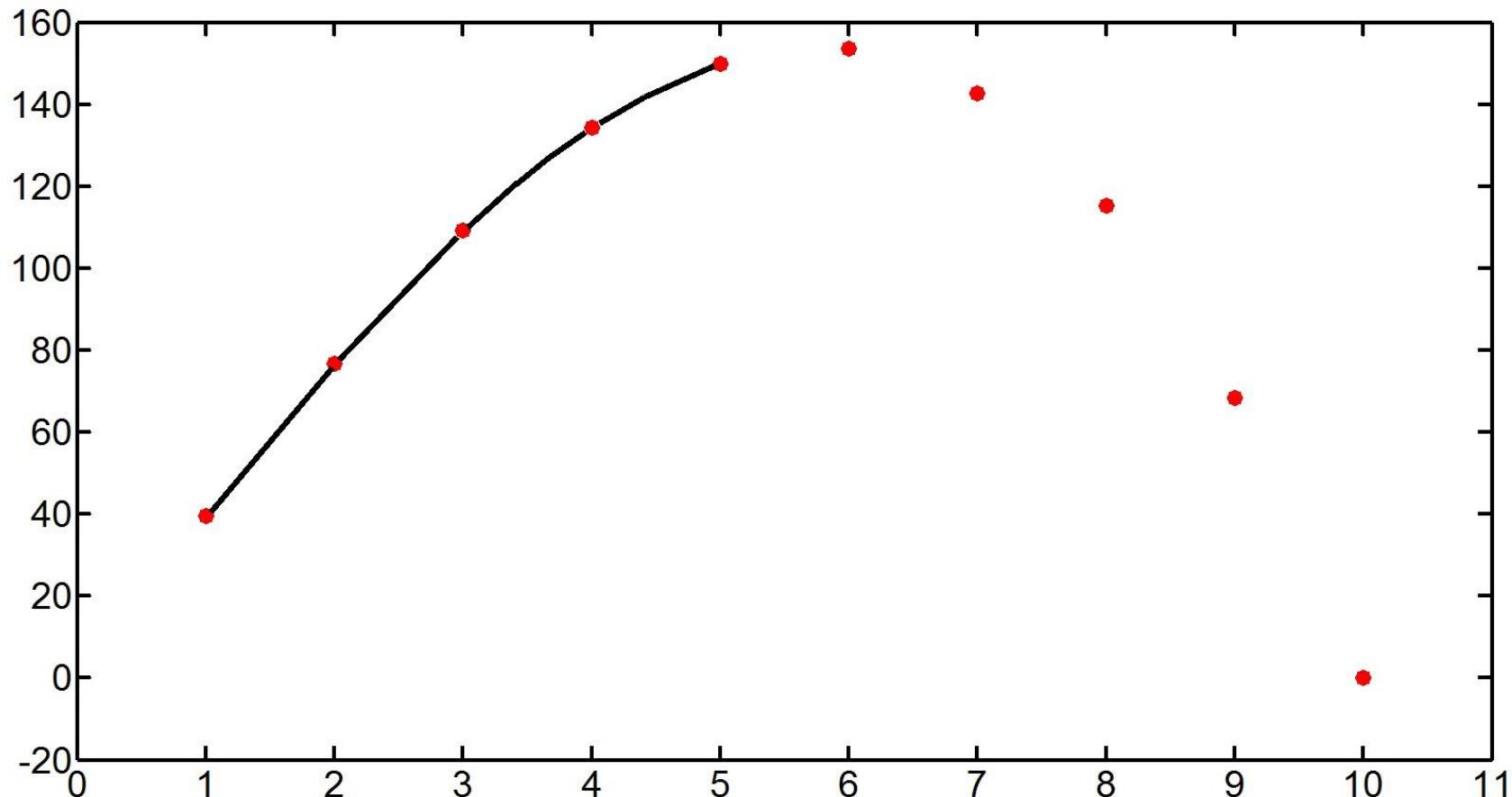
First, we consider quadratic splines

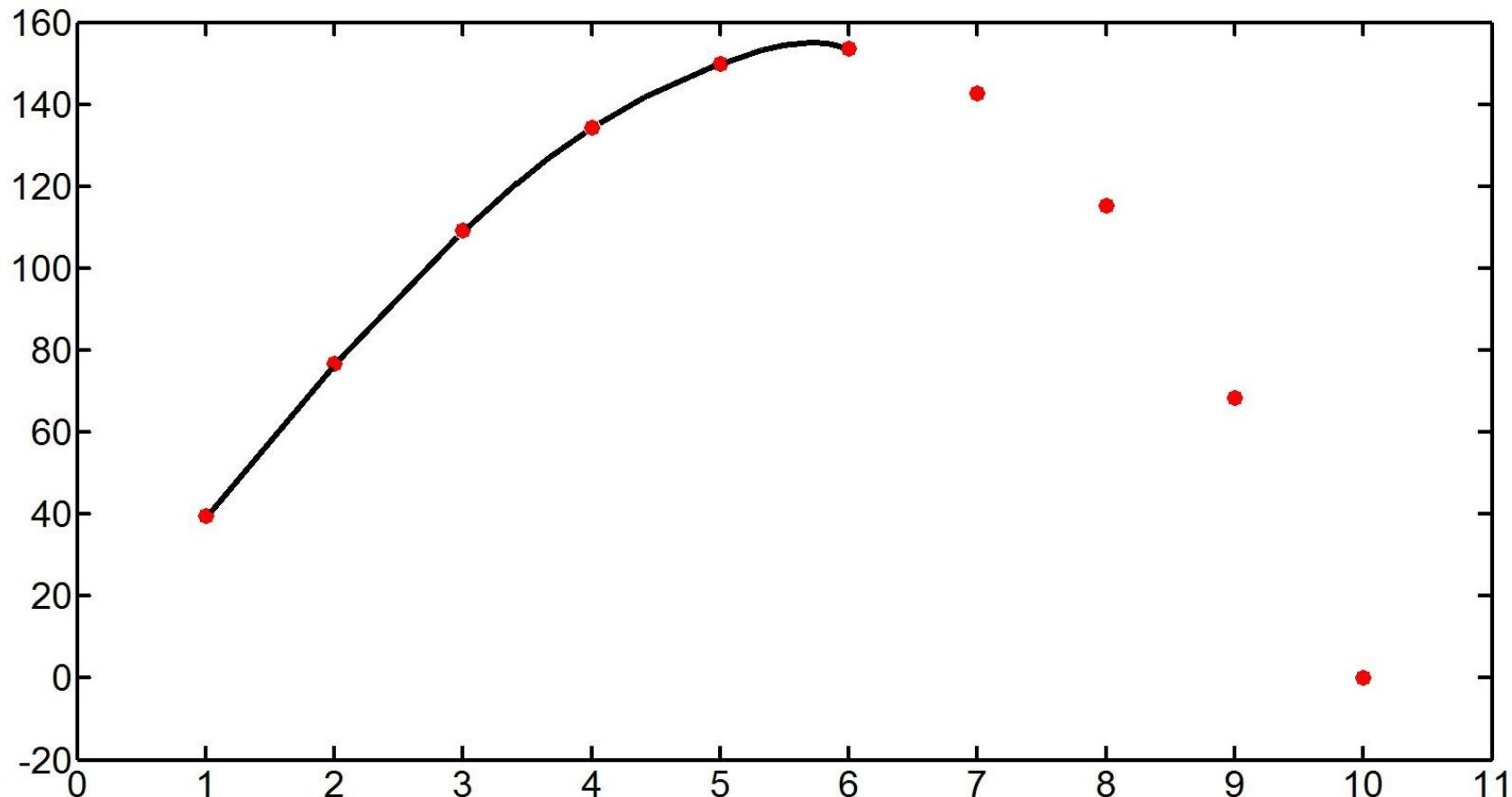
$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

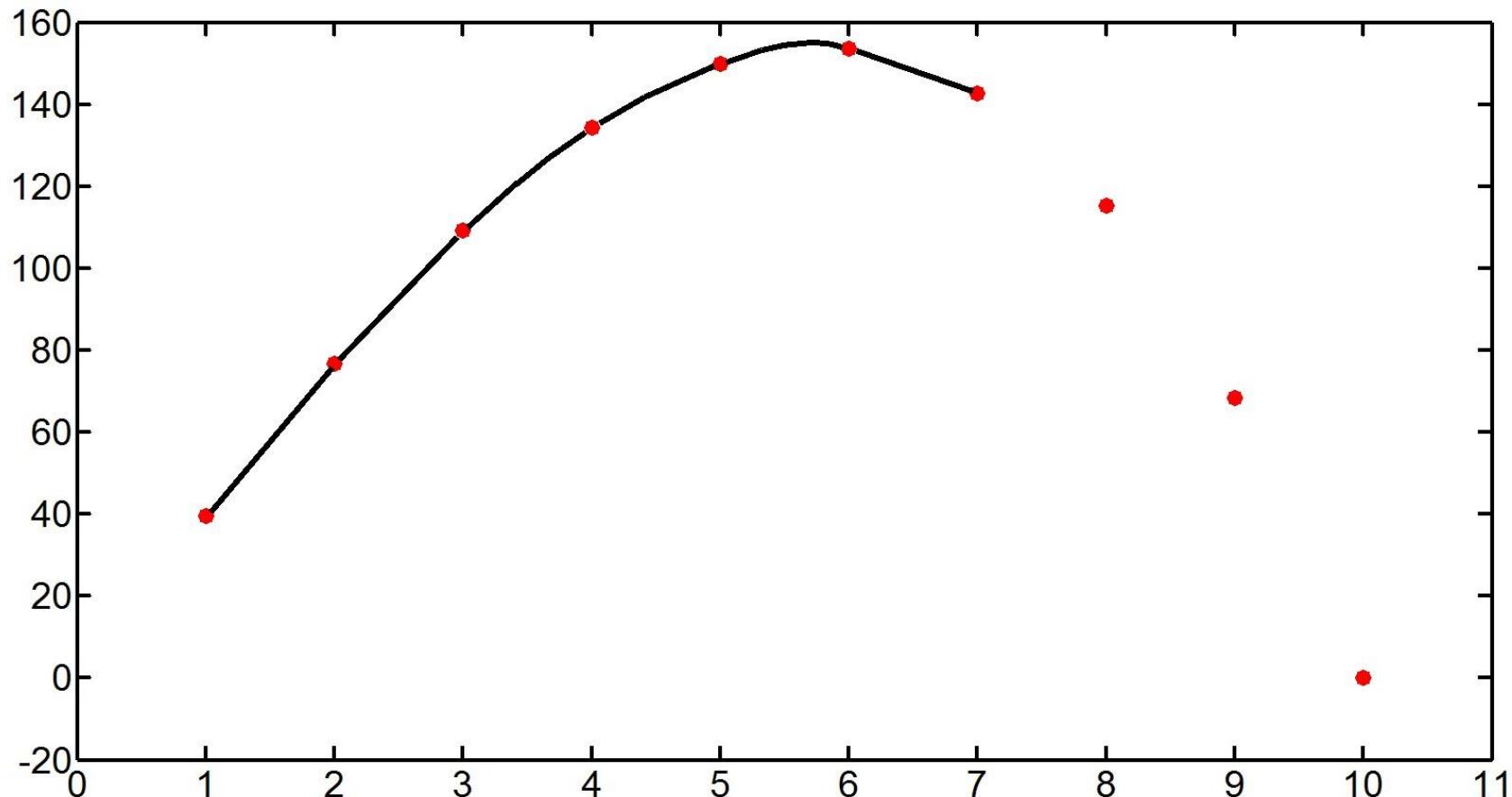
$$\text{at } x_i \leq x \leq x_{i+1} , \quad i=0, 1, \dots, n-1$$

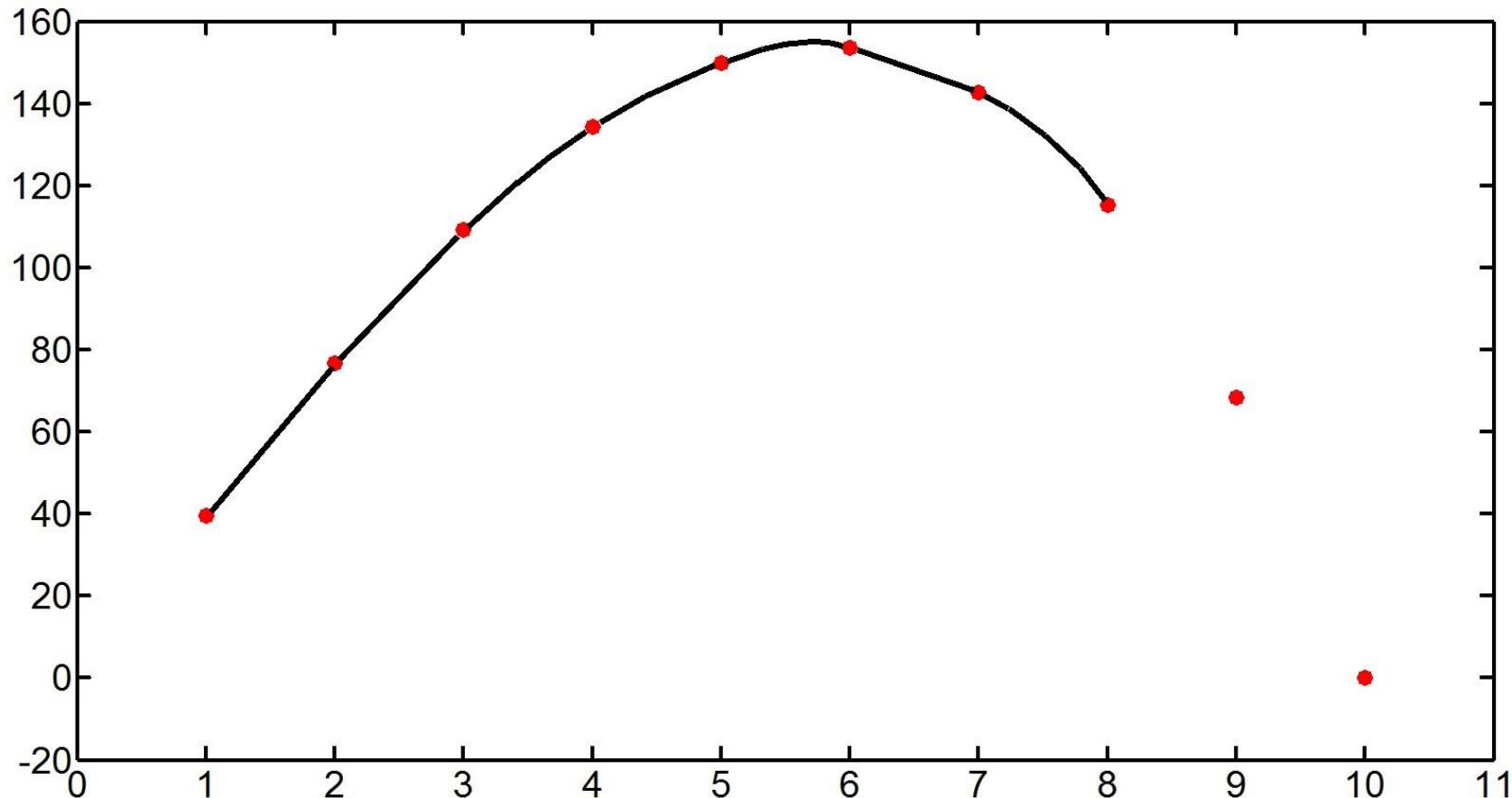


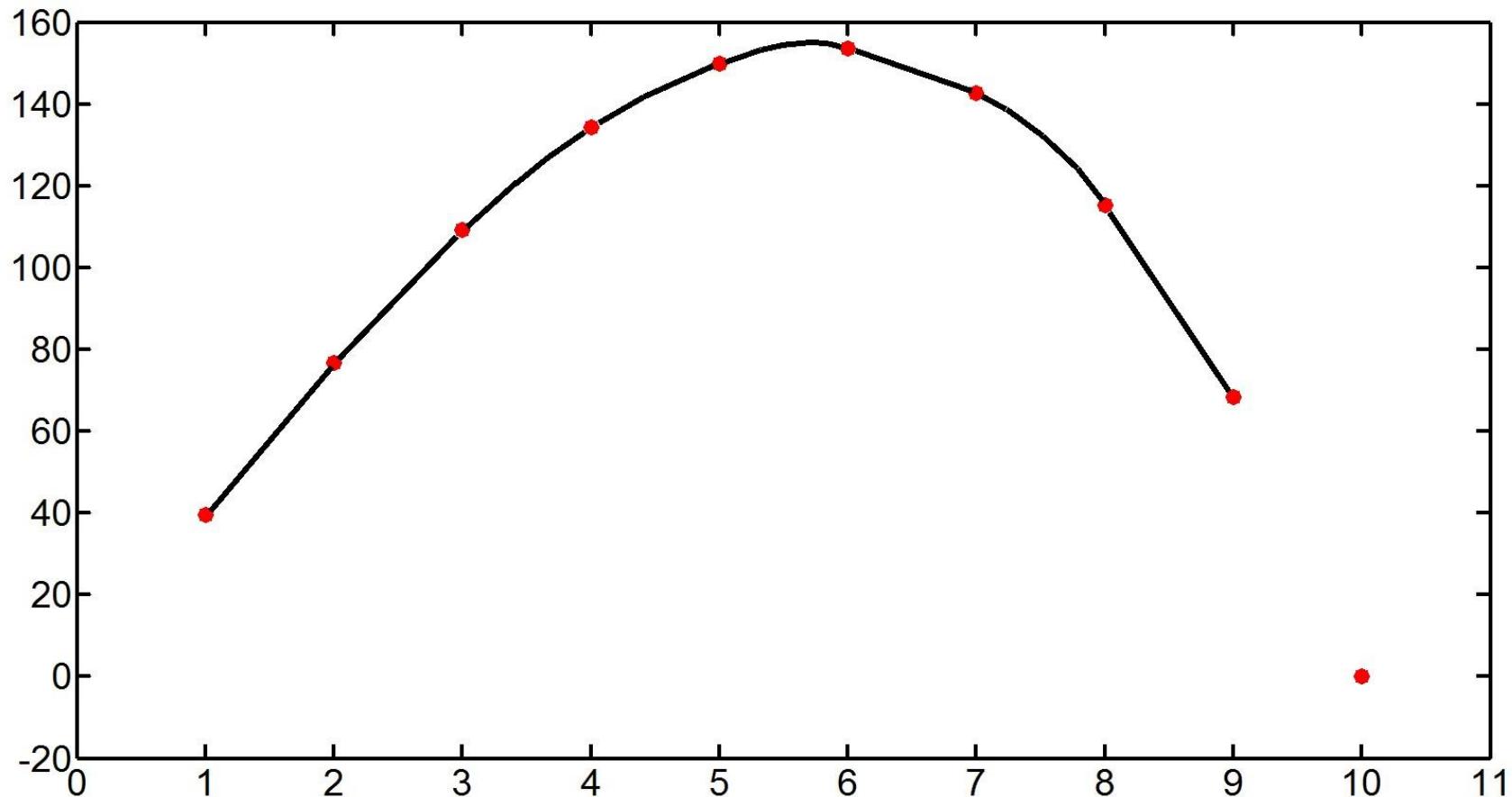


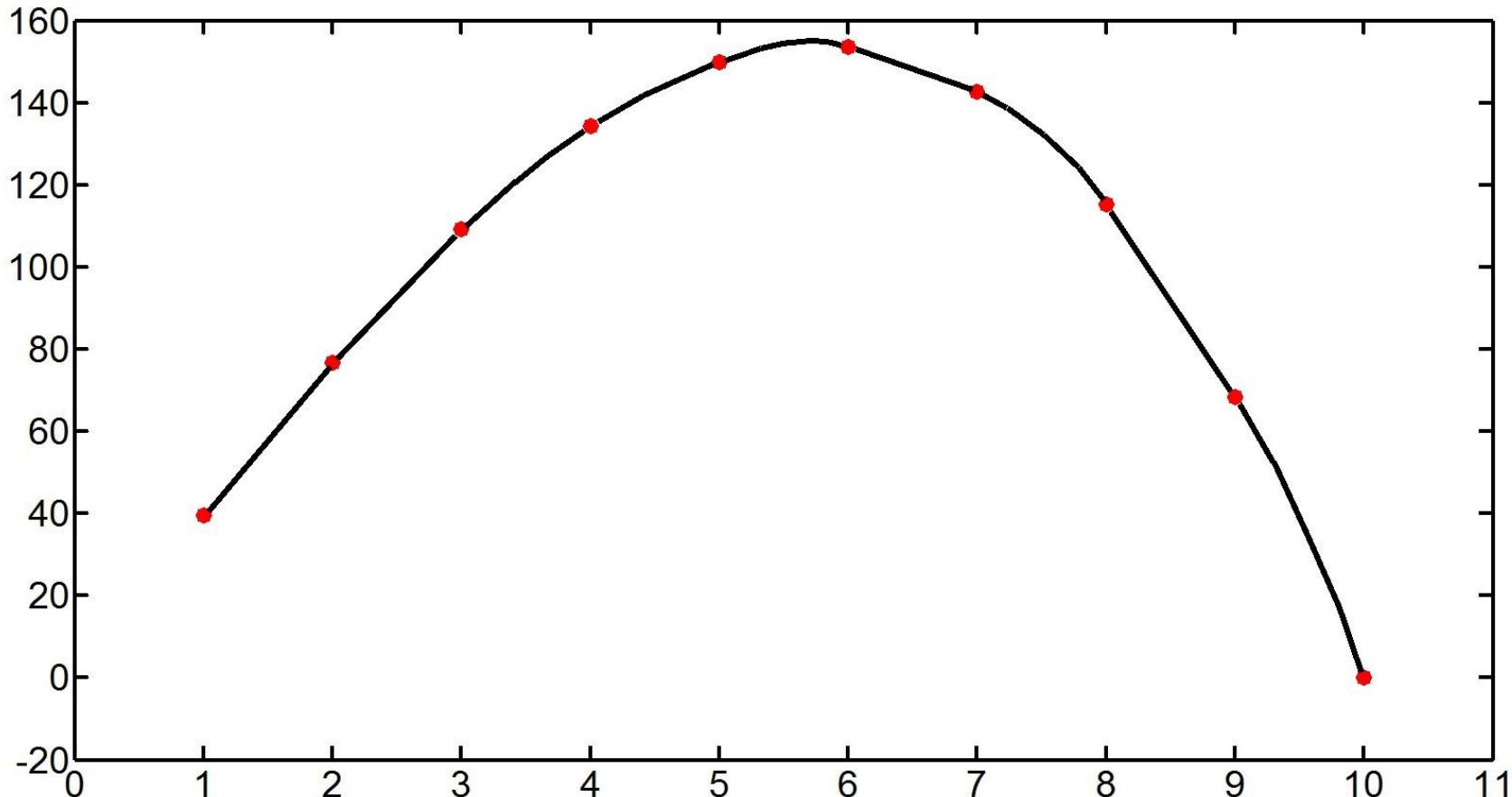












$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

at  $x_i \leq x \leq x_{i+1}$  ,  $i=0, 1, \dots, n-1$

$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

at  $x_i \leq x \leq x_{i+1}$  ,  $i=0, 1, \dots, n-1$

Let us find coefficients  $a_i, b_i, c_i$  using 3 conditions:

1) The condition that  $S(x)$  equals to given  $y_i$  at the left endpoint  $x_i$  :

$$S(x_i) = y_i \quad \Rightarrow \quad a_i = y_i \quad i=0, 1, \dots, n-1$$

$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

at  $x_i \leq x \leq x_{i+1}$  ,  $i=0, 1, \dots, n-1$

Let us find coefficients  $a_i, b_i, c_i$  using 3 conditions:

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$$S(x_i) = y_i \Rightarrow a_i = y_i \quad i=0, 1, \dots, n-1$$

2) The condition that  $S(x)$  equals to given value  $y_{i+1}$  at  $x_{i+1}$  :

$$S(x_{i+1}) = y_{i+1} \Rightarrow$$

$$\Rightarrow a_i + b_i \cdot (x_{i+1} - x_i) + c_i \cdot (x_{i+1} - x_i)^2 = y_{i+1}$$

$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

at  $x_i \leq x \leq x_{i+1}$ ,  $i=0, 1, \dots, n-1$

Let us find coefficients  $a_i, b_i, c_i$  using 3 conditions:

1) The condition that  $S(x)$  equals to given  $y_i$  at the left endpoint  $x_i$ :

$$S(x_i) = y_i \Rightarrow a_i = y_i \quad i=0, 1, \dots, n-1$$

2) The condition that  $S(x)$  equals to given value  $y_{i+1}$  at  $x_{i+1}$ :

$$S(x_{i+1}) = y_{i+1} \Rightarrow$$

$$\Rightarrow a_i + b_i \cdot (x_{i+1} - x_i) + c_i \cdot (x_{i+1} - x_i)^2 = y_{i+1}$$

3) The condition that  $dS/dx$  is continuous at node  $x_{i+1}$ :

$$dS/dx = b_i + 2 c_i \cdot (x - x_i) \quad \text{at } x_i \leq x \leq x_{i+1}$$

$$b_i + 2 c_i \cdot (x_{i+1} - x_i) = b_{i+1} + 2 c_{i+1} \cdot (x_{i+1} - x_{i+1})$$

$$\begin{cases} b_i + 2 c_i \cdot (x_{i+1} - x_i) = b_{i+1} & (1) \quad i=0, 1, \dots, n-2 \\ b_i \cdot (x_{i+1} - x_i) + c_i \cdot (x_{i+1} - x_i)^2 = y_{i+1} - y_i & (2) \end{cases}$$

see condition 2) above  $i=0, 1, \dots, n-1$

We get  $2n-1$  equations for  $2n$  coefficients  $b_i, c_i$

+ additional condition  $c_0 = 0$

$$( S(x) = a_0 + b_0 \cdot (x - x_0) \quad \text{at } x_0 \leq x \leq x_1 )$$

In fact, coefficients  $b_i, c_i$  can be calculated sequentially for  $i=1,2,\dots,n$  :

$\textcolor{red}{b}_1, \textcolor{red}{c}_1,$

*then  $\textcolor{red}{b}_2$  from (1),*

*then  $\textcolor{red}{c}_2$  from (2),*

*then  $\textcolor{red}{b}_3$  from (1), . . .*

## Theorem on the error of interpolation:

If  $f'''(x)$  is continuous on  $[x_0, x_n]$ , then

$$|f(x)-S(x)| \leq \max |f'''(x)| h^3/12,$$

where  $h=\max |x_{i+1}-x_i|$ .

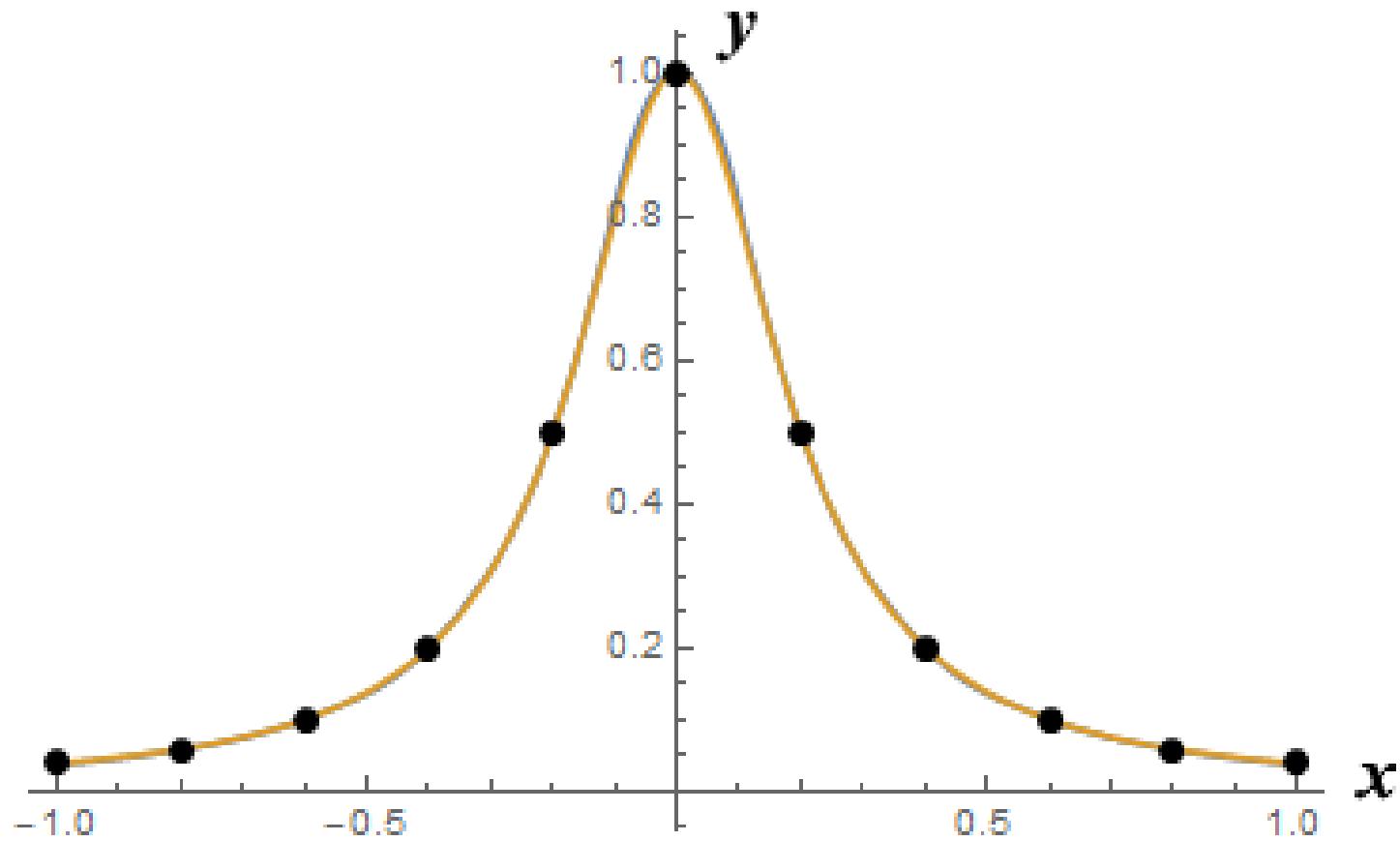
Proof see in:

“Optimal Error Bounds for Quadratic Spline Interpolation”

Francois Dubeau

JOURNAL OF MATHEMATICAL ANALYSIS AND APPLICATIONS,

Vol. 198, pages 49-63, 1996



Example: Interpolation of the function  $f(x) = 1/(1+25x^2)$  by a quadratic spline looks perfect (the curves  $S(x)$  and  $f(x)$  coincide)  
<https://engcourses-uofa.ca/books/numericalanalysis>  
in contrast to interpolation by a Lagrangian's polynomial.

We notice that the second-order derivative of

$$S(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2$$

is  $d^2 S/dx^2 = 2 c_i$

Therefore, it is constant at each interval

$$x_i \leq x \leq x_{i+1}$$

and usually jumps when  $x$  moves from  $x_i \leq x \leq x_{i+1}$  to next segment, since  $c_i$  are usually different in different segments.

## Using cubic splines

$$S_{cub}(x) = a_i + b_i \cdot (x - x_i) + c_i \cdot (x - x_i)^2 + d_i \cdot (x - x_i)^3$$

$$x_i \leq x \leq x_{i+1} , \quad i=0, 1, ..., n-1,$$

with coefficient  $d_i$ , we can provide the continuity of  $d^2 S_{cub}/dx^2$  at the nodes  $x_i$ .

Indeed, the derivative

$$d^2 S_{cub}/dx^2 = 2 c_i + 6 d_i \cdot (x - x_i)$$

changes in  $x_i \leq x \leq x_{i+1}$

and therefore its values at the ends of interval can be adjusted in a proper way.

We set 4 conditions:

1)  $S_{cub}(x)$  is equal to  $y_i$  at left end of segment

$$x_i \leq x \leq x_{i+1} : \quad y_i = a_i$$

2)  $S_{cub}(x)$  is equal to  $y_{i+1}$  at right end of the segment

$$y_{i+1} = a_i + b_i \cdot (x_{i+1} - x_i) + c_i \cdot (x_{i+1} - x_i)^2 + d_i \cdot (x_{i+1} - x_i)^3$$

3) First-order derivative:

$$dS_{cub}/dx = b_i + 2c_i \cdot (x - x_i) + 3d_i \cdot (x - x_i)^2$$

the condition of equal derivatives at the right endpoint:

$$b_i + 2c_i \cdot (x_{i+1} - x_i) + 3d_i \cdot (x_{i+1} - x_i)^2 = b_{i+1}$$

$$i=1, 2, \dots, n-1$$

#### 4) Second-order derivative :

$$d^2S_{cub}/dx^2 = 2c_i + 6d_i \cdot (x-x_i)$$

the condition of equal second derivatives at the right endpoint:

$$2c_i + 6d_i \cdot (x_{i+1}-x_i) = 2c_{i+1} \quad i=1,2,\dots,n-1$$

We have got  $2n+2(n-1)$  equations with respect to  $4n$  coefficients of spline  $S_{cub}$ .

We add two conditions  $d^2S_{cub}/dx^2 = 0$  at endpoints  $x_0, x_n$ :

$$c_0 = 0, \quad 2c_{n-1} + 6d_{n-1} \cdot (x_n - x_{n-1}) = 0$$

Finally, we have  $4n$  equations with respect to  $4n$  coefficients.

```
x0= 1
x1= 2
x2= 3
x3= 5
y0= 2
y1= 2.9
y2= 4.2
y3= 6
plot(x0,y0,'o',x1,y1,'o',x2,y2,'o',x3,y3,'o')

d=splin( [x0 x1 x2 x3], [y0 y1 y2 y3] )
x= 1: 0.02 : 5
y=interp(x, [x0 x1 x2 x3], [y0 y1 y2 y3], d)
plot(x,y,'b','LineWidth',3)
```

## **Interpolation of functions of two independent variables**

### **[DOUBLE INTERPOLATION]**

In the preceding sections we have derived interpolation formulae to approximate a function of a single variable. For a function of two or more variables, the formulae become complicated but a simpler procedure is to interpolate with respect to the first variable keeping the others constant, then interpolate with respect to the second variable, and so on. The method is illustrated below for a function of two variables.

The following table gives the values of  $z$  for different values of  $x$  and  $y$ . Find  $z$  when  $x = 2.5$  and  $y = 1.5$ .

$y$	$x$				
	0	1	2	3	4
0	0	1	4	9	16
1	2	3	6	11	18
2	6	7	10	15	22
3	12	13	16	21	28
4	18	19	22	27	34

We first interpolate with respect to  $x$  keeping  $y$  constant. For  $x = 2.5$ , we obtain the following table using *linear interpolation*.

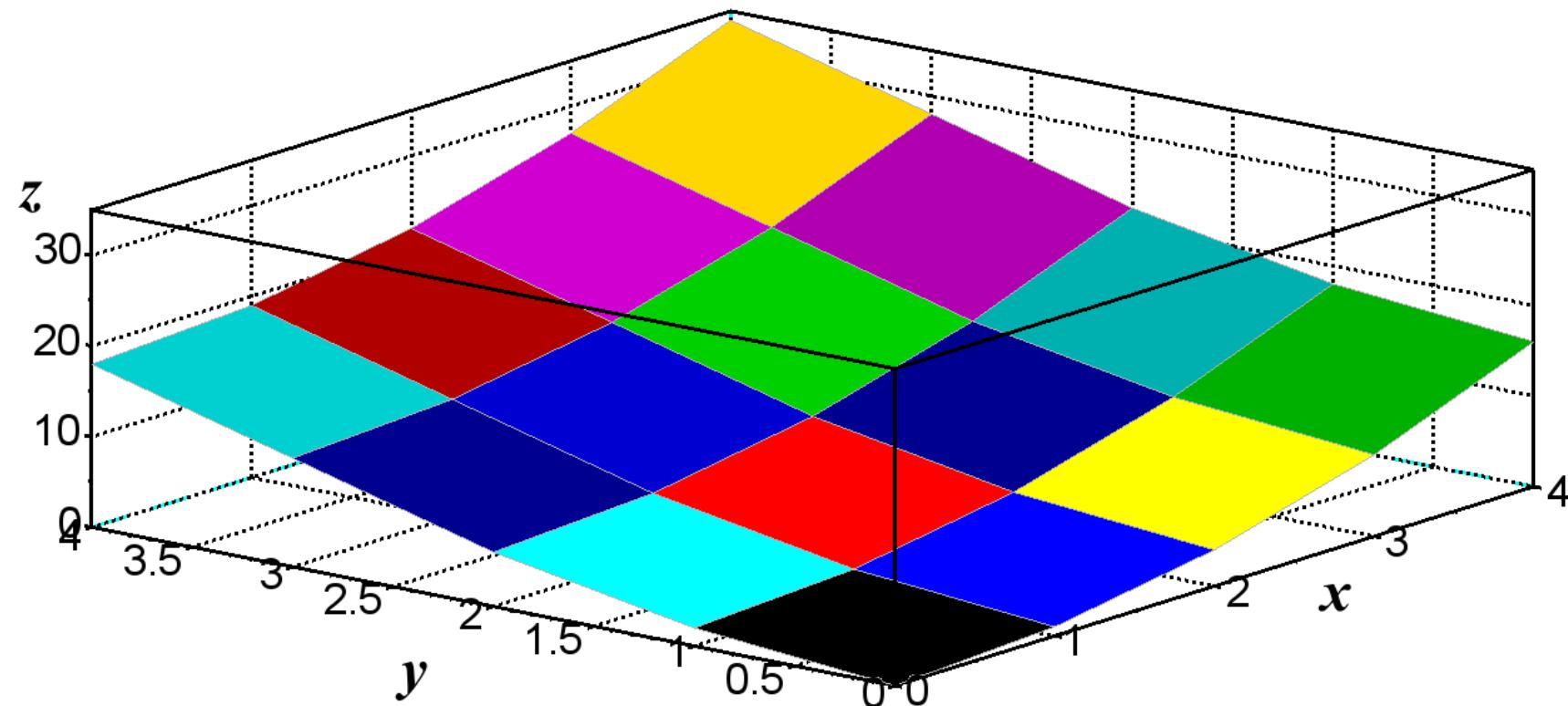
$y$	$z$
0	6.5
1	8.5
2	12.5
3	18.5
4	24.5

$y$	$z$
0	6.5
1	8.5
2	12.5
3	18.5
4	24.5

Now, we interpolate with respect to  $y$  using linear interpolation once again. For  $y = 1/5$ , we obtain

$$z = \frac{8.5 + 12.5}{2} = 10.5$$

so that  $z(2.5, 1.5) = 10.5$ . Actually, the tabulated function is  $z = x^2 + y^2 + y$  and hence  $z(2.5, 1.5) = 10.0$ , so that the computed value has an error of 5%.



**Scilab:**

**splin2d( )**  
**interp2d( )**