

# NUMERICAL METHODS

Week-7

25.03.2014

## Approximation by Spline functions & Smoothing of Data

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# Approximation by Spline functions

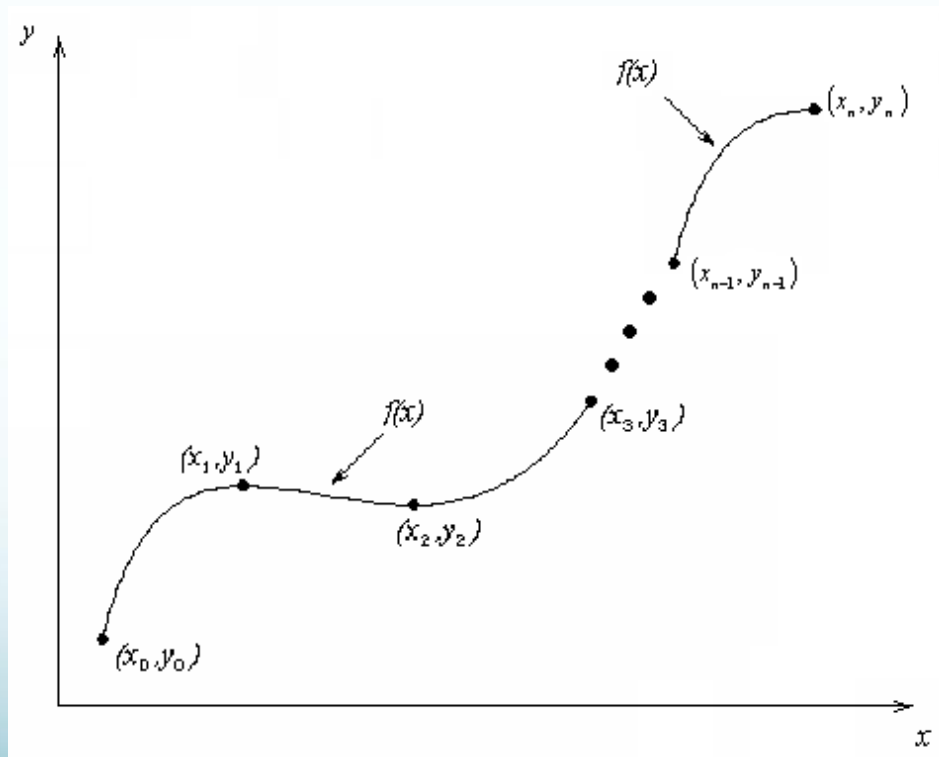
- **What** is a spline function?
- **Why** do we use approximation by splines?
- **How** do we solve spline equations?

# What is a Spline?

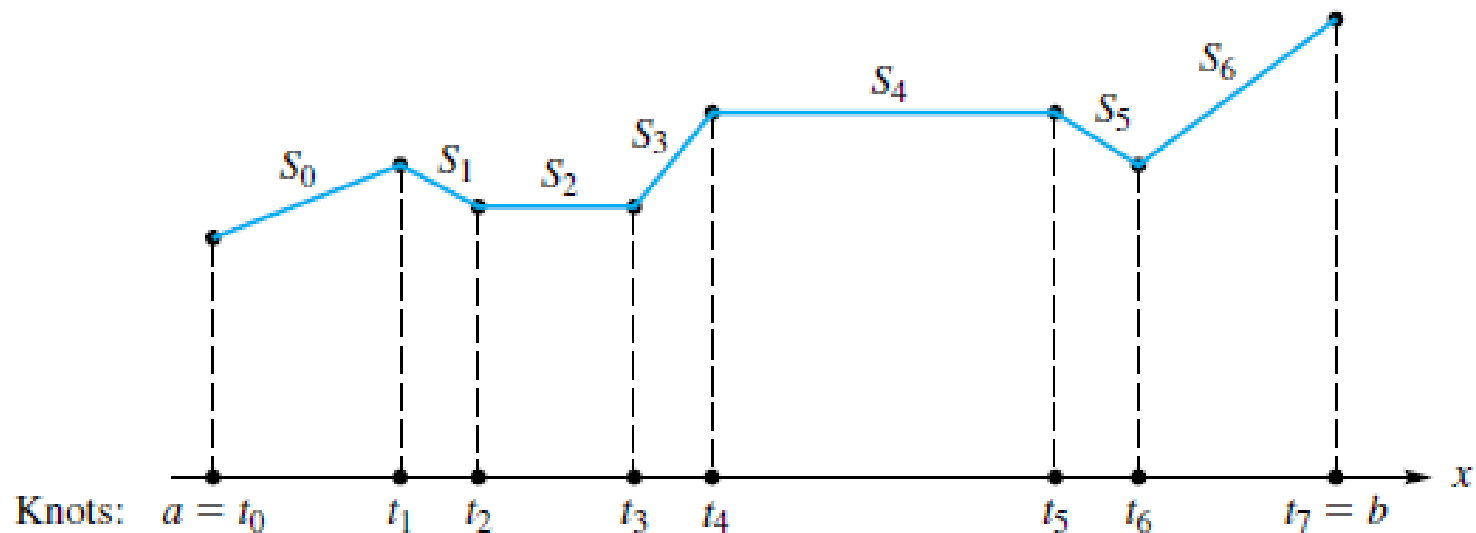
- A spline is a function that consisting of simple functions joined together.
- As with polynomial functions, splines are used to interpolate tabulated data as well as functions.
- A spline is different from a polynomial interpolation, which consists of a single well defined function that approximates a given shape; splines are normally **piecewise polynomial**.

# What is a spline function ?

Given  $(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)$ , find the value of 'y' at a value of 'x' that is not given.



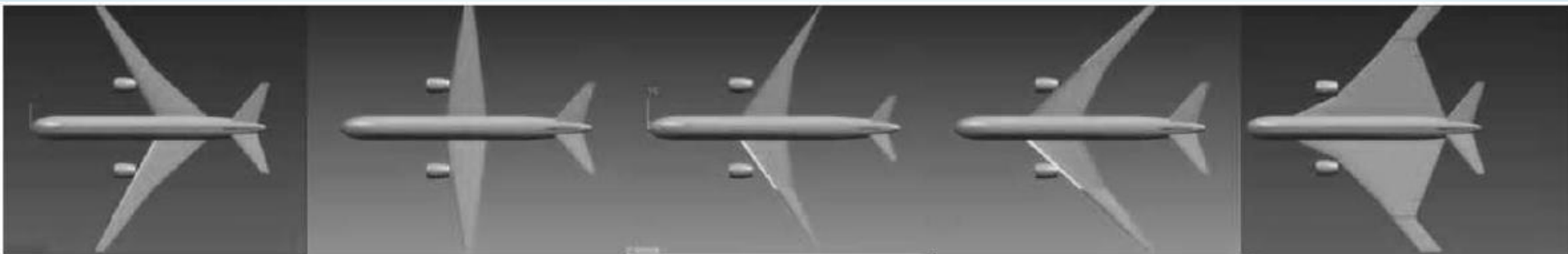
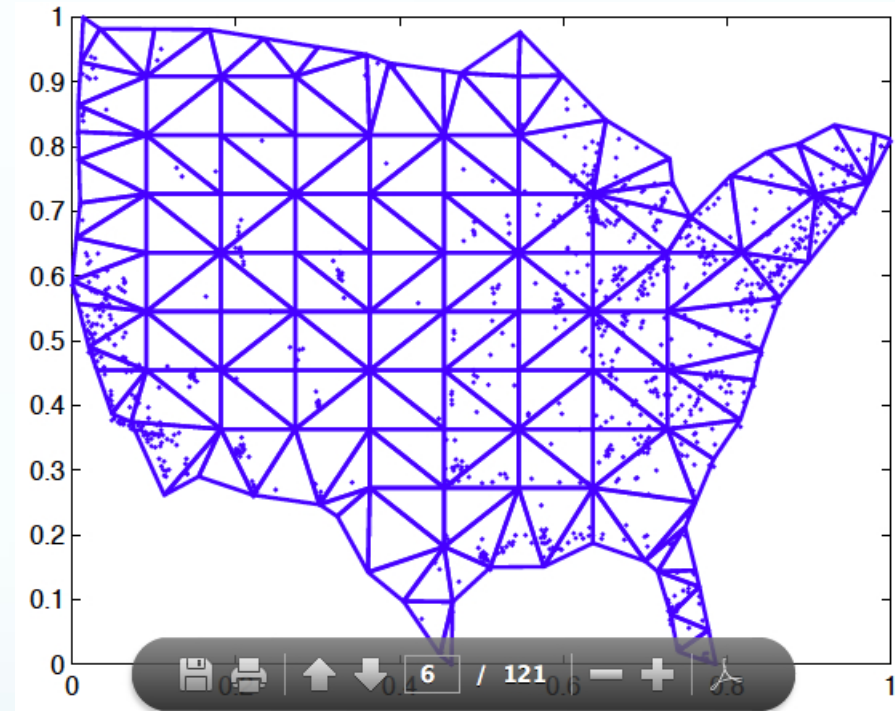
# What is a spline function ?



- A simple example is the **polygonal** function (or spline of degree 1), whose pieces are linear polynomials joined together to achieve **continuity**, as in figure. The points  $t_0, t_1, \dots, t_n$  at which the function changes its character are termed **knots** in the theory of splines.

# Some Applications

- Ship, aircraft building
- Detailed designs
- Computer Graphics
- Routing
- Movement Estimation



# Why Splines ?

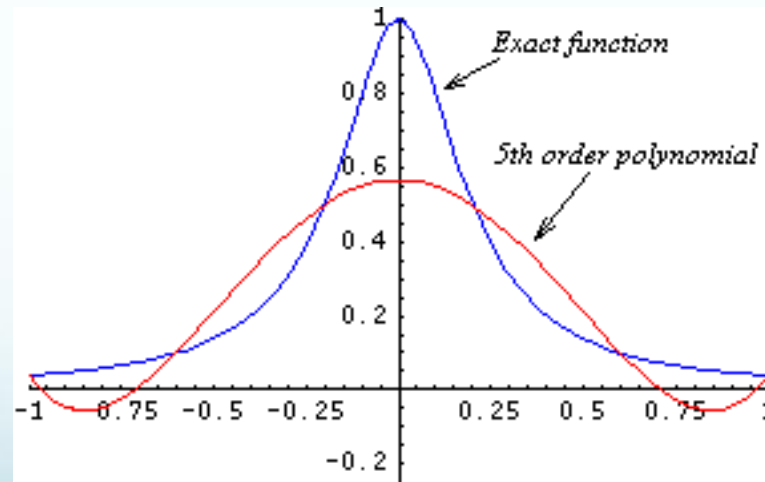
- Splines are used to approximate complex functions and shapes.
- Drawbacks of higher order polynomials in interpolating functions.
- Splines are normally piecewise polynomials so provides better approximation than polynomial interpolations.

# Why Splines ?

$$f(x) = \frac{1}{1 + 25x^2}$$

**Table : Six equidistantly spaced points in [-1, 1]**

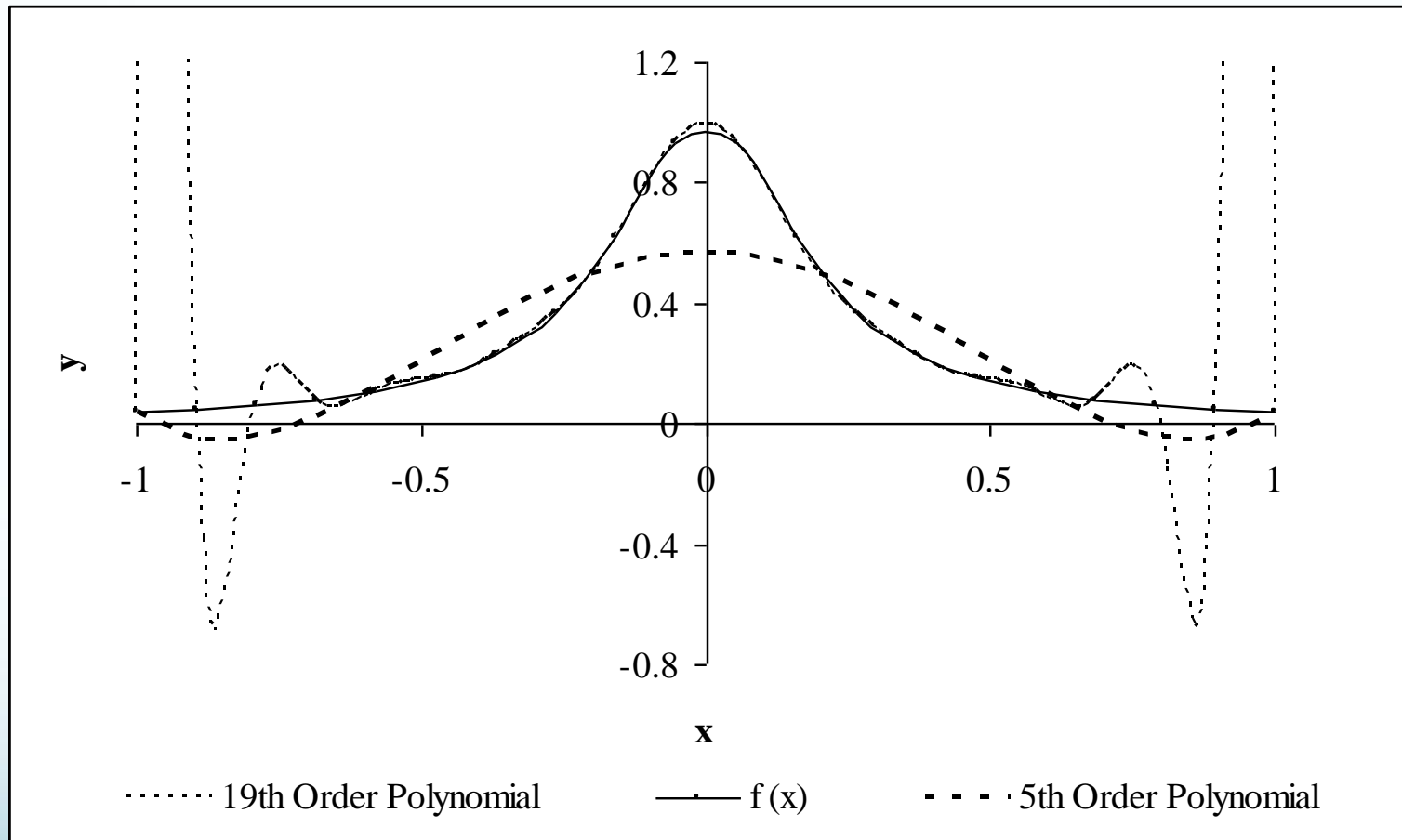
$x$	$y = \frac{1}{1 + 25x^2}$
-1.0	0.038461
-0.6	0.1
-0.2	0.5
0.2	0.5
0.6	0.1
1.0	0.038461



**Figure : 5<sup>th</sup> order polynomial vs. exact function**



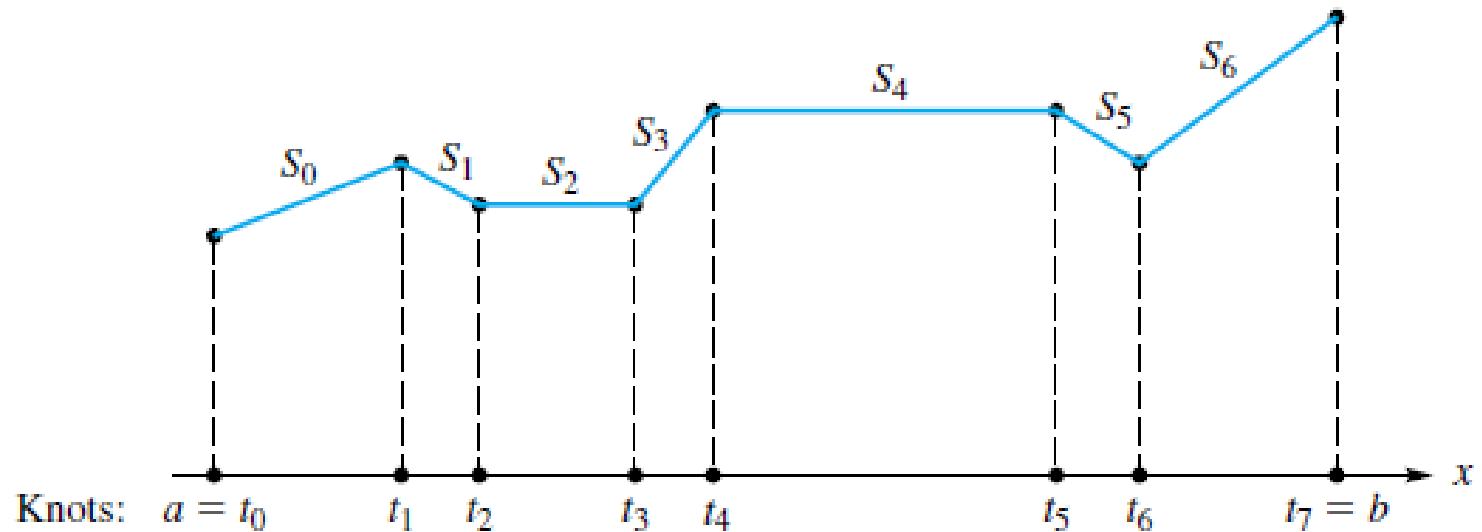
# Why Splines ?



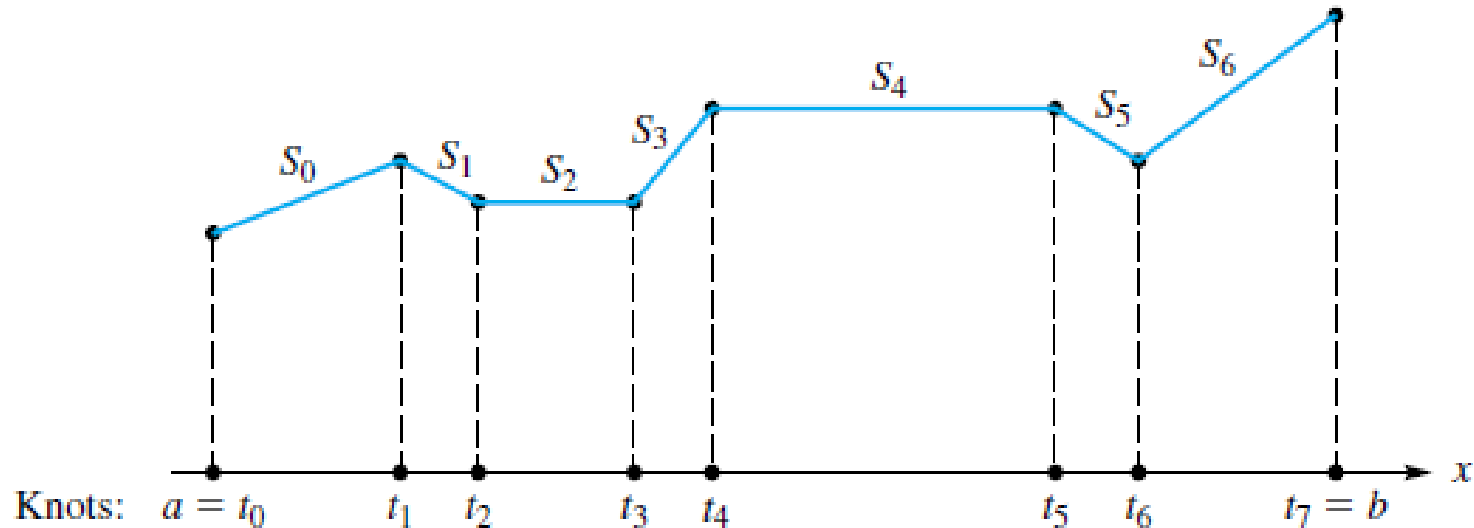
**Figure : Higher order polynomial interpolation is a bad idea**

# First Degree Splines

- Splines make use of partitions, which are a way of cutting an interval into a number of subintervals.
- The spline functions of degree 1 can be used for interpolation



# First Degree Splines



A function **S** is called a **spline of degree 1** if:

1. The domain of **S** is an interval  $[a, b]$ .
2. **S** is **continuous** on  $[a, b]$ .
3. There is a partitioning of the interval  $a = t_0 < t_1 < \dots < t_n = b$  such that **S** is a linear polynomial on each subinterval  $[t_i, t_{i+1}]$ .

# First Degree Splines

$$f(x) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x - x_0), \quad x_0 \leq x \leq x_1$$

$$= f(x_1) + \frac{f(x_2) - f(x_1)}{x_2 - x_1}(x - x_1), \quad x_1 \leq x \leq x_2$$

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$$= f(x_{n-1}) + \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}(x - x_{n-1}), \quad x_{n-1} \leq x \leq x_n$$

Note the terms of

$$\frac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}}$$

in the above function are simply slopes between  $x_{i-1}$  and  $x_i$ .



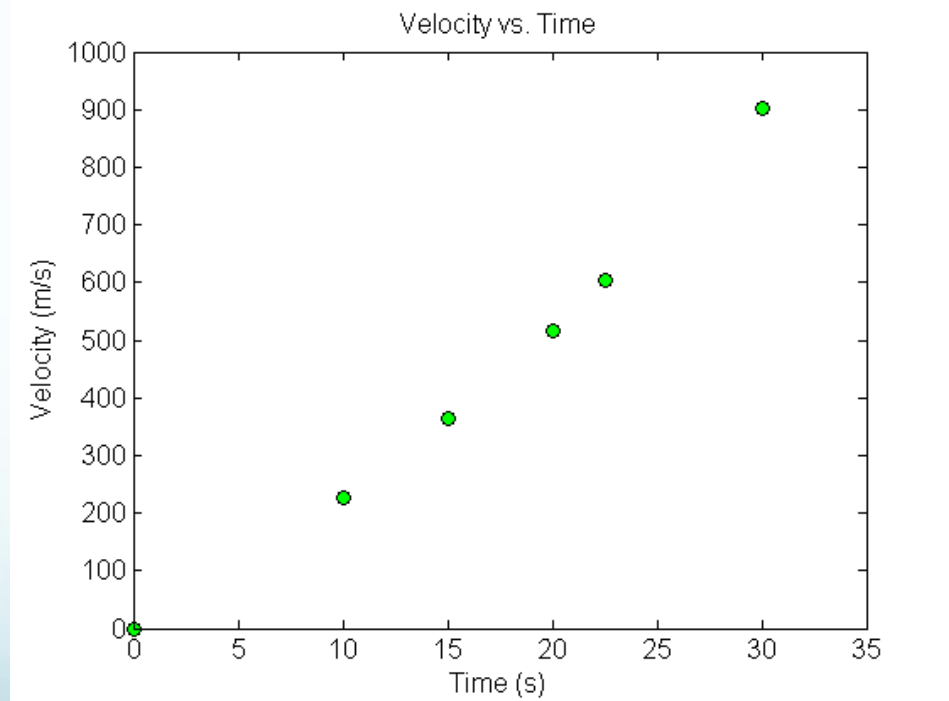
## Example

The upward velocity of a rocket is given as a function of time in table.

Find the velocity at  $t=16$  seconds using linear splines.

**Table :** Velocity as a function of time

$t$ (s)	$v(t)$ (m/s)
0	0
10	227.04
15	362.78
20	517.35
22.5	602.97
30	901.67



**Figure :** Velocity vs. time data for the rocket example

# Linear Splines

$$t_0 = 15, \quad v(t_0) = 362.78$$

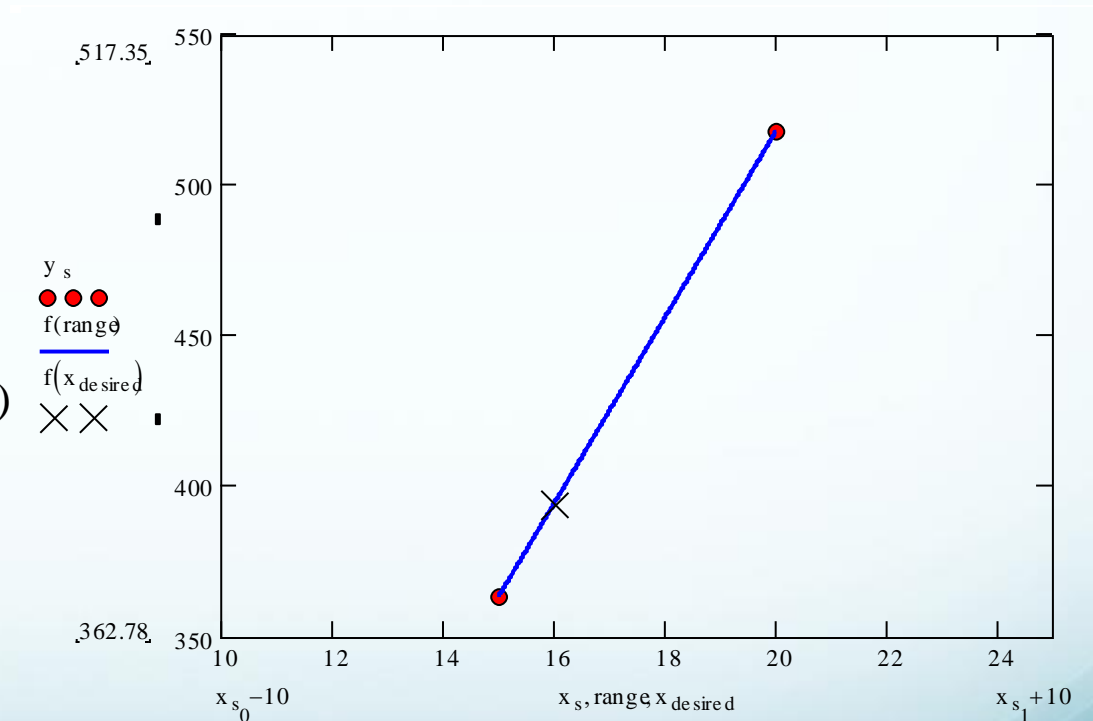
$$t_1 = 20, \quad v(t_1) = 517.35$$

$$\begin{aligned} v(t) &= v(t_0) + \frac{v(t_1) - v(t_0)}{t_1 - t_0} (t - t_0) \\ &= 362.78 + \frac{517.35 - 362.78}{20 - 15} (t - 15) \end{aligned}$$

$$v(t) = 362.78 + 30.913(t - 15)$$

At  $t = 16$ ,

$$\begin{aligned} v(16) &= 362.78 + 30.913(16 - 15) \\ &= 393.7 \text{ m/s} \end{aligned}$$



# Linear Spline Algorithm

```
real function Spline1( $n, (t_i), (y_i), x$ )  
integer  $i, n$ ; real  $x$ ; real array  $(t_i)_{0:n}, (y_i)_{0:n}$   
for  $i = n - 1$  to 0 step  $-1$  do  
    if  $x - t_i \geq 0$  then exit loop  
end for  
 $Spline1 \leftarrow y_i + (x - t_i)[(y_{i+1} - y_i)/(t_{i+1} - t_i)]$   
end function Spline1
```

# Quadratic Splines

A function  $Q$  is a **second-degree spline** if it has the following properties

A function  $Q$  is called a **spline of degree 2** if:

1. The domain of  $Q$  is an interval  $[a, b]$ .
2.  $Q$  and  $Q'$  are continuous on  $[a, b]$ .
3. There are points  $t_i$  (called **knots**) such that  $a = t_0 < t_1 < \cdots < t_n = b$  and  $Q$  is a polynomial of degree at most 2 on each subinterval  $[t_i, t_{i+1}]$ .



# Quadratic Splines

Given  $(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1}), (x_n, y_n)$ , fit quadratic splines through the data. The splines are given by

$$f(x) = a_1x^2 + b_1x + c_1, \quad x_0 \leq x \leq x_1$$

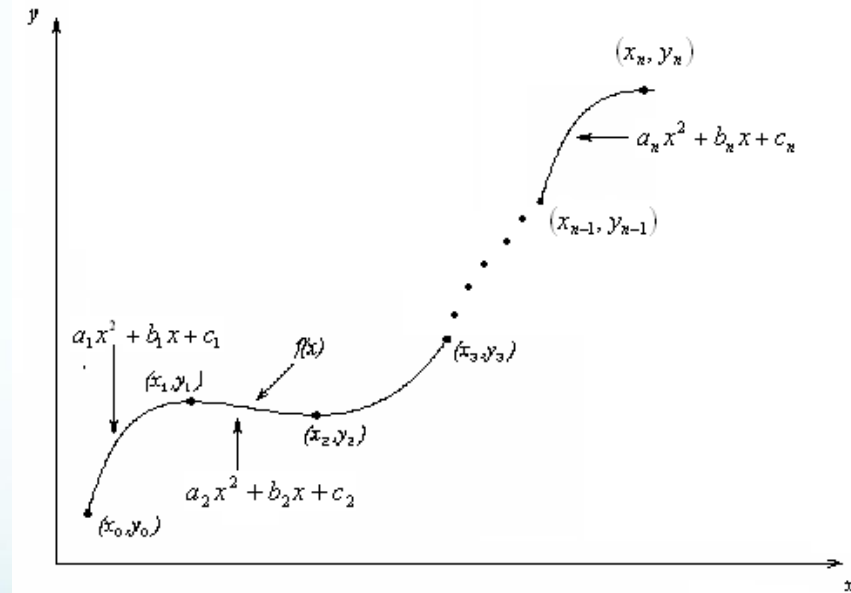
$$= a_2x^2 + b_2x + c_2, \quad x_1 \leq x \leq x_2$$

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$$= a_nx^2 + b_nx + c_n, \quad x_{n-1} \leq x \leq x_n$$



Find  $a_i, b_i, c_i, i = 1, 2, \dots, n$

# Quadratic Splines

Each quadratic spline goes through two consecutive data points

$$a_1 x_0^2 + b_1 x_0 + c_1 = f(x_0)$$

$$a_1 x_1^2 + b_1 x_1 + c_1 = f(x_1) \quad .$$

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$$a_i x_{i-1}^2 + b_i x_{i-1} + c_i = f(x_{i-1})$$

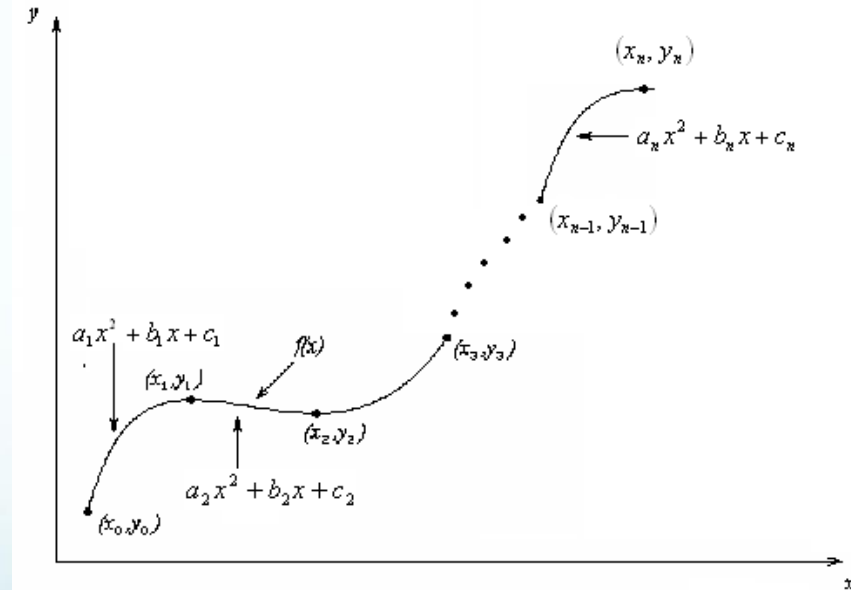
$$a_i x_i^2 + b_i x_i + c_i = f(x_i) \quad .$$

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$$a_n x_{n-1}^2 + b_n x_{n-1} + c_n = f(x_{n-1})$$

$$a_n x_n^2 + b_n x_n + c_n = f(x_n)$$



This condition gives  $2n$  equations

# Quadratic Splines

The first derivatives of two quadratic splines are continuous at the interior points.

For example, the derivative of the first spline

$$a_1x^2 + b_1x + c_1 \quad \text{is} \quad 2a_1x + b_1$$

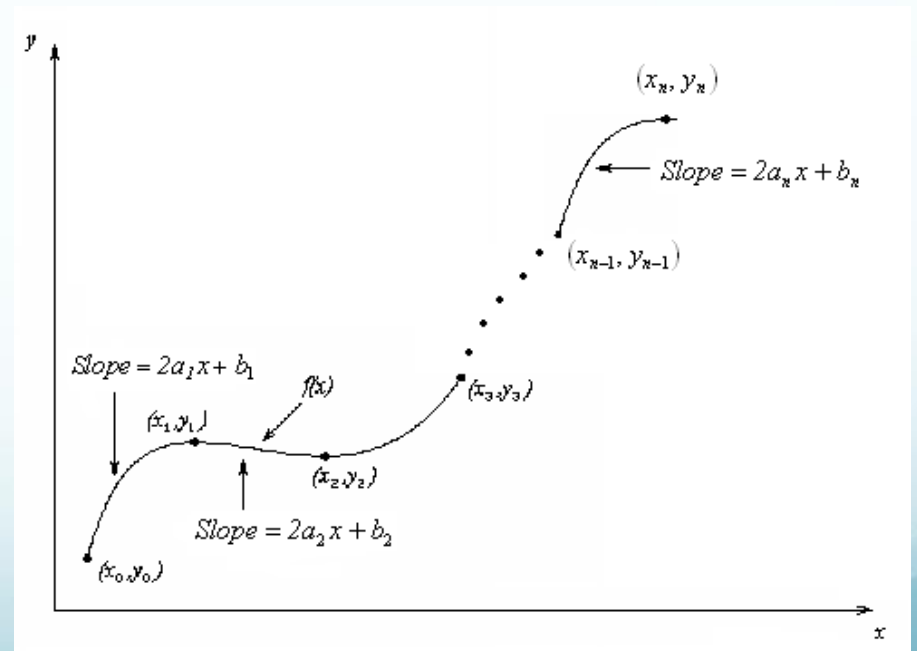
The derivative of the second spline

$$a_2x^2 + b_2x + c_2 \quad \text{is} \quad 2a_2x + b_2$$

and the two are equal at  $x = x_1$  giving

$$2a_1x_1 + b_1 = 2a_2x_1 + b_2$$

$$2a_1x_1 + b_1 - 2a_2x_1 - b_2 = 0$$



# Quadratic Splines

Similarly at the other interior points,

$$2a_2x_2 + b_2 - 2a_3x_2 - b_3 = 0$$

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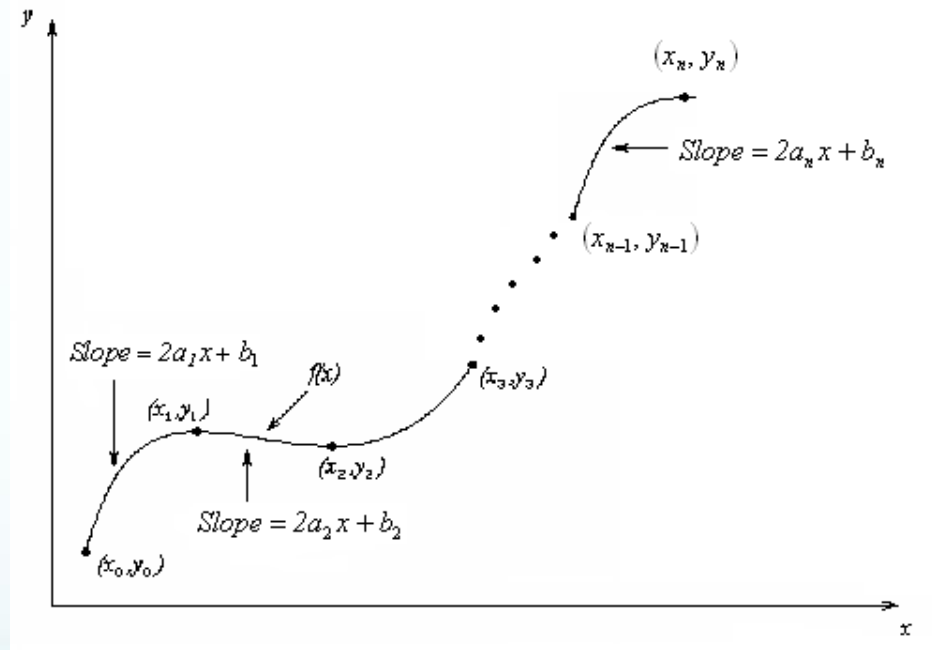
$$2a_ix_i + b_i - 2a_{i+1}x_i - b_{i+1} = 0$$

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$$2a_{n-1}x_{n-1} + b_{n-1} - 2a_nx_{n-1} - b_n = 0$$



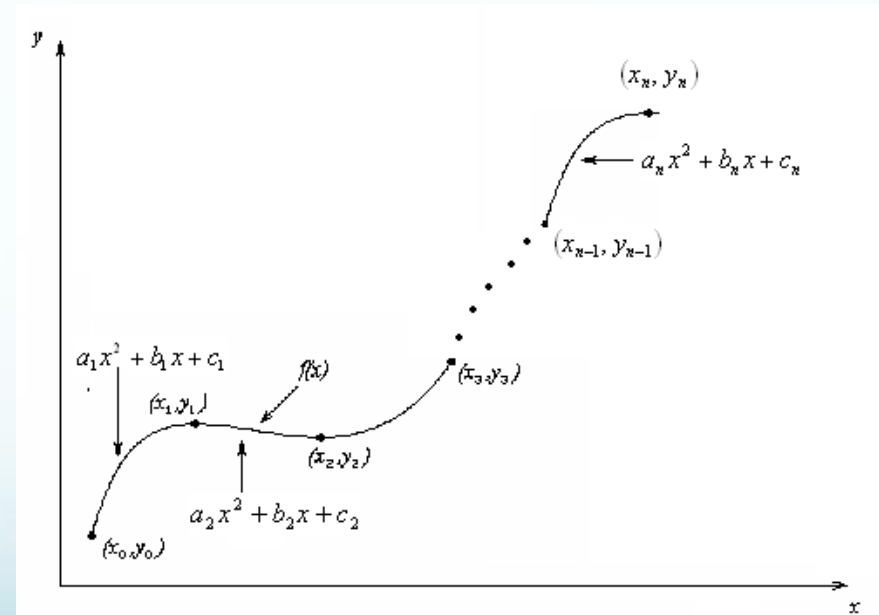
We have  $(n-1)$  such equations. The total number of equations is  $(2n) + (n-1) = (3n-1)$ .

We can assume that the first spline is linear, that is  $a_1 = 0$

# Quadratic Splines

This gives us '3n' equations and '3n' unknowns. Once we find the '3n' constants, we can find the function at any value of 'x' using the splines,

$$\begin{aligned}
 f(x) &= a_1x^2 + b_1x + c_1, & x_0 \leq x \leq x_1 \\
 &= a_2x^2 + b_2x + c_2, & x_1 \leq x \leq x_2 \\
 &\cdot \\
 &\cdot \\
 &\cdot \\
 &= a_nx^2 + b_nx + c_n, & x_{n-1} \leq x \leq x_n
 \end{aligned}$$



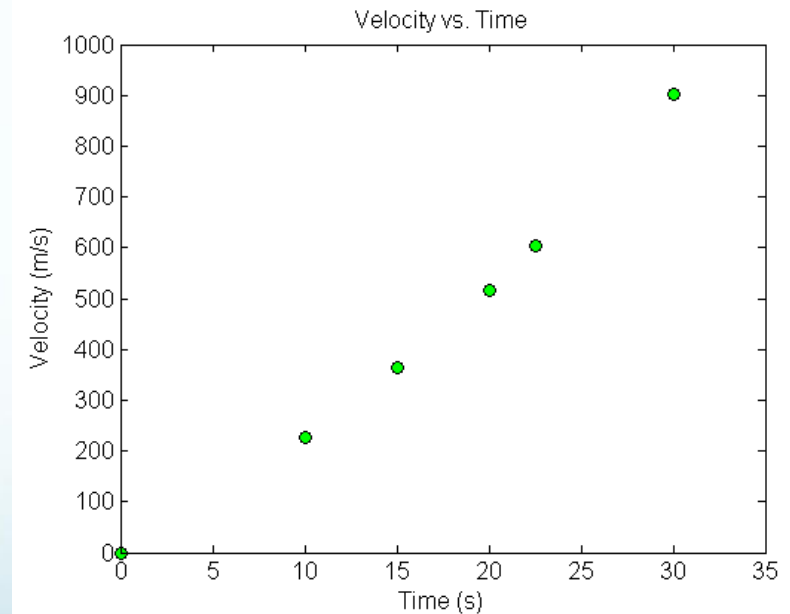
# Quadratic Spline Example

The upward velocity of a rocket is given as a function of time. Using quadratic splines

- a) Find the velocity at  $t=16$  seconds
- b) Find the acceleration at  $t=16$  seconds
- c) Find the distance covered between  $t=11$  and  $t=16$  seconds

**Table :** Velocity as a function of time

$t$ (s)	$v(t)$ (m/s)
0	0
10	227.04
15	362.78
20	517.35
22.5	602.97
30	901.67



**Figure :** Velocity vs. time data for the rocket example

## Solution

$$v(t) = a_1 t^2 + b_1 t + c_1, \quad 0 \leq t \leq 10$$

$$= a_2 t^2 + b_2 t + c_2, \quad 10 \leq t \leq 15$$

$$= a_3 t^2 + b_3 t + c_3, \quad 15 \leq t \leq 20$$

$$= a_4 t^2 + b_4 t + c_4, \quad 20 \leq t \leq 22.5$$

$$= a_5 t^2 + b_5 t + c_5, \quad 22.5 \leq t \leq 30$$

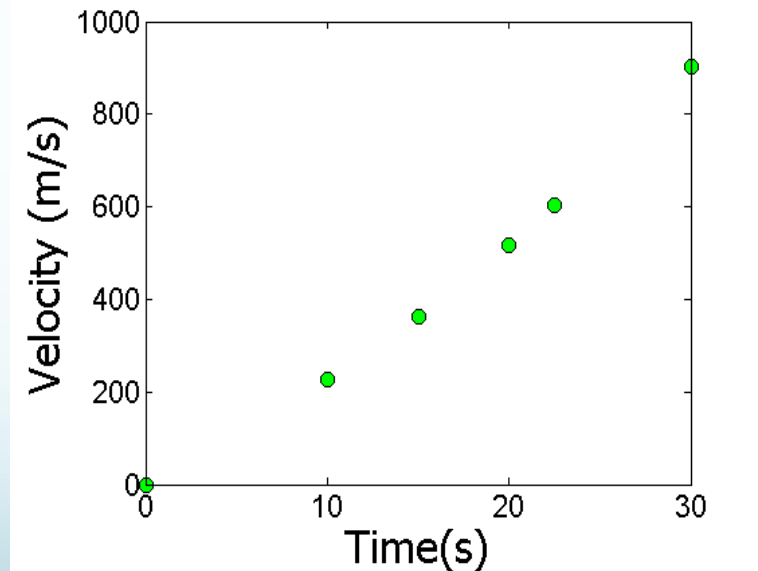
Let us set up the equations

# Each Spline Goes Through Two Consecutive Data Points

$$v(t) = a_1 t^2 + b_1 t + c_1, \quad 0 \leq t \leq 10$$

$$a_1(0)^2 + b_1(0) + c_1 = 0$$

$$a_1(10)^2 + b_1(10) + c_1 = 227.04$$





# Each Spline Goes Through Two Consecutive Data Points

<b>t</b>	<b>v(t)</b>
<b>s</b>	<b>m/s</b>
0	0
10	227.04
15	362.78
20	517.35
22.5	602.97
30	901.67

$$a_2(10)^2 + b_2(10) + c_2 = 227.04$$

$$a_2(15)^2 + b_2(15) + c_2 = 362.78$$

$$a_3(15)^2 + b_3(15) + c_3 = 362.78$$

$$a_3(20)^2 + b_3(20) + c_3 = 517.35$$

$$a_4(20)^2 + b_4(20) + c_4 = 517.35$$

$$a_4(22.5)^2 + b_4(22.5) + c_4 = 602.97$$

$$a_5(22.5)^2 + b_5(22.5) + c_5 = 602.97$$

$$a_5(30)^2 + b_5(30) + c_5 = 901.67$$

# Derivatives are Continuous at Interior Data Points

$$\begin{aligned}v(t) &= a_1 t^2 + b_1 t + c_1, 0 \leq t \leq 10 \\ &= a_2 t^2 + b_2 t + c_2, 10 \leq t \leq 15\end{aligned}$$

$$\left. \frac{d}{dt} (a_1 t^2 + b_1 t + c_1) \right|_{t=10} = \left. \frac{d}{dt} (a_2 t^2 + b_2 t + c_2) \right|_{t=10}$$

$$(2a_1 t + b_1)|_{t=10} = (2a_2 t + b_2)|_{t=10}$$

$$2a_1(10) + b_1 = 2a_2(10) + b_2$$

$$20a_1 + b_1 - 20a_2 - b_2 = 0$$

# Derivatives are Continuous at Interior Data Points

At  $t=10$

$$2a_1(10) + b_1 - 2a_2(10) - b_2 = 0$$

At  $t=15$

$$2a_2(15) + b_2 - 2a_3(15) - b_3 = 0$$

At  $t=20$

$$2a_3(20) + b_3 - 2a_4(20) - b_4 = 0$$

At  $t=22.5$

$$2a_4(22.5) + b_4 - 2a_5(22.5) - b_5 = 0$$

Last Equation  $a_1 = 0$

# Final Set of Equations

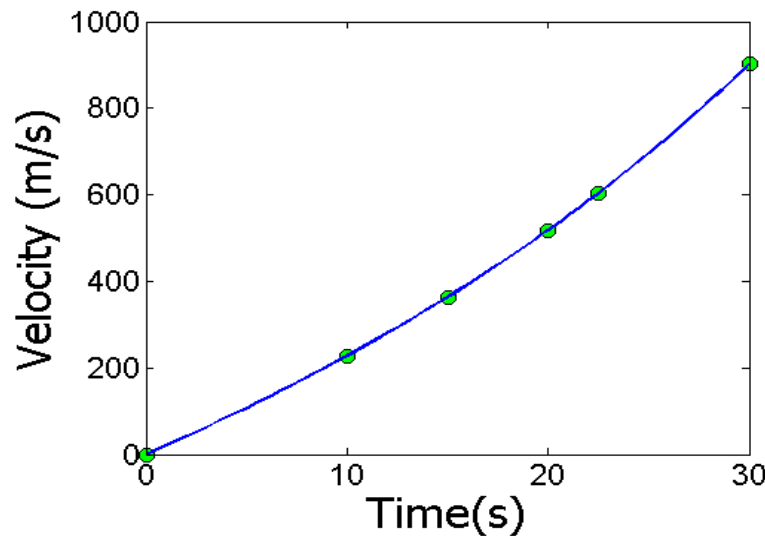
$$\begin{bmatrix}
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 100 & 10 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 100 & 10 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 225 & 15 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 225 & 15 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 400 & 20 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 400 & 20 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 506.25 & 22.5 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 506.25 & 22.5 & 1 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 900 & 30 & 1 & 0 \\
 20 & 1 & 0 & -20 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 30 & 1 & 0 & -30 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 40 & 1 & 0 & -40 & -1 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 45 & 1 & 0 & -45 & -1 & 0 \\
 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 a_1 \\
 b_1 \\
 c_1 \\
 a_2 \\
 b_2 \\
 c_2 \\
 a_3 \\
 b_3 \\
 c_3 \\
 a_4 \\
 b_4 \\
 c_4 \\
 a_5 \\
 b_5 \\
 c_5
 \end{bmatrix}
 =
 \begin{bmatrix}
 0 \\
 227.04 \\
 227.04 \\
 362.78 \\
 362.78 \\
 517.35 \\
 517.35 \\
 602.97 \\
 602.97 \\
 901.67 \\
 0 \\
 0 \\
 0 \\
 0 \\
 0
 \end{bmatrix}$$

# Coefficients of Spline

$i$	$a_i$	$b_i$	$c_i$
1	0	22.704	0
2	0.8888	4.928	88.88
3	-0.1356	35.66	-141.61
4	1.6048	-33.956	554.55
5	0.20889	28.86	-152.13

# Final Solution

$$\begin{aligned}v(t) &= 22.704t, & 0 \leq t \leq 10 \\&= 0.8888t^2 + 4.928t + 88.88, & 10 \leq t \leq 15 \\&= -0.1356t^2 + 35.66t - 141.61, & 15 \leq t \leq 20 \\&= 1.6048t^2 - 33.956t + 554.55, & 20 \leq t \leq 22.5 \\&= 0.20889t^2 + 28.86t - 152.13, & 22.5 \leq t \leq 30\end{aligned}$$



# Velocity at a Particular Point

a) Velocity at  $t=16$

$$\begin{aligned}v(t) &= 22.704t, & 0 \leq t \leq 10 \\&= 0.8888t^2 + 4.928t + 88.88, & 10 \leq t \leq 15 \\&= -0.1356t^2 + 35.66t - 141.61, & 15 \leq t \leq 20 \\&= 1.6048t^2 - 33.956t + 554.55, & 20 \leq t \leq 22.5 \\&= 0.20889t^2 + 28.86t - 152.13, & 22.5 \leq t \leq 30\end{aligned}$$

$$\begin{aligned}v(16) &= -0.1356(16)^2 + 35.66(16) - 141.61 \\&= 394.24 \text{ m/s}\end{aligned}$$

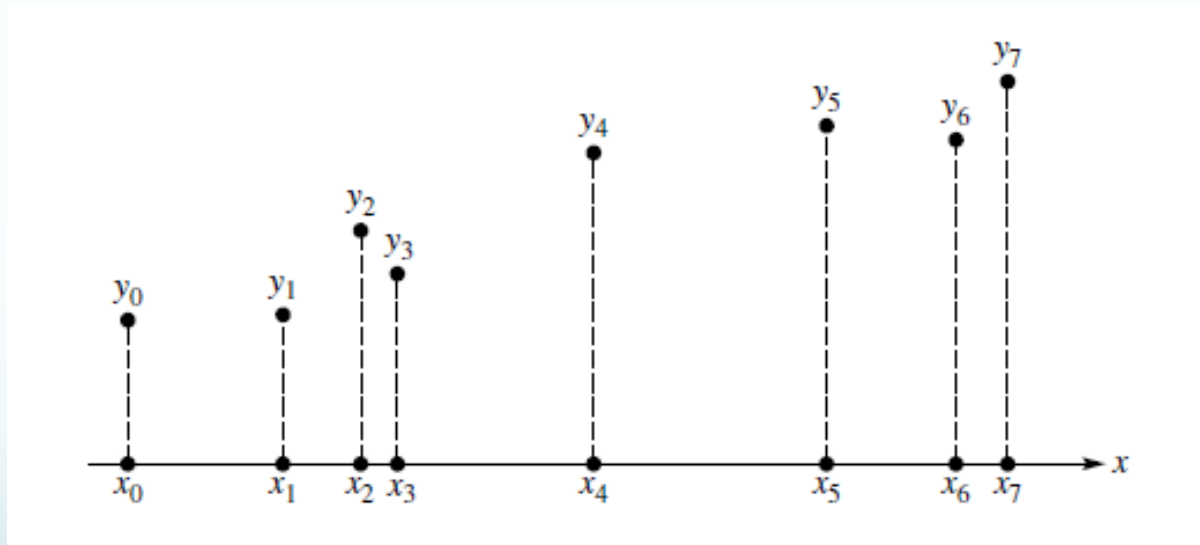
# Smoothing of Data & Least Squares Method

- **What** is data smoothing?
- **Why** do we use data smoothing & least squares?
- **How** do we solve method of least squares?



# data smoothing

- Fitting a smooth curve to tabulated data by experiments.
- Draw a curve that defines the characteristic of data best.
- Obtaining a function that express data.



- Dispersed data points

# Least Squares Method

- Using error function to fit into a curve. This is a method of **minimizing the error of function.**

## Linear Form

$$f(x) = a + bx$$

In this case, the function to be minimized is

$$S(a, b) = \sum_{i=0}^n [y_i - f(x_i)]^2 = \sum_{i=0}^n (y_i - a - bx_i)^2$$

# Linear Least Squares

$$S(a, b) = \sum_{i=0}^n [y_i - f(x_i)]^2 = \sum_{i=0}^n (y_i - a - bx_i)^2$$

$$\frac{\partial S}{\partial a} = \sum_{i=0}^n -2(y_i - a - bx_i) = 2 \left[ a(n+1) + b \sum_{i=0}^n x_i - \sum_{i=0}^n y_i \right] = 0$$

$$\frac{\partial S}{\partial b} = \sum_{i=0}^n -2(y_i - a - bx_i)x_i = 2 \left( a \sum_{i=0}^n x_i + b \sum_{i=0}^n x_i^2 - \sum_{i=0}^n x_i y_i \right) = 0$$

Dividing both equations by  $2(n+1)$ , rearrange terms;

$$a + \bar{x}b = \bar{y} \quad \bar{x}a + \left( \frac{1}{n+1} \sum_{i=0}^n x_i^2 \right) b = \frac{1}{n+1} \sum_{i=0}^n x_i y_i$$

# Linear Least Squares

$$a + \bar{x}b = \bar{y} \quad \bar{x}a + \left( \frac{1}{n+1} \sum_{i=0}^n x_i^2 \right) b = \frac{1}{n+1} \sum_{i=0}^n x_i y_i$$

Where;

$$\bar{x} = \frac{1}{n+1} \sum_{i=0}^n x_i \quad \bar{y} = \frac{1}{n+1} \sum_{i=0}^n y_i$$

*are the mean values of x and y data. The solution for parameters is;*

$$a = \frac{\bar{y} \sum x_i^2 - \bar{x} \sum x_i y_i}{\sum x_i^2 - n\bar{x}^2} \quad b = \frac{\sum x_i y_i - \bar{x} \sum y_i}{\sum x_i^2 - n\bar{x}^2}$$

# Linear Least Squares

*Data obtained by an experiment;*

$x$	4	7	11	13	17
$y$	2	0	2	6	7

*System of two equations;*

$$\begin{cases} 644a + 52b = 227 \\ 52a + 5b = 17 \end{cases}$$

*Corresponding values;*

$$a = 0.4864$$

$$b = -1.6589$$

