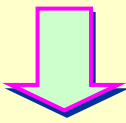
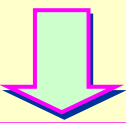


再提思想来源:

$n = 1:$

x	x_0	x_1
$f(x)$	$f(x_0)$	$f(x_1)$

x	x_0	x_1	\dots	x_n
$f(x)$	$f(x_0)$	$f(x_1)$	\dots	$f(x_n)$



2点式 $L_1(x) = \frac{x - x_1}{x_0 - x_1} f(x_0) + \frac{x - x_0}{x_1 - x_0} f(x_1)$

Lagrange 插值公式 $L_n(x)$

余项 $R_n(x)$

点斜式 $N_1(x) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x - x_0)$

Newton 插值公式 $N_n(x)$

待定系数法 $y = P_1(x) = \frac{x_1 f(x_0) - x_0 f(x_1)}{x_1 - x_0} + \frac{f(x_1) - f(x_0)}{x_1 - x_0} x$

唯一性
存在性

看来3个都有用。

§3 牛顿插值 /* Newton' s Interpolation */



Lagrange 插值虽然易算, 但若要增加一个节点时, 全部基函数 $l_i(x)$ 都需重新算过。



将 $L_n(x)$ 改写成 $c_0 + c_1(x - x_0) + c_2(x - x_0)(x - x_1) + \dots + c_n(x - x_0)\dots(x - x_{n-1})$ 的形式, 希望每加一个节点时, 只附加一项上去即可。

➤ 牛顿插值 /* Newton' s Interpolation */

1. 公式构造: $N_n(x) = ?$

推导思想: 递增式构造, 故具有承袭性。

$n = 0$: $N_0(x_0) = f(x_0) \Rightarrow N_0(x) = f(x_0)$ 0次多项式

$$n = 1: \begin{array}{c|ccc} x & x_0 & x_1 & x_2 & x_3 & \cdots & x_n \\ f(x) & f(x_0) & f(x_1) & f(x_2) & f(x_3) & \cdots & f(x_n) \end{array} \quad \begin{cases} N_1(x_0) = f(x_0) \\ N_1(x_1) = f(x_1) \end{cases}$$

设 $N_1(x) = N_0(x) + A(x - x_0) = f(x_0) + A(x - x_0)$ 则 $\Rightarrow \begin{cases} N_1(x_0) = f(x_0) \end{cases}$

由 $N_1(x_1) = f(x_0) + A(x_1 - x_0)$

得 $A = \frac{f(x_1) - f(x_0)}{x_1 - x_0} \triangleq f[x_0, x_1]$

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

有 $N_1(x) = f(x_0) + f[x_0, x_1](x - x_0)$

$n = 2:$

$$\begin{array}{c|ccc} x & x_0 & x_1 & x_2 & x_3 & \cdots & x_n \\ f(x) & f(x_0) & f(x_1) & f(x_2) & f(x_3) & \cdots & f(x_n) \end{array}$$

$$\begin{cases} N_2(x_0) = f(x_0) \\ N_2(x_1) = f(x_1) \\ N_2(x_2) = f(x_2) \end{cases}$$

设 $N_2(x) = N_1(x) + B(x - x_0)(x - x_1)$ 则 $\Rightarrow \begin{cases} N_2(x_0) = f(x_0) \\ N_2(x_1) = f(x_1) \end{cases}$

2024/10/21 由 $N_2(x_2) = N_1(x_2) + B(x_2 - x_0)(x_2 - x_1) = f(x_2)$

$n = 2$:

x	x_0	x_1	x_2	x_3	\cdots	x_n
$f(x)$	$f(x_0)$	$f(x_1)$	$f(x_2)$	$f(x_3)$	\cdots	$f(x_n)$

$$\begin{cases} N_2(x_0) = f(x_0) \\ N_2(x_1) = f(x_1) \\ N_2(x_2) = f(x_2) \end{cases}$$

设 $N_2(x) = N_1(x) + B(x - x_0)(x - x_1)$ 则 \rightarrow $\begin{cases} N_2(x_0) = f(x_0) \\ N_2(x_1) = f(x_1) \end{cases}$

由 $N_2(x_2) = N_1(x_2) + B(x_2 - x_0)(x_2 - x_1) = f(x_2)$

得 $B = \frac{f(x_2) - N_1(x_2)}{(x_2 - x_0)(x_2 - x_1)} = \frac{f(x_2) - (f(x_0) + f[x_0, x_1](x_2 - x_0))}{(x_2 - x_0)(x_2 - x_1)}$

$$= \frac{f(x_2) - f(x_0) - f[x_0, x_1](x_2 - x_0)}{(x_2 - x_0)(x_2 - x_0)} = \frac{\frac{f(x_2) - f(x_0)}{x_2 - x_0} - f[x_0, x_1]}{x_2 - x_1}$$

$$= \frac{f[x_0, x_2] - f[x_0, x_1]}{x_2 - x_1} \triangleq f[x_0, x_1, x_2]$$

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

得 $N_2(x) = f(x_0) + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1)$

一般地, 对于:

x	x_0	x_1	\cdots	x_n
$f(x)$	$f(x_0)$	$f(x_1)$	\cdots	$f(x_n)$

$$N_n(x_i) = f(x_i), i = 0, 1, \cdots, n$$

Newton插值是通过选取特殊的基函数实现的, 这时, 取

$$\varphi_0(x) = 1$$

$$f(x_0) = N_n(x_0) = c_0$$

$$\begin{aligned} f(x_1) &= N_n(x_1) = c_0 + c_1(x_1 - x_0) \\ &= f(x_0) + c_1(x_1 - x_0) \end{aligned}$$

$$\text{得 } c_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0} = f[x_0, x_1]$$

$$\begin{aligned} N_n(x) &= c_0 + c_1(x - x_0) + c_2(x - x_0)(x - x_1) \\ &\quad + \cdots + c_n(x - x_0)(x - x_1) \cdots (x - x_{n-1}) \end{aligned}$$

其中 c_0, c_1, \cdots, c_n 是待定系数, 由插值条件 (2.1) 决定。

➤ 差商(亦称均差) /* divided difference */

$$f[x_i, x_j] = \frac{f(x_i) - f(x_j)}{x_i - x_j} \quad (i \neq j, x_i \neq x_j)$$

1阶差商

什么是差商???

先相减?

后相除?

$$f[x_i, x_j, x_k] = \frac{f[x_i, x_j] - f[x_j, x_k]}{x_i - x_k} \quad (i \neq k)$$

2阶差商

(k+1)阶差商:

$$\begin{aligned} f[x_0, \dots, x_{k+1}] &= \frac{f[x_0, x_1, \dots, x_k] - f[x_1, \dots, x_k, x_{k+1}]}{x_0 - x_{k+1}} \\ &= \frac{f[x_0, \dots, x_{k-1}, x_k] - f[x_0, \dots, x_{k-1}, x_{k+1}]}{x_k - x_{k+1}} \end{aligned}$$

注: 阶差商必须由 $k + 1$ 个节点构成, k 个节点是构造不出 k 阶差商的。

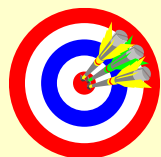
为统一 起见,补充定义函数 $f(x_0)$ 为零阶差商。

用归纳法易证, 一般地, 对于:

x	x_0	x_1	\cdots	x_n
$f(x)$	$f(x_0)$	$f(x_1)$	\cdots	$f(x_n)$

, $N_n(x_i) = f(x_i), i = 0, 1, \cdots, n$

通过插值条件运用数学归纳法 (P29-30) 可以求得

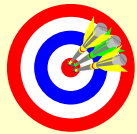


$$c_k = f[x_0, x_1, \cdots, x_k]$$

因此就得到下列的满足插值条件(2.1)的 n 次插值多项式



$$N_n(x) = f(x_0) + f[x_0, x_1](x - x_0) + \\ f[x_0, x_1, x_2](x - x_0)(x - x_1) + \cdots + \\ f[x_0, x_1, \cdots, x_n](x - x_0)(x - x_1) \cdots (x - x_{n-1})$$



$$N_n(x) = c_0 + c_1(x - x_0) + c_2(x - x_0)(x - x_1) + \cdots + c_n(x - x_0) \cdots (x - x_{n-1})$$

$$\left\{ \begin{array}{l} f(x) = f(x_0) + (x - x_0)f[x, x_0] \cdots \cdots \cdots \textcircled{1} \\ f[x, x_0] = f[x_0, x_1] + (x - x_1)f[x, x_0, x_1] \cdots \cdots \cdots \textcircled{2} \\ \dots \dots \dots \dots \\ f[x, x_0, \dots, x_{n-1}] = f[x_0, \dots, x_n] + (x - x_n)f[x, x_0, \dots, x_n] \cdots \cdots \textcircled{n-1} \end{array} \right.$$

$$\textcircled{1} + (x - x_0) \times \textcircled{2} + \dots \dots + (x - x_0) \cdots (x - x_{n-1}) \times \textcircled{n-1}$$

$$\begin{aligned} \Rightarrow f(x) = & f(x_0) + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1) + \dots \\ & + f[x_0, \dots, x_n](x - x_0) \cdots (x - x_{n-1}) \\ & + f[x, x_0, \dots, x_n](x - x_0) \cdots (x - x_{n-1})(x - x_n) \end{aligned}$$

$N_n(x)$

$$c_i = f[x_0, \dots, x_i]$$

$R_n(x)$

差商的值与 x_i 的顺序无关!



📖 差商性质:

1). k 阶差商可表为函数值 $f(x_0), f(x_1), \dots, f(x_k)$ 的线性组合, 即

$$f[x_0, \dots, x_k] = \sum_{j=0}^k \frac{f(x_j)}{(x_j - x_0) \cdots (x_j - x_{j-1})(x_j - x_{j+1}) \cdots (x_j - x_k)} = \sum_{j=0}^k \frac{f(x_j)}{\omega'_{k+1}(x_j)}$$

2). 差商具有对称性, 即

$$f[x_0, x_1, \dots, x_k] = f[x_1, x_0, x_2, \dots, x_k] = \cdots = f[x_1, \dots, x_k, x_0]$$

3). 若 $f(x)$ 是 n 次多项式, 则其 k 阶差商 $f[x_0, x_1, \dots, x_{k-1}, x]$

当 $k \leq n$ 时是一个 $n - k$ 次多项式, 而当 $k > n$ 时恒为零.

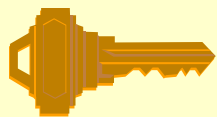
4).若 $f(x)$ 在 $[a, b]$ 上有 n 阶导数, 且节点 $x_i \in [a, b] (i = 0, 1, \dots, n)$, 则 n 阶差商与 n 阶导数有如下关系式:

$$f[x_0, x_1, \dots, x_n] = \frac{f^{(n)}(\xi)}{n!} \quad \xi \in [a, b]$$

注: 由唯一性可知 $N_n(x) \equiv L_n(x)$, 只是算法不同, 故其余项也相同, 即

$$f[x, x_0, \dots, x_n] \omega_{n+1}(x) = \frac{f^{(n+1)}(\xi_x)}{(n+1)!} \omega_{n+1}(x)$$

$$\Rightarrow f[x_0, \dots, x_n] = \frac{f^{(n)}(\xi)}{n!}, \quad \xi \in (x_{\min}, x_{\max})$$



差商计算可列差商表如下

x_i	$f(x_i)$	一阶差商	二阶差商	三阶差商
x_0	$f(x_0)$			
x_1	$f(x_1)$	$f[x_0, x_1]$		
x_2	$f(x_2)$	$f[x_1, x_2]$	$f[x_0, x_1, x_2]$	
x_3	$f(x_3)$	$f[x_2, x_3]$	$f[x_1, x_2, x_3]$	$f[x_0, x_1, x_2, x_3]$
\dots	\dots			\dots

$$\frac{f(x_1, x_2, x_3) - f(x_0, x_1, x_2)}{x_3 - x_0}$$

例 已知函数 $f(x)$ 在各节点处的函数值如下,用Newton插值法求 $f(0.596)$ 的值.

x_k	0.40	0.55	0.65	0.80
$f(x_k)$	<u>0.41075</u>	0.57815	0.69675	0.88811

解:

x_k	$f(x_k)$	一阶	二阶	三阶
0.40	<u>0.41075</u>			
0.55	0.57815	<u>1.11600</u>		
0.65	0.69675	1.18600	<u>0.28000</u>	
0.80	0.88811	1.27573	0.35893	<u>0.19733</u>

$$N_3(x) = 0.41075 + 1.11600(x - 0.4) + 0.28000(x - 0.4)(x - 0.55) + 0.19733(x - 0.4)(x - 0.55)(x - 0.65)$$



§4 埃尔米特插值 /* Hermite Interpolation */



不仅要求函数值重合，而且要求若干阶**导数**也重合。

即：要求插值函数 $\varphi(x)$ 满足 $\varphi(x_i) = f(x_i)$, $\varphi'(x_i) = f'(x_i)$,
..., $\varphi^{(m_i)}(x_i) = f^{(m_i)}(x_i)$.

注：  N 个条件可以确定 $N-1$ 阶多项式。

 要求在**1**个节点 x_0 处直到 m_0 阶导数都重合的插值多项式即为**Taylor多项式**

$$\varphi(x) = f(x_0) + f'(x_0)(x - x_0) + \dots + \frac{f^{(m_0)}(x_0)}{m_0!} (x - x_0)^{m_0}$$

其余项为 $R(x) = f(x) - \varphi(x) = \frac{f^{(m_0+1)}(\xi)}{(m_0+1)!} (x - x_0)^{(m_0+1)}$

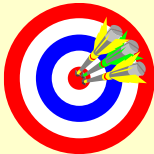
✂ 低次埃尔米特插值多项式

➤ 1. 二点三次埃尔米特插值多项式

设给定区间 $[x_0, x_1]$ 两端点处的函数值与导数值如下：

x	x_0	x_1
$f(x)$	y_0	y_1
$f'(x)$	m_0	m_1

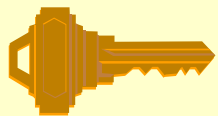
求插值多项式 $H_3(x)$, 使满足



$$\begin{cases} H_3(x_0) = y_0, H_3(x_1) = y_1 \\ H_3'(x_0) = m_0, H_3'(x_1) = m_1 \end{cases}$$

由于给出了四个条件，故可唯一确定出一个三次多项式，不妨设

$$H_3(x) = \alpha_0(x)y_0 + \alpha_1(x)y_1 + \beta_0(x)m_0 + \beta_1(x)m_1$$



因此问题归结为构造

$\alpha_0(x)$ 与 $\beta_0(x)$, $\alpha_1(x)$ 与 $\beta_1(x)$

首先, $\alpha_0(x)$ 与 $\beta_0(x)$, $\alpha_1(x)$ 与 $\beta_1(x)$ 应该为三次式。

再者, 由插值条件易知 $\alpha_i(x_j), \beta_i(x_j)$ $i, j = 1, 2$ 应该满足条件

$$\left\{ \begin{array}{l} \alpha_i(x_j) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases} \quad \alpha'_i(x_j) = 0 \quad (i, j = 0, 1) \\ \beta_i(x_j) = 0 \quad \beta'_i(x_j) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases} \end{array} \right.$$

取

$$l^2_i(x) = \left(\frac{x - x_j}{x_i - x_j} \right)^2, (i, j = 0, 1; i \neq j)$$

令

$$\alpha_i(x) = (a_i x + b_i) l_i^2(x)$$

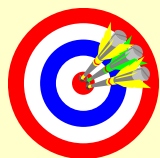
$$\beta_i(x) = c_i (x - x_i) l_i^2(x), i = 0, 1;$$

其中 a_i, b_i, c_i 为待定系数, 由 $\alpha_0(x_0) = 1, \alpha'_0(x_0) = 0$ 知 a_0, b_0

满足

$$\begin{cases} a_0 x_0 + b_0 = 1 \\ a_0 + 2(a_0 x_0 + b_0) l'_0(x_0) = 0 \end{cases}$$

解之得



$$a_0 = -2l'_0(x_0) = -2 \frac{1}{x_0 - x_1}$$

$$b_0 = 1 + 2x_0 l'_0(x_0) = 1 + 2x_0 \frac{1}{x_0 - x_1}$$

故

$$\alpha_0(x) = \left(1 - 2 \frac{x - x_0}{x_0 - x_1}\right) \left(\frac{x - x_1}{x_0 - x_1}\right)^2$$

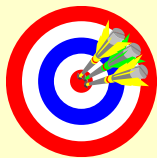
类似可以求得

$$\alpha_1(x) = \left(1 - 2 \frac{x - x_1}{x_1 - x_0}\right) \left(\frac{x - x_0}{x_1 - x_0}\right)^2,$$

$$\beta_0(x) = (x - x_0) \left(\frac{x - x_1}{x_0 - x_1}\right)^2, \quad \beta_1(x) = (x - x_1) \left(\frac{x - x_0}{x_1 - x_0}\right)^2$$

于是

$$\begin{aligned} H_3(x) = & y_0 \left(1 - 2 \frac{x - x_0}{x_0 - x_1}\right) l_0^2(x) + y_1 \left(1 - 2 \frac{x - x_1}{x_1 - x_0}\right) l_1^2(x) \\ & + m_0 (x - x_0) l_0^2(x) + m_1 (x - x_1) l_1^2(x) \end{aligned}$$



➤ 2.三点三次带一个导数值的插值多项式


假设给定的函数表如下：

x	x_0	x_1	x_2
$f(x)$	y_0	y_1	y_2
$f'(x)$		m_1	

要求三次多项式 $H_3(x)$,使

$$\begin{cases} H_3(x_i) = y_i \\ H'_3(x_1) = m_1 \end{cases} \quad (i = 0, 1, 2)$$

利用满足三个条件的Newton插值多项式，我们设



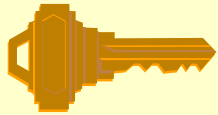
$$H_3(x) = y_0 + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1) + k(x - x_0)(x - x_1)(x - x_2)$$

其中 k 为待定系数，显然 $H_3(x)$ 满足前三个插值条件，利用第四个条件确定常数 k ，于是

$$H_3'(x_1) = f[x_0, x_1] + f[x_0, x_1, x_2](x_1 - x_0) + k(x_1 - x_0)(x_2 - x_1) = m_1$$



$$k = \frac{m_1 - f[x_0, x_1] - f[x_0, x_1, x_2](x_1 - x_0)}{(x_0 - x_1)(x_1 - x_2)}$$



将其代入，即可得到 $H_3(x)$ 的表达式：

$$H_3(x) = y_0 + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1) + \frac{m_1 - f[x_0, x_1] - f[x_0, x_1, x_2](x_1 - x_0)}{(x_0 - x_1)(x_1 - x_2)}(x - x_0)(x - x_1)(x - x_2)$$

例： 插值条件中有导数值，此时插值余项？

设有函数表

x	0	1	2
y	0	2	14
y'		5	

，求插值多项式，并写出余项表达式。

解： 设 $P_3(x) = a + bx + cx^2 + dx^3$ 则 $P'_3(x) = b + 2cx + 3dx^2$

由插值条件得：

$$\begin{cases} P_3(0) = a = 0 \\ P_3(1) = a + b + c + d \\ P'_3(1) = b + 2c + 3d = 5 \\ P_3(2) = a + 2b + 4c + 8d = 14 \end{cases} \quad \text{得} \quad \begin{cases} a = 0 \\ b = 1 \\ c = -1 \\ d = 2 \end{cases}$$

则 $P_3(x) = x - x^2 + 2x^3$

余项： 类似Lagrange插值余项(只是把导数点也用上)

$$R_3(x) = f(x) - P_3(x)$$

则 $R_3(0) = R_3(1) = R_3(2) = 0$, $R'_3(1) = f'(1) - P'_3(1) = 0$

故可设 $R_3(x) = k(x)x(x-1)^2(x-2)$

设 $\varphi(t) = f(t) - P_3(t) - k(x)t(t-1)^2(t-2) = R_3(t) - k(x)t(t-1)^2(t-2)$

则当 $t = 0, 1, 2, x$ 时 $\varphi(t) = 0$

当 $t = \xi_0, \xi_1, \xi_2, 1$ 时 $\varphi'(t) = 0$

当 $t = \xi$ 时 $\varphi^{(4)}(t) = 0, \xi \in [0, 2]$

即 $\varphi^{(4)}(\xi) = 0$

得 $f^{(4)}(\xi) - k(x)4! = 0$

有 $k(x) = \frac{f^{(4)}(\xi)}{4!}$

代入有: $R_3(x) = \frac{f^{(4)}(\xi)}{4!} x(x-1)^2(x-2)$

一般地, 已知 x_0, \dots, x_n 处有 y_0, \dots, y_n 和 y_0', \dots, y_n' , 求 $H_{2n+1}(x)$ 满足 $H_{2n+1}(x_i) = y_i$, $H'_{2n+1}(x_i) = y_i'$ 。

解: 设 $H_{2n+1}(x) = \sum_{i=0}^n y_i \alpha_i(x) + \sum_{i=0}^n y_i' \beta_i(x)$

$$l_i(x) = \prod_{j \neq i} \frac{(x - x_j)}{(x_i - x_j)}$$

其中 $\alpha_i(x_j) = \delta_{ij}$, $\alpha_i'(x_j) = 0$, $\beta_i(x_j) = 0$, $\beta_i'(x_j) = \delta_{ij}$

$\alpha_i(x)$ 有零点 $x_0, \dots, x_{i-1}, x_{i+1}, \dots, x_n$ 且都是2重零点 $\Rightarrow \alpha_i(x) = (A_i x + B_i) l_i^2(x)$

由余下条件 $\alpha_i(x_i) = 1$ 和 $\alpha_i'(x_i) = 0$ 可解 A_i 和 $B_i \Rightarrow$

$$\alpha_i(x) = [1 - \frac{1}{2} \frac{(x - x_i)^2}{l_i^2(x)}] l_i^2(x)$$

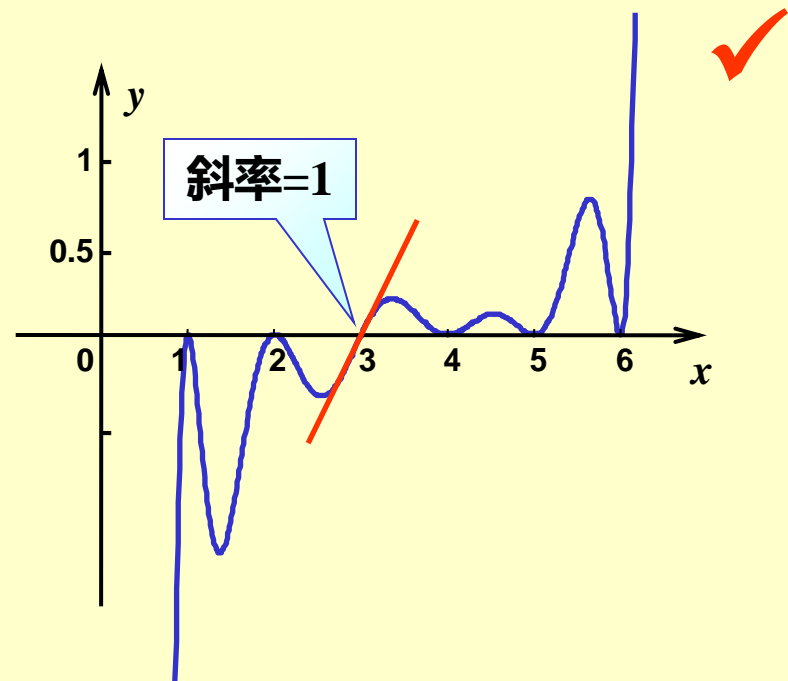
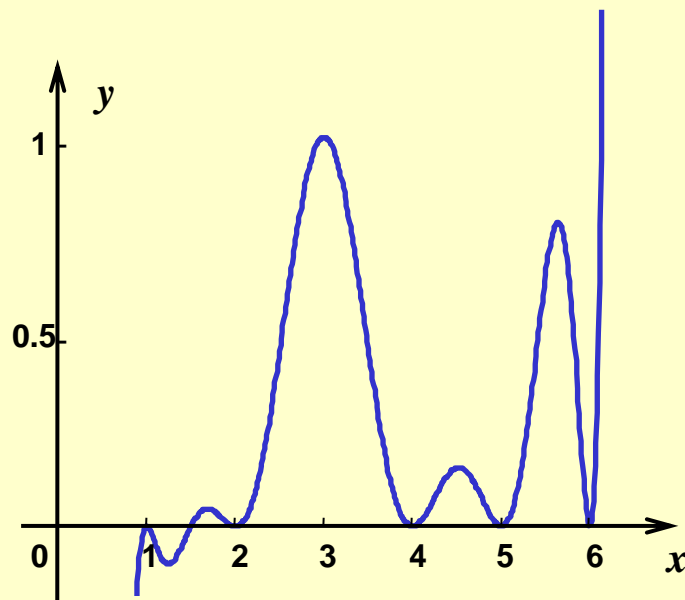
$\beta_i(x)$ 有零点 x_0, \dots, x_n 且都是2重零点 $\Rightarrow \beta_i(x) = C_i (x - x_i) l_i^2(x)$

这样的Hermite 插值唯一

$$\beta_i(x) = (x - x_i) l_i^2(x)$$

设 $a = x_0 < x_1 < \dots < x_n = b$, $f \in C^{2n}[a, b]$ 则 $R_n(x) = \frac{f^{(2n+2)}(\xi_x)}{(2n+2)!} \left[\prod_{i=0}^n (x - x_i) \right]^2$

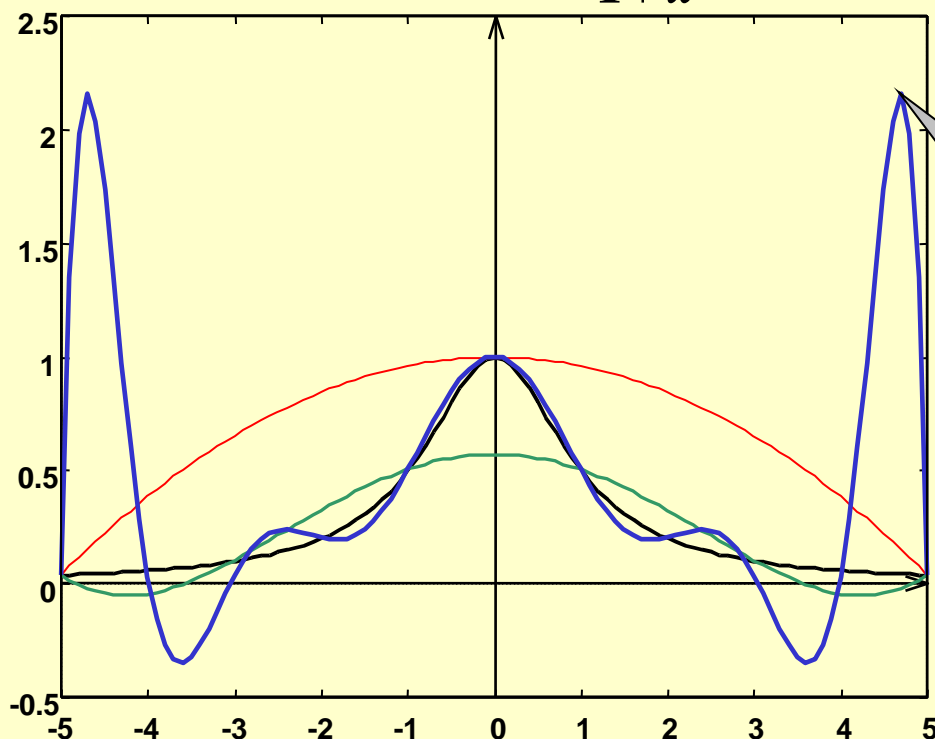
Quiz: 给定 $x_i = i + 1, i = 0, 1, 2, 3, 4, 5$. 下面哪个是 $\beta_2(x)$ 的图像?



§5 分段低次插值 /* piecewise polynomial approximation */

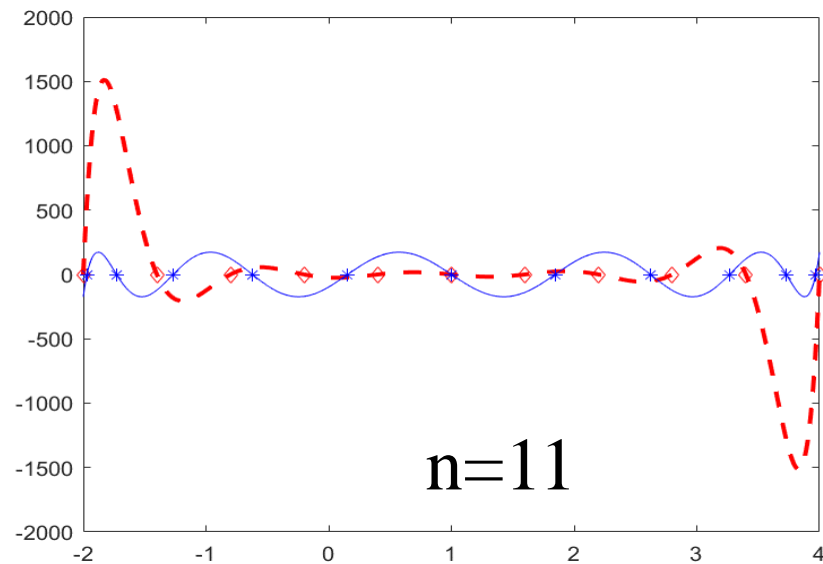
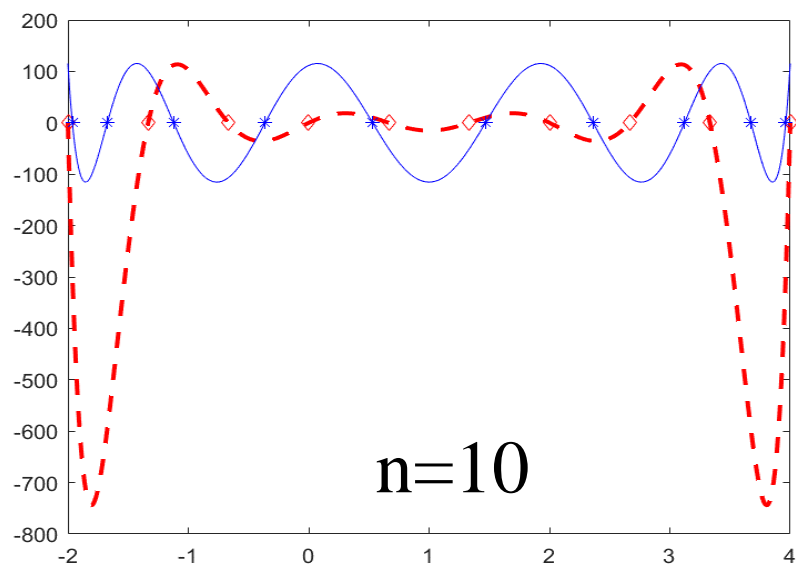
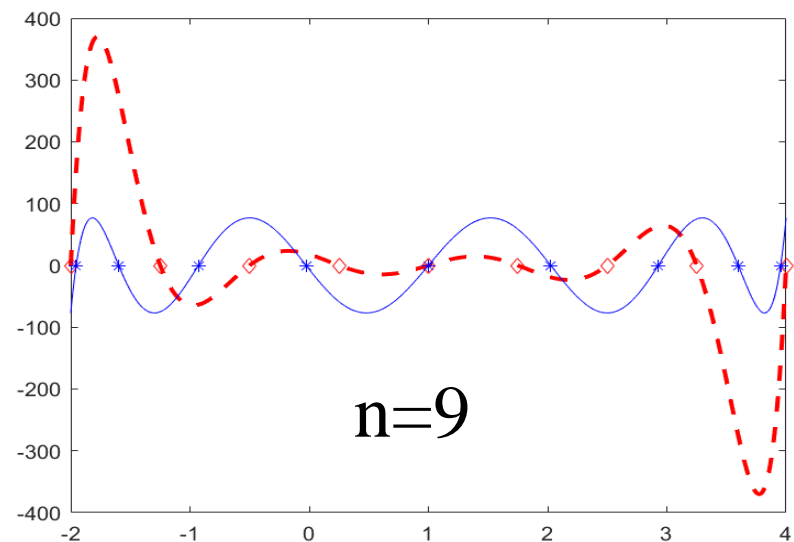
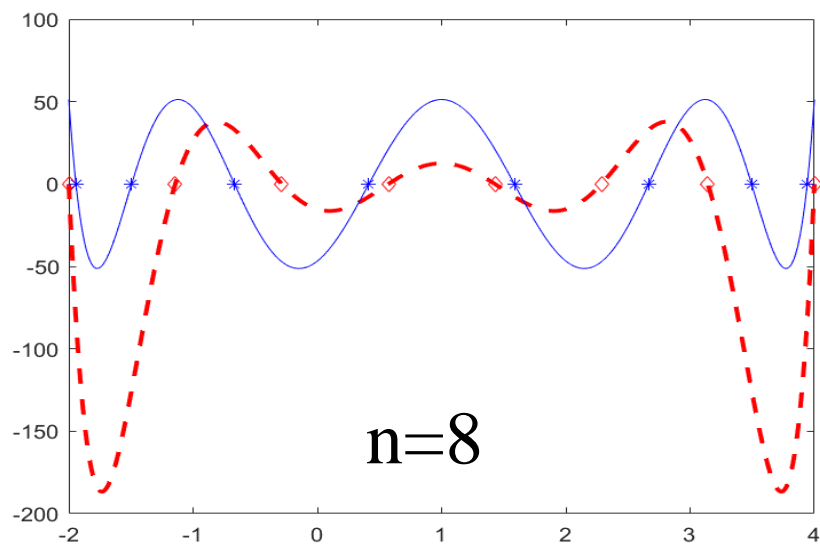
在区间 $[a, b]$ 上用插值多项式 $P_n(x)$ 近似函数 $f(x)$ ，是否 $P_n(x)$ 的次数越高，逼近效果越好呢，回答是否定的。由于次数越高计算工作量也越大，积累误差也越大；在整个区间上作高次多项式，当局部插值节点的值有微小误差时，就可能引起整个区间上函数值的较大变化，使计算不稳定。

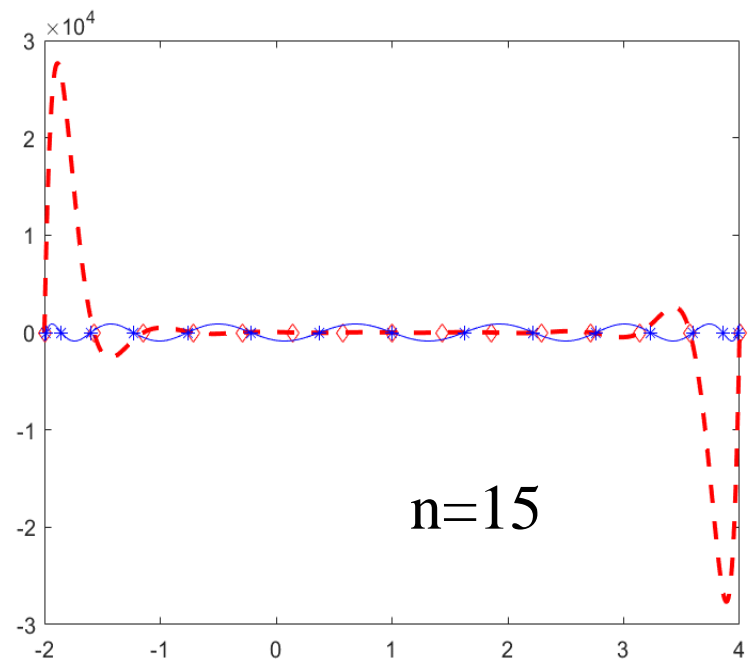
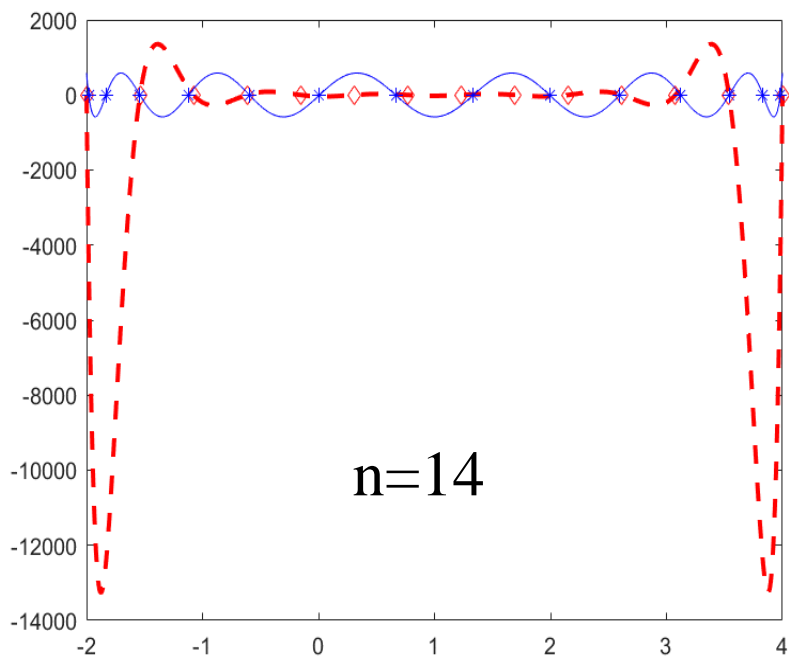
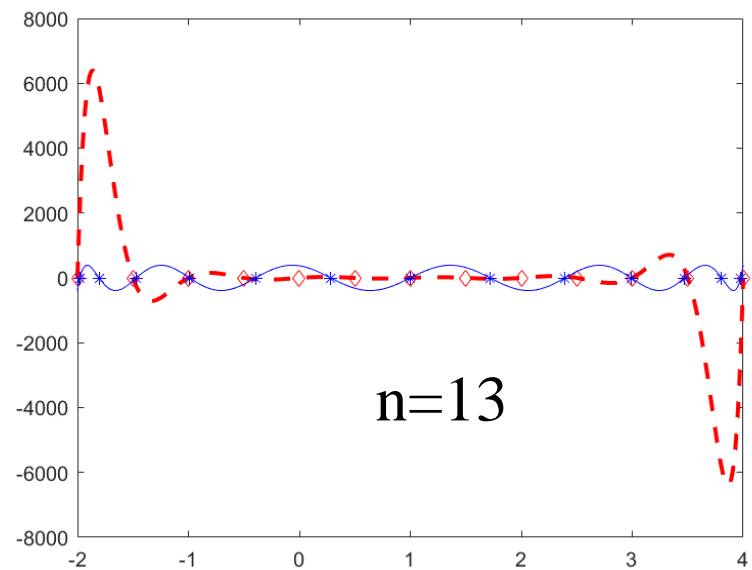
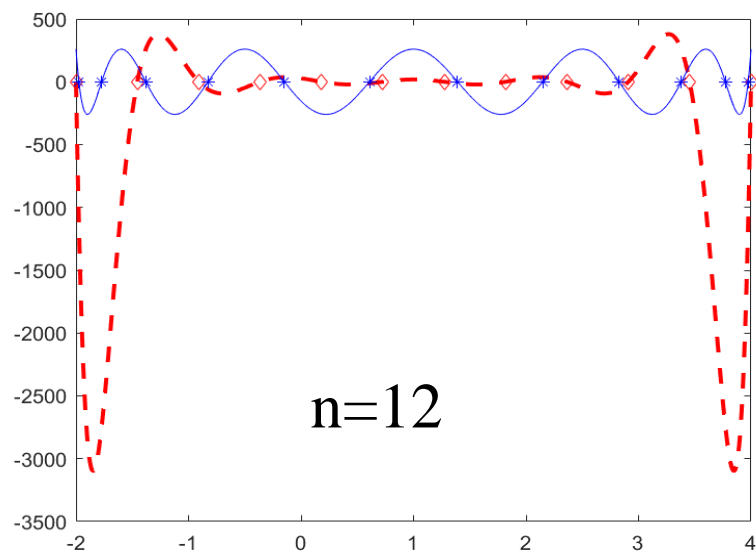
例：在 $[-5, 5]$ 上考察 $f(x) = \frac{1}{1+x^2}$ 的 $L_n(x)$ 。取 $x_i = -5 + \frac{10}{n}i$ ($i = 0, \dots, n$)



$L_n(x) \not\rightarrow f(x)$

n 越大，
端点附近抖动
越大，称为
Runge 现象







分段低次插值

➤ 分段线性插值 /* piecewise linear interpolation */

所谓的分段线性插值就是通过插值点用折线段连接起来逼近 $f(x)$ 设已知节点 $a < x_0 < x_1 < \cdots < x_n = b$ 上函数值 f_0, f_1, \cdots, f_n , 记

$$h_k = x_{k+1} - x_k, h = \max_k h_k$$

求一折线函数 $I_h(x)$ 满足

$$(1) I_h(x) \in C[a, b],$$

$$(2) I_h(x_k) = f_k (k = 0, 1, \cdots, n)$$

$$(3) I_h(x) \text{ 在每个小区间 } [x_k, x_{k+1}] \text{ 上都是线性函数。}$$

则称 $I_h(x)$ 为分段线性插值函数。

设

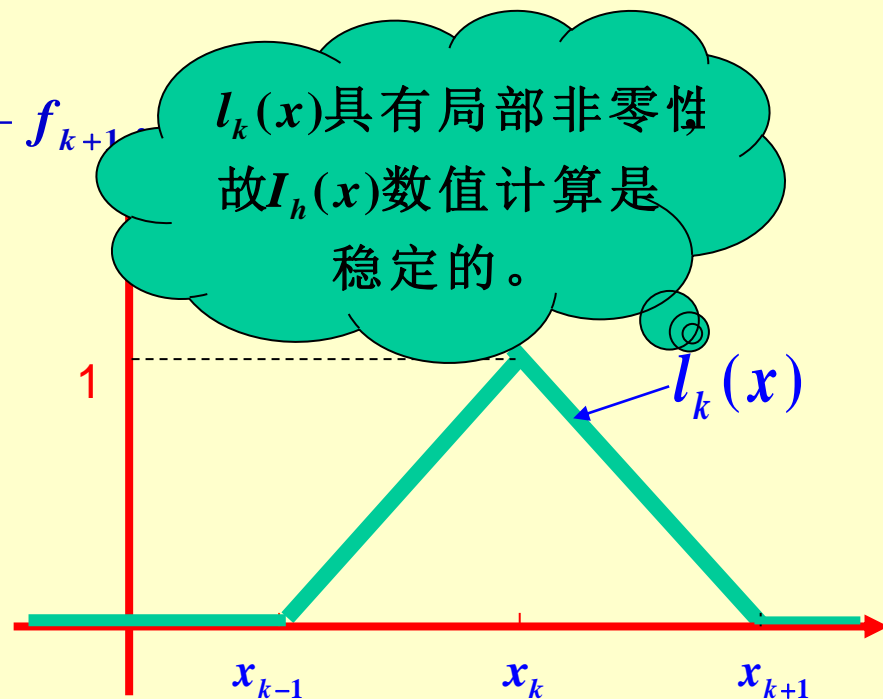
$$I_h(x) = \sum_{k=0}^{n-1} f_k l_k(x)$$

由定义可知 $I_h(x)$ 在每个小区间上可表示为

$$I_h(x) = \frac{x - x_{k-1}}{x_k - x_{k-1}} f_k + \frac{x - x_k}{x_{k-1} - x_k} f_{k-1}, x \in [x_{k-1}, x_k]$$

$$I_h(x) = \frac{x - x_{k+1}}{x_k - x_{k+1}} f_k + \frac{x - x_k}{x_{k+1} - x_k} f_{k+1}$$

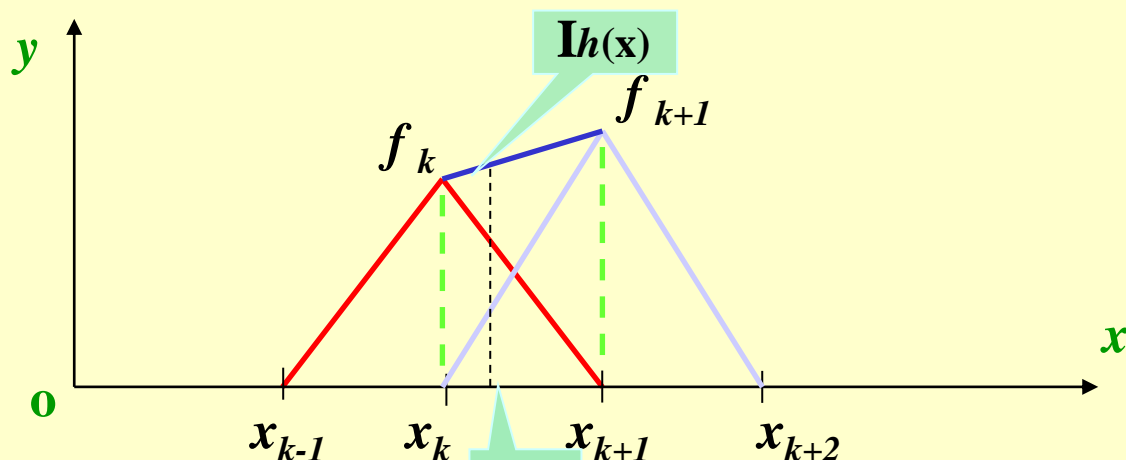
$$l_k(x) = \begin{cases} \frac{x - x_{k-1}}{x_k - x_{k-1}}, & x_{k-1} \leq x \leq x_k (k \neq 0) \\ \frac{x - x_{k+1}}{x_k - x_{k+1}}, & x_k \leq x \leq x_{k+1} (k \neq n) \\ 0, & x \in [a, b], \text{但} x \notin [x_{k-1}, x_{k+1}]. \end{cases}$$



注意 表达式 $I_h(x) = \sum_{k=0}^{n-1} f_k l_k(x)$ 在区间 $[x_k, x_{k+1}]$ 上, 只有 $l_k(x), l_{k+1}(x)$ 是非零的, 其它基函数均为零。即

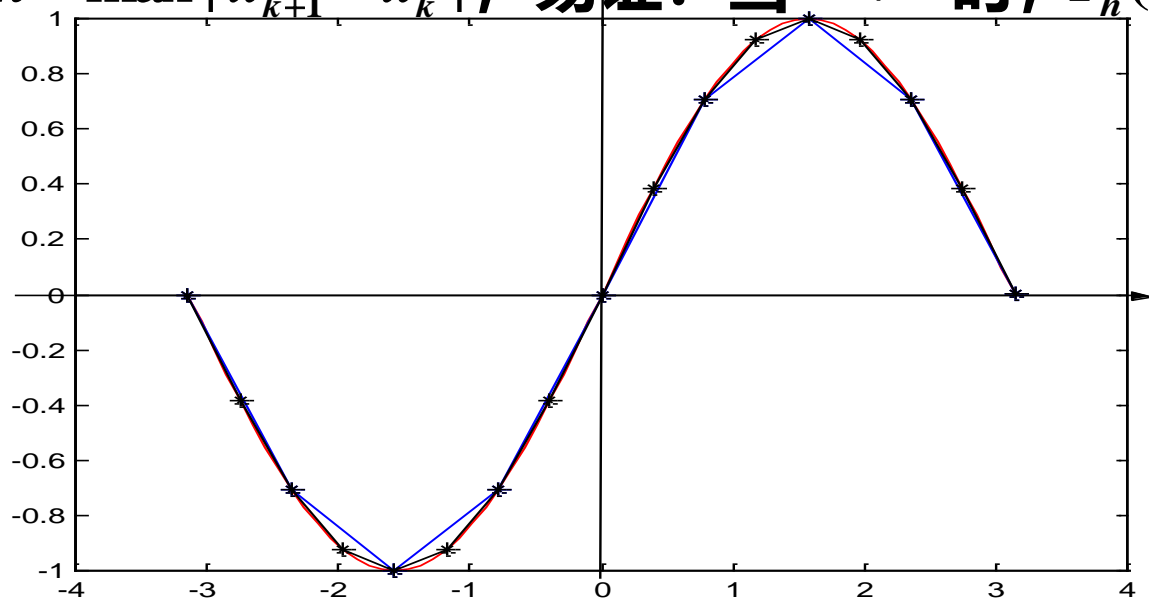
$$I_h(x) = f_k l_k(x) + f_{k+1} l_{k+1}(x) \quad x \in [x_k, x_{k+1}]$$

如图：



$y=I_h(x)$ 的图象实际上是连接点 (x_k, f_k) , $k=1,2,\dots,n$ 的一条折线, 也称**折线插值**, 如下图。如果增加节点数量, 会改善插值效果, 即:

记 $h = \max |x_{k+1} - x_k|$, 易证: 当 $h \rightarrow 0$ 时, $I_h(x) \xrightarrow{\text{一致}} f(x)$



分段线性插值的误差估计

根据拉格朗日1次插值函数的余项,可以得到分段线性插值函数的插值误差估计:

对 $x \in [a, b]$, 当 $x \in [x_k, x_{k+1}]$ 时,

$$R(x) = f(x) - I_h(x) = \frac{f''(x)}{2}(x - x_k)(x - x_{k+1})$$

$$\text{则 } |R(x)| \leq \frac{h^2}{8} M \quad \text{其中} \quad h = \max_{0 \leq i \leq n-1} |x_{i+1} - x_i|$$
$$M = \max_{x \in (a, b)} |f''(x)|$$

可以加密插值结点, 缩小插值区间, 使 h 减小, 从而减小误差

思考: 要构造对数表 $\log_{10}x$, $10 \leq x < 100$, 怎样选择步长 h , 才能使得分段线性插值具有六位有效数字?

➤ 分段三次Hermite插值 /* Hermite piecewise polynomials */

分段线性插值函数 $I_h(x)$ 的导数是间断的，若在节点 $x_k (k = 0, 1, \dots, n)$ 上除已知函数值 f_k 外还给出导数值 $f'_k = m_k$ ，这样就可以构造一个导数连续的分段插值多项式函数 $I_h(x)$ ，它满足

$$(1) I_h \in C^1[a, b],$$

$$(2) I_h(x_k) = f_k, I'_h(x_k) = f'_k (k = 0, 1, \dots, n)$$

$$(3) I_h(x) \text{ 在每个小区间 } [x_k, x_{k+1}] \text{ 是三次多项式}$$

设

$$I_h(x) = \sum_{k=0}^n [f(x_k)\alpha_k(x) + f'(x_k)\beta_k(x)]$$

根据两点三次插值多项式。可知，在区间 $[x_k, x_{k+1}]$ 上 $I_h(x)$ 的表达式为

$$\begin{aligned}
 I_h(x) = & \left(\frac{x - x_{k+1}}{x_k - x_{k+1}} \right)^2 \left(1 + 2 \frac{x - x_k}{x_{k+1} - x_k} \right) f_k \\
 & + \left(\frac{x - x_k}{x_{k+1} - x_k} \right)^2 \left(1 + 2 \frac{x - x_{k+1}}{x_k - x_{k+1}} \right) f_{k+1} \\
 & + \left(\frac{x - x_{k+1}}{x_k - x_{k+1}} \right)^2 (x - x_k) f'_k \\
 & + \left(\frac{x - x_k}{x_{k+1} - x_k} \right)^2 (x - x_{k+1}) f'_{k+1}
 \end{aligned}$$

于是:

$$\alpha_k(x) = \begin{cases} \left(\frac{x - x_{k-1}}{x_k - x_{k-1}} \right)^2 \left(1 + 2 \frac{x - x_k}{x_{k-1} - x_k} \right) & x \in [x_{k-1}, x_k] \\ \left(\frac{x - x_{k+1}}{x_k - x_{k+1}} \right)^2 \left(1 + 2 \frac{x - x_k}{x_{k+1} - x_k} \right) & x \in [x_k, x_{k+1}] \\ 0 & \text{else} \end{cases}$$

$$\beta_k(x) = \begin{cases} \left(\frac{x - x_{k-1}}{x_k - x_{k-1}} \right)^2 (x - x_k) & x \in [x_{k-1}, x_k] \\ \left(\frac{x - x_{k+1}}{x_k - x_{k+1}} \right)^2 (x - x_k) & x \in [x_k, x_{k+1}] \\ 0 & \text{else} \end{cases}$$

How can we make a
smooth interpolation
without asking too
much from f ?
Headache ...



导数一般不易得到。

