

Integration

Numerical integration of a function $f(x)$ from $x=a$ to $x=b$ is written as

$$I = \int_a^b f(x)dx = \sum_{i=0}^{n-1} w_i f(x_i) \quad (1)$$

where w_i are the weights and x_i are abscissas.

The integration schemes can be classified into two schemes:

Simple methods: Here we choose x_i as evenly spaced n points, and then compute w_i . If error is beyond admissible limit, then the number of points is increased.

Complex methods: Here x_i and w_i are chosen in such a way the integral has minimum error.

Newton-Cotes Formulas

Newton-Cotes scheme belongs to the class of simple methods. Here we divide the interval (a, b) into $(n - 1)$ equal divisions, that is, $(b - a)/(n - 1) = h$. The abscissas are located at $x_j = a + jh$ where $j = 0 : (n - 1)$.

We approximate the function $f(x)$ using the Lagrange interpolation:

$$P(x) = \sum L_j(x)f(x_j),$$

which is substituted in Eq. (1) that yields

$$\begin{aligned} I &= \int_a^b f(x)dx \approx \int_a^b P(x)dx = \sum_{j=0}^{n-1} f(x_j) \int_a^b L_j(x)dx \\ &= (b - a) \sum_{j=0}^{n-1} C_j^{(n)} f(x_j). \end{aligned}$$

where

$$C_j^{(n)} = \frac{1}{b - a} \int_a^b L_j(x)dx.$$

Since $L_j(x)$ are independent of the data $f(x_j)$, we conclude that $C_j^{(n)}$'s are independent of $f(x_j)$. Using $f(x) = 1$, we conclude that

$$\sum_j C_j^{(n)} = 1.$$

Note that $(b - a)C_j^{(n)}$ are the weights.

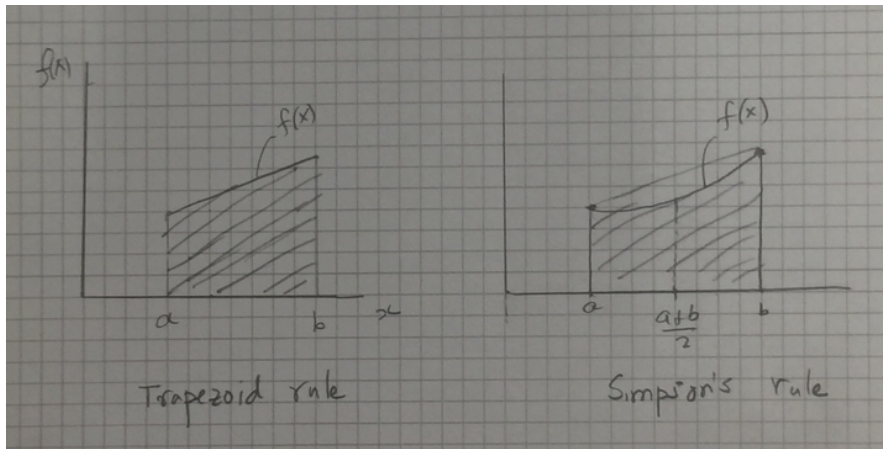
Now let us compute the $C_j^{(n)}$ for some of the Legendre polynomials. For two points, the polynomial is

$$P_2(x) = \frac{x-b}{a-b}f(a) + \frac{x-a}{b-a}f(b).$$

When we integrate the above, we obtain

$$I = \int_b^a P(x)dx = \frac{h}{2}(f(a) + f(b)).$$

This method, called trapezoid rule, is accurate for linear functions.



Note that the error in the above computation is

$$E = \int_a^b \frac{f''(\zeta)}{2!}(x-a)(x-b)dx = -\frac{1}{12}h^3f''(\zeta).$$

When we use $f(x) = P_3(x)$ using the points $x=a, (a+b)/2, b$, we obtain

$$I = \int_b^a P(x)dx = \frac{h}{3}(f(a) + 4f((a+b)/2) + f(b)),$$

This method is called Simpson's rule, and it is accurate for polynomials up to third order.

Figure

The higher-order terms get more complex. Here we list C_i in Table 1.

n	N							
2	2	1	1					
3	6	1	4	1				
4	8	1	3	3	1			
5	90	7	32	12	32	7		
6	288	19	75	50	50	75	19	
7	840	41	216	27	272	27	216	41

Error analysis for Newton-Cote's method

In an earlier chapter we derived that the error in Lagrange interpolation with n points is

$$E(x) = f(x) - P(x) = \frac{f^{(n)}(\zeta)}{n!} \prod (x - x_i)$$

Hence the error in the integration by Newton-Cotes scheme is

$$E = \int_a^b E(x)dx = \frac{f^{(n)}(\zeta)}{n!} \int_a^b dx \prod (x - x_i)$$

We observe that for even n , the above integration is proportional to h^{n+1} , but it vanishes for odd n . For odd n , the error comes from the next term of the expansion that passes through the points $(a, a+h, a+2h, \dots, b, b+h)$. The error estimate is

$$E = \int_a^b E(x)dx = \frac{f^{(n+1)}(\zeta)}{(n+1)!} \int_a^b dx \left[(x - b - h) \prod (x - x_i) \right].$$

The above integrals yield the errors as

n	Error
P_2	$(h^3/12)f^{(2)}(\zeta)$
P_3	$(h^5/90)f^{(4)}(\zeta)$
P_4	$(3h^5/80)f^{(4)}(\zeta)$
P_5	$(8h^7/945)f^{(6)}(\zeta)$
P_6	$(275h^7/12096)f^{(6)}(\zeta)$
P_7	$(9h^9/1400)f^{(8)}(\zeta)$

We illustrate the above error estimates using trapezoid and Simpson methods. For the trapezoid method

$$E_2 = \frac{f^{(2)}(\zeta)}{2!} \int_a^b dx (x - a)(x - b).$$

We choose $a=0$ and $b=1$, hence $h=1$ that yields

$$E_2 = \frac{f^{(2)}(\zeta)}{12}.$$

For $n=3$, the leading-order Legendre expansion yields

$$E_2 = \frac{f^{(3)}(\zeta)}{3!} \int_0^1 dx x(x - 1/2)(x - 1) = 0.$$

Hence, for estimation, we pick the next polynomial using points $(0, 1/2, 1, 3/2)$. The error is

$$E_2 = \frac{f^{(4)}(\zeta)}{4!} \int_0^1 dx x(x - 1/2)(x - 1)(x - 3/2) = \frac{h^5}{90} f^{(4)}(\zeta),$$

where $h=1/2$. Since $f^4(x)=0$ for a third-order polynomial, Simpson's method is accurate for polynomials up to cubic order.

We can compute the coefficients of Table 1 as well as error formula using Sympy.

```
x,a,b,h = symbols(('x','a','b','h'))
```

```
n = 4
```

```
h = (b-a)/(n-1)
```

```
xarray = []
```

```
for i in range(n):
```

```
    xarray.append(a+i*h)
```

```
def Newton_cotes_coeff(xarray):
```

```
    n = len(xarray)
```

```

coeff = []
for j in range(n):
    numr = 1; denr = 1;
    for i in range(n):
        if (j != i):
            numr *= (x-xarray[i])
            denr *= (xarray[j]-xarray[i])
    Cj = simplify(integrate(numr/denr, (x,a,b))/(b-a))
    coeff.append(Cj)

factor = 1
for i in range(n):
    factor *= (x-xarray[i])

if (n%2 == 1):
    error_factor = simplify(integrate(factor*(x-b-h), (x,a,b))/
(h**((n+2)*factorial(n+1))))
else:
    error_factor = simplify(integrate(factor, (x,a,b))/
(h**((n+1)*factorial(n))))

return coeff, error_factor

```

Example: $\int_0^{\pi/2} dx \sin x$

Newton_cotes.py

```

# assume a single interval
# Solve  $\int_a^b f(x)dx$  using m-th order Newton-Cotes method
def Newton_cotes(m, f, a, b):
    h=(b-a)/(m-1)

    if (m==2):
        return (b-a)*(f(a)+f(b))/2
    elif (m==3):
        return (b-a)*(f(a)+4*f((a+b)/2)+f(b))/6
    elif (m==4):
        return (b-a)*(f(a)+3*(f(a+h)+f(a+2*h))+f(b))/8
    elif (m==5):
        return (b-a)*(7*(f(a)+f(b)) + 32*(f(a+h)+f(b-h)) + 12*f(a+2*h))/
90
    elif (m==6):
        return (b-a)*(19*(f(a)+f(b)) + 75*(f(a+h)+f(b-h)) +
50*(f(a+2*h)+f(a+3*h)))/288

# (b-a) divided into n-1 interval; given n points
# Trapezoid rule
def Newton_cotes2(f, a, b, n):
    h = (b-a)/(n-1)
    ans = 0
    for i in range(n):
        ans += f(a+i*h)

    ans -= (f(a)+f(b))/2
    return h*ans

def f(x):
    return np.sin(x)

```

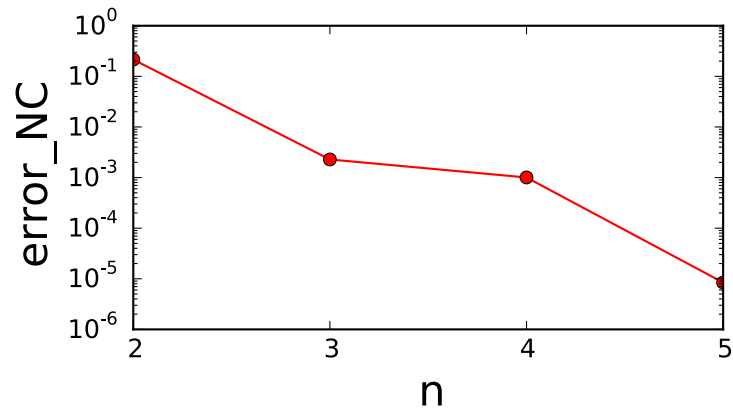
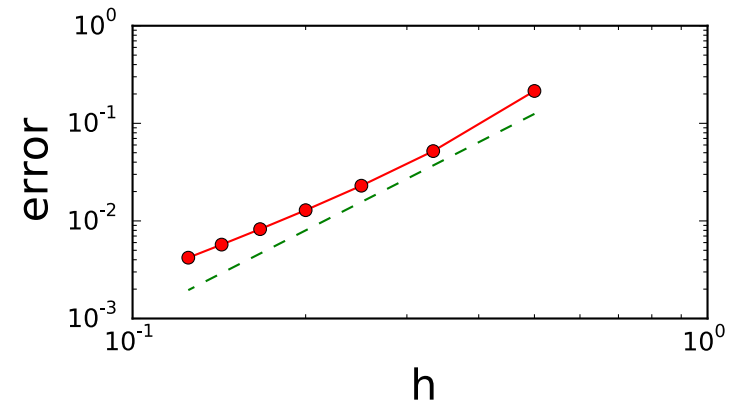


Fig: Error in Newton-Cotes method vs. degree of the polynomial, n , used.

The figure shows that error drops when we go from 2 to 3, but flattens from 3 to 4. See the error table for the reasons.

We can also integrate using smaller h , but summing over all the interval. We perform the above using trapezoid rule. The following figure illustrates how the error decreases with h as h^3 , as h is decreased. The green dashed line represents h^3 .



Gaussian Quadrature

In this method, we choose both x_m and w_m of

$$I = \int_a^b f(x)dx \approx \sum_{m=0}^{N-1} w_m f(x_m) \quad (1)$$

so that the integral is accurate. Thus we have $2N$ unknowns. If we demand exact quadrature for $f(x) = 1, x, x^2, \dots, x^{2N-1}$, we will have $2N$ equations using which we can obtain the aforementioned $2N$ unknowns.

Example: Work out for $N=2$ for $[-1,1]$.

Solution: We require that the integral is exact for $f(x) = 1, x, x^2, x^3$, which yields the following four equations:

$$2 = w_0 + w_1$$

$$0 = w_0 x_0 + w_1 x_1$$

$$\frac{2}{3} = w_0 x_0^2 + w_1 x_1^2$$

$$0 = w_0 x_0^3 + w_1 x_1^3,$$

whose solutions are $w_0 = w_1 = 1$, and $x_1 = -x_0 = 1/\sqrt{3}$.

Hence

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

The above integral is exact as long as $f(x)$ is a polynomial of degree 3 or less. We verify for $f(x) = x^2 + \frac{1}{2}x^3$, for which the RHS of the above equation yields $2/3$, which is the exact integral.

The above procedure however is quite tedious for a general function. In the following discussion, we provide a general formulation based on orthogonal polynomials. Note that our formulation should yield an exact quadrature for any polynomial of degree $2N-1$ or less. We denote this function by $f(x)$.

Suppose that the orthogonal polynomials used in our method is $\phi_i(x)$, and they satisfy the following orthogonality relation:

$$\int_a^b w(x) \phi_i(x) \phi_j(x) dx = \delta_{ij} \gamma_i,$$

where γ_i are constants, $i = 0 : (N-1)$, and $w(x)$ is the weight function. Note that not all the polynomials need to be of $(N-1)$ th order. However, we demand that at least one of them is of $(N-1)$ -order. In this scheme the integral is

$$I = \sum_{j=0}^{N-1} w_j f(x_j),$$

where x_j 's are the roots of $\phi_N(x)$, and w_j 's are

$$w_j = -\frac{a_N \gamma_N}{\phi'_N(x_j) \phi_{N+1}(x_j)}$$

We provide the proof of the above scheme in two steps:

Step 1: We write the function $f(x)$ as

$$f(x) = q_{N-1}(x)\phi_N(x) + r_{N-1}(x), \quad (2)$$

where $\phi_N(x)$ is the N-th order polynomial, q_{N-1} is the quotient, and r_{N-1} is the remainder. Both q_{N-1} and r_{N-1} are polynomials of order (N-1), hence

$$q_{N-1} = \sum_{j=0}^{N-1} a_j \phi_j(x),$$

where a_j 's are the coefficients. Integration of Eq. (2) yields

$$\begin{aligned} \int_a^b w(x)f(x)dx &= \int_a^b w(x)q_{N-1}(x)\phi_N(x)dx \\ &+ \int_a^b w(x)r_{N-1}(x)dx. \quad (3) \end{aligned}$$

The integral

$$\int_a^b w(x)q_{N-1}\phi_N(x)dx = \sum_j a_j \int_a^b w(x)\phi_j(x)\phi_N(x)dx = 0 \text{ due to or-}$$

thogonality property. Hence

$$\int_a^b w(x)f(x)dx = \int_a^b w(x)r_{N-1}(x)dx.$$

Step 2: The N-th order polynomial $\phi_N(x)$ has N roots. We choose these roots, $x_j (j = 0 : N)$, as the abscissa, hence $\phi_N(x_j) = 0$. The function $f(x)$ given at these x_j 's, i.e., $y_j = f(x_j)$. Using Eq. (2)

$$f(x_j) = q_{N-1}(x_j)\phi_N(x_j) + r_{N-1}(x_j) = r_{N-1}(x_j).$$

Expand $r_{N-1}(x)$ using Lagrange interpolation

$$r_{N-1}(x) = \sum_j r_{N-1}(x_j)L_j(x),$$

substitution of which in Eq. (3) yields

$$\begin{aligned} \int_a^b w(x)f(x)dx &= \int_a^b w(x)r_{N-1}(x)dx = \sum_j r_{N-1}(x_j) \int_a^b w(x)L_j(x)dx \\ &= \sum_j w_j r_{N-1}(x_j) = \sum_j w_j f(x_j), \end{aligned}$$

where $w_j = \int_a^b w(x)L_j(x)dx$. Using

$$\prod_{i,i \neq j} (x - x_i) = \frac{\prod (x - x_i)}{x - x_j} = \frac{\phi_N(x)}{a_N(x - x_j)},$$

where a_N is the coefficient of x^N in $\phi_N(x)$. Using L'Hospital rule, we take the limit $x \rightarrow x_j$, which yields

$$\prod_{i,i \neq j} (x_j - x_i) = \frac{\phi'_N(x_j)}{a_N}.$$

Therefore,

$$w_j = \int_a^b dx w(x) \frac{\prod_{i,i \neq j} (x - x_i)}{\prod_{i,i \neq j} (x_j - x_i)} = \frac{1}{\phi'_N(x_j)} \int_a^b \frac{w(x)\phi_N(x)}{x - x_j} dx. \quad (4)$$

Using Christoffel Darboux identity we can simplify it further to

$$\sum_{i=0}^N \frac{\phi_i(x)\phi_i(y)}{\gamma_i} = \frac{\phi_{N+1}(x)\phi_N(y) - \phi_N(x)\phi_{N+1}(y)}{a_N \gamma_N(x - y)} \quad (5)$$

where $\int_a^b w(x)\phi_i(x)\phi_j(x) = \delta_{ij}\gamma_i$ and

$$a_m = \frac{A_{m+1}}{A_m}$$

with A_m as the coefficient of x^m in $\phi_m(x)$. Choosing $y = x_j$, a zero of $\phi_N(x)$, substitution of which in Eq. (5) yields

$$\sum_{i=0}^N \frac{\phi_i(x)\phi_i(x_j)}{\gamma_i} = -\frac{\phi_N(x)\phi_{N+1}(x_j)}{a_N\gamma_N(x-x_j)}.$$

Integration of the above wrt $\int_a^b dx w(x)\phi_0(x) \dots$ and using the fact that $\phi_0(x) = \text{const}$ yields

$$\phi_0(x_j) = -\frac{\phi_{N+1}(x_j)}{a_N\gamma_N}\phi_0(x) \int_a^b dx \frac{w(x)\phi_N(x)}{(x-x_j)}.$$

Hence

$$\int_a^b dx \frac{w(x)\phi_N(x)}{(x-x_j)} = -\frac{a_N\gamma_N}{\phi_{N+1}(x_j)}.$$

Hence

$$w_j = -\frac{a_N\gamma_N}{\phi'_N(x_j)\phi_{N+1}(x_j)}. \quad (6)$$

We state without proof that the error in Gaussian quadrature is

$$E_{\text{int}} = \frac{f^{(2n)}(\zeta)}{(2n)!} \int_a^b w(x)W(x)dx \quad (7)$$

[NOT] **Alternate proof:**

We can also use Hermite's interpolation to derive the formula for Gaussian quadrature:

$$f(x) \approx P(x) = \sum_j U_j(x)y_j + \sum_j V_j(x)y'_j \quad (1)$$

with

$$U_j(x) = [1 - 2L'_j(x_j)(x - x_j)][L_j(x)]^2;$$

$$V_j(x) = (x - x_j)[L_j(x)]^2$$

The integral is

$$\begin{aligned} \int_a^b w(x)f(x)dx &\approx \int_a^b w(x)P(x)dx \\ &= \sum_j y_j \int_a^b w(x)U_j(x)dx + \sum_j y'_j \int_a^b w(x)V_j(x)dx \end{aligned}$$

We choose the abscissa in such a way that the second integral is zero, i.e.,

$$\int_a^b w(x)V_j(x)dx = \int_a^b w(x)(x - x_j)[L_j(x)]^2dx = 0.$$

Therefore,

$$\int_a^b w(x)P(x)dx = \sum_j y_j \int_a^b w(x)U_j(x)dx$$

$$= \sum_j y_j \int_a^b w(x) [L_j(x)]^2 dx = \sum_j y_j w_j$$

Thus the weight is

$$w_j = \int_a^b w(x) [L_j(x)]^2 dx$$

Compare it with the other formulas derived in the previous derivation.

Error:

We derived in the Hermite's polynomial that the error is

$$E = f(x) - P(x) = \frac{f^{(2n)}(\zeta)}{(2n)!} W(x),$$

where

$$W(x) = \prod_{i=0}^{n-1} (x - x_i)^2$$

Hence the error in the quadrature would be

$$E_{\text{int}} = \frac{f^{(2n)}(\zeta)}{(2n)!} \int_a^b w(x) W(x) dx$$

For integration of smooth functions within a limit, it is best to use **Legendre polynomials**, for which $w(x) = 1$, $a = -1$, and $b = 1$. We will show later how the integration scheme can be generalized to arbitrary a and b .

The leading Legendre polynomials are

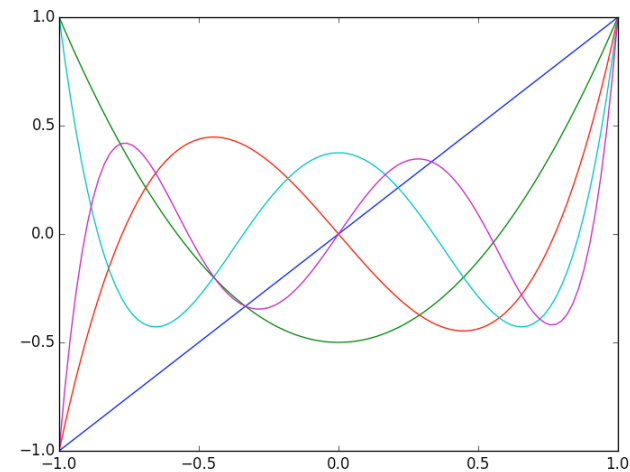
$$\phi_0(x) = 1$$

$$\phi_1(x) = x$$

$$\phi_2(x) = \frac{3x^2 - 1}{2}$$

$$\phi_3(x) = \frac{5x^3 - 3x}{2}$$

PLOTS OF LEGENDRE POL



Also,

$$\gamma_j = \frac{2}{2j+1}$$

$$A_j = \frac{(2j)!}{2j(j!)^2}, \quad a_N = \frac{2}{N+1}$$

Hence, using Eq. (6) we obtain

$$w_j = -\frac{2}{(j+1)\phi'_N(x_j)\phi_{N+1}(x_j)}$$

Using an identity

$$(1-x^2)\phi'_j(x) = (j+1)x\phi_j(x) - (j+1)\phi_{j+1}(x)$$

and substituting $j = N; x = x_j$ we obtain

$$w_j = \frac{2}{(1-x_j^2)[\phi'_N(x_j)]^2}$$

$$N=2: \phi_2(x) = \frac{3x^2-1}{2}, \quad \phi'_2(x) = 3x$$

The zeros of the polynomials at $x = -1/\sqrt{3}, 1/\sqrt{3}$. Also, $\phi'_2(\pm(1/\sqrt{3})) = \pm\sqrt{3}$. Therefore, $w_0 = w_1 = 1$.

$$N=3, \phi_2(x) = \frac{5x^3-3x}{2}.$$

The zeros of the polynomials at $x = -\sqrt{3/5}, 0, \sqrt{3/5}$, substitution of which yields:

$$w_0 = w_2 = 5/9, w_1 = 8/9.$$

We list the abscissa and weights in the following table:

n	x_i	w_i
2	± 0.5773503	1
3	0	0.888889
	± 0.774579	0.555556
4	± 0.339981	0.652145
	± 0.861136	0.347855
5	0	0.568889
	± 0.538469	0.478629
	± 0.90618	0.236927

Example code

```
# gaussian_quad.py
# sum(w_i f(x_i))

def my_gauss_quadrature(warray, xarray, f):

    yarray = np.zeros(len(xarray))

    for i in range(len(xarray)):

        yarray[i] = f(xarray[i])

    return sum(warray*yarray)
```

```
xarray = np.array([-1/np.sqrt(3),1/np.sqrt(3)])
```

```
warray = np.array([1,1])
```

Example 1: $\int_{-1}^1 x^2 dx = 0.6666667$, exact value with $n = 2$.

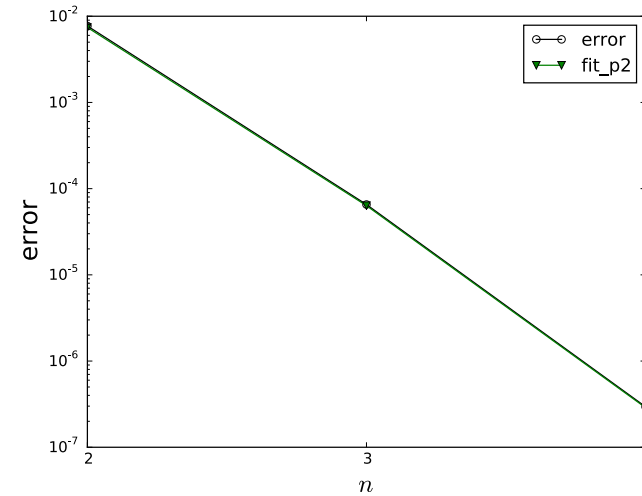
Example 2: $\int_{-1}^1 \exp(x) dx$ for $n = 2, 3, 4$

The integral converges quickly with N. In the following Table, we list the errors for N=2,3,4, and the builtin function. The error is plotted in Figure that shows that the error varies as

$$E_n = \exp(-0.318N^2 - 3.16N + 2.7)$$

Thus the convergence is exponential (for small N).

n	I	error
2	2.342	0.00770
3	2.35033	6.55E-05
4	2.3504021	2.87E-07
from Python Gauss quad	2.3504023872	8.23E-10



For a variable x' in an arbitrary interval $[a, b]$, we make a change of variable $x' = \alpha x + \beta$, with $(x' = -1, x = a)$ and $(x' = 1, x = b)$.. Hence $a = -\alpha + \beta$, and $b = \alpha + \beta$, leading to $\alpha = (b - a)/2$, $\beta = (b + a)/2$. Hence,

$$I = \int_a^b f(x') dx' = \frac{b - a}{2} \int_{-1}^1 f(\alpha x + \beta) dx.$$

Laguerre-Gauss quadrature

We often encounter integral of the form

$$\int_0^{\infty} e^{-x} f(x) dx,$$

$w(x) = \exp(-x)$ is the weight function. One of the example is Hydrogen atom. Note that the integration of the above function with Newton-Cote's scheme will be very expensive because we have to take many intervals to reach $x = \infty$.

we use Laguerre polynomials whose orthogonality relation is

$$\int_0^{\infty} e^{-x} L_n(x) L_m(x) dx = \delta_{m,n}$$

$$\int_0^{\infty} e^{-x} f(x) dx = \sum_{j=0}^{N-1} w_j f(x_j),$$

where x_i are the roots of $L_N(x)$. Also,

$$\gamma_j = 1$$

$$A_j = \frac{(-1)^j}{j!}, \quad a_N = -\frac{1}{N}$$

Hence

$$w_j = -\frac{1}{(N+1)\phi'_N(x_j)\phi_{N+1}(x_j)} = -\frac{1}{N\phi_{N-1}(x_j)\phi'_N(x_j)}$$

Using identities we obtain

$$w_j = \frac{1}{x_j[\phi'_N(x_j)]^2} = \frac{x_j}{(N+1)^2[L_{N+1}(x_j)]^2}$$

[REF: mathworld.wolfram.com]

Laguerre-Gauss quadrature

n	x_i	w_i
2	0.585786	0.853553
	3.41421	0.146447
3	0.415775	0.711093
	2.29428	0.278518
	6.28995	0.0103893
4	0.322548	0.603154
	1.74576	0.357419
	4.53662	0.0388879
	9.39507	0.000539295
5	0.26356	0.521756
	1.4134	0.398667
	3.59643	0.0759424
	7.08581	0.00361176
	12.6408	0.00002337

Hermit-Gauss quadrature

We also encounter integral of the form

$$\int_{-\infty}^{\infty} e^{-x^2} f(x) dx = \sum_j w_j f(x_j).$$

Here $w(x) = \exp(-x^2)$ is the weight function. For such integrals we employ Hermite's polynomials. Here x_i are the roots of $H_N(x)$, and w_i 's are given below. Recall the wavefunction of the harmonic oscillator.

Hermite-Gauss quadrature

n	x_i	w_i
2	± 0.707107	0.886227
3	0	1.18164
	± 1.22474	0.295409
4	± 0.524648	0.804914
	± 1.65068	0.0813128
5	0	0.945309
	± 0.958572	0.393619
	± 2.02018	0.0199532