

Lecture 14 Numerical integration: advanced topics

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Outline

Motivations

Gaussian quadrature

Adaptive quadrature

Gaussian quadrature

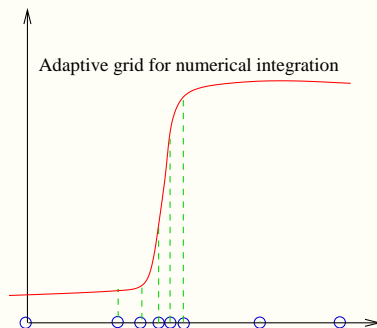
- ▶ 定积分的 n 点求积公式一般可写为

$$\int_a^b f(x) dx \approx \sum_{k=1}^n A_k f(x_k),$$

- ▶ How to obtain the **maximal accuracy** with fixed number of node points n and suitable choice of coefficients A_k ?

Adaptive integration

- ▶ Adaptive integration



- ▶ The information of the function itself must be taken into account.

Outline

Motivations

Gaussian quadrature

Adaptive quadrature

Interpolated integration

- ▶ 一般的数值积分可表示为

$$I = \int_a^b f(x)dx \approx \sum_{k=0}^n A_k f(x_k) = I_N$$

其中 x_k 称求积节点, A_k 称求积系数, 可定义误差

$$E_n(f) = I(f) - I_N(f).$$

- ▶ **定义** 如果 $\forall f \in \mathbb{P}_n, E_n(f) = 0$, 则称数值积分为插值型的。
- ▶ **定理** 下面两个命题是等价的:
 - (1) $I_N(f)$ 为插值型积分;
 - (2) $I_N(f)$ 可由对 x_0, x_1, \dots, x_n 的 n 次插值积分得到。

代数精度

- ▶ **定义:** 设 m 是一个正整数,如果数值积分公式的截断误差对

$$f(x) = 1, x, x^2, \dots, x^m$$

都为零,但对 $f(x) = x^{m+1}$ 不为零,则称数值积分公式的代数精度为 m .

- ▶ **例子:** 考察中点公式

$$\int_a^b f(x)dx \approx f\left(\frac{a+b}{2}\right)(b-a).$$

取 $f(x) = 1, x$, 等式均精确成立, 取 $f(x) = x^2$ 等式不成立, 故中点公式代数精度为1。

代数精度

- ▶ 下面定理精确刻划了 n 次代数精度与 $n + k$ 次代数精度的差异。

定理: 对 $\forall 0 \leq k \leq n + 1$, $I_N(f)$ 为 $d = n + k$ 次代数精度等价于

(1) $I_N(f)$ 为插值型的;

(2) $\omega_n(x) = \prod_{k=0}^n (x - x_k)$ 满足 $\int_a^b \omega_n(x)p(x)dx = 0$, $\forall p \in \mathbb{P}_{k-1}$.

- ▶ **定理:** Newton-Cotes求积公式至少具有 n 次代数精度, 如 n 为偶数, 则具有 $n + 1$ 次代数精度。

Gauss积分

- ▶ 关于定积分的 n 点求积公式一般可写为

$$\int_a^b f(x) dx \approx \sum_{k=1}^n A_k f(x_k), \quad (1)$$

- ▶ 更一般的如果考虑带有权函数的定积分的计算

$$\int_a^b \rho(x) f(x) dx \approx \sum_{k=1}^n A_k f(x_k), \quad (2)$$

其中 $\rho(x) > 0$ 为已知的权函数.

- ▶ 我们的目标是把节点个数 n 固定,而让节点的位置 x_k 与系数 A_k 都待定,而让求积公式的代数精度达到最高.这种在取定节点数目的条件下,代数精度达到最高的求积公式,我们称之为**Gauss**求积公式. 由于此时共有 $2n$ 个自由度,设想最大可能的代数精度为 $2n - 1$.

Gauss积分

- ▶ 不失一般性, 我们可以设积分区间 $[a, b] = [-1, 1]$, 这是因为有积分变量替换公式

$$\int_a^b f(x) dx = \frac{b-a}{2} \int_{-1}^1 f\left(\frac{a+b}{2} + \frac{b-a}{2}t\right) dt. \quad (3)$$

- ▶ 待定的参数有 $2n$ 个 (n 个节点, n 个系数), 取 $f(x)$ 分别为 $1, x, x^2, \dots, x^{2n-1}$, 并令求积公式对这些函数准确成立. 首先考虑 $\rho(x) \equiv 1$, 得到一个 $2n$ 个方程

$$\sum_{k=1}^n A_k f(x_k) = \frac{1}{j+1} (b^{j+1} - a^{j+1}), \quad j = 0, 1, 2, \dots, 2n-1.$$

非线性方程组, 难以求解!

Gauss积分

► $n = 1$ 情形

方程 $\int_{-1}^1 f(x) dx = A_1 f(x_1)$ 中, 令 $f(x) = 1$ 和 $f(x) = x$ 得

$$A_1 = 2, \quad x_1 = 0.$$

于是得到一个节点的Gauss求积公式为

$$\int_a^b f(x) dx \approx f\left(\frac{a+b}{2}\right)(b-a).$$

这就是中点公式.

Gauss积分

► $n = 2$ 情形

等式 $\int_{-1}^1 f(x) dx = A_1 f(x_1) + A_2 f(x_2)$ 中, 令 $f(x) = 1, f(x) = x, f(x) = x^2, f(x) = x^3$ 得方程组

$$\begin{cases} A_1 + A_2 = 2, \\ A_1 x_1 + A_2 x_2 = 0, \\ A_1 x_1^2 + A_2 x_2^2 = \frac{2}{3}, \\ A_1 x_1^3 + A_2 x_2^3 = 0. \end{cases}$$

- 此方程组有唯一解 $A_1 = A_2 = 1, x_1 = -\frac{1}{\sqrt{3}}, x_2 = \frac{1}{\sqrt{3}}$. 于是得到两个节点的 Gauss 求积公式为

$$\int_a^b f(x) dx \approx \frac{b-a}{2} \left(f\left(\frac{a+b}{2} - \frac{b-a}{2\sqrt{3}}\right) + f\left(\frac{a+b}{2} + \frac{b-a}{2\sqrt{3}}\right) \right).$$

易证此求积公式的代数精度为3.

正交多项式

- ▶ Gauss积分一般情形难以求解!
- ▶ 定义: 设 $\rho(x)$ 为 (a, b) 区间上的一个给定的非负函数, 任取两个函数 $f, g \in C[a, b]$, 我们称数值

$$(f, g) = \int_a^b \rho(x) f(x) g(x) dx$$

为函数 f 与 g 的带权内积, 并称

$$\|f\|_2 = \sqrt{(f, f)} = \sqrt{\int_a^b \rho(x) f(x) f(x) dx}$$

为函数 f 的带权2范数, 其中的 $\rho(x)$ 称为权函数.

- ▶ 定义: 函数 $f, g \in C[a, b]$ 称为是正交的, 如果它们的内积

$$(f, g) = \int_a^b \rho(x) f(x) g(x) dx = 0$$

正交多项式

- ▶ 定义 设 (a, b) 区间上实系数多项式函数

$$\{\varphi_0(x), \varphi_1(x), \varphi_2(x), \dots, \},$$

满足

$$\int_a^b \rho(x) \varphi_j(x) \varphi_l(x) dx = 0, \quad \text{当 } j \neq l \text{ 时}, \quad (4)$$

则称之为区间 $[a, b]$ 上关于权函数 $\rho(x)$ 的正交多项式系.

- ▶ 按照给定的内积, 可用**Gram-Schmidt**正交化方法把基函数 $\{1, x, x^2, \dots, x^k, \dots\}$ 化成正交多项式系

$$\begin{cases} \varphi_0(x) = 1 \\ \varphi_{j+1}(x) = x^{j+1} - \sum_{i=0}^j \frac{(x^{j+1}, \varphi_i(x))}{(\varphi_i(x), \varphi_i(x))} \varphi_i(x), \quad \text{对 } j = 0, 1, \dots \end{cases}$$

正交多项式

- ▶ (1) Legendre 多项式: $[a, b] = [-1, 1]$, $\rho(x) \equiv 1$,

$$\begin{cases} \varphi_0(x) = 1, & \varphi_1(x) = x, \\ \varphi_{k+1}(x) = \frac{2k+1}{k+1}x\varphi_k(x) - \frac{k}{k+1}\varphi_{k-1}(x), \\ \text{对 } k = 1, 2, \dots \end{cases}$$

- ▶ (2) Chebyshev 多项式: $[a, b] = [-1, 1]$, $\rho(x) = 1/\sqrt{1-x^2}$

$$\begin{cases} T_0(x) = 1, & T_1(x) = x, \\ T_{k+1}(x) = 2xT_k(x) - T_{k-1}(x), \\ \text{对 } k = 1, 2, \dots \end{cases}$$

- ▶ (3) Laguerre 多项式: $[a, b] = [0, +\infty)$, $\rho(x) = \exp(-x)$

$$\begin{cases} Q_0(x) = 1, & Q_1(x) = 1 - x, \\ Q_{k+1}(x) = (1 + 2k - x)Q_k(x) - k^2Q_{k-1}(x), \\ \text{对 } k = 1, 2, \dots \end{cases}$$

Hermite 多项式等等.

Gauss积分

- ▶ **定理:** 区间 $[a, b]$ 上的 n 阶带权正交多项式必有 n 个实单根. 取这些实单根做节点 x_j 即可得到**Gauss**积分.
- ▶ **Gauss-Legendre**公式: 此时 $[a, b] = [-1, 1], \rho(x) \equiv 1$, 求积节点 x_j 为 n 阶**Legendre**多项式的零点. 几个低阶具体情形为:

$$n = 1 \text{ 时 } \int_{-1}^1 f(x) dx \approx 2f(0)$$

$$n = 2 \text{ 时 } \int_{-1}^1 f(x) dx \approx f\left(-\frac{1}{\sqrt{3}}\right) + f\left(\frac{1}{\sqrt{3}}\right)$$

$$n = 3 \text{ 时 } \int_{-1}^1 f(x) dx \approx \frac{5}{9}f\left(-\sqrt{\frac{3}{5}}\right) + \frac{8}{9}f(0) + \frac{5}{9}f\left(\sqrt{\frac{3}{5}}\right)$$

其它的可以从数学手册中查到.

- ▶ **Gauss-Chebyshev**公式

$$\int_{-1}^1 \frac{f(x)}{\sqrt{1-x^2}} dx \approx \frac{\pi}{n} \sum_{j=1}^n f(x_j)$$

其中 x_j 是 $[-1, 1]$ 区间上的 n 阶**Chebyshev**多项式 $T_n(x)$ 的零点

$$x_j = \cos\left(\frac{2j-1}{2n}\pi\right)$$

Gauss积分

► Gauss-Laguerre公式

$$\int_0^{\infty} e^{-x} f(x) dx \approx \sum_{j=1}^n A_j f(x_j)$$

节点 x_j 为 $(-\infty, 0]$ 区间上的 n 阶Laguerre多项式 $Q_n(x)$ 的零点.

当 $n = 2$ 时,Gauss-Laguerre公式为

$$\int_0^{\infty} e^{-x} f(x) dx \approx \frac{2 + \sqrt{2}}{4} f(2 - \sqrt{2}) + \frac{2 - \sqrt{2}}{4} f(2 + \sqrt{2})$$

其它的可以从数学手册中查到.

- **定理:** 设权函数 $\rho(x) = 1$ 相应 $[a, b]$ 上首1的 $n + 1$ 次正交多项式记为 $q_{n+1}(x)$ 则当 $f \in C^{2n+2}[a, b]$ 时, Gauss-Legendre积分误差:

$$E(f) = \frac{f^{(2n+2)}(\xi)}{(2n+2)!} \int_a^b (\omega_n(x))^2 dx.$$

Outline

Motivations

Gaussian quadrature

Adaptive quadrature

Composite trapezoid rule

- ▶ Consider, for example, the general composite trapezoid rule

$$I_N(f) = \sum_{i=1}^N \frac{x_i - x_{i-1}}{2} [f(x_{i-1}) + f(x_i)]$$

and the error estimate

$$I(f) - I_N(f) = - \sum_{i=1}^N \frac{f''(\eta_i)(x_i - x_{i-1})^3}{12}$$

where the nodes x_i are not necessarily equally spaced.

- ▶ The contribution

$$- \frac{f''(\eta_i)(x_i - x_{i-1})^3}{12}, \quad \eta_i \in (x_{i-1}, x_i)$$

depends on both the size of $f''(x)$ on the interval (x_{i-1}, x_i) and the size of $|x_i - x_{i-1}|$.

Adaptive idea

- ▶ In those parts of the interval of integration (a, b) where $|f''(x)|$ is “small”, we can take subintervals of “large” size, while in regions where $|f''(x)|$ is “large”, we can take subintervals of “small” size.
- ▶ This **equi-partition policy of errors** is best if the goal is to minimize the number of subintervals and hence the number of function evaluations, necessary to calculate $I(f)$ to a given accuracy.

Adaptive quadrature

- Suppose the object is to compute an approximation P to the integral $I(f) = \int_a^b f(x)dx$ such that

$$|P - I(f)| \leq \epsilon$$

and to do this using as small a number of function evaluations as possible.

- Define $I_i(f) = \int_{x_i}^{x_{i+1}} f(x)dx$, and

$$S_i = \frac{h}{6} \left(f(x_i) + 4f\left(x_i + \frac{h}{2}\right) + f(x_{i+1}) \right)$$

$$\bar{S}_i = \frac{h}{12} \left(f(x_i) + 4f\left(x_i + \frac{h}{4}\right) + 2f\left(x_i + \frac{h}{2}\right) + 4f\left(x_i + \frac{3h}{4}\right) + f(x_{i+1}) \right)$$

Adaptive quadrature

- ▶ From error estimates

$$I_i(f) - S_i = -\frac{f^{(4)}(\eta_1)}{90} \left(\frac{h}{2}\right)^5$$

$$I_i(f) - \bar{S}_i = -\frac{2f^{(4)}(\eta_2)}{90} \left(\frac{h}{4}\right)^5$$

- ▶ Assume $\eta_1 = \eta_2$, then

$$\bar{S}_i - S_i = \frac{f^{(4)} \cdot h^5}{2^5 \cdot 90} \left(\frac{1 - 2^4}{2^4}\right)$$

thus

$$\frac{f^{(4)} \cdot h^5}{2^5 \cdot 90} = \frac{2^4(\bar{S}_i - S_i)}{1 - 2^4}$$

- ▶ Substitute it back we have

$$I_i(f) - \bar{S}_i = \frac{\bar{S}_i - S_i}{1 - 2^4} = \frac{1}{15}(\bar{S}_i - S_i)$$

Adaptive quadrature

- ▶ So the principle of equipartition of errors gives

$$E_i = \frac{1}{15}(\bar{S}_i - S_i) \leq \frac{h}{b-a}\epsilon$$

and the approximate integration is taken as

$$P = \sum_{i=1}^N \bar{S}_i$$

- ▶ Adaptive quadrature essentially consists of applying the Simpson's rule to each of the subintervals covering $[a, b]$ until the above inequality of equipartition of errors is satisfied. If the inequality is not satisfied, those subintervals must be further subdivided and the entire process repeated.

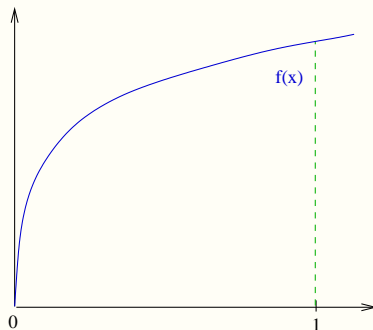
Adaptive quadrature

- ▶ Example

$$I(f) = \int_0^1 \sqrt{x} dx$$

and the error $\epsilon = 0.0005$.

- ▶ Sketch of function $f(x) = \sqrt{x}$



The curve is very steep in the vicinity of 0 while it is fairly flat as $x \rightarrow 1$.

Adaptive quadrature

- ▶ Step 1: Divide the interval $[0, 1]$ into $[0, \frac{1}{2}]$ and $[\frac{1}{2}, 1]$.

$$S[\frac{1}{2}, 1] = 0.43093403$$

$$\bar{S}[\frac{1}{2}, 1] = 0.43096219$$

then

$$E[\frac{1}{2}, 1] = \frac{1}{15}(\bar{S} - S) = 0.0000018775 < \frac{1/2}{1}(0.0005) = 0.00025$$

The error criterion is satisfied, we accept the value $\bar{S}[\frac{1}{2}, 1]$.

- ▶ Step 2: Compute the integral in $[0, \frac{1}{2}]$:

$$S[0, \frac{1}{2}] = 0.22449223$$

$$\bar{S}[0, \frac{1}{2}] = 0.23211709$$

and

$$E[0, \frac{1}{2}] = 0.00043499 > 0.00025$$

The error criterion is not satisfied, we must subdivide $[0, \frac{1}{2}]$ again.

Adaptive quadrature

- ▶ Step 3: Compute the integral in $[\frac{1}{4}, \frac{1}{2}]$:

$$S[\frac{1}{4}, \frac{1}{2}] = 0.15235819$$

$$\bar{S}[\frac{1}{4}, \frac{1}{2}] = 0.15236814$$

and

$$E[\frac{1}{4}, \frac{1}{2}] = 0.664 \cdot 10^{-6} < \frac{1/4}{1}(0.0005) = 0.000125$$

The error criterion is satisfied, we accept the value $\bar{S}[\frac{1}{4}, \frac{1}{2}]$.

- ▶ Step 4: Compute the integral in $[0, \frac{1}{4}]$:

$$S[0, \frac{1}{4}] = 0.07975890$$

$$\bar{S}[0, \frac{1}{4}] = 0.08206578$$

and

$$E[0, \frac{1}{4}] > \frac{1/4}{1}(0.0005) = 0.000125$$

The error criterion is not satisfied, we must subdivide $[0, \frac{1}{4}]$ again.

Adaptive quadrature

- ▶ Step 5: We repeat the procedure above again and again until all of the subintervals satisfies the condition. The subintervals are

$$\left[0, \frac{1}{8}\right], \left[\frac{1}{8}, \frac{1}{4}\right], \left[\frac{1}{4}, \frac{1}{2}\right], \left[\frac{1}{2}, 1\right],$$

- ▶ So we have the approximate integration values

$$P = \sum_i \bar{S}_i = 0.66621524$$

We have $I(f) = \frac{2}{3}$, so

$$|P - F(f)| = 0.00045142 < 0.0005$$

- ▶ The subroutine for adaptive quadrature is very complicate in general. But it is very useful for large scale computations.