# Lecture 14 Interpolation/Splines

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# Given n + 1 distinct points $x_0, \ldots, x_n$ , and values $y_0, \ldots, y_n$ , there exists a unique polynomial p(x) of degree at most n so that

$$p(x_i) = y_i \quad i = 0, \ldots, n$$

#### **Recall: Monomials**

Obvious attempt: try picking

$$p(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$$

So for each  $x_i$  we have

$$p(x_i) = a_0 + a_1 x_i + a_2 x_i^2 + \dots + a_n x_i^n = y_i$$

OR

$$\begin{bmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ & & \vdots & \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_n \end{bmatrix}$$

That is,

$$a = M^{-1}y$$

Very bad matrix: terribly ill-conditioned, inverse entries are **large** Very bad evaluation: values are **huge** 

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Evaluating

$$p(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$$

may have huge values. Partial fix:

$$p(x) = a_0 + a_1(x - \bar{x}) + a_2(x - \bar{x})^2 + \dots + a_n(x - \bar{x})^n$$

Then  $M = vander(x - \bar{x})$  and

$$a = M^{-1}y$$

Still, a very bad matrix.

The general Lagrange form is

$$\ell_k(x) = \prod_{i=0, i \neq k}^n \frac{x - x_i}{x_k - x_i}$$

The resulting interpolating polynomial is

$$p(x) = \sum_{k=0}^{n} \ell_k(x) y_k$$

Find the equation of a quadratic passing through the points (0,-1), (1,-1), and (2,7).

$$x_0 = 0, x_1 = 1, x_2 = 2$$
  $y_0 = -1, y_1 = -1, y_2 = 7$ 

• Form the Lagrange basis functions,  $\ell_i(x)$  with  $\ell_i(x_j) = \delta_{ij}$ 

Ombine the Lagrange basis functions

$$p_2(x) = y_0 \ell_0(x) + y_1 \ell_1(x) + y_2 \ell_2(x)$$
  
=  $(-1) \frac{(x-1)(x-2)}{2} + (-1) \frac{x(x-2)}{-1} + (7) \frac{x(x-1)}{2}$ 

Evaluate is nice, but expensive: no easy nested form.

# **Recall: Newton Polynomials**

Newton Polynomials are of the form

$$p_n(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)(x - x_1) + a_3(x - x_0)(x - x_1)(x - x_2) + \dots$$

The basis used is thus

function	order
1	0
$x - x_0$	1
$(x - x_0)(x - x_1)$	2
$(x - x_0)(x - x_1)(x - x_2)$	3

- More stable evaluation than monomials
- More computationally efficient (nested iteration) than using Lagrange

# How bad is polynomial interpolation?

Let's take something very smooth function



How does interpolation behave?

#### Some analysis...

what can we say about

$$e(t) = f(t) - p_n(t)$$

at some point x? Consider p = 1: linear interpolation of a function at  $x = x_0, x_1$ 

- want: error at *x*, *e*(*x*)
- Iook at

$$g(t) = e(t) - \frac{(t - x_0)(t - x_1)}{(x - x_0)(x - x_1)}e(x)$$

- g(t) is 0 at  $t = x_0, x_1, x$
- so g'(t) is zero at two points
- so g''(t) is zero at one point, call it c

then

$$0 = g''(c) = e''(t) - 2\frac{e(x)}{(x - x_0)(x - x_1)}$$
$$= f''(t) - 2\frac{e(x)}{(x - x_0)(x - x_1)}$$
$$e(x) = \frac{(x - x_0)(x - x_1)}{2}f''(c)$$

#### Theorem: Interpolation Error I

If  $p_n(x)$  is the (at most) *n* degree polynomial interpolating f(x) at n + 1 distinct points and if  $f^{(n+1)}$  is continuous, then

$$e(x) = f(x) - p_n(x) = \frac{1}{(n+1)!} f^{(n+1)}(c) \prod_{i=0}^n (x - x_i)$$

#### Theorem: Bounding Lemma

Suppose  $x_i$  are equispaced in [a, b] for i = 0, ..., n. Then

$$\prod_{i=0}^{n} |x - x_i| \leqslant \frac{h^{n+1}}{4}n!$$

#### Theorem: Interpolation Error II

Let  $|f^{(n+1)}(x)| \leq M$ , then with the above,

$$|f(x) - p_n(x)| \leq \frac{Mh^{n+1}}{4(n+1)}$$



We have two options:

- move the nodes: Chebychev nodes
- Piecewise polynomials (splines)

Option #1: Chebychev nodes in [-1, 1]

$$x_i = cos(\pi \frac{2i+1}{2n+2}), \quad i = 0, \dots, n$$

Option #2: piecewise polynomials...

# **Chebychev Nodes**



- Can obtain nodes from equidistant points on a circle projected down
- Nodes are non uniform and non nested

High degree polynomials using equispaced points suffer from many oscillations



- Points are bunched at the ends of the interval
- Error is distributed more evenly

# Why Splines?







- truetype fonts, postscript, metafonts
- graphics surfaces
- smooth surfaces are needed
- how do we interpolate smoothly a set of data?
- keywords: Bezier Curves, splines, B-splines, NURBS
- basic tool: piecewise interpolation



#### **Piecewise Polynomial**

A function f(x) is considered a piecewise polynomial on [a, b] if there exists a (finite) partition *P* of [a, b] such that f(x) is a polynomial on each  $[t_i, t_{i+1}] \in P$ .

#### Example

$$f(x) = \begin{cases} x^3 & x \in [0, 1] \\ x & x \in (1, 2) \\ 3 & x \in [2, 3] \end{cases}$$



- we would like the piecewise polynomial to do two things
  - interpolate (or be close to) some set of data points
  - Iook nice (smooth)
- one option is called a *spline*

- A spline is a piecewise polynomial with a certain level of smoothness.
- take Matlab: plot(1:7,rand(7,1))
- this is linear and continuous, but not very smooth
- the function changes behavior at *knots*  $t_0, \ldots, t_n$



### degree 1 spline

#### definition

A function S(x) is a spline of degree 1 if:

- The domain of S(x) is an interval [a, b]
- 2 S(x) is continuous on [a, b]
- There is a partition  $a = t_0 < t_1 < \cdots < t_n = b$  such that S(x) is linear on each subinterval  $[t_i, t_{i+1}]$ .





Given data  $t_0, \ldots, t_n$  and  $y_0, \ldots, y_n$ , how do we form a spline?

We need two things to hold in the interval  $[a, b] = [t_0, t_n]$ :

• 
$$S(t_i) = y_i$$
 for  $i = 0, ..., n$ 

2 
$$S_i(x) = a_i x + b_i$$
 for  $i = 0, ..., n$ 

Write  $S_i(x)$  in point-slope form

$$S_i(x) = y_i + m_i(x - t_i)$$
  
=  $y_i + \frac{y_{i+1} - y_i}{t_{i+1} - t_i}(x - t_i)$ 

Done.

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input t, y vectors of data
input evaluation location x
find interval i with x \in [t_i, t_{i+1}]
S = y_i + (x-t_i) m_i
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Interesting:

- input n + 1 data points  $t_0, \ldots, t_n, y_0, \ldots, y_n$
- in each interval we have  $S_i(x) = a_i x + b_i$
- 2 unknowns per interval [t<sub>i</sub>, t<sub>i+1</sub>]
- or 2n total unknowns
- the *n* + 1 pieces of input contraints *S*(*t<sub>i</sub>*) = *y<sub>i</sub>* gives 2 constraints per interval
- or 2n total constraints

#### definition

A function S(x) is a spline of degree 2 if:

- The domain of S(x) is an interval [a, b]
- 2 S(x) is continuous on [a, b]
- **③** S'(x) is continuous on [a, b]
- **③** There is a partition  $a = t_0 < t_1 < \cdots < t_n = b$  such that S(x) is quadratic on each subinterval  $[t_i, t_{i+1}]$ .

$$S(x) = \begin{cases} S_0(x) & x \in [t_0, t_1] \\ S_1(x) & x \in [t_1, t_2] \\ \vdots & \vdots \\ S_{n-1}(x) & x \in [t_{n-1}, t_n] \end{cases}$$

for each *i* we have

$$S_i(x) = a_i x^2 + b_i x + c_i$$

What are  $a_i$ ,  $b_i$ ,  $c_i$ ?

- 3 unknowns in each interval
- 3n total unknowns
- 2n constraints for matching up the input data (2 per interval)
- n-1 interior points require continuity of the derivative:  $S'_i(x_{i+1}) = S'_{i+1}(x_{i+1})$
- but this is just n-1 constraints
- total of 3n 1 constraints
- extra constraint:  $S'(x_0)$  =given, for example.

#### definition

A function S(x) is a spline of degree 3 if:

- The domain of S(x) is an interval [a, b]
- 2 S(x) is continuous on [a, b]
- 3 S'(x) is continuous on [a, b]
- S''(x) is continuous on [a, b]
- So There is a partition  $a = t_0 < t_1 < \cdots < t_n = b$  such that S(x) is cubic on each subinterval  $[t_i, t_{i+1}]$ .

$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4*n* unknowns

$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4n unknowns
- 2*n* constraints by continuity



$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4n unknowns
- 2n constraints by continuity
- n-1 constraints by continuity of S'(x)

$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4n unknowns
- 2n constraints by continuity
- n-1 constraints by continuity of S'(x)
- n-1 constraints by continuity of S''(x)

$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4n unknowns
- 2n constraints by continuity
- n-1 constraints by continuity of S'(x)
- n-1 constraints by continuity of S''(x)
- 4n-2 total constraints

$$S_i(x) = a_{0,i} + a_{1,i}x + a_{2,i}x^2 + a_{3,i}x^3$$

- n intervals, n + 1 knots, 4 unknowns per interval
- 4n unknowns
- 2n constraints by continuity
- n-1 constraints by continuity of S'(x)
- n-1 constraints by continuity of S''(x)
- 4n-2 total constraints
- This leaves 2 extra degrees of freedom. The cubic spline is not yet unique!

Some options:

- natural cubic spline:  $S''(t_0) = S''(t_n) = 0$
- fixed-slope:  $S'(t_0) = a$ ,  $S'(t_n) = b$
- not-a-knot: S'''(x) continuous at  $t_1$  and  $t_{n-1}$
- periodic: S' and S" are periodic at the ends

How do we find  $a_{0,i}, a_{1,i}, a_{2,i}, a_{3,i}$  for each *i*? Consider knots  $t_0, \ldots, t_n$ . Follow our

example with the following steps:

- Assume we knew  $S''(t_i)$  for each i
- 2  $S_i''(x)$  is linear, so construct it
- Get  $S_i(x)$  by integrating  $S''_i(x)$  twice
- Impose continuity
- **O** Differentiate  $S_i(x)$  to impose continuity on S'(x)

Assume we knew  $S''(t_i)$  for each i

#### We know S''(x) is continuous. So assume

$$z_i = S''(t_i)$$

(we don't actually know  $z_i$ , not yet at least)

 $S_i''(x)$  is linear, so construct it

Since  $S_i''(x)$  is linear, and

$$S_i''(t_i) = z_i$$
  
 $S_i''(t_{i+1}) = z_{i+1}$ 

we can write  $S_i''(x)$  as

$$S_i''(x) = z_i \frac{t_{i+1} - x}{t_{i+1} - t_i} + z_{i+1} \frac{x - t_i}{t_{i+1} - t_i}$$
$$= \frac{z_i}{h_i} (t_{i+1} - x) + \frac{z_{i+1}}{h_i} (x - t_i)$$

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Get  $S_i(x)$  by integrating  $S''_i(x)$  twice

Take

$$S_i''(x) = \frac{z_i}{h_i}(t_{i+1} - x) + \frac{z_{i+1}}{h_i}(x - t_i)$$

and integrate once:

$$S'_{i}(x) = -\frac{z_{i}}{2h_{i}}(t_{i+1} - x)^{2} + \frac{z_{i+1}}{2h_{i}}(x - t_{i})^{2} + \hat{C}_{i}$$

twice:

$$S_i(x) = \frac{z_i}{6h_i}(t_{i+1} - x)^3 + \frac{z_{i+1}}{6h_i}(x - t_i)^3 + \hat{C}_i x + \hat{D}_i$$

adjust:

$$S_i(x) = \frac{z_i}{6h_i}(t_{i+1} - x)^3 + \frac{z_{i+1}}{6h_i}(x - t_i)^3 + C_i(x - t_i) + D_i(t_{i+1} - x)$$

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Impose continuity

For each interval  $[t_i, t_{i+1}]$ , we require  $S_i(t_i) = y_i$  and  $S_i(t_{i+1}) = y_{i+1}$ :

$$y_{i} = S_{i}(t_{i}) = \frac{z_{i}}{6h_{i}}(t_{i+1} - t_{i})^{3} + \frac{z_{i+1}}{6h_{i}}(t_{i} - t_{i})^{3} + C_{i}(t_{i} - t_{i}) + D_{i}(t_{i+1} - t_{i})$$
$$= \frac{z_{i}}{6}h_{i}^{2} + D_{i}h_{i}$$
$$D_{i} = \frac{y_{i}}{h_{i}} - \frac{h_{i}}{6}z_{i}$$

and

$$y_{i+1} = S_i(t_{i+1}) = \frac{z_i}{6h_i}(t_{i+1} - t_{i+1})^3 + \frac{z_{i+1}}{6h_i}(t_{i+1} - t_i)^3 + C_i(t_{i+1} - t_i) + D_i(t_{i+1} - t_{i+1})$$
$$= \frac{z_{i+1}}{6}(h_i)^2 + C_ih_i$$
$$C_i = \frac{y_{i+1}}{h_i} - \frac{h_i}{6}z_{i+1}$$

Image: Image:

Impose continuity

#### So far we have

$$S_{i}(x) = \frac{z_{i}}{6h_{i}}(t_{i+1}-x)^{3} + \frac{z_{i+1}}{6h_{i}}(x-t_{i})^{3} + \left(\frac{y_{i+1}}{h_{i}} - \frac{h_{i}}{6}z_{i+1}\right)(x-t_{i}) + \left(\frac{y_{i}}{h_{i}} - \frac{h_{i}}{6}z_{i}\right)(t_{i+1}-x)^{3} + \frac{z_{i+1}}{6}(x-t_{i})^{3} + \left(\frac{y_{i+1}}{h_{i}} - \frac{h_{i}}{6}z_{i+1}\right)(x-t_{i}) + \frac{z_{i+1}}{6}(x-t_{i})^{3} + \frac{z$$

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Differentiate  $S_i(x)$  to impose continuity on S'(x)

$$S'_{i}(x) = -\frac{z_{i}}{2h_{i}}(t_{i+1} - x)^{2} + \frac{z_{i+1}}{2h_{i}}(x - t_{i})^{2} + \frac{y_{i+1}}{h_{i}} - \frac{h_{i}}{6}z_{i+1} - \frac{y_{i}}{h_{i}} + \frac{h_{i}}{6}z_{i}$$
  
need  $S'_{i}(t_{i}) = S'_{i-1}(t_{i})$ :

We need  $S'_{i}(t_{i}) = S'_{i-1}(t_{i})$ :

$$S'_{i}(t_{i}) = -\frac{h_{i}}{6}z_{i+1} - \frac{h_{i}}{3}z_{i} + \underbrace{\frac{1}{h_{i}}(y_{i+1} - y_{i})}_{b_{i}}$$

$$S_{i-1}'(t_i) = \frac{h_{i-1}}{6} z_{i-1} + \frac{h_{i-1}}{3} z_i + \underbrace{\frac{1}{h_{i-1}}(y_i - y_{i-1})}_{\underbrace{b_{i-1}}}$$

Thus  $z_i$  is defined by

$$h_{i-1}z_{i-1} + 2(h_i + h_{i-1})z_i + h_i z_{i+1} = 6(b_i - b_{i-1})$$

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 $z_i$  is defined by

$$h_{i-1}z_{i-1} + 2(h_i + h_{i-1})z_i + h_i z_{i+1} = 6(b_i - b_{i-1})$$

• This is n-1 equations, n-1 unknowns ( $z_0 = z_n = 0$  already) • an  $(n-1) \times (n-1)$  tridiagonal system  $\begin{bmatrix} 1 & & & & & & \\ h_0 & u_1 & h_1 & & & & \\ & h_1 & u_2 & h_2 & & & & \\ & h_2 & u_3 & h_3 & & & \\ & & \ddots & \ddots & \ddots & & \\ & & & h_{n-3} & u_{n-2} & h_{n-2} & \\ & & & & & & 1 \end{bmatrix} \begin{bmatrix} z_0 \\ z_1 \\ z_2 \\ z_3 \\ \vdots \\ z_{n-2} \\ z_{n-1} \\ z_n \end{bmatrix} = \begin{bmatrix} 0 \\ v_1 \\ v_2 \\ v_3 \\ \vdots \\ v_{n-2} \\ v_{n-1} \\ 0 \end{bmatrix}$  $u_i = 2(h_i + h_{i-1})$  $v_i = 6(b_i - b_{i-1})$ 

# example

Find the natural cubic spline for 
$$\begin{array}{c|c} x & -1 & 0 & 1 \\ \hline y & 1 & 2 & -1 \end{array}$$

**Determine** 
$$h_i$$
,  $b_i$ ,  $u_i$ ,  $v_i$ 

$$h = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad b = \begin{bmatrix} 1 \\ -3 \end{bmatrix} \quad u = \begin{bmatrix} 4 \end{bmatrix} \quad v = \begin{bmatrix} -24 \end{bmatrix}$$

Solve
$$\begin{bmatrix}
1 & & \\
1 & 4 & 1 \\
& & 1
\end{bmatrix}
\begin{bmatrix}
z_0 \\
z_1 \\
z_2
\end{bmatrix} =
\begin{bmatrix}
0 \\
-24 \\
0
\end{bmatrix}$$

Result:

$$\begin{bmatrix} z_0 \\ z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} 0 \\ -6 \\ 0 \end{bmatrix}$$

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#### example

Find the natural cubic spline for  $\frac{x \mid -1 \quad 0 \quad 1}{y \mid 1 \quad 2 \quad -1}$ 

• Plug  $z_i$  into

$$S_{i}(x) = \frac{z_{i}}{6h_{i}}(t_{i+1}-x)^{3} + \frac{z_{i+1}}{6h_{i}}(x-t_{i})^{3} + \left(\frac{y_{i+1}}{h_{i}} - \frac{h_{i}}{6}z_{i+1}\right)(x-t_{i}) + \left(\frac{y_{i}}{h_{i}} - \frac{h_{i}}{6}z_{i}\right)(t_{i+1}-x)$$

$$S(x) = \begin{cases} -(x+1)^3 + 3(x+1) - x & -1 \le x < 0\\ -(1-x)^3 - x + 3(1-x) & 0 \le x < 1 \end{cases}$$

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Compute for 
$$i = 0, \ldots, n-1$$

$$h_i = t_{i+1} - t_i$$
  $b_i = \frac{1}{h_i}(y_{i+1} - y_i)$ 

- 2 Set *u*, *v*:
- tridiagonal solve to get z
- Substitute into the nested form for S(x) equation 12, page 392 NMC6 (NMC5: equation 10 page 404)

- more on the cubic algorithm
- the B-splines/Bezier Curves