Computational Astrophysics

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 - Simpson's Rule (non-equidistant intervals)
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Integration

We want to evaluate numerically the integral

$$Q = \int_{a}^{b} f(x) dx , \qquad (1)$$

where f(x) may be a well behaved analytical function or a function given as discrete data $f(x_i)$.

Suppose that we know the values of f(x) at a finite set of points (nodes) $\{x_j\}$, $(j=0,\cdots,n)$, in the interval [a,b].

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It is possible to replace f(x) with an interpolated polynomial p(x) whose analytic integral is known.

This integration, based on interpolation polynomials, is generally called *Newton-Cotes quadrature formulae*.

Now, consider that the full interval over which we intend to integrate can be broken down into N sub-intervals $[a_i,b_i]$ that encompass N+1 nodes x_i $(i=0,\cdots,N)$ at which we know the integrand f(x).

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The full integral Q can be written as the sum of the sub-integrals Q_i :

$$Q = \sum_{i=0}^{N-1} Q_i = \sum_{i=0}^{N-1} \int_{a_i}^{b_i} f(x) dx .$$
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There are many ways to evaluate these sub-integrals.

The simplest approximation is to assume that the function f(x) is constant on the interval $[a_i, b_i]$.

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Hence, using its central value we write

$$Q_i = \int_{a_i}^{b_i} f(x) dx = (b_i - a_i) f\left(\frac{a_i + b_i}{2}\right) . \tag{3}$$

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Note that we need to be able to evaluate f(x) at the midpoint (knowing the value or using an interpolating polynomial).

The error in the midpoint quadrature can be estimated using a Taylor expansion about the midpoint $m_i = (a_i + b_i)/2$,

$$f(x) = f(m_i) + f'(m_i)(x - m_i) + \frac{f''(m_i)}{2}(x - m_i)^2 + \frac{f'''(m_i)}{6}(x - m_i)^3 + \dots$$
(4)

Integrating this expression from a_i to b_i , the odd-order terms drop out, and what is left is

$$Q_i = \int_{a_i}^{b_i} f(x) dx = f(m_i)(b_i - a_i) + \frac{f''(m_i)}{24}(b_i - a_i)^3 + \dots$$
 (5)

$$Q_i = \int_{a_i}^{b_i} f(x) dx = f(m_i)(b_i - a_i) + \mathcal{O}(h^3) + \dots$$
 (6)

where $h = (b_i - a_i) \approx \frac{(b-a)}{N}$.



Trapezoidal Rule

This method approximates f(x) with a linear polynomial on $[a_i, b_i]$,

$$Q_{i} = \int_{a_{i}}^{b_{i}} f(x) dx = \frac{1}{2} (b_{i} - a_{i}) [f(b_{i}) + f(a_{i})] . \qquad (7)$$

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Using a Taylor expansion it is shown that the trapezoidal rule has a local error that, like the midpoint rule, goes with h^3 .

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$$Q_{i} = \frac{b_{i} - a_{i}}{6} \left[f(a_{i}) + 4f\left(\frac{a_{i} + b_{i}}{2}\right) + f(b_{i}) \right] + \mathcal{O}([b_{i} - a_{i}]^{5}) .$$
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This method has an error going with h^5

Simpson's Rule (non-equidistant intervals)

If f(x) is non-equidistantly sampled, i.e. $h = [a_i, b_i]$ is not constant, Simpson's rule modifies.

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Write

$$f(x) = ax^2 + bx + c$$
 for $x \in [x_{i-1}, x_{i+1}]$, (9)

where,

$$a = \frac{h_{i-1}f(x_{i+1}) - (h_{i-1} + h_i)f(x_i) + h_if(x_{i-1})}{h_{i-1}h_i(h_{i-1} + h_i)},$$

$$b = \frac{h_{i-1}^2f(x_{i+1}) + (h_i^2 - h_{i-1}^2)f(x_i) - h_i^2f(x_{i-1})}{h_{i-1}h_i(h_{i-1} + h_i)},$$

$$c = f(x_i),$$

$$(10)$$

with $h_i = x_{i+1} - x_i$ and $h_{i-1} = x_i - x_{i-1}$.

Simpson's Rule (non-equidistant intervals)

Since the integral

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx$$

is independent of the origin of the coordinates, we can choose $x_i = 0$, and hence $-h_{i-1} = x_{i-1}$, $h_i = x_{i+1}$ to obtain

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx = \int_{-h_{i-1}}^{n_i} f(x) dx = \alpha f(x_{i+1}) + \beta f(x_i) + \gamma f(x_{i-1}),$$
(11)

with

$$\alpha = \frac{2h_i^2 + h_i h_{i-1} - h_{i-1}^2}{6h_i} , \qquad \beta = \frac{(h_i + h_{i-1})^3}{6h_i h_{i-1}} ,$$

$$\gamma = \frac{-h_i^2 + h_i h_{i-1} + 2h_{i-1}^2}{6h_{i-1}} ,$$
(12)

Gaussian Quadrature

Gaussian quadrature is an integration method in which there are chosen two quantities, nodes and weights, to maximize the degree of the resulting integration rule.

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Gaussian quadrature is an integration method in which there are chosen two quantities, nodes and weights, to maximize the degree of the resulting integration rule.

Using 2n degrees of freedom, it is possible to integrate polynomials of degree 2n-1. The condition is that, for a set of n nodes x_i and n weights w_i at which a function f is known,

$$\int_{a}^{b} f(x)dx \approx \sum_{i=1}^{n} w_{i}f(x_{i})$$
 (13)

2-point Gaussian Quadrature

In the 2-point Gaussian quadrature we consider 2 nodes x_i and 2 weights w_i on the interval [a,b] to integrate exactly a polynomial of degree 3. Hence

$$\int_{a}^{b} f(x)dx = w_1 f(x_1) + w_2 f(x_2). \tag{14}$$

We can find the unknowns w_1 , w_2 , x_1 , and x_2 by demanding that the formula give exact results for integrating a general polynomial of degree 3,

$$\int_{a}^{b} f(x)dx = \int_{a}^{b} (c_{0} + c_{1}x + c_{2}x^{2} + c_{3}x^{3})dx ,$$

$$= c_{0}(b - a) + c_{1}\left(\frac{b^{2} - a^{2}}{2}\right) + c_{2}\left(\frac{b^{3} - a^{3}}{3}\right) + c_{3}\left(\frac{b^{4} - a^{4}}{4}\right) .$$
(15)

2-point Gaussian Quadrature

However, we have

$$\int_{a}^{b} f(x)dx = w_{1}f(x_{1}) + w_{2}f(x_{2})$$

$$= w_{1}(c_{0} + c_{1}x_{1} + c_{2}x_{1}^{2} + c_{3}x_{1}^{3}) + w_{2}(c_{0} + c_{1}x_{2} + c_{2}x_{2}^{2} + c_{3}x_{2}^{3}).$$

Therefore, comparing these results gives

$$w_1 = \frac{b-a}{2} , \qquad w_2 = \frac{b-a}{2} , \qquad (16)$$

$$x_1 = \left(\frac{b-a}{2}\right)\left(-\frac{1}{\sqrt{3}}\right) + \left(\frac{b+a}{2}\right) , \qquad (17)$$

$$x_2 = \left(\frac{b-a}{2}\right) \left(\frac{1}{\sqrt{3}}\right) + \left(\frac{b+a}{2}\right) . \tag{18}$$

2-point Gaussian Quadrature

In the special case of the interval [-1, 1]:

$$w_1 = w_2 = 1$$
, $x_1 = -\frac{1}{\sqrt{3}}$, $x_2 = \frac{1}{\sqrt{3}}$, (19)

and

$$\int_{-1}^{1} f(x)dx = f\left(-\frac{1}{\sqrt{3}}\right) + f\left(\frac{1}{\sqrt{3}}\right) . \tag{20}$$

Change of Interval

Using an affine transformation that maps [a,b]
ightarrow [-1,1] as

$$t = \frac{b - a}{2}x + \frac{a + b}{2} , \qquad (21)$$

and a change of variables

$$dt = \frac{b-a}{2}dx , \qquad (22)$$

it is possible to change the interval of integration,

$$\int_{a}^{b} f(x)dx = \frac{b-a}{2} \int_{-1}^{1} f\left(\frac{b-a}{2}x + \frac{a+b}{2}\right) dx . \tag{23}$$



Integration of this interpolated second derivative two times and using $f_i = p_i(x_i) = f(x_i)$ gives

$$p_i(x) = \alpha_i(x - x_i)^3 + \beta_i(x - x_{i+1})^3 + \gamma_i(x - x_i) + \eta_i(x - x_{i+1}), \quad (24)$$

where

$$\alpha_i = \frac{p_{i+1}''}{6h_i}$$
 $\beta_i = \frac{-p_i''}{6h_i}$, (25)

$$\gamma_i = \frac{f_{i+1}}{h_i} - \frac{h_i p_{i+1}''}{6} \qquad \qquad \eta_i = \frac{h_i p_i''}{6} - \frac{f_i}{h_i} , \qquad (26)$$

with $h_i = x_{i+1} - x_i$.

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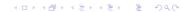
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with $h_i = x_{i+1} - x_i$.

Hence, all we need to find the spline is to know all of its second derivatives $p_i'' = p_i''(x_i)$.



Now apply the condition $p'_{i-1}(x_i) = p'_i(x_i)$ to the interpolated polynomial to obtain the equation

$$h_{i-1}p_{i-1}'' + 2(h_{i-1} + h_i)p_i'' + h_ip_{i+1}'' = 6\left(\frac{g_i}{h_i} - \frac{g_{i-1}}{h_{i-1}}\right)$$
, (27)

where $g_i = f_{i+1} - f_i$.

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where $g_i = f_{i+1} - f_i$.

This is a linear system with n-1 unknowns: p_i'' for $i=1,\cdots,n-1$ (remember that $p_0''=p_n''=0$, as set by the natural spline condition.

Defining
$$d_i = 2(h_{i-1} + h_i)$$
 and $b_i = 6\left(\frac{g_i}{h_i} - \frac{g_{i-1}}{h_{i-1}}\right)$, we can write
$$Ap'' = b \quad \text{with } A_{ij} = \begin{cases} d_i & \text{if } i = j, \\ h_i & \text{if } i = j-1, \\ h_{i-1} & \text{if } i = j+1, \\ 0 & \text{otherwise.} \end{cases}$$
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The coefficient matrix A_{ij} is real, symmetric, and tri-diagonal. The solution of such a linear system of equations will be presented later in the course.