

# Engineering 7: Introduction to computer programming for scientists and engineers

## Interpolation

- Recap
- Polynomial interpolation
- Spline interpolation

# Regression and Interpolation: *“learning” functions from data*

# Regression and Interpolation:

## Regression

Find a function, from a specified class, that best fits the data. For example,

$$\min_{f \in \mathbf{F}_R} \sum_{i=1}^N |f(x_i) - y_i|^2$$

Often use a squared-error criterion to score a function's performance.

Let  $f^R$  denote the best-fit function.

Training Data:  $(x_i, y_i)_{i=1, \dots, N}$

## Interpolation

Find a function, from a specified class, that exactly matches the data. For example, find  $f \in \mathbf{F}_I$  with

$$f(x_i) = y_i \text{ for all } 1 \leq i \leq N$$

Let  $f^I$  denote such an interpolator.

The obtained function ( $f^R$  or  $f^I$ ) is often used for *prediction*: given another value of  $x$  (with  $y$  unknown), estimate/predict “what is the corresponding value of  $y$ ?”

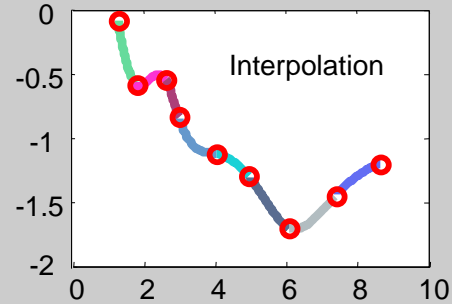
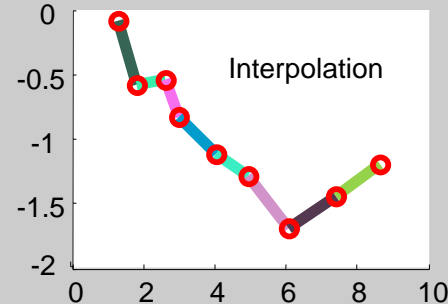
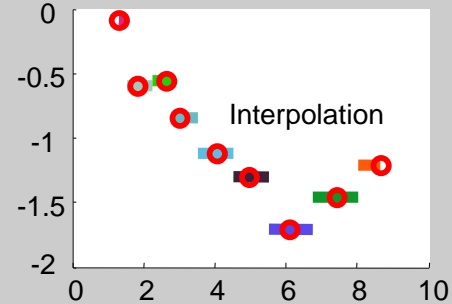
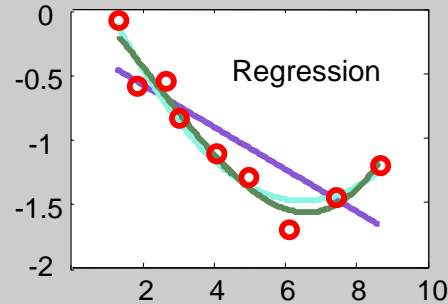
# Perspectives: Differences/similarities

## Regression:

- Measured data is inaccurate in a random fashion.
- Data seems to exhibit “complex” variability, but that variability is not truly indicative of the underlying phenomena that produces the data
- Choose a “model” class of functions to represent the phenomena. It should be no more complex than necessary
- Get approximate fit from this simple class of functions.

## Interpolation:

- Measured data is believed to be accurate.
- Any complex variability is thought to be truly indicative of the underlying phenomena that produces the data
- Functional representation should pass through all data points; no reason to expect wild variability between points.
- The smoothness of the interpolating function used depends on assumptions about the phenomena that produces the data.



# Regression or Interpolation?

Accurate, fast approximation to

$$f(r) := \int_0^r e^{-x^{3/2}} \tan^{-1} x \, dx$$

Interpolation: it makes sense that the fitting function should pass through the training points

over the range  $r \in [0, 5]$ . Data,  $(r_i, f_i)_{i=1, \dots, N}$  is obtained by careful numerical integration, is repeatable, and not subject to any appreciable error.

Regression: the “experimental” aspect hints at errors and some degree of non-repeatability in the training data due to uncontrolled/unknown factors.

Rebound velocity, as a function of impact velocity for a composite baseball bat and baseball,  $r(v)$ . Data,  $(v_i, r_i)_{i=1, \dots, N}$  is obtained from experimental lab apparatus.

# Polynomial Interpolation

# Linear interpolation


A linear (affine) polynomial has 2 parameters

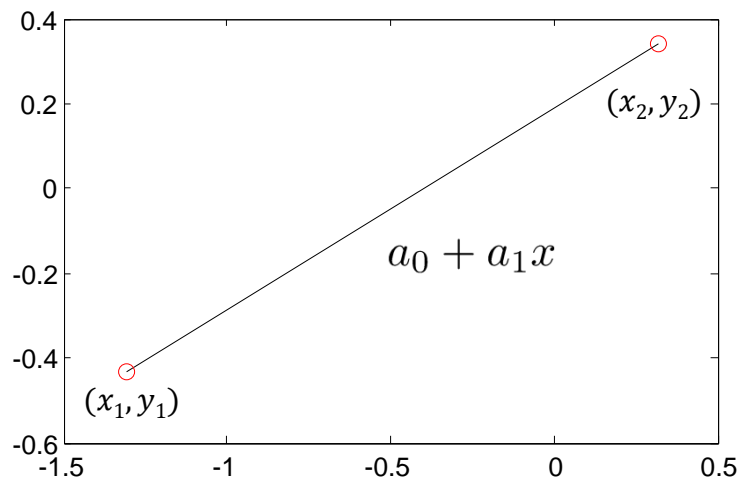
$$p_a(x) = a_0 + a_1x$$

Given 2 data points,  $(x_1, y_1), (x_2, y_2)$  with the  $x_1 \neq x_2$ , there exists a unique choice of parameters  $(a_0, a_1)$  so that  $p_a(x_i) = y_i$  for  $i = 1, 2$ .

$$\begin{bmatrix} 1 & x_1 \\ 1 & x_2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

$Ax = b$





# quadratic interpolation

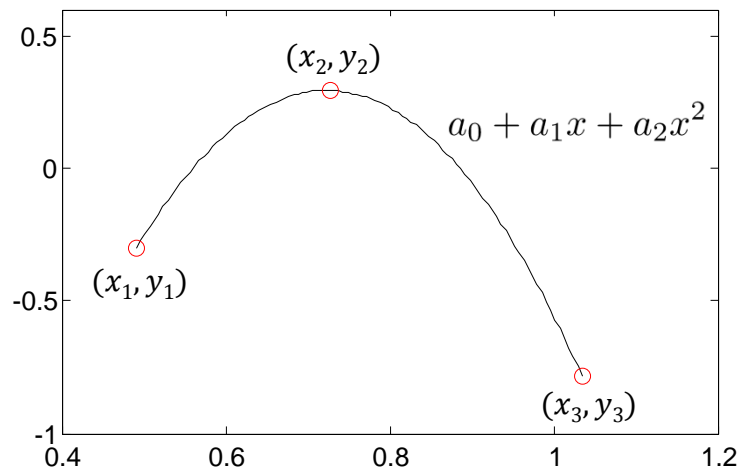
A quadratic polynomial has 3 parameters

$$p_a(x) = a_0 + a_1x + a_2x^2$$

Given 3 data points,  $(x_1, y_1)$ ,  $(x_2, y_2)$ ,  $(x_3, y_3)$ , with the  $\{x_i\}_{i=1,2,3}$  all distinct from one another, there exists a unique choice of parameters  $(a_0, a_1, a_2)$  so that  $p_a(xi) = y_i$  for  $i = 1, 2, 3$ .

$$\begin{bmatrix} 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \\ 1 & x_3 & x_3^2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

$\swarrow \quad \swarrow \quad \swarrow$   
 $Ax = b$



# Cubic interpolation

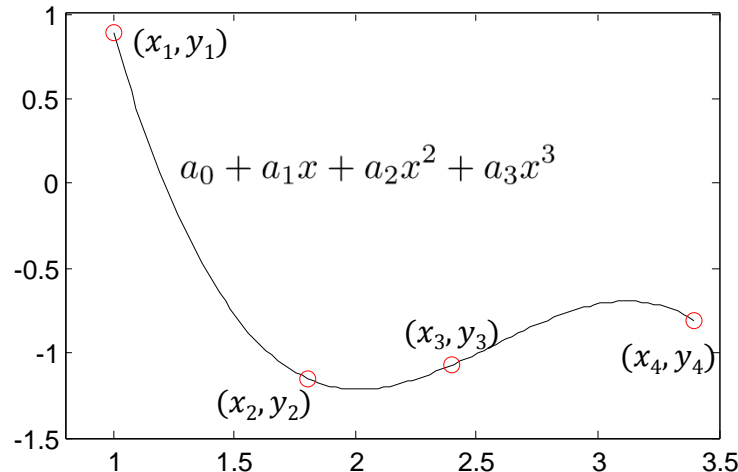
A cubic polynomial has 4 parameters

$$p_a(x) = a_0 + a_1x + a_2x^2 + a_3x^3$$

Given 4 data points,  $(x_1, y_1), (x_2, y_2), (x_3, y_3), (x_4, y_4)$ , with the  $\{x_i\}_{i=1,2,3,4}$  all distinct from one another, there exists a unique choice of parameters  $(a_0, a_1, a_2, a_3)$  so that  $p_a(x_i) = y_i$  for  $i = 1, 2, 3, 4$ .

$$\begin{bmatrix} 1 & x_1 & x_1^2 & x_1^3 \\ 1 & x_2 & x_2^2 & x_2^3 \\ 1 & x_3 & x_3^2 & x_3^3 \\ 1 & x_4 & x_4^2 & x_4^3 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix}$$

$Ax = b$



# Polynomial interpolation

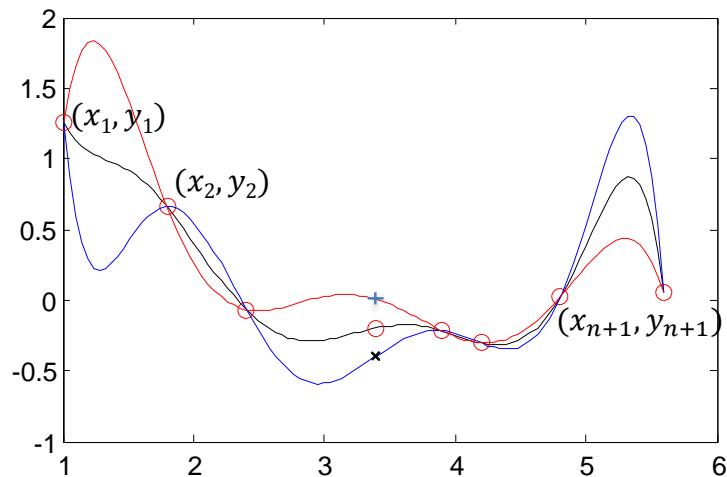
More generally, an  $n^{\text{th}}$  degree polynomial has  $n + 1$  parameters

$$p_a(x) = a_0 + a_1x + a_2x^2 + \cdots + a_nx^n$$

Given  $n + 1$  data points,  $(x_1, y_1), (x_2, y_2), \dots, (x_{n+1}, y_{n+1})$ , with the  $\{x_i\}_{i=1,2,\dots,(n+1)}$  all distinct from one another, there exists a unique choice of parameters  $(a_0, a_1, a_2, \dots, a_n)$  so that  $p_a(x_i) = y_i$  for  $i = 1, 2, \dots, n + 1$ .

$$\begin{bmatrix} 1 & x_1 & \cdots & x_1^n \\ 1 & x_2 & \cdots & x_2^n \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n+1} & \cdots & x_{n+1}^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{n+1} \end{bmatrix}$$

$Ax = b$



# Behavior of Polynomial Interpolation

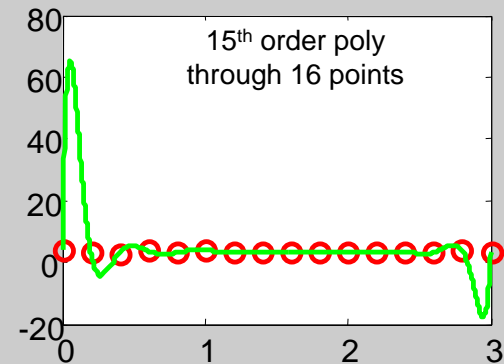
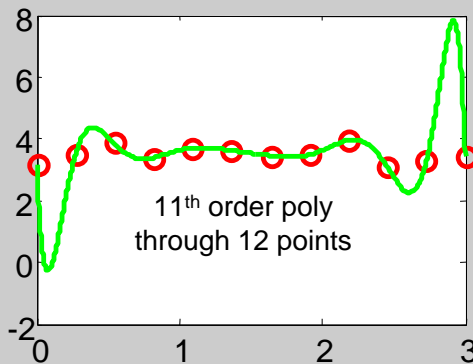
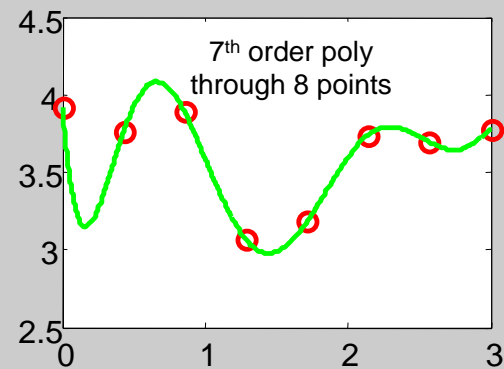
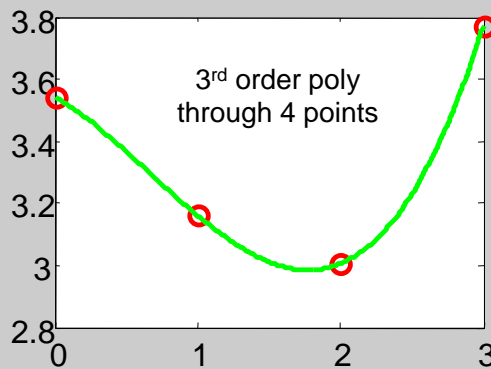
For large  $n$ , this polynomial tends to have large oscillations near the end points.

The behavior of the interpolating function, between points is unexpected and non-intuitive.

For evenly-spaced  $x$ , the prediction (ie., interpolating polynomial's value) at the midpoint between  $x_1$  and  $x_2$  becomes highly sensitive to  $y_{i=n/2}$ .

$n$	5	7	9	11	13	15
$Sens$	1.2	1.5	3.3	8.6	24	72

Polynomial interpolating functions using large  $n$  are generally thought of as *bad predictors*



# Spline interpolation: linear and cubic

# linear-spline

Given: data,  $(x_i, y_i)_{i=1, \dots, n}$ . The  $(x_i)$  are distinct and sorted, so that  $x_i < x_{i+1}$ .

A linear-spline interpolation consists of  $(n - 1)$  linear (affine) functions, defined separately on each interval  $[x_i \ x_{i+1}]$ .

- 1<sup>st</sup> function connects  $(x_1, y_1)$  to  $(x_2, y_2)$  with a straight line,
- 2<sup>nd</sup> function connects  $(x_2, y_2)$  to  $(x_3, y_3)$  with a straight line,
- k<sup>th</sup> function connects  $(x_k, y_k)$  to  $(x_{k+1}, y_{k+1})$  with a straight line
  - For  $X$  in  $[x_k \ x_{k+1}]$  interpolation function (relating  $X$  to  $Y$ ) is easily expressed as

$$Y = y_k + \frac{y_{k+1} - y_k}{x_{k+1} - x_k}(X - x_k)$$

- (n-1)<sup>th</sup> function connects  $(x_{n-1}, y_{n-1})$  to  $(x_n, y_n)$  with a straight line.

Viewed as a whole: the interpolation function is piecewise-linear

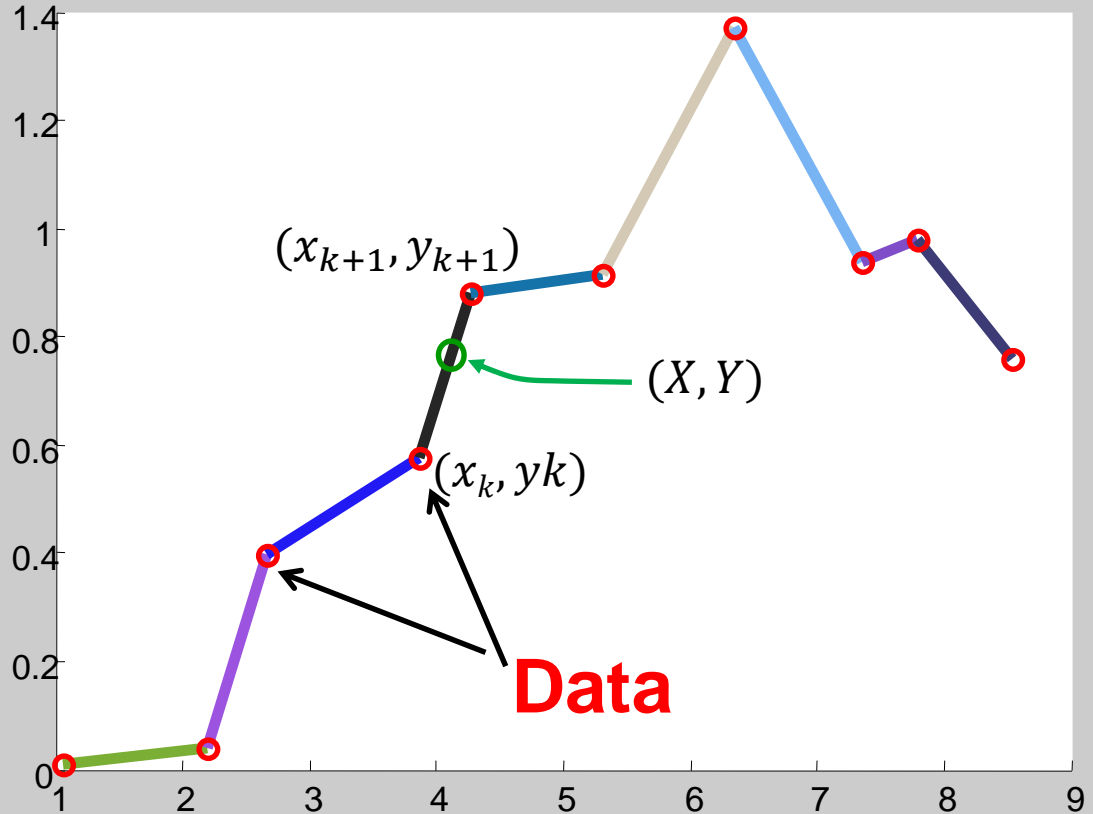
# Linear-spline (or just “linear”) interpolation

$$Y = y_k + \frac{y_{k+1} - y_k}{x_{k+1} - x_k} (X - x_k)$$

Starting height

Slope in this interval

“Run”



# Cubic-spline

Given: data,  $(x_i, y_i)_{i=1, \dots, n}$ . The  $(x_i)$  are distinct and sorted, so that  $x_i < x_{i+1}$ .

A cubic-spline interpolation consists of cubic polynomial functions, defined on each interval  $[x_i, x_{i+1}]$ .

- 1<sup>st</sup> function connects  $(x_1, y_1)$  to  $(x_2, y_2)$  with cubic polynomial,
- 2<sup>nd</sup> function connects  $(x_2, y_2)$  to  $(x_3, y_3)$  with cubic polynomial,
- $k^{\text{th}}$  function connects  $(x_k, y_k)$  to  $(x_{k+1}, y_{k+1})$  with cubic polynomial,
- $(n-1)^{\text{th}}$  function connects  $(x_{n-1}, y_{n-1})$  to  $(x_n, y_n)$  with cubic polynomial

## Constraints

$$\begin{array}{lll} @ x_1 & f_1(x_1) = y_1 & @ x_k \quad f_{k-1}(x_k) = y_k \quad f'_{k-1}(x_k) = f'_k(x_k) \\ @ x_n & f_{n-1}(x_n) = y_n & (2 \leq k \leq n-1) \quad f_k(x_k) = y_k \quad f''_{k-1}(x_k) = f''_k(x_k) \end{array}$$

$k^{\text{th}}$  function is parametrized by 4 coefficients (to be determined by the constraints)

$$f_k(x) = a_0 + a_1(x - x_k) + a_2(x - x_k)^2 + a_3(x - x_k)^3$$

# Cubic spline

$(n - 1)$  cubic polynomial functions, each defined on  $[x_i, x_{i+1}]$ .

$k^{\text{th}}$  function connects  $(x_k, y_k)$  to  $(x_{k+1}, y_{k+1})$

## Constraints

@  $x_1$

1 equation

$$f_1(x_1) = y_1$$

@  $x_k$  ( $2 \leq k \leq n - 1$ )

$$f_{k-1}(x_k) = y_k$$

$$f_k(x_k) = y_k$$

$$f'_{k-1}(x_k) = f'_k(x_k)$$

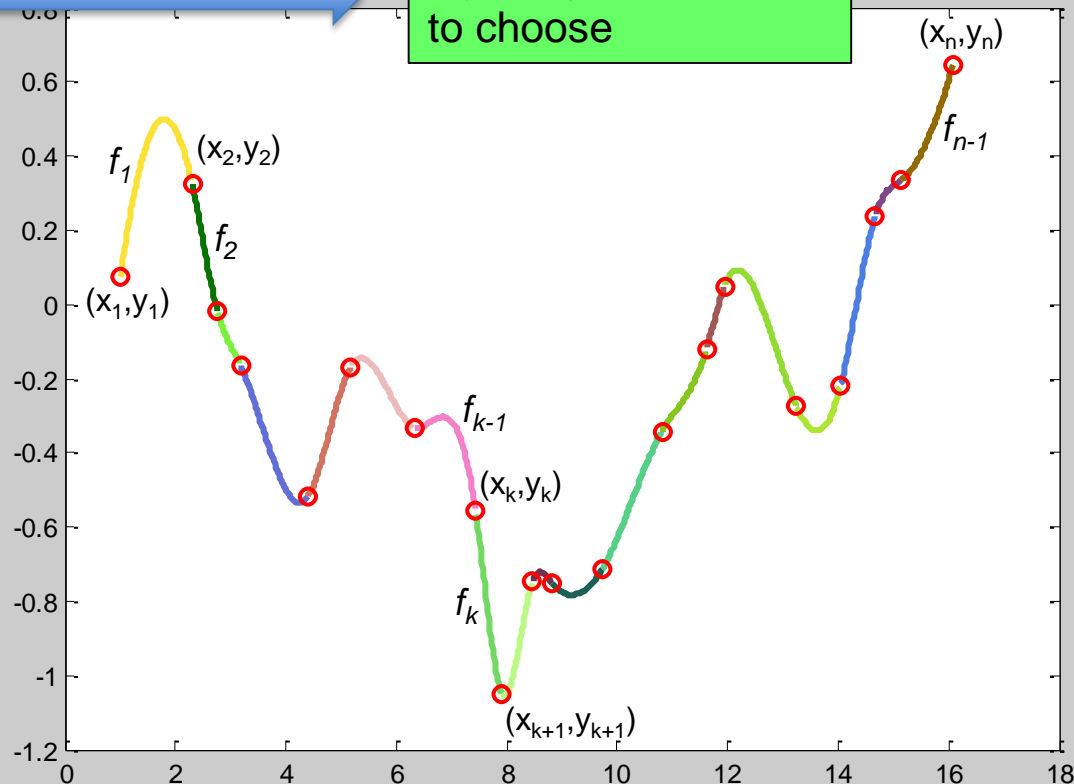
$$f''_{k-1}(x_k) = f''_k(x_k)$$

@  $x_n$

1 equation

$$f_{n-1}(x_n) = y_n$$

4(n - 1) parameters to choose



4(n - 2)  
equations

4(n - 2) + 2 equations

# Two Extra conditions

As there are fewer equations than unknowns (coefficients of the cubic functions), the matching conditions (constraints) are not enough to uniquely determine the cubic functions.

Two more conditions can be imposed. Four common approaches are

– Natural

$$f_1''(x_1) = 0, \quad f_{n-1}''(x_n) = 0$$

– Periodic

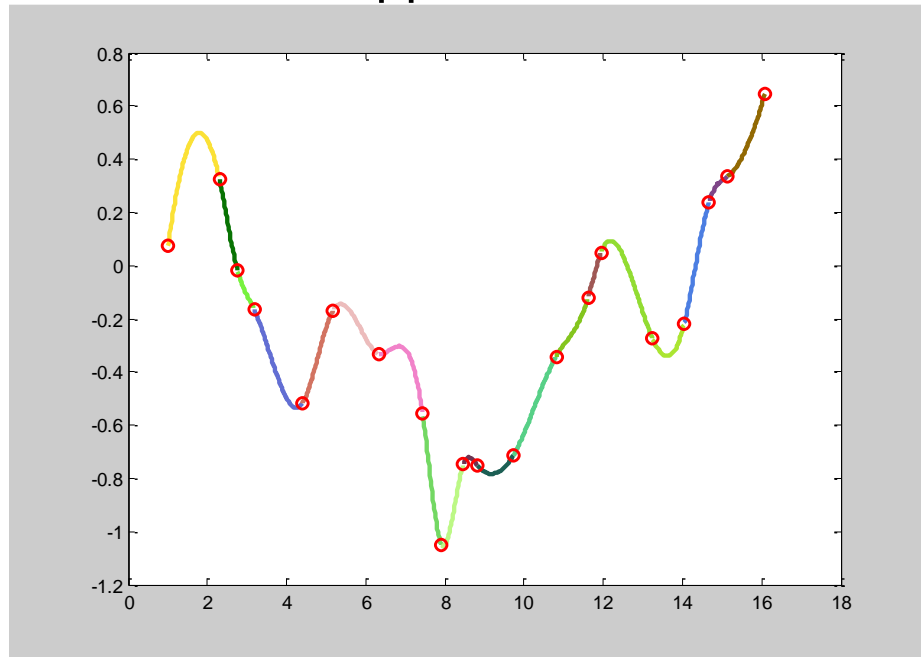
$$f_1'(x_1) = f_{n-1}'(x_n), \quad f_1''(x_1) = f_{n-1}''(x_n)$$

– Not-a-knot  $f_1'''(x_2) = f_2'''(x_2),$

$$f_{n-2}'''(x_{n-1}) = f_{n-1}'''(x_{n-1})$$

– Specified end-slope, given values  $v_1, v_n$

$$f_1'(x_1) = v_1, \quad f_{n-1}'(x_n) = v_n$$



# Setting up the equations

# Values and derivatives of cubic

What are the values and derivative of a cubic

$$f(x) = a_0 + a_1(x - \bar{x}) + a_2(x - \bar{x})^2 + a_3(x - \bar{x})^3$$

$$f'(x) = a_1 + 2a_2(x - \bar{x}) + 3a_3(x - \bar{x})^2$$

$$f''(x) = 2a_2 + 6a_3(x - \bar{x})$$

$$f'''(x) = 6a_3$$

$$\begin{bmatrix} f(x) \\ f'(x) \\ f''(x) \\ f'''(x) \end{bmatrix} = \begin{bmatrix} 1 & x - \bar{x} & (x - \bar{x})^2 & (x - \bar{x})^3 \\ 0 & 1 & 2(x - \bar{x}) & 3(x - \bar{x})^2 \\ 0 & 0 & 2 & 6(x - \bar{x}) \\ 0 & 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}$$

# Values and derivatives of $f_k$

For  $f_k$ , the value of  $\bar{x} = x_k$ , so

$$f_k(x) = a_0 + a_1(x - x_k) + a_2(x - x_k)^2 + a_3(x - x_k)^3$$

$$f'_k(x) = a_1 + 2a_2(x - x_k) + 3a_3(x - x_k)^2$$

$$f''_k(x) = 2a_2 + 6a_3(x - x_k)$$

$$f'''_k(x) = 6a_3$$

$$\begin{bmatrix} f_k(x_k) \\ f'_k(x_k) \\ f''_k(x_k) \\ f'''_k(x_k) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_k \qquad \begin{bmatrix} f_k(x_{k+1}) \\ f'_k(x_{k+1}) \\ f''_k(x_{k+1}) \\ f'''_k(x_{k+1}) \end{bmatrix} = \begin{bmatrix} 1 & h_k & h_k^2 & h_k^3 \\ 0 & 1 & 2h_k & 3h_k^2 \\ 0 & 0 & 2 & 6h_k \\ 0 & 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_k$$

$$@x = x_k$$

$$@x = x_{k+1}$$

$$h_k := x_{k+1} - x_k$$

# Constraints

$$@x_1, f_1(x_1) = y_1$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_1 = y_1$$

$$(a_0)_1 = y_1$$

$$@x_n, f_{n-1}(x_n) = y_n$$

$$\begin{bmatrix} 1 & h_{n-1} & h_{n-1}^2 & h_{n-1}^3 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_{n-1} = y_n$$

$$@x_k, 2 \leq k \leq (n-1)$$

$$f_{k-1}(x_k) = y_k$$

$$f_k(x_k) = y_k$$

$$f'_{k-1}(x_k) = f'_k(x_k)$$

$$f''_{k-1}(x_k) = f''_k(x_k)$$

$$\begin{bmatrix} 1 & h_{k-1} & h_{k-1}^2 & h_{k-1}^3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 2h_{k-1} & 3h_{k-1}^2 & 0 & -1 & 0 & 0 \\ 0 & 0 & 2 & 6h_{k-1} & 0 & 0 & -2 & 0 \end{bmatrix} \begin{bmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{pmatrix}_{k-1} \\ \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{pmatrix}_k \end{bmatrix} = \begin{bmatrix} y_k \\ y_k \\ 0 \\ 0 \end{bmatrix}$$

$$(a_0)_k = y_k$$

# Rewrite Constraints

$$(a_0)_1 = y_1$$

$$(a_0)_k = y_k, 2 \leq k \leq (n-1)$$

$$(a_0)_k = y_k, \quad \forall k \quad (1 \leq k \leq (n-1))$$

$$@x_n, f_{n-1}(x_n) = y_n$$

$$\begin{bmatrix} h_{n-1} & h_{n-1}^2 & h_{n-1}^3 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}_{n-1} = y_n - y_{n-1}$$

$$@x_k, 2 \leq k \leq (n-1)$$

$$f_{k-1}(x_k) = y_k$$

$$f'_{k-1}(x_k) = f'_k(x_k)$$

$$f''_{k-1}(x_k) = f''_k(x_k)$$

$$\begin{bmatrix} h_{k-1} & h_{k-1}^2 & h_{k-1}^3 & 0 & 0 & 0 \\ 1 & 2h_{k-1} & 3h_{k-1}^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_{k-1} & 0 & -2 & 0 \end{bmatrix} \begin{bmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_{k-1} \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_k \end{bmatrix} = \begin{bmatrix} y_k - y_{k-1} \\ 0 \\ 0 \end{bmatrix}$$

# Assembling the equations ( $n = 6$ )

$$\begin{bmatrix}
 \begin{bmatrix} h_1 & h_1^2 & h_1^3 & 0 & 0 & 0 \\ 1 & 2h_1 & 3h_1^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_1 & 0 & -2 & 0 \end{bmatrix} & & & & & \\
 & \begin{bmatrix} h_2 & h_2^2 & h_2^3 & 0 & 0 & 0 \\ 1 & 2h_2 & 3h_2^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_2 & 0 & -2 & 0 \end{bmatrix} & & & & & \\
 & & \begin{bmatrix} h_3 & h_3^2 & h_3^3 & 0 & 0 & 0 \\ 1 & 2h_3 & 3h_3^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_3 & 0 & -2 & 0 \end{bmatrix} & & & & \\
 & & & \begin{bmatrix} h_4 & h_4^2 & h_4^3 & 0 & 0 & 0 \\ 1 & 2h_4 & 3h_4^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_4 & 0 & -2 & 0 \end{bmatrix} & & & \\
 & & & & \begin{bmatrix} h_5 & h_5^2 & h_5^3 \end{bmatrix} & & & \\
 & & & & & \text{Two more conditions} & 
 \end{bmatrix}
 \begin{bmatrix}
 \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_1 \\
 \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_2 \\
 \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_3 \\
 \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_4 \\
 \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_5
 \end{bmatrix}
 =
 \begin{bmatrix}
 y_2 - y_1 \\
 0 \\
 0 \\
 y_3 - y_2 \\
 0 \\
 0 \\
 y_4 - y_3 \\
 0 \\
 0 \\
 y_5 - y_4 \\
 0 \\
 0 \\
 y_6 - y_5
 \end{bmatrix}$$

## Assembling the equations (arbitrary $n$ )

$$\begin{bmatrix} h_1 & h_1^2 & h_1^3 & 0 & 0 & 0 \\ 1 & 2h_1 & 3h_1^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_1 & 0 & -2 & 0 \end{bmatrix} \quad \begin{matrix} \text{Cols=} \\ 3*(k-1)+1:3k+3 \end{matrix}$$

$$\begin{matrix} \text{Rows=} \\ 3*(k-1)+1:3k \end{matrix} \quad \begin{bmatrix} h_k & h_k^2 & h_k^3 & 0 & 0 & 0 \\ 1 & 2h_k & 3h_k^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_k & 0 & -2 & 0 \end{bmatrix}$$

$$\begin{bmatrix} h_{n-2} & h_{n-2}^2 & h_{n-2}^3 & 0 & 0 & 0 \\ 1 & 2h_{n-2} & 3h_{n-2}^2 & -1 & 0 & 0 \\ 0 & 2 & 6h_{n-2} & 0 & -2 & 0 \end{bmatrix}$$

Two more conditions

$$\begin{bmatrix} h_{n-1} & h_{n-1}^2 & h_{n-1}^3 \end{bmatrix}$$

# Incorporating extra conditions

# Two Extra conditions

As there are fewer equations than unknowns (coefficients of the cubic functions), the matching conditions are not enough to uniquely determine the cubic functions.

Two more conditions can be imposed. Four common approaches are

– Natural

$$f_1''(x_1) = 0, \quad f_{n-1}''(x_n) = 0$$

– Periodic

$$f_1'(x_1) = f_{n-1}'(x_n), \quad f_1''(x_1) = f_{n-1}''(x_n)$$

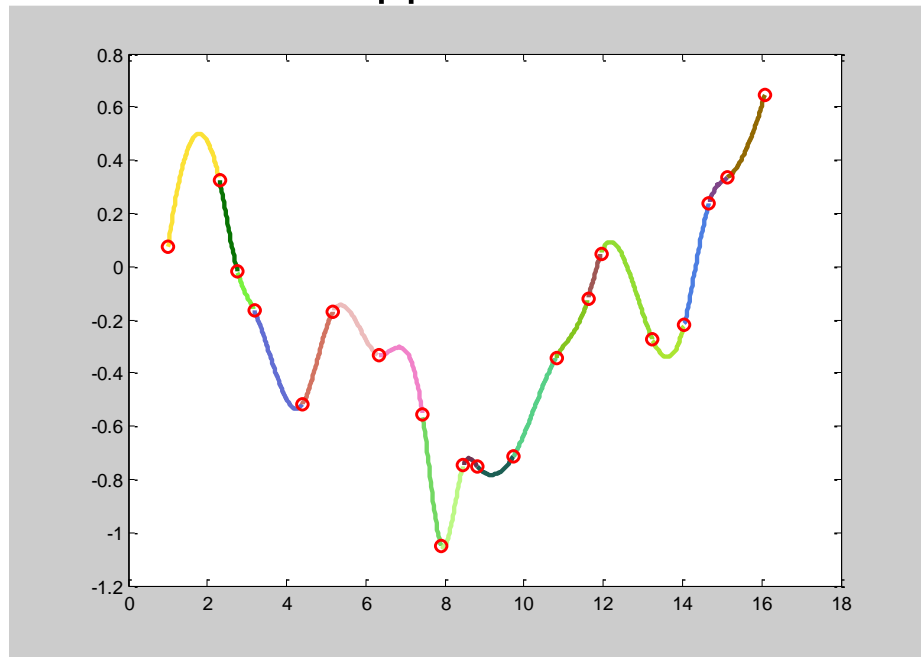
– Not-a-knot

$$f_1'''(x_2) = f_2'''(x_2),$$

$$f_{n-2}'''(x_{n-1}) = f_{n-1}'''(x_{n-1})$$

– Specified end-slope, given values  $v_1, v_n$

$$f_1'(x_1) = v_1, \quad f_{n-1}'(x_n) = v_n$$



# Values and derivatives of $f_k$

For  $f_k$ , the value of  $\bar{x} = x_k$ , so

$$f_k(x) = a_0 + a_1(x - x_k) + a_2(x - x_k)^2 + a_3(x - x_k)^3$$

$$f'_k(x) = a_1 + 2a_2(x - x_k) + 3a_3(x - x_k)^2$$

$$f''_k(x) = 2a_2 + 6a_3(x - x_k)$$

$$f'''_k(x) = 6a_3$$

$$\begin{bmatrix} f_k(x_k) \\ f'_k(x_k) \\ f''_k(x_k) \\ f'''_k(x_k) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_k \qquad \begin{bmatrix} f_k(x_{k+1}) \\ f'_k(x_{k+1}) \\ f''_k(x_{k+1}) \\ f'''_k(x_{k+1}) \end{bmatrix} = \begin{bmatrix} 1 & h_k & h_k^2 & h_k^3 \\ 0 & 1 & 2h_k & 3h_k^2 \\ 0 & 0 & 2 & 6h_k \\ 0 & 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{bmatrix}_k$$

$$@x = x_k$$

$$@x = x_{k+1}$$

$$h_k := x_{k+1} - x_k$$

# Extra equations for Natural spline

## Conditions for Natural Spline

$$f_1''(x_1) = 0 \quad f_{n-1}''(x_n) = 0$$

## Expressions

$$f_1''(x_1) = 2(a_2)_1$$

$$f_{n-1}''(x_n) = 2(a_2)_{n-1} + 6h_{n-1}(a_3)_{n-1}$$

$$\begin{bmatrix} 0 & 2 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 2 & 6h_{n-1} \end{bmatrix}$$

$$\begin{bmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_1 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_2 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_3 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_4 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_5 \end{bmatrix}$$

=

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

# Extra equations for Periodic spline

## Conditions for Periodic Spline

$$f_1'(x_1) = f_{n-1}'(x_n) \quad f_1''(x_1) = f_{n-1}''(x_n)$$

## Expressions

$$f_1'(x_1) = (a_1)_1$$

$$f_{n-1}'(x_n) = (a_1)_{n-1} + 2h_{n-1}(a_2)_{n-1} + 3h_{n-1}^2(a_3)_{n-1}$$

$$f_1''(x_1) = 2(a_2)_1$$

$$f_{n-1}''(x_n) = 2(a_2)_{n-1} + 6h_{n-1}(a_3)_{n-1}$$

$$\begin{bmatrix} -1 & 0 & 0 \\ 0 & -2 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2h_{n-1} & 3h_{n-1}^2 \\ 0 & 2 & 6h_{n-1} \end{bmatrix}$$

$$\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_1$$

$$\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_2$$

$$\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_3$$

$$\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_4$$

$$\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_5$$

=

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

# Extra equations for Not-a-knot

## Conditions for Not-a-knot spline

$$f_1'''(x_2) = f_2'''(x_2) \quad f_{n-2}'''(x_{n-1}) = f_{n-1}'''(x_{n-1})$$

## Expressions

$$f_1'''(x_2) = 6(a_3)_1$$

$$f_2'''(x_2) = 6(a_3)_2$$

$$f_{n-2}'''(x_{n-1}) = 6(a_3)_{n-2}$$

$$f_{n-1}'''(x_{n-1}) = 6(a_3)_{n-1}$$

$$\begin{bmatrix} 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} 0 & 0 & -6 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 6 \end{bmatrix} \begin{bmatrix} 0 & 0 & -6 \end{bmatrix}$$

$$\begin{bmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_1 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_2 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_3 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_4 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_5 \end{bmatrix}$$

=

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

# Extra equations for Specified end-slope

Conditions for specified end-slope

$$f'_1(x_1) = v_1, \quad f'_{n-1}(x_n) = v_n$$

Expressions

$$f'_1(x_1) = (a_1)_1$$

$$f'_{n-1}(x_n) = (a_1)_{n-1} + 2h_{n-1}(a_2)_{n-1} + 3h_{n-1}^2(a_3)_{n-1}$$

$$\begin{bmatrix} 1 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2h_{n-1} & 3h_{n-1}^2 \end{bmatrix}$$

$$\begin{bmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_1 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_2 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_3 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_4 \\ \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}_5 \end{bmatrix}$$

=

$$\begin{bmatrix} v_1 \\ v_n \end{bmatrix}$$

# Implementation

Need one function to transform  $(x_i, y_i)_{i=1, \dots, n}$  into the coefficients  $(\{a_0, a_1, a_2, a_3\}_{i=1, \dots, (n-1)})$ . Just build the arrays, and use \ (backslash)

Need another function to evaluate the “spline” at an arbitrary value of  $x$ . This function needs

- the coefficients,  $(\{a_0, a_1, a_2, a_3\}_{i=1, \dots, (n-1)})$ ,
- the  $(x_i)_{i=1, \dots, n}$  samples, and
- the  $x$ -values for which the evaluation should take place.

The commands

**spline, ppval**

implement the ideas put forth here, in a more efficient manner. Only the “not-a-knot” and “specified end-slope” conditions are available.