

计算方法实验一

Error, Interpolation,
Curve Fitting, Integration

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- 实验内容：
 - 编写程序调试，用作业题验证算法
 - 基于Lagrange插值的秘密图像共享
- 考核方式：
 - 实验报告 (90%)
 - 实验考勤 (10%)
- 提交方式：
 - 文件命名：学号-XXX-LAB1
 - 以班级为单位提交电子版至下面邮箱：
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1.1 有效数字丢失现象观察

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例：下面两个理论上等价的函数，当 x 不断增大

$$f_1(x) = \sqrt{x}(\sqrt{x+1} - \sqrt{x}), f_2(x) = \frac{\sqrt{x}}{\sqrt{x+1} + \sqrt{x}}$$

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At x=          1, f1(x)=0.414213562373095150, f2(x)=0.414213562373095090
At x=         10, f1(x)=0.488088481701514750, f2(x)=0.488088481701515480
At x=        100, f1(x)=0.498756211208899460, f2(x)=0.498756211208902730
At x=       1000, f1(x)=0.499875062461021870, f2(x)=0.499875062460964860
At x=      10000, f1(x)=0.499987500624854420, f2(x)=0.499987500624960890
At x=     100000, f1(x)=0.499998750005928860, f2(x)=0.499998750006249940
At x=    1000000, f1(x)=0.499999875046341910, f2(x)=0.499999875000062490
At x=   10000000, f1(x)=0.499999987401150920, f2(x)=0.499999987500000580
At x=  100000000, f1(x)=0.500000005558831620, f2(x)=0.499999998749999950
At x= 1000000000, f1(x)=0.500000077997506340, f2(x)=0.499999999874999990
At x= 10000000000, f1(x)=0.499999441672116520, f2(x)=0.499999999987500050
At x= 100000000000, f1(x)=0.500004449631168080, f2(x)=0.499999999998750000
At x= 1000000000000, f1(x)=0.500003807246685030, f2(x)=0.499999999999874990
At x= 10000000000000, f1(x)=0.499194546973835970, f2(x)=0.499999999999987510
At x= 100000000000000, f1(x)=0.502914190292358400, f2(x)=0.499999999999998720

sqrt(x+1) = 31622776.6016838100000, sqrt(x) = 31622776.6016837920000
diff=0.00000001862645149230957, sum=63245553.2033676060000000000000000
```

1.2 n 次插值的Lagrange形式

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□ Lagrange型 n 次插值算法

① 读入插值节点 $x_i, y_i (i = 0, 1, \dots, n)$ 和点 xx

② $yy = 0$

③ 对 $i = 0, 1, \dots, n$ 做如下工作

□ $t = 1$

□ 对 $k = 0, 1, \dots, i - 1, i + 1, \dots, n$ 做

□ $t = t \times \frac{xx - x_k}{x_i - x_k}$

□ $yy = yy + t \times y_i$

④ 输出点 xx 相应的函数近似值 yy

□ Newton型 n 次插值算法

① 读入插值节点 $x_i, y_i (i = 0, 1, \dots, n)$ 和点 xx

② $\omega = 1, V_0 = y_0, yy = y_0$

③ 对 $k = 1, 2, \dots, n$ 做如下工作

□ $V_k = y_k$

□ 对 $i = 0, 1, \dots, k - 1$ 做

□ $V_k = \frac{V_i - V_k}{x_i - x_k}$

□ $\omega = \omega \times (xx - x_{k-1})$

□ $yy = yy + \omega \times V_k$

④ 输出插值点 xx 相应的函数近似值 yy

1.4 三次样条插值

□ 三次样条插值算法

- ① 读入插值节点 $x_i, y_i (i = 0, 1, \dots, n)$ 和点 x
- ② 对 $i = 0, 1, \dots, n - 1$ 计算 $h_i = x_{i+1} - x_i$
- ③ 计算 α_i 和 $\beta_i (i = 0, 1, \dots, n)$

□ 对第一种边界条件:

$$\alpha_0 = 0, \beta_0 = 2m_0;$$

$$\alpha_n = 1, \beta_n = 2m_n$$

□ 对第二种边界条件:

$$\alpha_0 = 1, \beta_0 = \frac{3}{h_0} (y_1 - y_0);$$

$$\alpha_n = 0, \beta_n = \frac{3}{h_{n-1}} (y_n - y_{n-1})$$

□ 三次样条插值算法

□ 对 $i = 1, 2, \dots, n-1$ 做

$$\alpha_i = \frac{h_{i-1}}{h_{i-1} + h_i}; \beta_i = 3 \left[\frac{1 - \alpha_i}{h_{i-1}} (y_i - y_{i-1}) + \frac{\alpha_i}{h_i} (y_{i+1} - y_i) \right]$$

④ 计算 a_i 和 b_i ($i = 0, 1, \dots, n$)

$$a_0 = -\frac{\alpha_0}{2}; b_0 = \frac{\beta_0}{2}$$

□ 对 $i = 1, 2, \dots, n$ 做

$$a_i = -\frac{\alpha_i}{2 + (1 - \alpha_i)a_{i-1}}; b_i = \frac{\beta_i - (1 - \alpha_i)b_{i-1}}{2 + (1 - \alpha_i)a_{i-1}}$$

□ 三次样条插值算法

⑤ 计算 $m_i (i = 0, 1, \dots, n)$

$$\begin{aligned} m_n &= b_n \\ m_i &= a_i m_{i+1} + b_i \quad (i = n-1, \dots, 1, 0) \end{aligned}$$

⑥ 判别点 xx 所在区间 $[x_i, x_{i+1}] (i = 0, 1, \dots, n-1)$, 然后求出 $s(x)$ 的值并输出

$$\begin{aligned} yy &= s(xx) \\ &= \left(1 + 2 \frac{xx - x_i}{x_{i+1} - x_i}\right) \left(\frac{xx - x_{i+1}}{x_i - x_{i+1}}\right)^2 y_i \\ &\quad + \left(1 + 2 \frac{xx - x_{i+1}}{x_i - x_{i+1}}\right) \left(\frac{xx - x_i}{x_{i+1} - x_i}\right)^2 y_{i+1} + (xx - x_i) \left(\frac{xx - x_{i+1}}{x_i - x_{i+1}}\right)^2 m_i \\ &\quad + (xx - x_{i+1}) \left(\frac{xx - x_i}{x_{i+1} - x_i}\right)^2 m_{i+1} \end{aligned}$$

1.5 单变量数据拟合（最小二乘法）

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□ 单变量线性拟合法算法

① 读入数据 x_i 和 y_i ($i = 1, 2, \dots, n$)

② 计算

$$s_x = \sum_{i=1}^n x_i, s_y = \sum_{i=1}^n y_i, s_{xx} = \sum_{i=1}^n x_i^2, s_{xy} = \sum_{i=1}^n x_i y_i$$

③ 解正规方程组

$$\begin{cases} na + s_x b = s_y \\ s_x a + s_{xx} b = s_{xy} \end{cases}$$

④ 输出 a 和 b

$$a = \frac{s_{xx}s_y - s_x s_{xy}}{n s_{xx} - s_x^2}, b = \frac{n s_{xy} - s_x s_y}{n s_{xx} - s_x^2}$$

1.6 自动选取步长梯形法

- ① 输入 a, b 和 ϵ
- ② 计算 $h = \frac{b-a}{2}$, $T1 = (f(a) + f(b)) * h, n = 1$
- ③ 计算 $T0 = T1, S = 0$ ($T0$ 表示前次积分近似值, $T1$ 表示后次积分近似值)
- ④ 对 $k = 1, 2, \dots, n$ 计算
$$S = S + f(a + (2 * k - 1) * h/n)$$
- ⑤ $T1 = \frac{T0}{2} + S * \frac{h}{n}$
- ⑥ 若 $|T1 - T0| < 3\epsilon$, 则输出 $T1$ 的值, 结束计算, 否则 $n = 2n$, 返回③

① 输入 a, b 和 ϵ

② 计算 $T_0^{(0)}$:

$$T(0,0) = \frac{b-a}{2} [f(a) + f(b)]$$

③ $k = 1$ (其中 k 用来记录把积分区间 $[a, b]$ 2等分的次数)

④ 按复化梯形公式计算 $T_0^{(k)}$:

$$T(0,k) = \frac{1}{2} \left[T(0,k-1) + \frac{b-a}{2^{k-1}} \sum_{i=1}^{2^{k-1}} f\left(a + (2i-1) \times \frac{b-a}{2^k}\right) \right]$$

⑤ 计算第 $k + 1$ 行元素 $T_m^{(k-m)}$:

$$T(m, k - m) = \frac{4^m T(m - 1, k - m + 1) - T(m - 1, k - m)}{4^m - 1}$$

其中 $m = 1, 2, \dots, k$

⑥ 精度控制:

- 对指定的精度 ϵ , 若 $|T(k, 0) - T(k - 1, 0)| < \epsilon$,
即 $|T_k^{(0)} - T_{k-1}^{(0)}| < \epsilon$, 则终止计算, 并取 $T(k, 0)$ 即
 $T_k^{(0)}$ 作为满足精度要求的积分近似值; 否则 $k = k + 1$, 转回④继续计算.

- 参考文献 C.-C. Thien and J.-C. Lin, "Secret image sharing," Computers & Graphics, vol. 26, no. 5, pp. 765–770, 2002.
- 实现基于Lagrange插值的秘密图像共享
 - 输入一幅灰度图像, n , k , $k \leq n$
 - 输出 n 幅子图
 - 任意选择 k 幅子图重构原图像
 - 思考不截断的实现方法, 即保持灰度图像像素值范围0-255

Thank You!

