

Lecture 6

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Last time we had done osculatory interpolation.

x_0, x_1, \dots, x_m not necc. distinct pts

we say $f(x)$ and $g(x)$ agree on x_0, \dots, x_m if

$$f^{(j)}(z) = g^{(j)}(z) \quad \text{for } j=0, 1, \dots, r-1$$

for any z which appears r times

in x_0, x_1, \dots, x_m

Theorem If $f(x)$ has r continuous derivatives and no point in the sequence x_0, x_1, \dots, x_m occur more than r times, then there exists exactly one polynomial $P_m(x)$ of degree $\leq m$ which agrees with $f(x)$ at x_0, x_1, \dots, x_m .

$f[x_0, x_1, \dots, x_m] = \text{coeff of } x^m \text{ in } P_m(x)$

$$P_m(x) = P_{m-1}(x) + f[x_0, \dots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

Special cases

1) $x_0 = x_1 = x_2 = \dots = x_m$

$P_m(x) =$ Taylor polynomial with center x_0

$$= f(x_0) + f'(x_0)(x-x_0) + \frac{f''(x_0)}{2}(x-x_0)^2 \\ + \dots + \frac{f^{(m)}(x_0)}{m!}(x-x_0)^m$$

$$\text{Thus } \underbrace{f[x_0, x_0, \dots, x_0]}_{m+1 \text{ times}} = \frac{f^{(m)}(x_0)}{m!}$$

This case is $\underbrace{x_0, x_0, \dots, x_0}_{m+1 \text{ times}}$

Special case 2

x_0, x_1, \dots, x_k

$k+1$ distinct
pts

$f(x_0), f(x_1), \dots, f(x_k)$

$f'(x_0), f'(x_1), \dots, f'(x_k)$ is also given

need to find polynomial $p(x)$ of degree $\leq 2k+1$
such that $p(x_i) = f(x_i)$ & $p'(x_i) = f'(x_i) \forall i$

Algorithm

Let $y_0 = x_0, y_1 = x_0$

$y_2 = x_1, y_3 = x_1$

\vdots

$y_{2i} = x_i, y_{2i+1} = x_i$

\vdots

$y_{2k} = x_k, y_{2k+1} = x_k$

We are looking
for a polynomial which
agrees with $f(x)$ at

$y_0, y_1, \dots, y_{2k}, y_{2k+1}$

Construct divided diff table

"use $f[a, a] = f'(a)$ "

$$p(x) = f[y_0] + \sum_{i=1}^{2k+1} f[y_0, y_1, \dots, y_i] \prod_{j=0}^{i-1} (x - y_j)$$

Example a, b distinct pts

$$f(a), f(b) \quad f'(a), f'(b)$$

$$P_3(x) = f(a) + f[a, a](x-a) + f[a, a, b](x-a)^2 + f[a, a, b, b](x-a)^2(x-b)$$

We prove by direct computation
that $P_3(x)$ agree with $f(x)$ at a, a, b, b

$$f[a, a] = f'(a)$$

clearly $P_3(a) = f(a)$, $P_3'(a) = \frac{f[a, a]}{1} = f'(a)$

$$f[a, b] = \frac{f(b) - f(a)}{b - a}$$

$$f[a, a, b] = \frac{f[a, b] - f[a, a]}{b - a}$$

$$= \frac{\frac{f(b) - f(a)}{b - a} - f'(a)}{b - a}$$

$$= \frac{f(b) - f(a) - (b - a)f'(a)}{(b - a)^2}$$

$$\begin{aligned}
 P_3(b) &= f(a) + f'(a)(b-a) + \\
 &\quad + \frac{f(b)-f(a)-(b-a)f'(a)}{(b-a)^2} \cdot \cancel{(b-a)^2} \\
 &= f(b)
 \end{aligned}$$

$$\begin{aligned}
 f[a, b, b] &= \frac{f[b, b] - f[a, b]}{b-a} \\
 &= \frac{f'(b) - \frac{f(b)-f(a)}{b-a}}{b-a} \\
 &= \frac{f'(b)(b-a) - (f(b)-f(a))}{(b-a)^2}
 \end{aligned}$$

$$\begin{aligned}
 f[a, a, b, b] &= \frac{f[a, b, b] - f[a, a, b]}{b-a} \\
 &= \frac{f'(b)(b-a) - (f(b)-f(a)) - (f'(a)(b-a) - (f(b)-f(a)))}{(b-a)^3} \\
 &= \frac{(f'(b) + f'(a))(b-a) - 2(f(b)-f(a))}{(b-a)^3}
 \end{aligned}$$

$$P_3(x) = f(a) + f[a,a](x-a) + f[a,a,b](x-a)^2 + f[a,a,b,b](x-a)^2(x-b)$$

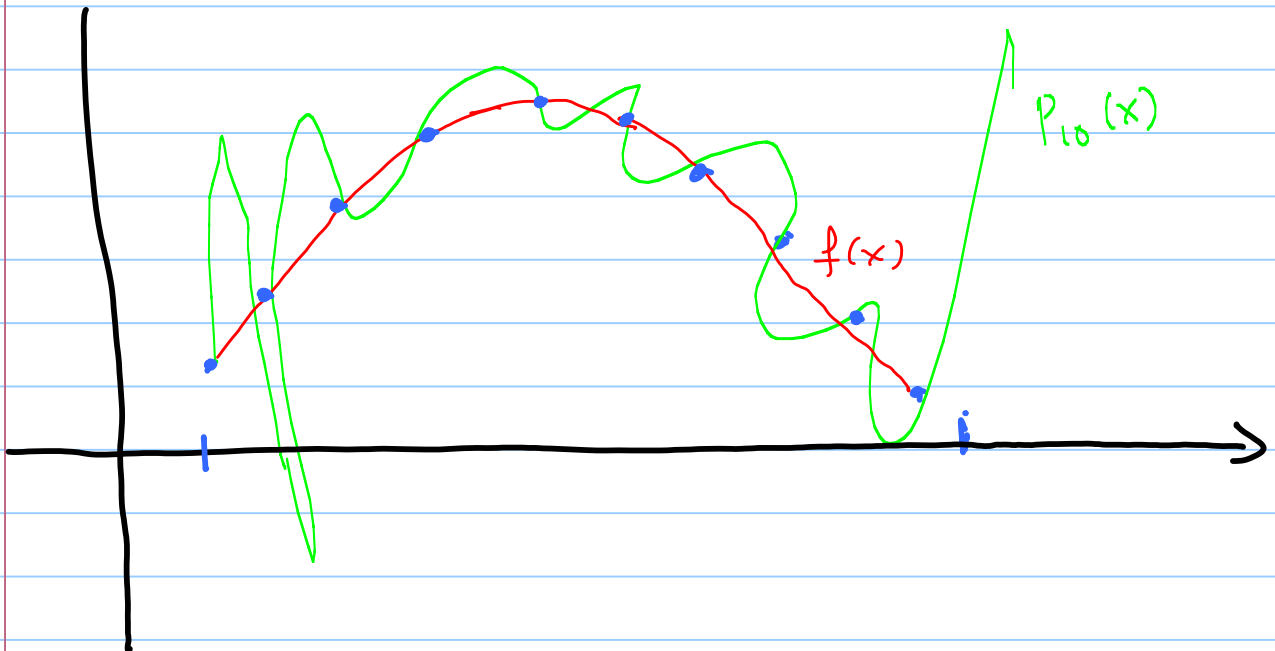
$$\begin{aligned} P_3'(b) &= f'(a) + 2f[a,a,b](b-a) + f[a,a,b,b](b-a)^2 \\ &= f'(a) + 2 \underbrace{\{f(b) - f(a) - (b-a)f'(a)\}}_{b-a} \\ &\quad + \underbrace{\{f'(b) + f'(a)(b-a) - 2(f(b) - f(a))\}}_{b-a} \\ &= f'(b) \end{aligned}$$

Disadvantages of interpolation

Note that if x_0, x_1, \dots, x_k are
pts in $[a, b]$
then interpolating polynomial
has degree k .

In practice k is large

However a polynomial of degree k
(k large) oscillates a lot



for example if there are 101 data pts, then it is not advisable to work with a degree 100 interpolating formula as this also creates lot of round-off error

Strategy

Use piecewise-polynomial approximation

Simplest-case

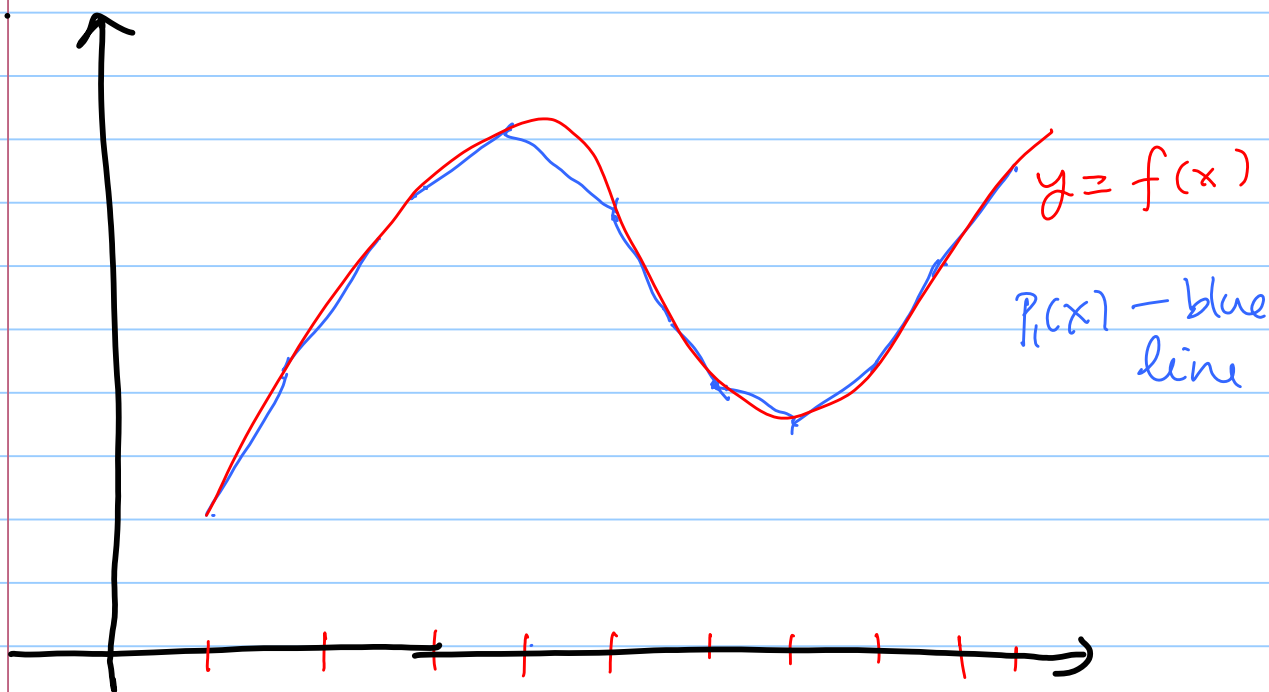
Piecewise-linear interpolation

$$a = x_1 < x_2 < x_3 < \dots < x_{N+1} = b$$

$f(\bar{x})$ is approximated at a pt \bar{x} by locating the interval $[x_k, x_{k+1}]$ containing \bar{x} and then taking

$$P_1(\bar{x}) = f(x_k) + f[x_k, x_{k+1}](\bar{x} - x_k)$$

Graphical representation



note that if $|x_i - x_{i+1}|$ is small
then $P_i(x)$ approximates $f(x)$ to a
a high degree.

If we use higher degree (say cubic)
piecewise-polynomial approximation then
we get a better approximation.

Construction of piecewise-cubic
function $g_3(x)$ which interpolates
 $f(x)$ at the pts x_1, \dots, x_{N+1} where

$$a = x_1 < x_2 < \dots < x_{N+1} = b$$

on each $[x_i, x_{i+1}]$ we construct
 $g_3(x)$ as a cubic polynomial $P_i(x)$
 $i=1, \dots, N$

$$P_i(x) = C_{1,i} + C_{2,i}(x-x_i) + C_{3,i}(x-x_i)^2 \\ + C_{4,i}(x-x_i)^3 \\ i=1, \dots, N$$

Since $g_3(x_i) = f(x_i)$ for $i=1, \dots, N+1$

We have
 $P_i(x_i) = f(x_i)$ & $P_i(x_{i+1}) = f(x_{i+1})$
 $i=1, \dots, N$

In particular

$$P_{i-1}(x_i) = P_i(x_i) = f(x_i) \\ i = 2, \dots, N$$

So $g_3(x)$ is continuous on $[a, b]$.

Only constraint for $P_i(x)$

$$i, P_i(x_i) = f(x_i)$$

$$\text{and } P_i(x_{i+1}) = f(x_{i+1})$$

So we have some freedom in choosing the $P_i(x)$

We study 2 cases

1) Piecewise-cubic Hermite interpolation

2) Cubic-spline interpolation

1) piecewise-cubic Hermite interpolation

One determines $P_i(x)$ so as to interpolate $f(x)$ at $x_i, x_i, x_{i+1}, x_{i+1}$

i.e., $P_i'(x_i) = f'(x_i)$ & $P_i'(x_{i+1}) = f'(x_{i+1})$

By Newton's formula

$$P_i(x) = f(x_i) + f[x_i, x_i](x - x_i) + f[x_i, x_i, x_{i+1}](x - x_i)^2 + f[x_i, x_i, x_{i+1}, x_{i+1}](x - x_i)^2(x - x_{i+1})$$

note $(x - x_{i+1}) = (x - x_i) + (x_i - x_{i+1})$

$$\begin{aligned} \text{So } P_i(x) &= f(x_i) + f'(x_i)(x - x_i) + \\ &+ \left(f[x_i, x_i, x_{i+1}] - f[x_i, x_i, x_{i+1}, x_{i+1}] \Delta x_i \right) (x - x_i)^2 \\ &+ f[x_i, x_i, x_{i+1}, x_{i+1}](x - x_i)^3 \end{aligned}$$

Algorithm

$$f_i = f(x_i)$$

$$\Delta x_i = x_{i+1} - x_i$$

$$g_i = f'(x_i)$$

$$i = 1, \dots, N+1$$

$$C_{1,i} = f_i$$

$$C_{2,i} = g_i$$

$$C_{3,i} = f[x_i, x_i, x_{i+1}] - f[x_i, x_i, x_{i+1}, x_{i+1}] \Delta x_i$$

$$= \frac{f[x_i, x_{i+1}] - g_i}{\Delta x_i} - C_{4,i} \Delta x_i$$

$$C_{4,i} = f[x_i, x_i, x_{i+1}, x_{i+1}]$$

$$= \frac{f[x_i, x_{i+1}, x_{i+1}] - f[x_i, x_i, x_{i+1}]}{\Delta x_i}$$

$$= \frac{g_{i+1} + g_i - 2f[x_i, x_{i+1}]}{\Delta x_i}$$

Example

$a = x_1$	x_2	x_3	$x_4 = b$
$f(x_1)$	$f(x_2)$	$f(x_3)$	$f(x_4)$
$f'(x_1)$	$f'(x_2)$	$f'(x_3)$	$f'(x_4)$

Construction of $g_3(x)$
in $[x_1, x_2]$

$$g_3(x) = p_1(x) = f[x_1] + f[x_1, x_1](x - x_1) + f[x_1, x_1, x_2](x - x_1)^2 + f[x_1, x_1, x_2, x_2](x - x_1)^2(x - x_2)$$

$g_4[x_2, x_3]$

$$g_4(x) = p_2(x) = f[x_2] + f[x_2, x_2](x - x_2) + f[x_2, x_2, x_3](x - x_2)^2 + f[x_2, x_2, x_3, x_3](x - x_2)^2(x - x_3)$$

$g_5[x_3, x_4]$

$$g_5(x) = p_3(x) = f[x_3] + f[x_3, x_3](x - x_3) + f[x_3, x_3, x_4](x - x_3)^2 + f[x_3, x_3, x_4, x_4](x - x_3)^2(x - x_4)$$

This method is used in numerical solution to first order differential equations.

Example

$$\frac{dy}{dx} = y - x^2 + 1 \quad 0 \leq x \leq 1$$

$$y(0) = 0.5$$

→ computed by a numerical method

x_i	$y(x_i)$	$y'(x_i)$
0	0.500	1.5
0.2	0.826	1.786
0.4	1.207	2.047
0.6	1.637	2.277
0.8	2.110	2.470
1.0	2.618	2.618

find $y(0.7)$, $y(0.9)$

usual Osculatory interpolating polynomial
has degree 11

So we use piecewise Hermite interpolation

x	$y(x)$	$f[,]$	$f[, ,]$	$f[, , ,]$
0.6	1.637	2.277	4.4 E-1	4.25 E-1
0.6	1.637	2.365	5.25 E-1	
0.8	2.110	2.470		
0.8	2.110			

$$p_4(x) = 1.637 + 2.277(x-0.6) + 4.4 \text{ E-1} (x-0.6)^2 + 4.25 \text{ E-1} (x-0.6)^2 (x-0.8)$$

$$p_4(0.7) = 1.869$$

$$p_4(0.9) = ?$$

x	$y(x)$	$f[.]$	$f[.,.]$	$f[.,.,.]$
0.8	2.11	2.47	0.35	0.2
0.8	2.11	2.54	0.39	
1.0	2.618	2.618		
1.0	2.618			

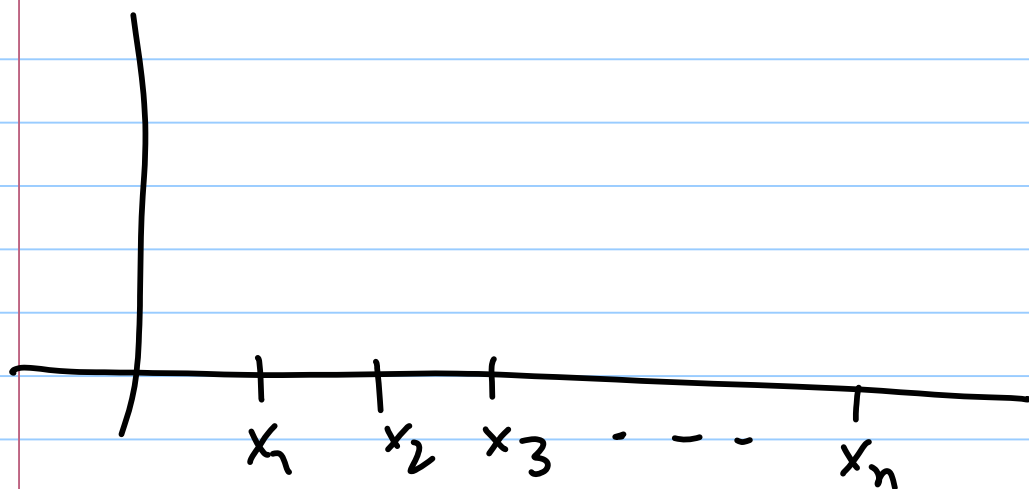
$$\begin{aligned}
 P_5(x) = & f[0.8] + f[0.8, 0.8](x-0.8) \\
 & + f[0.8, 0.8, 1.0](x-0.8)^2 \\
 & + f[0.8, 0.8, 0.9, 0.9](x-0.8)^2(x-1)
 \end{aligned}$$

$$\begin{aligned}
 P_5(0.9) &= 2.3603 \\
 &= 2.360 \text{ in 4 sig digits}
 \end{aligned}$$

Disadvantages of Hermite interpolation

It requires knowledge of $f'(x_1), \dots, f'(x_{n+1})$.

This is not always available.



$g_3(x)$ cubic spline
 $g_3(x)$ is continuously twice differentiable

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