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Computational Geodynamics

Numerical Integration

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The basic problem in numerical integration is to compute an approximate solution to a definite integral

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

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

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If f(x) is a smooth function, and the domain of integration is bounded, there are many methods for approximating the integral to the desired precision.





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- ► A formula for the integrand may be known, but it may be difficult or impossible to find an antiderivative that is an elementary function. An example of such an integrand is $f(x) = exp(-x^2)$, the antiderivative of which (the error function, times a constant) cannot be written in elementary form.



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- It may be possible to find an antiderivative symbolically, but it may be easier to compute a numerical approximation than to compute the antiderivative. That may be the case if the antiderivative is given as an infinite series or product, or if its evaluation requires a special function that is not available.

The simplest method of this type is to let the interpolating function be a constant function (a polynomial of degree zero) that passes through the point ((a+b)/2, f((a+b)/2)).

This is called the midpoint rule or rectangle rule.

$$\int_{a}^{b} f(x) dx \simeq (b-a) f(\frac{a+b}{2})$$





If we divide the interval into 4 subintervals:

$$\int_{0}^{1} f(x)dx$$

$$= \int_{0}^{0.25} f(x)dx + \int_{0.25}^{0.5} f(x)dx + \int_{0.5}^{0.75} f(x)dx + \int_{0.75}^{1} f(x)dx$$

$$\simeq 0.25f(0.125) + 0.25f(0.375) + 0.25f(0.625) + 0.25f(0.875)$$

The interpolating function may be a straight line (an affine function, i.e. a polynomial of degree 1) passing through the points (a, f(a)) and (b, f(b)).

This is called the trapezoidal rule.

$$\int_{a}^{b} f(x) dx \simeq (b-a) \frac{f(a) + f(b)}{2}$$





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$$= 0.25 \frac{f(0) + f(0.25)}{2} + 0.25 \frac{f(0.25) + f(0.5)}{2} + 0.25 \frac{f(0.75) + f(1)}{2}$$

$$= 0.25 \left(\frac{f(0)}{2} + f(0.25) + f(0.5) + f(0.75) + \frac{f(1)}{2}\right)$$



For either one of these rules, we can make a more accurate approximation by breaking up the interval [a, b] into some number n of subintervals, computing an approximation for each subinterval, then adding up all the results. The midpoint rule can be stated as

$$\int_{a}^{b} f(x) dx \simeq h \sum_{k=0}^{n-1} f(a + (k+1/2)h) \qquad h = (b-a)/n$$

where the subintervals are [kh, (k + 1)h], k = 0, 1, ..., n - 1. The composite trapezoidal rule can be stated as

$$\int_a^b f(x)dx \simeq h\left(\frac{f(a)}{2} + \sum_{k=1}^{n-1} f(a+kh) + \frac{f(b)}{2}\right)$$

<u>Exercise 1:</u> Write a program which uses the midpoint rule to compute (subdivide the interval in *n* subintervals)

$$I = \int_0^{\pi/2} f(x) \, dx$$
 $f(x) = x$ and $f(x) = \cos(x)$

Compute and plot the (absolute) error between the measured I_n and the analytical value I as a function of the subinterval size h.

Exercise 2: Same exercise as above but with the trapezoidal rule.

Which method is the most accurate?

Bonus: Repeat Ex.1 with
$$I = \int_{1}^{3} \int_{2}^{4} (x^{2}y^{3} + xy + 1) dx dy$$



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- Interpolation with polynomials evaluated at equally spaced points in [a, b] yields the Newton–Cotes formulas, of which the rectangle rule and the trapezoidal rule are examples.
- If we allow the intervals between interpolation points to vary, we find another group of quadrature formulas, such as the Gaussian quadrature formulas.
- A Gaussian quadrature rule is typically more accurate than a Newton–Cotes rule, which requires the same number of function evaluations, if the integrand is smooth (i.e., if it is sufficiently differentiable).



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- ► An *n*-point Gaussian quadrature rule, named after Carl Friedrich Gauss, is a quadrature rule constructed to yield an exact result for polynomials of degree 2*n* − 1 or less by a suitable choice of the points *x_i* and weights *w_i* for *i* = 1,..., *n*.



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- The domain of integration for such a rule is conventionally taken as [-1, 1], so the rule is stated as

$$\int_{-1}^{+1} f(x) dx = \sum_{i_q=1}^{n} w_{i_q} f(x_{i_q})$$

Numerical integration Gaussian guadrature





Johann Carl Friedrich Gauss (30 April 1777 – 23 February 1855)



► Gaussian quadrature will only produce good results if the function *f*(*x*) is well approximated by a polynomial function within the range [-1, 1].

Numerical integration Gaussian quadrature



- ► Gaussian quadrature will only produce good results if the function *f*(*x*) is well approximated by a polynomial function within the range [-1, 1].
- The method is not, for example, suitable for functions with singularities.





Gauss-Legendre points and their weights

Number of points, n	Points, x _i	Weights, w_i
1	0	2
2	$\pm \sqrt{\frac{1}{3}}$	1
3	0	$\frac{8}{9}$
	$\pm \sqrt{\frac{3}{5}}$	$\frac{5}{9}$
4	$\pm\sqrt{\tfrac{3}{7}-\tfrac{2}{7}\sqrt{\tfrac{6}{5}}}$	$\tfrac{18+\sqrt{30}}{36}$
	$\pm\sqrt{\tfrac{3}{7}+\tfrac{2}{7}\sqrt{\tfrac{6}{5}}}$	$\tfrac{18-\sqrt{30}}{36}$
5	0	$\frac{128}{225}$
	$\pm \tfrac{1}{3} \sqrt{5-2\sqrt{\tfrac{10}{7}}}$	$\frac{322{+}13\sqrt{70}}{900}$
	$\pm \tfrac{1}{3} \sqrt{5 + 2 \sqrt{\tfrac{10}{7}}}$	$\tfrac{322-13\sqrt{70}}{900}$

 x_i is the *i*-th root of the Legendre polynomial $P_n(x)$

Numerical integration Gaussian guadrature



Comparison between 2-point Gaussian and trapezoidal quadrature.



Blue line: polynomial $y(x) = 7x^3 - 8x^2 - 3x + 3$ with $\int_{-1}^{+1} y(x) dx = 2/3$



Comparison between 2-point Gaussian and trapezoidal quadrature.



Blue line: polynomial $y(x) = 7x^3 - 8x^2 - 3x + 3$ with $\int_{-1}^{+1} y(x) dx = 2/3$ The trapezoidal rule returns the integral of the orange dashed line, equal to y(-1) + y(1) = -10



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The trapezoidal rule returns the integral of the orange dashed line, equal to y(-1) + y(1) = -10The 2-point Gaussian quadrature rule returns the integral of the black dashed curve, equal to $y(-\sqrt{1/3}) + y(\sqrt{1/3}) = 2/3$



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Aussian quadrature

 ± 1

An integral over [a, b] must be changed into an integral over [-1, 1] before applying the Gaussian quadrature rule.

-1

Numerical integration Gaussian guadrature





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- This change of interval can be done in the following way:

$$r = \frac{2}{b-a}(x-a) - 1$$
 $x = \frac{b-a}{2}(1+r) + a$ $dx = \frac{b-a}{2}dr$

Numerical integration Gaussian guadrature





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 $x = \frac{b-a}{2}(1+r) + a$ $dx = \frac{b-a}{2}dr$

$$\int_{a}^{b} f(x) dx = \frac{b-a}{2} \int_{-1}^{+1} f(r) dr \simeq \frac{b-a}{2} \sum_{i_{q}=1}^{n} w_{i_{q}} f(r_{i_{q}})$$



For simplicity, a = -1, b = +1. Let us take $f(x) = \pi$

$$I = \int_{-1}^{+1} f(x) dx = \pi \int_{-1}^{+1} dx = 2\pi$$



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$$I_{gq} = \int_{-1}^{+1} f(x) dx = \sum_{i_q=1}^{n_q} w_{i_q} f(x_{i_q}) = \sum_{i_q=1}^{n_q} w_{i_q} \pi = \pi \sum_{i_q=1}^{n_q} w_{i_q} = 2\pi$$

since $\sum_{i_q} w_{i_q} = 2$!



Let us now take f(x) = mx + p

$$I = \int_{-1}^{+1} f(x) dx = \int_{-1}^{+1} (mx + p) dx = \left[\frac{1}{2}mx^2 + px + C\right]_{-1}^{+1} = 2p$$

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since the quadrature points are symmetric w.r.t. to zero on the x-axis.



$$I = \int_{-1}^{+1} f(x) dx = \int_{-1}^{+1} x^2 dx = \left[\frac{1}{3}x^3\right]_{-1}^{+1} = \frac{2}{3}$$



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$$I_{gq} = \int_{-1}^{+1} f(x) dx = \sum_{i_q=1}^{n_q} w_{i_q} f(x_{i_q}) = \sum_{i_q=1}^{n_q} w_{i_q} x_{i_q}^2$$

•
$$nq = 1$$
: $x_{iq}^{(1)} = 0$, $w_{iq} = 2$. $I_{gq} = 0$



$$I = \int_{-1}^{+1} f(x) dx = \int_{-1}^{+1} x^2 dx = \left[\frac{1}{3}x^3\right]_{-1}^{+1} = \frac{2}{3}$$

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•
$$nq = 1$$
: $x_{iq}^{(1)} = 0$, $w_{i_q} = 2$. $I_{gq} = 0$
• $nq = 2$: $x_q^{(1)} = -1/\sqrt{3}$, $x_q^{(2)} = 1/\sqrt{3}$, $w_q^{(1)} = w_q^{(2)} = 1$. $I_{gq} = \frac{2}{3}$



$$I = \int_{-1}^{+1} f(x) dx = \int_{-1}^{+1} x^2 dx = \left[\frac{1}{3}x^3\right]_{-1}^{+1} = \frac{2}{3}$$

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▶
$$nq = 1$$
: $x_{iq}^{(1)} = 0$, $w_{i_q} = 2$. $I_{gq} = 0$
▶ $nq = 2$: $x_q^{(1)} = -1/\sqrt{3}$, $x_q^{(2)} = 1/\sqrt{3}$, $w_q^{(1)} = w_q^{(2)} = 1$. $I_{gq} = \frac{2}{3}$
▶ It also works for all $n_q > 2$!

Gaussian quadrature, example



```
program integration
implicit none
integer, parameter:: ng=2
real(8), dimension(nq), parameter:: xq=(/-1.d0/sqrt(3.d0),+1.d0/sqrt(3.d0)/)
real(8).dimension(ng).parameter:: wg=(/1.d0.1.d0/)
real(8) I
integer ig
I=0.d0
do ig=1,ng
  I=I+wq(iq)*fct(xq(iq))
end do
write(*,*) 'I=',I
contains
function fct(x)
implicit none
real(8) x,fct
!fct=3.14
!fct=x+1
fct=x**2
end function
end program
```

Gaussian quadrature, example



(to copy-paste)

```
program integration
implicit none
integer, parameter:: nq=2
real(8),dimension(nq),parameter:: xq=(/-1.d0/sqrt(3.d0),+1.d0/sqrt(3.d0)/)
real(8),dimension(nq),parameter:: wq=(/1.d0,1.d0/)
real(8) I
integer ig
I=0.d0
do iq=1,nq
I=I+wq(iq)*fct(xq(iq))
end do
write(*,*) 'I=',I
contains
function fct(x)
implicit none
real(8) x,fct
fct=x**2
end function
end program
```



Exercise 3:

- Modify the previous program to use 5 quadrature points instead of two.
- Integrate the functions

 $f_1(x) = \sin(x\pi + \pi/2)$ $f_2(x) = \sqrt{x+1}$ $f_3(x) = x^4 - x^3$

with the 2-point and the 5-point quadrature rules.

Compare the results with the analytical values.

Numerical integration Gauss quadrature



- The fundamental theorem of Gaussian quadrature states that the optimal abscissas of the n_q—point Gaussian quadrature formulas are precisely the roots of the orthogonal polynomial for the same interval and weighting function.
- Gaussian quadrature is optimal because it fits all polynomials up to degree 2n_q - 1 exactly.



Numerical integration Gauss guadrature



An important property of the Legendre polynomials is that they are orthogonal with respect to the L^2 inner product on the interval [-1, 1]

$$\int_{-1}^{+1} P_m(x) P_n(x) dx = \frac{2}{2n+1} \delta_{mn}$$



Let us now turn to a two-dimensional integral of the form

$$I = \int_{a}^{b} \int_{c}^{d} f(x, y) dx dy$$

The equivalent Gaussian quadrature writes:

$$I_{gq} \simeq \sum_{i_q=1}^{n_q} \sum_{j_q}^{n_q} f(x_{i_q}, y_{j_q}) w_{i_q} w_{j_q}$$



Exercise 4:

Compute analytically the following integral:

$$f(x,y)=x^2+4y$$

over the domain $\Omega = [11, 14] \times [7, 10]$

Write a code which integrates this function by means of a 2x2, 3x3 or 4x4 Gauss-Legendre quadrature algorithm.



Applying this to Finite Elements



initialisation & setup mesh domain (fill icon array) timestepping loop do iel=1.nel loop over elements use icon array to retrieve node # which make up iel do iq=1,nq _ loop over quad. pts $A^e = A^e + \dots$ $b^{e} = b^{e} + ...$ assemble in A and b apply b.c. solve

Before

Now

Verify the analytically derived expressions for K_a^e , K_d^e , and M^e .



$$\begin{split} \mathbf{M}^{e} &= \int_{\Omega_{e}} \rho c_{p} \mathbf{N}^{T} \mathbf{N} d\Omega \\ &= \int_{x_{1}}^{x_{3}} \int_{y_{1}}^{y_{3}} \rho(x, y) c_{p}(x, y) \mathbf{N}^{T}(x, y) \mathbf{N}(x, y) dx dy \\ &= \frac{h_{x} h_{y}}{4} \int_{-1}^{+1} \int_{-1}^{+1} \rho(r, s) c_{p}(r, s) \mathbf{N}^{T}(r, s) \mathbf{N}(r, s) dr ds \\ &= \frac{h_{x} h_{y}}{4} \sum_{i_{q}} \sum_{j_{q}} \rho(r_{i_{q}}, s_{j_{q}}) c_{p}(r_{i_{q}}, s_{j_{q}}) \mathbf{N}^{T}(r_{i_{q}}, s_{j_{q}}) \mathbf{N}(r_{i_{q}}, s_{j_{q}}) w_{i_{q}} w_{j_{q}} \end{split}$$



$$\mathbf{K}_{a}^{e} = \frac{\rho c_{p}}{3} \begin{pmatrix} -\frac{1}{2}uh_{y} - \frac{1}{2}vh_{x} & \frac{1}{2}uh_{y} - \frac{1}{4}vh_{x} & \frac{1}{4}uh_{y} + \frac{1}{4}vh_{x} & -\frac{1}{4}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{2}vh_{x} & \frac{1}{4}uh_{y} + \frac{1}{2}vh_{x} & -\frac{1}{4}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{4}uh_{y} + \frac{1}{2}vh_{x} & -\frac{1}{4}uh_{y} + \frac{1}{4}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{2}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{2}vh_{x} & \frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{4}vh_{x} & \frac{1}{4}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{2}vh_{x} & \frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{4}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{4}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{4}vh_{x} & \frac{1}{2}uh_{y} + \frac{1}{4}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{4}vh_{x} & \frac{1}{4}uh_{y} - \frac{1}{4}vh_{x} & \frac{1}{2}uh_{y} + \frac{1}{4}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{4}vh_{x} & \frac{1}{4}vh_{x} & \frac{1}{2}uh_{y} + \frac{1}{4}vh_{x} & -\frac{1}{2}uh_{y} + \frac{1}{2}vh_{x} & \frac{1}{4}vh_{x} & \frac{1$$



$$\boldsymbol{K}_{d}^{e} = k \frac{h_{x}h_{y}}{6} \begin{pmatrix} \frac{2}{h_{x}^{2}} + \frac{2}{h_{y}^{2}} & -\frac{2}{h_{x}^{2}} + \frac{1}{h_{y}^{2}} & -\frac{1}{h_{x}^{2}} - \frac{1}{h_{y}^{2}} & \frac{1}{h_{x}^{2}} - \frac{2}{h_{y}^{2}} \\ \cdot & \frac{2}{h_{x}^{2}} + \frac{2}{h_{y}^{2}} & \frac{1}{h_{x}^{2}} - \frac{2}{h_{y}^{2}} & -\frac{1}{h_{x}^{2}} - \frac{1}{h_{y}^{2}} \\ \cdot & \cdot & \frac{2}{h_{x}^{2}} + \frac{2}{h_{y}^{2}} & -\frac{2}{h_{x}^{2}} + \frac{1}{h_{y}^{2}} \\ \cdot & \cdot & \frac{2}{h_{x}^{2}} + \frac{2}{h_{y}^{2}} & -\frac{2}{h_{x}^{2}} + \frac{1}{h_{y}^{2}} \\ \cdot & \cdot & \frac{2}{h_{x}^{2}} + \frac{2}{h_{y}^{2}} & -\frac{2}{h_{x}^{2}} + \frac{1}{h_{y}^{2}} \end{pmatrix}$$



$$\boldsymbol{M}_{d}^{e} = \rho c_{p} \frac{h_{x} h_{y}}{9} \begin{pmatrix} 1 & 1/2 & 1/4 & 1/2 \\ . & 1 & 1/2 & 1/4 \\ . & . & 1 & 1/2 \\ . & . & 1 & 1/2 \\ . & . & . & 1 \end{pmatrix}$$