Quadrature based on polynomial interpolation

#### Newton-Cotes rules

- equidistant quadrature points  $x_i = a + ih$  where h = (b a)/n and i = 0, ..., n
- quadrature rule:

$$Q(f,a,b) = \sum_{i=0}^{n} w_i f(x_i)$$

▶ Newton-Cotes method: choose *quadrature weights w<sub>i</sub>* such that

$$Q(p, a, b) = \int_{a}^{b} p(x) dx$$

for all polynomials p of degree up to n

#### special cases

▶ rectangle rule n = 0

$$Q(f,a,b)=(b-a)f(x_0)$$

for any quadrature point  $x_0 \in [a, b]$ , weight  $w_0 = b - a$ 

- midpoint rule  $x_0 = (a+b)/2$
- **trapezoidal rule** n=1

$$Q(f,a,b) = \frac{b-a}{2}f(a) + \frac{b-a}{2}f(b)$$

quadrature points  $x_0 = a$ ,  $x_1 = b$ , weights  $w_0 = w_1 = (b - a)/2$ 

### integer quadrature points for $n \geq 1$

approximation of

$$I(f) = \int_0^n f(x) \, dx$$

quadrature rule

$$Q(f) = \sum_{k=0}^{n} w_k f(\underline{k})$$

▶ note that this does not give a rule for n = 0 as  $\int_0^0 f(x) dx = 0$ , for this we consider "open" rules which approximate

$$I(f) = \int_0^{n+1} f(x) \, dx$$

by Q(f) with the same points  $x_k = k$  but adapted weights  $w_k$  for k = 0, ..., n

#### transformation formula

▶ introduce variable  $z \in [0, n]$  such that

$$x = a + zh$$

where h = (b - a)/n

▶ transformed function g(z) satisfying

$$g(z) = f(a + zh)$$

▶ it follows that

$$\int_{a}^{b} f(x) dx = h \int_{0}^{n} g(z) dz$$

and one gets the transformed quadrature rule

$$Q(f) = h \sum_{k=0}^{n} w_k f(a + kh)$$

where  $w_k$  are the Newton-Cotes weights for the interval [0, a]

#### composite Newton-Cotes rules



- ► choose N = nm quadrature points  $x_k = a + kh$  where h = (b a)/N
- composite formula

$$Q(f) = h \sum_{j=0}^{m} \sum_{k=0}^{n} w_{k} f(a + x_{k+jn})$$

where  $w_k$  are the weights for the Newton Cotes on the interval [0, n]

# computing the weights $w_k$ from the Lagrange interpolation formula

- we only need to consider the interval [0, n]
- choose quadrature formula defined by

$$Q(f) = \int_0^n p(x) \, dx$$

where p is the interpolating polynomial at  $x_k = k$  for  $k = 0, \dots, n$ 

► Lagrange interpolation formula

$$p(x) = \sum_{k=0}^{n} (I_k(x)) f(k)$$

▶ integrate this formula to get the weights

$$w_k = \int_0^n l_k(x) \, dx$$

#### example n = 1 – trapezoidal rule

Lagrange functions

$$l_0(x) = 1 - x$$
,  $l_1(x) = x$ 

weights

$$w_0 = \int_0^1 (1-x) dx = 1/2, \quad w_1 = \int_0^1 x dx = 1/2$$

### example n = 2 - Simpson's rule

Lagrange (or cardinal) functions

$$l_0(x) = (x-1)(x-2)/2$$
,  $l_1(x) = -x(x-2)$ ,  $l_2(x) = x(x-1)/2$ 

weights

$$w_0 = \int_0^2 l_0(x) dx = \int_0^2 (x^2 - 3x + 2)/2 = 1/3$$

$$w_1 = \int_0^2 l_1(x) dx = 4/3$$

$$w_2 = \int_0^2 l_2(x) dx = 1/3$$

# computing the weights using the method of unknown coefficients

▶ set up linear system of n + 1 equations for the  $w_j$  from the conditions

$$Q(x^j) = \int_0^n x^j dx, \quad j = 0, \dots, n$$

equations

$$\sum_{k=0}^n k^j w_k = \frac{n^{j+1}}{j+1}, \quad j=0,\ldots,n$$

matrix is Vandermonde matrix

#### example n = 1 trapezoidal rule

equations

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1/2 \end{bmatrix}$$

weights

$$w_1 = 1/2, \quad w_0 = 1 - 1/2 = 1$$

### example n = 2 Simpson's rule

equations

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 1 & 4 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 8/3 \end{bmatrix}$$

with Gauss elimination we get

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 2/3 \end{bmatrix}$$

▶ the solution by back substitution is as before  $w_0 = 1/3$ ,  $w_1 = 4/3$  and  $w_2 = 1/3$ 

```
# compute Newton-Cotes weights with sympy
w = np.zeros((4,9))
for i, n in enumerate((1,2,4,8)):
   x = sy.Symbol('x')
   print("\n n = {}:".format(n),end=' ')
   for j in range(n+1):
       li = 1
       for k in range(n+1): # compute Lagrangian polynom
           if (k!=j): lj *= (x-k)/(j-k)
       w[i,j] = float(sy.integrate(lj,(x,0,n)))
       print("w{} = {:4.2f}".format(j,w[i,j]),end="
                                                     1)
n = 1: w0 = 0.50 w1 = 0.50
n = 2: w0 = 0.33 w1 = 1.33 w2 = 0.33
n = 4: w0 = 0.31 w1 = 1.42 w2 = 0.53 w3 = 1.42
n = 8: w0 = 0.28 w1 = 1.66 w2 = -0.26 w3 = 2.96
```

# example: use weights to compute $I = \int_0^1 \exp(-x) dx$

- exact value  $I = 1 e^{-1}$
- approximations with Newton-Cotes

```
n = 1, Q = 0.6839397, Error = 5.2e-02

n = 2, Q = 0.6323337, Error = 2.1e-04

n = 4, Q = 0.6321209, Error = 3.2e-07

n = 8, Q = 0.6321206, Error = 3.6e-13
```

### **Errors**

## general formula for interval [0, n]

• error of n degree interpolating polynomial p for interpolation points  $x_k = k$ 

$$e(x) = p(x) - f(x) = -\frac{1}{(n+1)!} f^{(n+1)}(\xi_x) w(x)$$

where  $w(x) = \prod_{i=0}^{n} (x - k)$ 

error of quadrature equals integral of interpolation error

$$E = \int_0^n e(x) dx = \frac{1}{(n+1)!} \int_0^n f^{(n+1)}(\xi_x) w(x) dx$$

mean value theorem gives

$$|E| \le \frac{1}{(n+1)!} |f^{(n+1)}(\underline{\xi})| \int_0^n |w(x)| dx$$

for some  $\xi \in [0, n]$ 

## example: $\int_0^1 \exp(-x) dx$

▶ transformation from [0, n] to [0, 1]

$$f(x) = \exp(-x/n)$$

value of integral

$$I = \int_0^1 \exp(-x) \, dx = n^{-1} \int_0^n \exp(-x/n) \, dx$$

• quadrature  $(w_k$  are weights for [0, n])

$$Q = n^{-1} \sum_{k=0}^{n} w_k \exp(-k/n)$$

derivatives for error bounds

$$f^{(k)}(x) = (-n)^{-k} \exp(-x/n)$$

example: error for n = 1,  $f(x) = \exp(-x)$  and  $x \in [0, 1]$ 

▶ one has for some  $\xi \in [0,1]$ :

$$|E| \le \frac{1}{2} \exp(-\xi) \int_0^1 x(1-x) \, dx \le \frac{1}{12} \approx 0.08$$

where the actual error is 0.05 (see computation done previously)

# example error for $f(x) = \exp(-x)$ , $x \in [0, 1]$ and general n

error bound

$$|E| \le \frac{h^{n+2}}{(n+1)!} \int_0^n |w(x)| dx$$

• compute  $\int_0^n |w(x)| dx$  with sympy

## compute the weights for n = 1, 2, 3, 4, 5

#### compute the error bounds for n = 1, 2, 3, 4, 5

```
Ebound = np.zeros(5); E = np.zeros(5)
for i,n in enumerate ((1,2,3,4,5)):
    h = 1.0/n
    Ebound[i] = h**(n+2)/math.factorial(n+1)*wint[n-1]
    E[i] = h*np.sum(w[n-1,:n+1]*)
        np.exp(-np.linspace(0,1,n+1))) - 1.0 + 1.0/np.e
    print("n = {}), error = {:4.2e}, bound = {:4.2e}"
                 .format(n.E[i]. Ebound[i]))
n = 1, error = 5.18e-02, bound = 8.33e-02
n = 2, error = 2.13e-04, bound = 5.21e-03
n = 3, error = 9.50e-05, bound = 2.80e-04
n = 4, error = 3.16e-07, bound = 1.29e-05 \checkmark
n = 5, error = 1.78e-07, bound = 5.20e-07
```

#### error bounds for even n

- ▶ note: error bounds for n = 2 and n = 4 are bad
- ▶ as  $\int_0^n w(x) dx = 0$  for even n, quadrature exact for polynomials of degree n + 1 in this case
- ► Taylor expansion for  $f^{(n+1)}(\xi)$ :

$$f^{(n+1)}(\xi) = f^{(n+1)}(1/2) + (\xi - 1/2)f^{(n+2)}(\eta)$$

▶ in the case of even n the first (constant) term does not contribute to the error and one gets for a general function f(x)

$$|E| \le \frac{h^{n+3}}{2(n+1)!} |f^{(n+2)}(\xi)| \int_0^n |w(x)| dx$$

and for our example  $f(x) = \exp(-x)$ :

$$|E| \le \frac{h^{n+3}}{2(n+1)!} \int_0^n |w(x)| dx$$

#### recompute the error bounds for n = 2, 4

```
Ebound = np.zeros(2); E = np.zeros(2)
for i,n in enumerate ((2,4)):
    h = 1.0/n
    Ebound[i] = h**(n+3)/(2*math.factorial(n+1))
                  *wint[n-1]
    E[i] = h*np.sum(w[n-1,:n+1])
     *np.exp(-np.linspace(0,1,n+1))) - 1.0 + 1.0/np.e
    print("n = {}), error = {:4.2e}, bound = {:4.2e}"
            .format(n,E[i], Ebound[i]))
n = 2, error = 2.13e-04, bound = 1.30e-03 

n = 4, error = 3.16e-07, bound = 1.61e-06
```