

- HERMITE INTERPOLATION
— match $f(x_i) \approx f'(x_i)$

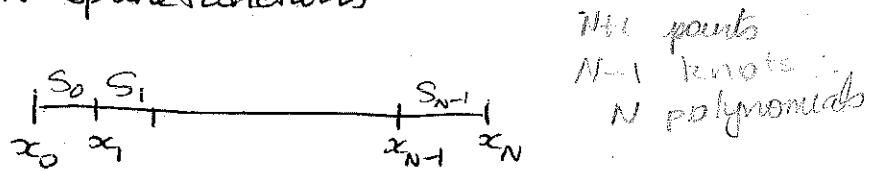
- SPLINES : 3.4

A spline function is a function that consists of polynomial pieces joined together by certain smoothness conditions.

Points where the function changes character - KNOTS

$N-1$ knots, N spline functions

LINEAR SPLINE



$2N$ unknowns \rightarrow 2 unknowns per line $\times n$ lines

Continued $\left\{ \begin{array}{l} S_k(x_k) = y_k \quad \text{for } k=0, \dots, N-1 \quad \text{left edge of line} \\ S_k(x_{k+1}) = y_{k+1} \quad \text{for } k=0, \dots, N-1 \quad \text{right node of line} \end{array} \right.$

$2N$ eqns.

QUADRATIC SPLINES

Continued $\left\{ \begin{array}{l} S_k(x_k) = y_k, \quad k=0, \dots, N-1 \\ S_k(x_{k+1}) = y_{k+1}, \quad k=0, \dots, N-1 \end{array} \right.$ — 3 unknowns per quadratic $\times n$ quadratics

N eq'n's
 N eq'n's

Continued $\left\{ S'_k(x_{k+1}) = S'_{k+1}(x_{k+1}), \quad k=0, \dots, N-2 \right.$ $N-1$ eq'n's
(interior knots)

Add the eq'n. $S'(x_0) = 0$ for example

or $S'(x_n) = f'(x_n)$

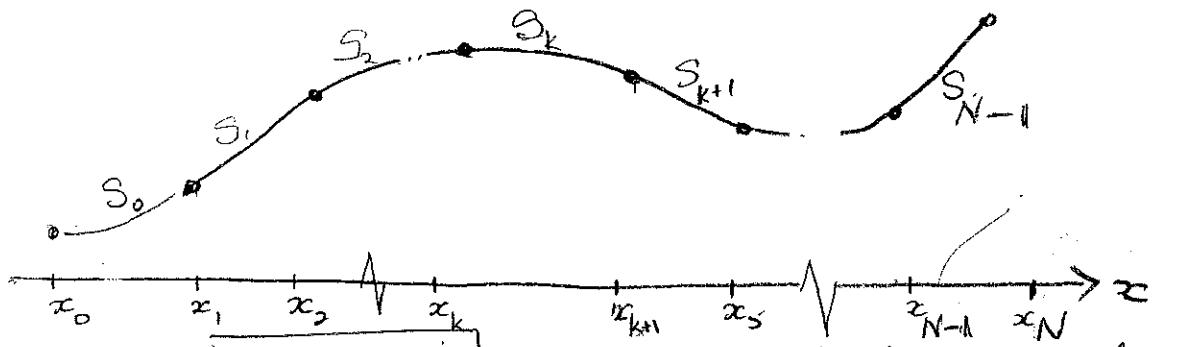
SPINE

①

CUBIC SPLINE INTERPOLATION

Oscillatory nature of high-degree polynomials restricts their use.

Divide interval $[a, b]$ into subintervals and use approx. polynomial on each subinterval — PIECEWISE POLYNOMIAL APPROXIMATION



splines: $\boxed{4N \text{ unknowns}}$ — 4 unknowns/cubic $\times N$ cubics

$$\left\{ S_k(x_k) = y_k \quad \text{For } k=0, 1, 2, \dots, N-1 \quad (\text{N eqns}) \right.$$

— CONTINUE S $\left\{ S_k(x_{k+1}) = y_{k+1} \quad \text{For } k=0, 1, 2, \dots, N-1 \quad (\text{N eqns}) \right.$

CONTINUE S' $S'_k(x_{k+1}) = S'_{k+1}(x_{k+1}) \quad \text{For } k=0, 1, 2, \dots, N-2 \quad (\text{N-1 eqns})$
Interior knots

CONTINUE S'' $S''_k(x_{k+1}) = S''_{k+1}(x_{k+1}) \quad \text{For } k=0, 1, 2, \dots, N-2 \quad (\text{N-1 eqns})$

$4N-2$ constraints — Need a more

Add : $S''(x_0) = 0 ; S''(x_N) = 0$ FREE OR NATURAL BOUNDARY
or

$S'(x_0) = \text{given} ; S'(x_N) = \text{given}$ CLAMPED BOUNDARY

FREE / NATURAL forcing a flexible elastic rod through the data points letting the stored at the ends be free to equilibrate to the position that minimizes oscillatory behaviour of curve.

WHY Cubic Splines?

- Discontinuities in the third derivative cannot be detected visually.
- Experience has shown that using splines of degree > 3 seldom yields any advantage.

Cubic Spline Smoothness

If S is the natural cubic spline function that interpolates $f \in C^2$ at the knots, then

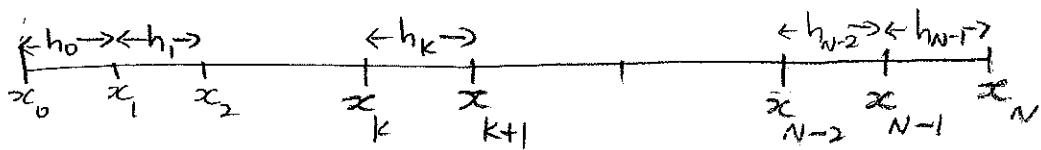
$$\int_a^b [S''(x)]^2 dx \leq \int_a^b [f''(x)]^2 dx$$

Avg Curvature of spline S is no larger than average curvature of any function which agrees w/ S at the knots

Oscillation \Rightarrow rapid change in first deriv \Rightarrow large 2nd deriv.
Cubic splines have "controlled" 2nd derivs.

Look at
`polyshow.m`
`splinetx.m`
`pchip.m`

- Changing data point has a global effect in that the entire curve is affected.



Given data (x_k, y_k) for $k=0, 1, \dots, N-1$

$$\text{Find } S_k(x) = a_k + b_k(x-x_k) + c_k(x-x_k)^2 + d_k(x-x_k)^3$$

which we use for $x_k \leq x \leq x_{k+1}$

for $k=0, 1, \dots, N-1$

i.e. Find the $4N$ unknowns a_k, b_k, c_k, d_k for $k=0, 1, \dots, N-1$
subject to the cubic spline conditions.

$$(i) \quad S_k(x_k) = y_k \text{ for } k=0, 1, \dots, N-1$$

$$\Rightarrow \boxed{a_k = y_k \text{ for } k=0, 1, \dots, N-1} \quad \text{Equation (i)}$$

$$(ii) \quad S_k(x_{k+1}) = y_{k+1} \quad \text{for } k=0, 1, \dots, N-1$$

$$\Rightarrow \boxed{a_k + h_k b_k + h_k^2 c_k + h_k^3 d_k = y_{k+1} \quad \text{for } k=0, 1, \dots, N-1} \quad \text{Equation (ii)}$$

$$(iii) \quad S'_k(x) = b_k + 2c_k(x-x_k) + 3d_k(x-x_k)^2$$

$$\text{So } S'_k(x_{k+1}) = S'_{k+1}(x_{k+1}) \text{ for } k=0, 1, \dots, N-2$$

$$\Rightarrow \boxed{b_k + 2c_k h_k + 3d_k h_k^2 = b_{k+1} \text{ for } k=0, 1, \dots, N-2} \quad \text{Equation (iii)}$$

$$(iv) \quad S''_k(x) = 2c_k + 6d_k(x-x_k)$$

$$\text{So } S''_k(x_{k+1}) = S''_{k+1}(x_{k+1}) \text{ for } k=0, 1, \dots, N-2$$

$$\Rightarrow 2c_k + 6d_k h_k = 2c_{k+1} \Rightarrow \boxed{c_k + 3d_k h_k = c_{k+1} \quad \text{for } k=0, \dots, N-2}$$

and $k=N-1$

Define $c_N \equiv \frac{1}{2} S''(x_N)$, so since $S''_{N-1}(x_N) = S''(x_N)$

Spline 4

$$\Rightarrow 2c_{N-1} + 6d_{N-1} h_{N-1} = 2c_N$$

$$\text{So } d_k = \frac{c_{k+1} - c_k}{3h_k} \quad \text{for } k=0, 1, \dots, N-2, N-1$$

eqn (iv')

where

$$c_N \equiv \frac{1}{2} S''(x_N)$$

So if you know c_k 's you can get your d_k 's

Putting (iv') & (i) in (ii) gives

$$y_k + h_k b_k + h_k^2 c_k + h_k^2 \frac{c_{k+1} - c_k}{3} = y_{k+1} \quad \text{for } k=0, 1, \dots, N-1$$

\Rightarrow

$$b_k = \frac{y_{k+1} - y_k}{h_k} - \left(\frac{c_{k+1} + 2c_k}{3} \right) h_k$$

for $k=0, 1, \dots, N-1$

(eqn ii')

So if you know
your c_k 's you
can get your
 b_k 's

$$\text{Eqn (iii) is } b_k + 2c_k h_k + 3d_k h_k^2 = b_{k+1} \quad \text{for } k=0, 1, \dots, N-2$$

$$\left[\frac{y_{k+1} - y_k}{h_k} - \left(\frac{c_{k+1} + 2c_k}{3} \right) h_k \right] + 2c_k h_k + \left[h_k (c_{k+1} - c_k) \right] = \frac{y_{k+2} - y_{k+1}}{h_{k+1}} - \left(\frac{c_{k+2} + c_{k+1}}{3} \right) h_k$$

$$\Rightarrow \frac{h_k}{3} c_k + \frac{2(h_k + h_{k+1})}{3} c_{k+1} + \frac{h_{k+1}}{3} c_{k+2} = \frac{y_{k+2} - y_{k+1}}{h_{k+1}} - \frac{y_{k+1} - y_k}{h_k}$$

for $k=0, 1, \dots, N-2$

These are $N-1$ linear eqns involving $N+1$ unknowns $c_0, c_1, \dots, c_{N-1}, c_N$

Add 2 more conditions - eg Boundary Conditions -
and Problem may give unique solution

$$h_k c_k + 2(h_k + h_{k+1})c_{k+1} + h_{k+1}c_{k+2} = \frac{3}{h_{k+1}}(y_{k+2} - y_{k+1})$$

$$h_0 c_0 + 2(h_0 + h_1)c_1 + h_1 c_2 = \frac{3}{h_1}(y_2 - y_1) - \frac{3}{h_0}(y_{k+1} - y_k)$$

for $k=0, 1, \dots, N-2$

$N-1$ eqns in $N+1$ unknowns c_0, c_1, \dots, c_N

where $c_N = \frac{1}{2} S''(x_N)$.
 (In general, $c_k = \frac{1}{2} S''(x_k)$)

Need 2 more conditions.

What will things look like in matrix-vector form?

$$\text{Solve } A\vec{x} = \vec{b}$$

$$\begin{matrix}
 & * \\
 & h_0 & 2(h_0+h_1) & h_1 & 0 \\
 & & h_1 & 2(h_1+h_2) & h_2 \\
 & & & \ddots & \vdots \\
 & & & h_{N-2} & 2(h_{N-2}+h_{N-1}) & h_{N-1} \\
 & & & & * & * \\
 & & & & & *
 \end{matrix}
 \quad
 \begin{bmatrix}
 c_0 \\
 c_1 \\
 c_2 \\
 \vdots \\
 c_{N-1} \\
 c_N
 \end{bmatrix}$$

\vec{x} \vec{b}
 $N+1$ eqns

$$\begin{array}{l}
 * \\
 \frac{y_2 - y_1}{h_1} - \frac{y_1 - y_0}{h_0} \\
 \frac{y_3 - y_2}{h_2} - \frac{y_2 - y_1}{h_1} \\
 \vdots \\
 \frac{y_N - y_{N-1}}{h_{N-1}} - \frac{y_{N-1} - y_{N-2}}{h_{N-2}}
 \end{array}$$

The other 2 conditions
 will complete the
 $(N+1) \times (N+1)$ matrix-vector
 system.

```
function [v,sigma] = splinetx(x,y,u)
%SPLINETX Textbook spline function.
% v = splinetx(x,y,u) finds the piecewise cubic interpolatory
% spline S(x), with S(x(j)) = y(j), and returns v(k) = S(u(k)).
% [v,sigma] = splinetx(...) also returns sigma(k) = S''(x(k))/6.
%
% Set up tridiagonal system.
% h = offdiagonal, d = diagonal, b = right hand side.
%
h = diff(x);
n = length(x);
k = (2:n-1)';
d = zeros(n,1);
d(k) = 2*(h(k-1) + h(k));
b = zeros(n,1);
b(k) = diff(diff(y)./h);
A = diag(h,-1) + diag(d,0) + diag(h,1);
%
% Not-a-knot end conditions.
%
A(1,1:3) = [h(2) -(h(2)+h(1)) h(1)];
A(n,n-2:n) = [h(n-1) -(h(n-2)+h(n-1)) h(n-2)];
%
% Solve tridiagonal system.
%
sigma = A\b;
%
% Find interval indices.
%
k = ones(size(u));
for j = 2:n-1
    k(u > x(j)) = j;
end
%
% Evaluate spline.
%
s = (u - x(k))./h(k);
t = 1-s;
sigma = reshape(sigma,size(y));
v = s.*y(k+1) + t.*y(k) + ...
    ((s.^3-s).*sigma(k+1) + (t.^3-t).*sigma(k)).*h(k).^2;
```