

# Computational Techniques and Numerical Calculations

MASTER STUDIES IN THEORETICAL CHEMISTRY AND COMPUTATIONAL  
MODELLING

## NUMERICAL INTEGRATION, ROOT FINDING AND FUNCTION OPTIMIZATION

*Evaluation exercises*

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## Part I

# Numerical integration

To calculate the integral

$$I = \int_1^3 \sin(x^2) - \cos(2x) dx, \quad (1)$$

the following numerical integrations methods have been used.

## 1 Composite Simpson rule

The composite Simpson rule (CSR) is one of the Newton-Cotes integration rules, an interpolation method based on equidistant interpolation basissas in a  $[a, b]$  interval

$$I_{\text{rule}} = \int_a^b P_{n-1}(x) dx = \sum_{i=1}^n f(x_i) \underbrace{\int_a^b \prod_{j=1, j \neq i}^n \frac{x - x_j}{x_i - x_j} dx}_{\omega_i}, \quad (2)$$

where  $n$  is the number of interpolating points (abscissa points). For each value of  $n$ , there is a named rule. For the Simpson rule,  $n = 3$ .

Also, for each one of these methods, there are two rules: simple rule and composite rule.

The simple rule considers the whole interval,  $dx = b - a = h$  or  $dx = (b - a)/2 = h$  for the Simpson method.

For the composite rule, a number of subintervals,  $N$ , is used to evaluate the integral. Defining the subinterval spacing,  $h$ , as

$$h = \frac{b - a}{2N}, \quad (3)$$

the abscissa points,  $x_i$ , where the function is evaluated are given by

$$x_i = a + ih, \quad i = 0, \dots, 2N \quad (4)$$

Then, the integral is evaluated as

$$I_{\text{CS}} = \frac{h}{3} \left[ f(a) + f(b) + 2 \sum_{i=2 \text{ (even)}}^{2N-2} f(x_i) + 4 \sum_{i=1 \text{ (odd)}}^{2N-1} f(x_i) \right], \quad (5)$$

with an error of the order  $\mathcal{O}(h^5)$

$$E_{\text{CS}} = -N \times \frac{f^4(\xi)}{90} \times h^5 = \mathcal{O}(h^5), \quad \xi \in [a, b]. \quad (6)$$

To compute the integral in eq. (1), the calculation starts with  $N = 1$  subinterval points, doubling in each iteration until reaching a convergence (difference between a result and the preceding one) of  $10^{-8}$ .

Then, the basic structure of the algorithm is the following (algorithm 1).

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**Algorithm 1** Composite Simpson rule

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1: procedure SIMPSONCOMPOSITENCM( $a, b, N_0, \varepsilon, I, t$ )  
   Input. Integration limits,  $a, b$  ; starting number of subinterval points,  $N_0$  ; threshold for  
   convergence,  $\varepsilon$   
   Output. Value of the integral,  $I$  ; total number of iterations,  $t$   
2:   Initialize the number of subintervals,  $N = N_0$ , value of the integral,  $I = 0$ , and number of  
   iterations  $t = 0$   
3:   while  $\Delta I < \varepsilon$  do  
4:     Calculate subinterval spacing,  $h$ , and abscissa points,  $x_i$   
5:     Calculate the new value of the integral,  $I'$   
6:     Calculate the convergence,  $\Delta I = |I' - I|$   
7:     Update value of the integral,  $I = I'$   
8:     Update number of iterations,  $t = t + 1$ 
```

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**Table 1:** Results from the composite Simpson rule

Iteration, $t$	Subinterval number, $N$	Integral value, $I_{CS}$	Difference, $\Delta I$
1	1	0.09897684	0.09897684
2	2	1.05135757	0.95238073
3	4	1.06287028	0.01151271
4	8	1.05794291	0.00492737
5	16	1.05766834	0.00027457
6	32	1.05765178	0.00001656
7	64	1.05765076	0.00000103
8	128	1.05765069	0.00000006

The results are collected in table 1.

Summarizing, the results are

- Number of iterations needed: 9
- Final subinterval value: 0.00390625
- Number of subintervals: 256
- Number of abscissa points: 257
- Final value of the quadrature: 1.05765069

## 2 Romberg's method

The Romberg integration is an extrapolation method based in the terative Richardson extrapolation applied on the Composite Trapezoidal Rule  $n = 2$  function.

The Richardson extrapolation method aims to speed up the convergence of a sequence, based on the extrapolation of two function values calculated at  $\lim_{h \rightarrow 0} R(h)$  and  $\lim_{h \rightarrow 0} R(\frac{h}{2})$ , eliminating the errors of the form  $E(h) = Ch^n$ . A function,  $G(h)$ , is approximated by  $R(h_1)$  (where  $h_1 = x_a - x_0$ ) and by  $R(h_2)$  (where  $h_2 = \frac{h_1}{2} = x_b - x_0$ ), resulting in less rounding errors and/or less number of

calculations

$$G(h) = \frac{2^n R(h/2) - R(h)}{2^n - 1}. \quad (7)$$

The Romberg method provides two mechanisms to improve the accuracy: reduce the value of the subinterval spacing,  $h$ , and apply Richardson extrapolation.

Then, the value of the integral is given by

$$I_{CT} = \int_a^b f(x) dx \approx \frac{h_k}{2} \left[ f(a) + f(b) + 2 \sum_{i=1}^{N-1} f(x_i) \right], \quad (8)$$

where the spacing  $h = h_k$  is now reduced as

$$h_k = \frac{b-a}{N} = \frac{b-a}{2^{k-1}}, \quad k = 1, 2, 3, \dots \quad (9)$$

Richardson extrapolation is applied  $k - 1$  times for each  $k$  value to the previously computed approximation, so both mechanisms are applied simultaneously. Basically,  $k$  is the number of different values of  $h$  computed initially related to the number of trapezoids (subintervals).

Then, the Romberg matrix,  $\mathbf{R}$ , is computed as

$$R_{11} = \frac{h_1}{2} [f(a) + f(b)], \quad (10)$$

$$R_{k,1} = \frac{1}{2} \left[ R_{k-1,1} + h_{k-1} \sum_{i=1}^{2^{k-2}} f(a + (2i-1)h_k) \right], \quad k > 1 = j, \quad (11)$$

$$R_{k,j} = R_{k,j-1} + \frac{R_{k,j-1} - R_{k-1,j-1}}{4^{j-1} - 1}, \quad k \geq j > 1, \quad (12)$$

or

$$\mathbf{R} = \begin{pmatrix} R_{11} & & & & \\ R_{21} & R_{22} & & & \\ R_{31} & R_{32} & R_{33} & & \\ R_{41} & R_{42} & R_{43} & R_{44} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \\ \uparrow & \uparrow & \uparrow & \uparrow & \\ \mathcal{O}(h^2) & \mathcal{O}(h^4) & \mathcal{O}(h^6) & \mathcal{O}(h^8) & \dots \end{pmatrix}. \quad (13)$$

The algorithm is the following (algorithm 2).

To compute the integral,  $n = 10$  is chosen so the  $10 \times 10$  Romberg matrix is computed, although the convergence criteria of  $\varepsilon = 10^{-8}$  is followed to store both the converged value and its position in the matrix.

The resulting  $10 \times 10$  Romberg matrix is printed in table 2, and the converged value is

- $R_{7,5} = 1.05765069$
- Number of iterations: 25

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**Algorithm 2** Romberg method

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```
1: procedure ROMBERGIA( $a, b, \varepsilon, n, \mathbf{R}, I, r_1, r_2, t$  )  
   Input. Integration limits,  $a, b$ ; threshold for convergence,  $\varepsilon$ ; dimensions of the Romberg matrix,  
    $n$   
   Output. Romberg matrix,  $\mathbf{R}$ ; Converged value of the integral,  $I$  ; position of the converged  
   element,  $R(r_1, r_2)$ ; number of iterations until convergence,  $t$   
2:   Initialize Romberg matrix,  $\mathbf{R} = 0$ ; number of iterations,  $t' = 1$ ; dummy variable  $d = 0$   
3:   Compute the first element,  $R_{1,1}$   
4:   for  $i \leftarrow 2, n$  do  
5:     Compute  $R_{i,1}$   
6:     for  $j \leftarrow 2, i$  do  
7:       Compute  $R_{i,j}$   
8:       Calculate the difference with the previous value,  $\Delta R = |R_{i,j} - R_{i,j-1}|$   
9:       if  $\Delta R < \varepsilon$  and  $d = 0$  then  
10:        Store the converged value,  $I = R_{i,j}$   
11:        Store the indexes of the converged value,  $r_1 = i$  and  $r_2 = j$   
12:        Store number of iterations until convergence,  $t = t'$   
13:        Update a dummy variable to not reassign  $r_1$  and  $r_2$  after convergence,  $d = 1$   
14:      Update number of iterations,  $t' = t' + 1$ 
```

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**Table 2:** Resulting  $10 \times 10$  Romberg matrix.

$\mathbf{R}$	1	2	3	4	5	6	7	8	9	10
1	0.70956602									
2	0.25162414	0.09897684								
3	0.85142421	1.05135757	1.11484962							
4	1.01000877	1.06287028	1.06363780	1.06282491						
5	1.04595938	1.05794291	1.05761442	1.05751881	1.05749800					
6	1.05474110	1.05766834	1.05765004	1.05765060	1.05765112	1.05765127				
7	1.05692411	1.05765178	1.05765068	1.05765069	1.05765069	1.05765069	1.05765069			
8	1.05746909	1.05765076	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069		
9	1.05760529	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	
10	1.05763934	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069	1.05765069

### 3 Gauss-Legendre method

The Gaussian Quadrature is an interpolation method that, unlike the Newton-Cotes methods, picks optimal abscissa points,  $x_i$ , at which the function is evaluated,  $f(x_i)$ , unequally spaced.

The integration rule is exact for polynomials of degree up to  $2n - 1$ , where  $n$  are the appropriately chosen abscissa points. Therefore, the maximum degree of accuracy is  $2n - 1$ . Also, not only the abscissa points can be chosen, but also the weights  $\omega_i$ .

The integral is evaluated as

$$\int_a^b f(x) dx \approx \int_a^b P_{2n-1}(x) dx = \sum_{i=1}^n \omega_i f(x_i), \quad i = 1, 2, \dots, n \quad (14)$$

The abscissa points can be found for any particular case based in the Fundamental theorem of Gaussian Quadrature [1].

**Theorem 1 (Fundamental theorem of Gaussian Quadrature)** *The abscissas of the  $n$ -point Gaussian quadrature formulas with weighting function  $W(x)$  in the interval  $(a, b)$  are precisely the roots of the orthogonal polynomial  $P_n(x)$  for the same interval and weighting function.*

The solution relates to the orthogonal polynomials generated by the weight function  $W(x)$

$$\int_a^b f(x) dx \approx \int_c^d W(t) P_{2n-1}(t) dt = \sum_{i=1}^n \omega_i P(t_i). \quad (15)$$

where  $t_i$  are the roots of the orthogonal polynomial  $P_{2n-1}$  and are not equally spaced. Therefore,  $P_{2n-1}(t)$  is optimized for the specific range  $t \in [c, d]$ , and only the roots  $t_i$  (abscissa points) and weights,  $\omega_i$ , have to be determined.

In this case, the Gauss-Legendre polynomials are used. The weight functions for these polynomials is, simply

$$W(x) = 1, \quad (16)$$

the interval is

$$-1 < x < 1, \quad (17)$$

and the recurrence relation, needed to compute the polynomials

$$(k+1) P_{k+1}(x) = (2k+1) x P_k(x) - k P_{k-1}(x). \quad (18)$$

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**Algorithm 3** Gauss-Legendre method

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```

1: procedure GAUSSQUADRATURE( $a, b, N_0, N_{\text{tot}}, \varepsilon, I, t$ )
   Input. Integration limits,  $a, b$ ; starting number of quadrature points,  $N_0$ ; total number of
   quadrature points,  $N_{\text{tot}}$ ; threshold for convergence,  $\varepsilon$ 
   Output. Value of the integral,  $I$ ; total number of iterations,  $t$ 
2:   Initialize the value of the integral,  $I = 0$ 
3:   Calculate  $c = (a + b) / 2$  and  $m = (b - a) / 2$ 
4:   for  $n \leftarrow N_0, N_{\text{tot}}$  do
5:     Compute the weights,  $\omega_i = (\omega)_i$ , and  $t_i = (\mathbf{t})_i$  values with GAULEG( $i, \mathbf{t}, \omega$ )
6:     Compute the integral,  $I' = m \sum_{i=1}^n \omega_i f(c + m t_i)$ 
7:     Calculate the difference with the previous value,  $\Delta I = |I' - I|$ 
8:     if  $\Delta I < \varepsilon$  and  $\Delta I > 0$  then exit
9:     Update the value of the integral,  $I = I'$ 
10:    Update total number of iterations,  $t = n$ 

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The results are collected in table 3, and

- Number of quadrature points employed: 10
- Final value of the quadrature: 1.05765069

## 4 Results

The results from the three methods can be summarized as in table 4 and fig. 1.

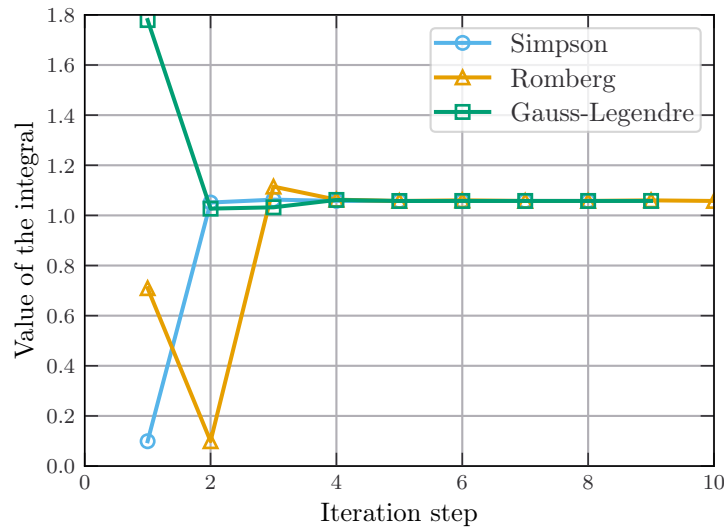
From the table 4, the fastest method to reach convergence is the Composite Simpson rule, followed by the Romberg's method (15.3% slower than Simpson's) and, lastly, the Gauss-Legendre (437.2% slower than Simpson's).

**Table 3:** Results from the Gauss-Legendre method.

Quadrature points	Integral values	Difference, $\Delta I =  I_{n+1} - I_n $
2	1.77932651	1.77932651
3	1.02706235	0.75226416
4	1.03224500	0.00518265
5	1.06169102	0.02944602
6	1.05749649	0.00419453
7	1.05763955	0.00014306
8	1.05765184	0.00001229
9	1.05765067	0.00000118
10	1.05765069	0.00000002

**Table 4:** Results from the Composite Simpson rule, Romberg's method and Gauss-Legendre method.

Method	Total number of iterations	Execution time (ms)
Composite Simpson rule	9	1.83
Romberg's method	25	2.11
Gauss-Legendre method	10	9.83

**Figure 1:** Integral value with increasing iterations for each of the methods used.

## Part II

# Root finding and function optimization

## References

1. Press, WH; Teukolsky, SA; Vetterling, WT; Flannery, BP Numerical Recipes in Fortran 77, 1986.