# Integration

$$S = \int_{a}^{b} f(x) dx$$

- ⇒ Exact integration
- ⇒ Simple numerical methods
- ⇒ Advanced numerical methods

# Part 1

## Exact integration

#### Three possible ways for exact integration

- Standard techniques of integration substitution rule, integration by parts, using identities, ...
- → Tables of integrals
- → Computer algebra systems

#### Tables of integrals

#### Table of Integrals, Series and Products

by Gradshteyn I. S. and Ryzhik I. M. Academic Press, 1994 (many editions since 195x) (most referenced in physics)

#### Integral and Series, vol.1-3,

by Prudnikov A P, Brychkov Yu A and Marichev A I Gordon and Breach, New York, 1986 (most sophisticated)

**Tables of Integrals and Other Mathematical Data**by Herbert B. Dwight(very simple integrals)

and many more ...

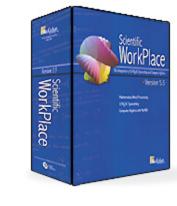
#### Computer algebra systems

- ➡ Maple
- ➡ Mathematica
- ➡ MathCad
- ⇒ Scientific Workplace

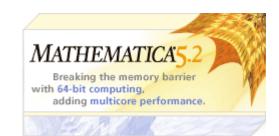
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➡ Derive



Maple (





## Part 2

## Basic ideas

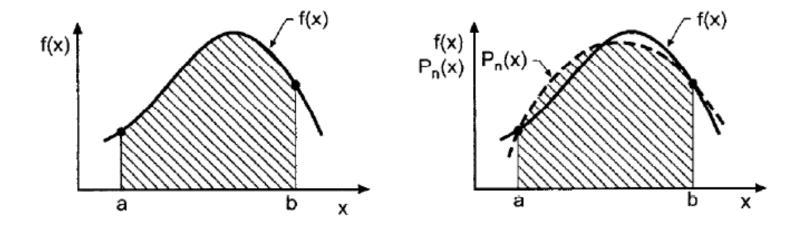
# Quite often we need numerical integrations

- if you can not get an analytic answer using Tables of integrals
   Computer algebra systems
   ... and various calculus books
- ⇒ if you have a discrete set of data points, i.e.  $f_i(x_i)$ as a result of measurements or calculations

### Integrating approximating functions

Numerical integration can be based on fitting approximating functions (polynomials) to discrete data and integrating approximating functions

$$I = \int_{a}^{b} f(x) dx \cong \int_{a}^{b} P_{n}(x) dx$$



## Integrating approximating functions

Case 1:

The function to be integrated is known only at a finite set of discrete points

Parameters under control – the degree of approximating polynomial

## Integrating approximating functions

Case 2:

The function to be integrated is known.

Parameters under control

- ⇒ The total number of discrete points
- The degree of the approximating polynomial to represent the discrete data.
- The locations of the points at which the known function is discretized

## Part 3a

## Direct fit polynomials

## Direct fit polynomials

The procedure can be used for both unequally and equally spaced data It is based on fitting the data by a direct fit polynomial and integrating that polynomial.

$$f(x) \cong P_n(x) = a_0 + a_1 x + a_2 x^2 + \dots$$
  
then

$$I = \int_{a}^{b} f(x) dx \cong \int_{a}^{b} P_{n}(x) dx = \left(a_{0}x + a_{1}\frac{x^{2}}{2} + a_{2}\frac{x^{3}}{3} + \dots\right)_{a}^{b}$$

## Part 3b

# Quadrature methods on equal subintervals

#### Riemann Integral

If f(x) is a continuous function defined for  $a \le x \le b$ and we divide the interval into n subintervals of equal width  $\Delta x = (b-a)/n$  then the definite integral is

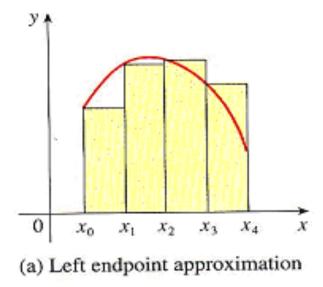
$$\int_{a}^{b} f(x)dx = \lim_{n \to \infty} \sum_{i=1}^{n} f(x_{i}^{*})\Delta x_{i}$$

The Riemann integral can be interpreted as a net area under the curve y = f(x) from a to b

Bernhard Riemann, 1826-1866, German mathematician

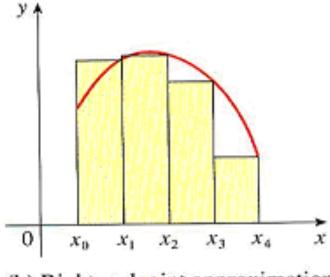
#### 1. The left endpoint Riemann sum

$$\int_{a}^{b} f(x)dx = L_{n} = \lim_{n \to \infty} \sum_{i=1}^{n} f(x_{i-1})\Delta x, \quad \Delta x = (b-a)/n$$



#### 2. The right endpoint Riemann sum

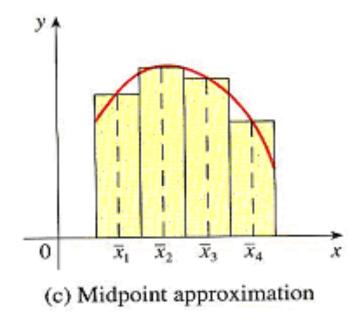
$$\int_{a}^{b} f(x)dx = R_n = \lim_{n \to \infty} \sum_{i=1}^{n} f(x_i)\Delta x, \quad \Delta x = (b-a)/n$$



(b) Right endpoint approximation

3. The midpoint Riemann sum

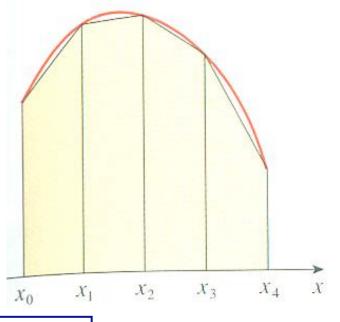
$$\int_{a}^{b} f(x)dx = R_{n} = \lim_{n \to \infty} \sum_{i=1}^{n} f(\frac{x_{i-1} + x_{i}}{2})\Delta x, \quad \Delta x = (b-a)/n$$



#### 4. The trapezoidal rule

The area of the trapezoid that lies above the i-th subinterval

$$S_i = \frac{\Delta x}{2} \left( f(x_{i-1}) + f(x_i) \right)$$



$$\int_{a}^{b} f(x)dx = \frac{\Delta x}{2} [f(x_0) + 2f(x_1) + 2f(x_2) + \dots \\ 2f(x_{n-1}) + f(x_n)]$$

### Example: C++

```
// Integration by the trapezoidal rule of f(x) on [a,b]
// f - Function to integrate (supplied by a user)
// a - Lower limit of integration
// b - Upper limit of integration
// r - Result of integration (out)
// n - number of intervals
double int trap(double(*f)(double),
                double a, double b, int n)
Ł
   double r, dx, x;
    r = 0.0;
    dx = (b-a)/n;
    for (int i = 1; i \le n-1; i=i+1)
    {
       x = a + i * dx;
        r = r + f(x);
    }
    r = (r + (f(a)+f(b))/2.0) * dx;
    return r;
```

#### Let us talk about ... interpolation

First-order interpolation for the i-th subinterval

$$f(x) = f(x_{i-1}) + \frac{f(x_i) - f(x_{i-1})}{\Delta x} (x - x_{i-1})$$

Integral for the i-th subinterval

$$\int_{x_{i-1}}^{x_i} f(x) dx = \int_{x_{i-1}}^{x_i} \left[ f(x_{i-1}) + \frac{f(x_i) - f(x_{i-1})}{\Delta x} (x - x_{i-1}) \right] dx =$$
$$= \frac{\Delta x}{2} \left( f(x_{i-1}) + f(x_i) \right)$$

Trapezoidal approximation is application of the 1st order interpolation for the each subinterval

# Integration and the second order interpolation

Using the three-point interpolation one may write Simpson's Rule for integration

$$\int_{a}^{b} f(x)dx = \frac{\Delta x}{3} [f(x_{0}) + 4f(x_{1}) + 2f(x_{2}) + 4f(x_{3}) + \dots + 2f(x_{n-2}) + 4f(x_{n-1}) + f(x_{n})]$$

Number of n intervals should be even. If n is odd then the last interval should be treated by some other way



Useful exercise: Derive the Simpson rule with a pair of slices with an equal interval by using second-order interpolation for f(x) in the region  $[x_{i-1}, x_{i+1}]$ .

Thomas Simpson, 1710-1761, England

```
Example: C++
```

```
// integration by Simpson rule of f(x) on [a,b]
// f - Function to integrate (supplied by a user)
// s - Result of integration (out)
// n - number of intervals
double int simp(double(*f)(double),
                 double a, double b, int n)
  double s, dx, x;
// if n is odd we add +1 interval to make it even
    if ((n/2) * 2 != n) \{n=n+1;\}
    s = 0.0;
   dx = (b-a)/n;
 for ( int i=0; i <n/2; i=i+1)</pre>
    \{ x = a + 2 * i * dx; \}
        s = s + 4.0/3.0*f(x);
 for ( int i=1; i <n/2; i=i+1)</pre>
    \{ x = a + 2 * i * dx; \}
        s = s + 2.0/3.0 f(x);
    s = (s + (f(a) + f(b))/3.0) * dx;
    return s;}
```

## Part 3c

### Newton-Cotes formulas

#### Equally spaced points -> Newton forward difference polynomials

Newton forward/backward/centered difference polynomial are very useful in interpolation and numerical differentiation. Let's consider numerical integration

$$I = \int_{a}^{b} f(x) dx \cong \int_{a}^{b} P_{n}(x) dx$$

where

$$P_n(x) = f_0 + s\Delta f_0 + \frac{s(s-1)}{2!}\Delta^2 f_0 + \frac{s(s-1)(s-2)}{3!}\Delta^3 f_0 + \dots + \frac{s(s-1)(s-2)\dots(s-[n-1])}{n!}\Delta^n f_0 + \text{Error}$$

 $s = (x - x_0)/h, \quad x = x_0 + sh$ <sup>24</sup>

#### Comments

using s as a variables

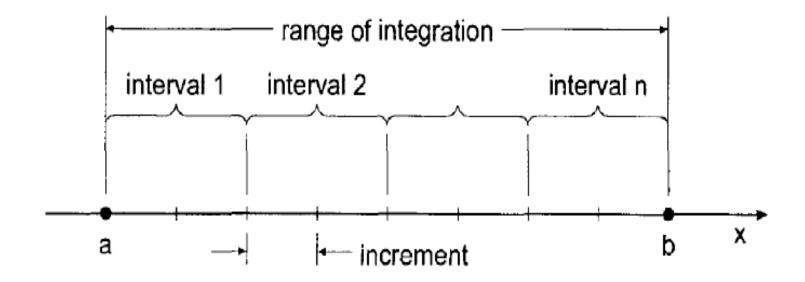
$$I = \int_{a}^{b} f(x)dx \cong \int_{a}^{b} P_{n}(x)dx = h \int_{s(a)}^{s(b)} P_{n}(s)ds = h \int_{0}^{s} P_{n}(x_{0} + sh)ds$$

where

dx = hds

Each choice of the degree n of the interpolating polynomial yields a different Newton-Cotes formula. See Abramowitz and Stegun (1964) for many high-order formulas





#### The trapezoid rule (revisited)

A first degree polynomial for a single interval (two points)

$$\Delta I = h \int_{0}^{1} (f_0 + s \Delta f_0) ds = h \left( s f_0 + \frac{s^2}{2} \Delta f_0 \right)_{0}^{1}$$

where

$$s = 1, \quad h = \Delta x, \quad \Delta f_0 = (f_1 - f_0)$$
$$\Delta I = h(f_0 + \frac{1}{2}\Delta f_0) = \frac{1}{2}h(f_0 + f_1)$$

applying over all the intervals

$$I = \sum_{i=0}^{n-1} \Delta I_i = \sum_{i=0}^{n-1} \frac{1}{2} h(f_i + f_{i+1}) = \frac{1}{2} h(f_0 + 2f_1 + 2f_2 + \dots + 2f_{n-1} + f_n)$$

#### The trapezoid rule (revisited)

The error estimation can be done by integrating the error term in the polynomial for a single interval

$$Error = h \int_{0}^{1} \frac{s(s-1)}{2} h^{2} f''(\zeta) ds = \frac{1}{12} h^{3} f''(\zeta) = O(h^{3})$$

the total error

Total error = 
$$\sum_{i=0}^{n-1} Error_i = -\frac{1}{12}(x_n - x_0)h^2 f''(\zeta) = O(h^2)$$

where

$$x_0 \leq \zeta \leq x_n$$

#### The Simpson's rule (revisited)

Simpson's 1/3 rule = a second degree polynomial for two intervals (three equally spaced points)

upper limit of integration for a single interval s=2

$$\Delta I = h \int_{0}^{2} \left[ f_{0} + s \Delta f_{0} + \frac{s(s-1)}{2!} \Delta^{2} f_{0} \right] ds$$

using definitions for  $\Delta f_0, \Delta^2 f_0$ 

$$\Delta I = \frac{1}{3}h(f_0 + 4f_1 + f_2)$$

applying over all the intervals

$$I = \frac{1}{3}h(f_0 + 4f_1 + 2f_2 + 4f_3 + \dots + 4f_{n-1} + f_n)$$

#### The Simpson's rule (revisited)

The error estimation can be done by integrating the error term in the polynomial

for a single interval

$$Error = h \int_{0}^{2} \frac{s(s-1)(s-2)}{6} h^{3} f'''(\zeta) ds = 0$$

the next order

$$Error = h \int_{0}^{2} \frac{s(s-1)(s-2)(s-3)}{24} h^{4} f''''(\zeta) ds = -\frac{1}{90} h^{5} f''''(\zeta)$$

and

*Total error* =  $O(h^4)$ 

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#### The Simpson's 3/8 rule

Simpson's 3/8 rule = a third degree polynomial for four equally spaced points

upper limit of integration for a single interval s=3

$$\Delta I = h \int_{0}^{3} \left[ f_{0} + s \Delta f_{0} + \frac{s(s-1)}{2!} \Delta^{2} f_{0} + \frac{s(s-1)(s-2)}{6} \Delta^{3} f \right] ds$$

using definitions for  $\Delta f_0$ ,  $\Delta^2 f_0$ ,  $\Delta^3 f$ 

$$\Delta I = \frac{3}{8}h(f_0 + 3f_1 + 3f_2 + f_3)$$

applying over all the intervals

$$I = \frac{3}{8}h(f_0 + 3f_1 + 3f_2 + 2f_3 + \dots + 3f_{n-1} + f_n)$$

#### The Simpson's 3/8 rule

total number of increments must be a multiple of three

Error for a single interval

$$Error = h \int_{0}^{3} \frac{s(s-1)(s-2)(s-3)}{24} h^{4} f''''(\zeta) ds = -\frac{3}{80} h^{5} f''''(\zeta)$$

and

Total error =  $O(h^4)$ 

the same order of the error as Simpson's 1/3 rule!

#### Other view on numerical integration

Quadrature is weighted sum of finite number of sample values of integrand function

$$\int_{a}^{b} f(x) dx \approx \sum_{i=1}^{n} f(x_i) W_i$$

trapezoid approximations Simpson's rule

$$w_i = (\Delta x / 2, \Delta x / 2)$$
$$w_i = (\Delta x / 3, 4\Delta x / 3, \Delta x / 3)$$

### Equal interval integration (quadrature)

Elementary weights for uniform-step integration rules

name	degre	e elementary weights
trapezoid	1	(h/2, h/2)
Simpson's	2	(h/3, 4h/3, h/3)
3/8	3	(3h/8, 9h/8, 9h/8, 3h/8)
Milne	4	(14h/45, 64h/45, 24h/45, 64h/45, 14h/45)

#### Integration error

Generally as N increases, the precision of a method increases However, as N increases, the round-off error increases

Some evaluations (not exact but gives an idea)\*:

#### Number of steps for highest accuracy:

trapezoid rule	steps	error
single precision	631	3*10 <sup>-6</sup>
double precision	10 <sup>6</sup>	10 <sup>-12</sup>
Simpson's rule	steps	error
single precision	36	6*10 <sup>-7</sup>
double precision	2154	5*10 <sup>-14</sup>

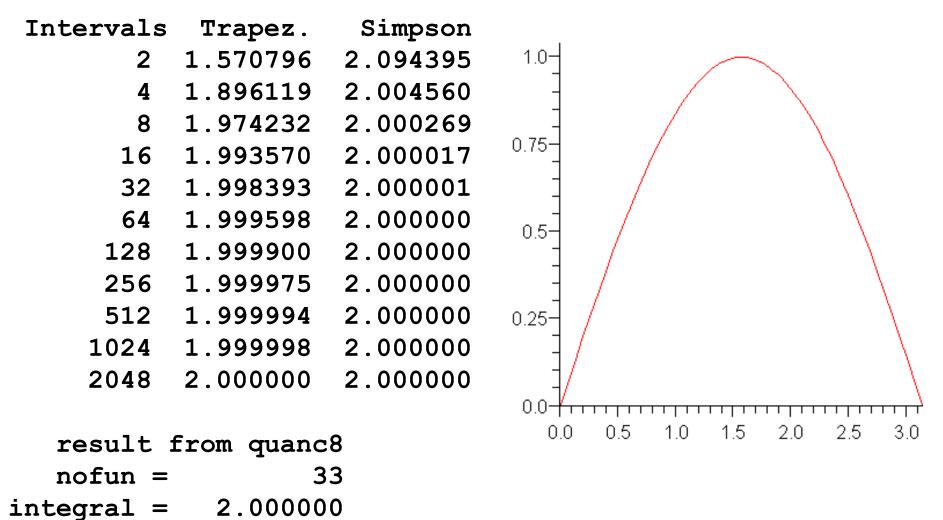
\* see details in R.H. Landau & M.J.Paez, An Introduction to Computational Physics

#### Integration error (cont.)

The best numerical evaluation of an integral can be obtained with a relatively small number is sub-intervals (N~1000) (not with N  $\rightarrow \infty$ )

It is possible to get an error close to machine precision with Simpson's rule and with other higher-order methods (Newton-Cotes quadrature)

# Example $\int_{0}^{\pi} \sin(x) dx = 2.0$



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Example 
$$\int_{0}^{\pi} \frac{x}{x^{2}+1} \cos(10x^{2}) dx = 0.0003156$$
  
Intervals Trapez. Simpson  
2 0.578769 0.811200  
4 0.813285 0.891458  
8 0.688670 0.647131  
16 0.285919 0.151669  
32 0.049486 -0.029325  
64 0.004360 -0.010682  
128 0.001183 0.000124  
256 0.000526 0.000306  
512 0.000368 0.000315  
1024 0.000329 0.000316  
2048 0.000319 0.000316  
4096 0.000316 0.000316  
8192 0.000316 0.000316  
16384 0.000316 0.000316  
16384 0.000316 0.000316  
16384 0.000316 0.000316  
result from quanc8  
nofun = 1601  
integral = 0.0003156

### Richardson Extrapolation and Romberg Integration

$$I = \int_{a}^{b} f(x) dx$$

Key idea – use the error estimation to extrapolate integrals' values

$$Error(h/R) = \frac{1}{R^n - 1} [I(h/R) - I(h)]$$

where

- R is the ratio of the increment size
- n is the global order of the algorithm

Extrapolated value = f(h/R) + Error(h/R)

When extrapolation is applied to numerical integration by the trapezoid rule, the result is called Romberg integration

### **Romberg Integration**

Error in the trapezoid rule has the functional form

*Error* = 
$$C_1h^2 + C_2h^4 + C_3h^6 +$$
  
so  $n = 2$  and for  $R = 2$   
 $error(h/2) = \frac{1}{2^n - 1}[I(h/2) - I(h)] = \frac{1}{3}[I(h/2) - I(h)]$   
Extrapolated value =  $f(h/2) + error(h/2) + O(h^4)$ 

# Part 4

## Gaussian quadrature



Gaussian quadrature

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

Flexibility for known functions – to choose n points  $x_i$  and  $c_i$  so that the integral of a polynomial of degree 2n-1 is exact.

Gaussian integration produces higher accuracy than the Newton-Cotes formulas with the same number of function evaluations.

If the function to integrate is not smooth, then Gaussian quadrature may give lower accuracy

Example for n=2 and 2n-1=3 
$$\int_{a}^{b} f(x)dx \approx \sum_{i=1}^{n} c_{i}f(x_{i})$$
  
for  $f(x) = a_{0} + a_{1}x + a_{2}x^{2} + a_{3}x^{3}$   
$$\int_{a}^{b} f(x)dx \approx c_{1}f(x_{1}) + c_{2}f(x_{2})$$
  
$$= c_{1}\left(a_{0} + a_{1}x_{1} + a_{2}x_{1}^{2} + a_{3}x_{1}^{3}\right) + c_{2}\left(a_{0} + a_{1}x_{2} + a_{2}x_{2}^{2} + a_{3}x_{2}^{3}\right)$$
  
$$\int_{a}^{b} f(x)dx = \int_{a}^{b} \left(a_{0} + a_{1}x + a_{2}x^{2} + a_{3}x^{3}\right) dx$$
  
$$= \left[a_{0}x + a_{1}\frac{x^{2}}{2} + a_{2}\frac{x^{3}}{3} + a_{3}\frac{x^{4}}{4}\right]_{a}^{b}$$

$$= a_0(b-a) + a_1\left(\frac{b^2 - a^2}{2}\right) + a_2\left(\frac{b^3 - a^3}{3}\right) + a_3\left(\frac{b^4 - a^4}{4}\right)$$

Example for n=2 
$$\int_{a}^{b} f(x)dx \approx \sum_{i=1}^{n} c_{i}f(x_{i})$$
  
and

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

$$b - a = c_1 + c_2$$

$$\frac{b^2 - a^2}{2} = c_1 x_1 + c_2 x_2$$

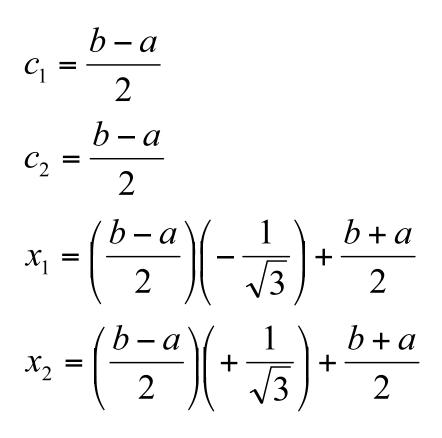
$$\frac{b^3 - a^3}{3} = c_1 x_1^2 + c_2 x_2^2$$

$$\frac{b^4 - a^4}{4} = c_1 x_1^3 + c_2 x_2^3$$

four unknowns and four equations

### Example for n=2

and after some tedious work ...



 $\int f(x)dx \approx \sum_{i=1}^{n} c_i f(x_i)$ 

Another way. Choose  $t_1$ ,  $t_2$ ,  $c_1$ ,  $c_2$  so that *I* is exact for the following four polynomials: f(t) = 1, t,  $t^2$ ,  $t^3$  (and use a=-1, b=1)

$$I[F(t) = 1] = \int_{-1}^{1} (1) dt = t \Big|_{-1}^{1} = 2 = C_{1}(1) + C_{2}(1) = C_{1} + C_{2}$$
$$I[F(t) = t] = \int_{-1}^{1} t dt = \frac{1}{2}t^{2} \Big|_{-1}^{1} = 0 = C_{1}t_{1} + C_{2}t_{2}$$
$$I[F(t) = t^{2}] = \int_{-1}^{1} t^{2} dt = \frac{1}{3}t^{3} \Big|_{-1}^{1} = \frac{2}{3} = C_{1}t_{1}^{2} + C_{2}t_{2}^{2}$$
$$I[F(t) = t^{3}] = \int_{-1}^{1} t^{3} dt = \frac{1}{4}t^{4} \Big|_{-1}^{1} = 0 = C_{1}t_{1}^{3} + C_{2}t_{2}^{3}$$

Solving the system gives

$$C_1 = C_2 = 1$$
 and  $t_1 = -\frac{1}{\sqrt{3}}$ ,  $t_2 = \frac{1}{\sqrt{3}}$ 

$$I = \int_{-1}^{1} F(t)dt = F(-\frac{1}{\sqrt{3}}) + F(\frac{1}{\sqrt{3}})$$
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# Gaussian quadratures for $I = \int_{a}^{b} f(x) dx$

transformation between x and t space

x = mt + cwhere  $x = a \rightarrow t = -1$ ,  $x = b \rightarrow t = 1$  and dx = m dtthen a = m(-1) + c and b = m(1) + c $m = \frac{b-a}{2}, c = \frac{b+a}{2}$  and  $x = \frac{b-a}{2}t + \frac{b+a}{2}$  $I = \int_{-\infty}^{b} f(x)df = \int_{-\infty}^{1} f[x(t)]dt = \int_{-\infty}^{1} f(mt+c)mdt$ F(t) = f[x(t)] = f(mt + c) $I = \frac{b-a}{2} \int F(t) dt = \frac{b-a}{2} \sum_{i=1}^{n} C_{i} F(t_{i})$ 

	$C_i$	
$-1/\sqrt{3}$	1	
$1/\sqrt{3}$	1	
$-\sqrt{0.6}$	5/9	
0	8/9	
$\sqrt{0.6}$	5/9	
-0.8611363116	0.3478548451	
-0.3399810436	0.6521451549	
0.3399810436	0.6521451549	
0.8611363116	0.3478548451	

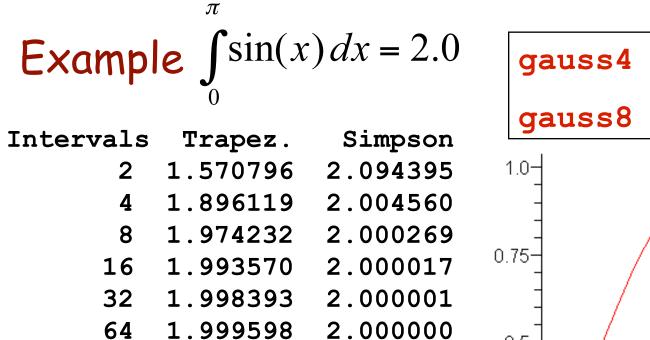
#### Example: Gaussian quadrature parameters

## Example: C++

```
/* Numerical integration of f(x) on [a,b]
    method: Gauss (4 points)
input:
       - a single argument real function
    f
    a,b - the two end-points (interval of integration)
output: r - result of integration
*/
  double gauss4(double(*f)(double), double a, double b)
Ł
    const int n = 4;
    double ti[n] = \{-0.8611363116, -0.3399810436,
                     0.3399810436, 0.8611363116};
    double ci[n] = \{ 0.3478548451, 0.6521451549, \}
                     0.6521451549, 0.3478548451;
    double r, m, c;
    r = 0.0;
    m = (b-a)/2.0;
    c = (b+a)/2.0;
    for (int i = 1; i <= n; i=i+1)</pre>
    {r = r + ci[i-1]*f(m*ti[i-1] + c); }
    r = r*m;
    return r;
```

## Example: C++

```
/* Numerical integration of f(x) on [a,b]
   method: Gauss (8 points using symmetry)
input:
      - a single argument real function
   f
   a,b - the two end-points (interval of integration)
output: r - result of integration */
 double gauss8(double(*f)(double), double a, double b)
   const int n = 4;
   double ti[n] = \{0.1834346424, 0.5255324099,
                  0.7966664774, 0.9602898564};
   0.2223810344, 0.1012285362;
   double r, m, c;
   r = 0.0;
   m = (b-a)/2.0;
   c = (b+a)/2.0;
   for (int i = 1; i \le n; i=i+1)
   {r=r+ci[i-1]*(f(m*(-1.0)*ti[i-1]+c)+f(m*ti[i-1]+c));
   r = r m;
   return r;
```



1.999900

1.999975

1.999994

1.999998

2.000000

2.000000

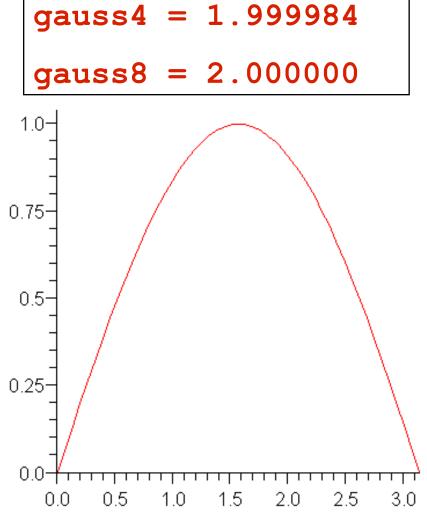
2.000000

2.000000

2.000000

2.000000

2.000000



result fr	om quanc8
nofun =	33
integral =	2.000000

64

128

256

512

1024

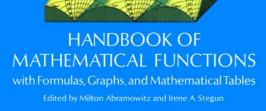
2048

51

### Gaussian quadratures

w(x)Gauss-Legendre[-1, +1]1Gauss-Jacobi(-1, +1) $(1-x)^{\alpha}(1+x)^{\beta}$ Gauss-Chebyshev(-1, +1) $1/\sqrt{1-x^2}$ Gauss-Hermite $(-\infty, +\infty)$  $\exp(-x^2)$ Gauss-Laguerre $[0, +\infty)$  $\exp(-x)$ 

Tables with coefficients can be found in "Handbook of Mathematical Functions, With Formulas, Graphs, and Mathematical Tables" by Abramowitz and Stegun.



b

The aim of an automatic integration scheme is to relieve the person who has to compute an integral of any need to think.

Davis P. J., and P. Rabinowitz, Methods of Numerical Integration (Dover, 2nd edition) (2007)

# Part 5

## Automatic and Adaptive Integration



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### Automatic and Adaptive Integration

A favorite research topic since the 1960s.

User-friendly routines where the user enters 1) the limits of integration, 2) selects the routine for computation of f(x), 3) provides a tolerance  $\varepsilon$ , and 4) enters the upper bound N for the number of functional computations.

$$\left|I_{a}^{b}(f) - I\right| \le \varepsilon \quad \text{or} \frac{\left|I_{a}^{b}(f) - I\right|}{I_{a}^{b}(\left|f\right|)} \le \varepsilon$$

Then the program exits either

 with the computed value which is correct within the ε
 with a statement that the upper bound N was attained but the tolerance was not achieved, and the computed result may be the "best" value of the integral determined by the program.

### Automatic and Adaptive Integration

Automatic integration falls into two classes: iterative or noniterative, and adaptive or non-adaptive.

The iterative schemes: computing successive approximations to the integral until an agreement with the tolerance is achieved,

The non-iterative schemes: the information from the first approximation is carried over to generate the second approximation, which then becomes the final result.

### Adaptive Integration

The adaptive schemes: the points at which the integration is carried out are chosen in a manner that is dependent on the nature of the integrand – the domain of integration is selectively refined to reflect behavior of particular integrand function on a specific subinterval.

The non-adaptive schemes: the integration points are chosen in a fixed manner which is independent of the nature of the integrand, although the number of these points depends on the integrand - continue to subdivide all subintervals, say by half, until overall error estimate falls below desired tolerance (not an inefficient way). 56

# Adaptive programs tend to be effective in practice ... but it can be fooled

Interval of integration may be very wide but "interesting" behavior of integrand is confined to narrow range

Sampling by automatic routine may miss interesting part of integrand behavior, and resulting value for integral may be completely wrong

# Part 6

# "Special cases"



### Integrals with oscillating functions

```
\int_{a}^{b} f(x) \cos^{n}(\omega x) dx
```

Use methods or programs specially designed to calculate integrals with oscillating functions:

# Filon's method

# Clenshaw-Curtis method

$$\frac{x}{x^2+1}\cos(1.2x^2)$$

#### Improper Integrals: Type 1 - Infinite Intervals $\infty$

 $\sim$ 

$$\int_{a}^{\infty} f(x)dx \qquad \int_{-\infty}^{\infty} f(x)dx$$

- 1. Transform variable of integration so that the new interval is finite: examples: y=exp(-x), then  $[0,\infty]$  into [0,1](but: not to introduce singularities)
- 2. Replace infinite limits of integration by carefully chosen finite values.
- 3. Use asymptotic behavior (if possible) to evaluate the "tail" contribution.
- 4. Use nonlinear quadrature rules designed for infinite intervals

# example: replace infinite limits of integration by finite values

$$\int_{0}^{\infty} f(x)dx = \lim_{r \to \infty} \int_{0}^{r} f(x)dx$$
  

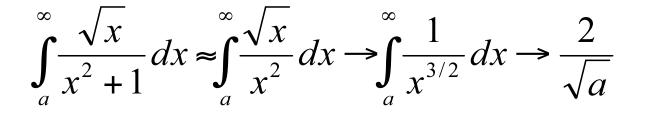
$$I_{n} = \int_{0}^{r_{n}} f(x)dx \quad \text{where } r_{n} = 2^{n} \qquad I_{n} = \int_{0}^{r_{n}} \frac{e^{-x}}{1+x^{4}}dx, \quad r_{n} = 2^{n}.$$

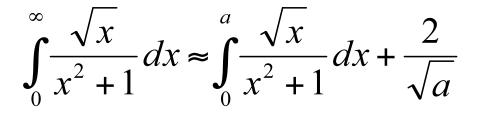
n	I <sub>n</sub>	Number of functional evaluations
0	.5720 2582	35
1	.6274 5952	52
2	.6304 3990	100
3	.6304 7761	178
4	.6304 7766	322
Exact	.6304 7783	

#### example: using asymptotic behavior

$$\int_{0}^{\infty} \frac{\sqrt{x}}{x^{2} + 1} dx = \int_{0}^{a} \frac{\sqrt{x}}{x^{2} + 1} dx + \int_{a}^{\infty} \frac{\sqrt{x}}{x^{2} + 1} dx$$

for a >> 1 we use the asymptotic behavior of the function





exact value

$$\int_{0}^{\infty} \frac{\sqrt{x}}{x^{2} + 1} dx = \pi \frac{\sqrt{2}}{2}$$

Example 
$$\int_{0}^{\infty} \frac{\sqrt{x}}{x^{2}+1} dx = \pi \frac{\sqrt{2}}{2} = 2.221441469$$

#### upper limit = 100 no "tail"

with the "tail" = 0.20000

Interva	ls Trapez	. Simpso	on Interva	als Trape	z. Simpson
2	0.166362	0.205151	2	0.366362	0.405151
4	0.321345	0.373006	4	0.521345	0.573006
8	0.536673	0.608449	8	0.736673	0.808449
16	0.833630	0.932615	16	1.033630	1.132615
32	1.218034	1.346168	32	1.418034	1.546168
64	1.619001	1.752657	64	1.819001	1.952657
128	1.873848	1.958797	128	2.073848	2.158797
256	1.970354	2.002522	256	2.170354	2.202522
512	2.003473	2.014513	512	2.203473	2.214513
1024	2.015099	2.018974	1024	2.215099	2.218974
2048	2.019202	2.020570	2048	2.219202	2.220570
4096	2.020652	2.021136	4096	2.220652	2.221136
8192	2.021165	2.021336	8192	2.221165	2.221336
16384	2.021346	2.021407	16384	2.221346	2.221407
32768	2.021410	2.021432	32768	2.221410	2.221432
result from	quanc8		result from	quanc8	
nofun =	76	9	nofun =	76	63 63
integral =	2.02144	5	integral =	2.22144	5

**Improper Integrals: Type 2 - Discontinuous Integrands**  $\int_{0}^{1} f(x) dx$  when f(x) is discontinuous at 0

Formal definition

. . .

$$\int_{0}^{1} f(x)dx = \lim_{t \to 0+} \int_{t}^{1} f(x)dx$$
Proceeding to the limit 
$$\int_{0}^{1} f(x)dx = \int_{R_{1}}^{1} f(x)dx + \int_{R_{2}}^{R_{1}} f(x)dx + \dots$$
Change variables
Elimination of the singularity
Gauss type quadratures

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Improper integrals 3: Integrals with integrable singularity

$$\int_{a}^{b} \frac{f(x)}{x-c} dx \qquad a \le c \le b$$

Method 1:

 $f(x) = \varphi(x) + \psi(x)$ 

where  $\varphi(x)$  can be integrated numerically and  $\psi(x)$  can be integrated analytically

example: 
$$f(x) = \frac{1}{\sqrt{x(1+x^2)}}$$
 (problem at  $x = 0$ )  
$$f(x) = \left(\frac{1}{\sqrt{x(1+x^2)}} - \frac{1}{\sqrt{x}}\right) + \frac{1}{\sqrt{x}}$$

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Improper integrals 3: Integrals with integrable singularity

$$\int_{a}^{b} \frac{f(x)}{x-c} dx \qquad a \le c \le b$$

Method 2:

 $f(x) = \varphi(x) \cdot \rho(x)$ 

then for some cases one of following quadrature rules can be used:

Gauss-Christoffel

Jacoby,

Chebyshev

Improper integrals 3: Integrals with integrable singularity  $\int_{-\infty}^{b} \frac{f(x)}{dx} dx \qquad a \le c \le b$ 

$$\frac{dx}{x-c} = ax \qquad a \le c$$

Method 3:

using non-standard quadrature rules allowing explicitly for the singularity

Method 4:

Use programs from trusted numerical libraries or books.

### Principal value integrals

$$\int_{0}^{\infty} \frac{f(x)}{x - x_0 \pm i\varepsilon} dx \quad \text{where } \varepsilon \to 0$$

$$\lim_{\varepsilon \to 0} \int_{0}^{\infty} \frac{f(x)}{x - x_0 \pm i\varepsilon} dx = P \int_{0}^{\infty} \frac{f(x)}{x - x_0} dx \mp i\pi f(x_0)$$

the definition of the principal value integral

$$P\int_{0}^{\infty} \frac{f(x)}{x - x_{0}} dx = \lim_{\mu \to 0} \left[ \int_{0}^{x_{0} - \mu} \frac{f(x)}{x - x_{0}} dx + \int_{x_{0} + \mu}^{\infty} \frac{f(x)}{x - x_{0}} dx \right]$$

it is possible to calculate it using regular methods, but... there is another way

### Principal value integrals (part 2)

$$\int_{0}^{\infty} \frac{f(x)}{x - x_{0}} dx \approx \int_{0}^{R} \frac{f(x)}{x - x_{0}} dx$$

$$\int_{0}^{R} \frac{f(x)}{x - x_{0}} dx = \int_{0}^{R} \frac{[f(x) - f(x_{0})]f(x)}{x - x_{0}} dx + f(x_{0}) \int_{0}^{R} \frac{1}{x - x_{0}} dx$$

$$\ln \frac{R - x_{0}}{-x_{0}} = \left[ \ln(-1) + \ln \frac{R - x_{0}}{x_{0}} \right] = \pm i\pi + \ln \frac{R - x_{0}}{x_{0}}$$

$$\int_{0}^{R} \frac{f(x)}{x - x_{0} \mp i\varepsilon} dx = \int_{0}^{R} \frac{[f(x) - f(x_{0})]f(x)}{x - x_{0}} dx + f(x_{0}) \ln \frac{R - x_{0}}{x_{0}} \pm i\pi f(x_{0})$$
then regular methods can be used since

$$\frac{\left[f(x) - f(x_0)\right]f(x)}{x - x_0} \quad \text{is smooth around} \quad x_0$$

### Double and multiple integrals

Use automatic one-dimensional quadrature routine for each dimension, one for outer integral and another for inner integral

Monte-Carlo method (effective for large dimensions)

$$\int_{0}^{1} dx_{1} \int_{0}^{1} dx_{2} \int_{0}^{1} dx_{3} \int_{0}^{1} dx_{4} \int_{0}^{1} dx_{5} \int_{0}^{1} dx_{6} \int_{0}^{1} (x_{1} + x_{2} + \dots + x_{7})^{2} dx_{7}$$

### Integrating Tabular Data

Reasonable approach is to integrate piecewise interpolant

Cubic spline interpolation could be a good method.



HANDBOOK OF Computational Methods for Integration

Prem K. Kythe Michael R. Schäferkotter



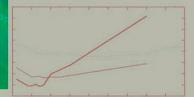
### PHILIP RABINOWITZ Methods of Numerical Integration Second Edition

Lecture Notes in Computer Science

848

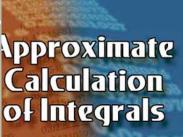
Arnold R. Krommer Christoph W. Ueberhuber

Numerical Integration on Advanced Computer Systems





Springer-Verlag



V. I. Krylov Translated by Arthur H. Stroud

### too sad

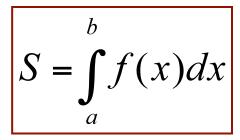
95 % of all practical work in numerical analysis boiled down to applications of Simpson's rule and linear interpolation.

Milton Abramowitz

from Davis P. J., and P. Rabinowitz, Methods of Numerical Integration (Dover, 2nd edition) (2007)



### Conclusion



- Analyze first: the existence of the integral
- ⇒ Transform the integral to a simpler form (if possible)
- Analyze the function: smooth or oscillating,
   functions with singularities, narrow peaks, ...
- Analyze to type of the integral (regular, improper, …)
- ⇒ Select a method that fits the function and the integral
- Always test any program for integration before using for real calculations.