

Introduction to Scientific Computing

(Lecture 9: Numerical integration)

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In last lecture we have shown that the proces of solving

$$\dot{x}(t)=f(t,x(t)),$$

for $t \in [t_0, T], x(t_0) = x_0$, where $f : \mathbb{R}_+ \times \mathbb{R}^n \to \mathbb{R}^n$ is a given function and x_0 is the given initial condition, is equivalent to the process of evaluating the solution of the following integral equation

$$x(T) = x_0 + \int_{t_0}^T f(t, x(t)) dt$$

The main issue appearing in the previous equation is that the integral cannot be always computed in an analytical way. In order to avoid this, we may perform numerical integration

integral = area under curve

which further means that we need to compute the area under curve

$$\int_{t_0}^T f(t, x(t)) dt$$

However, this is not an easy task since the term under integral f(t, x(t)) depends on the unknown x. In practice, there are many situations in which the previous integrand can be respesented as a function of time only

$$\int_{t_0}^T f(t, x(t)) dt = \int_{t_0}^T g(t) dt$$

which further makes our problem easier to treat. As analytical solution of the previous integral is not always possible, one may try to numerically compute the value by approximating the function g by something simpler such as a polynomial $P_n(t)$ of order n—polynomial interpolation.

Polynomial interpolation

The polynomial

$$P_n(t) = a_0 + a_1t + \ldots + a_nt^n$$

can be integrated instead of g(t) in one of following ways

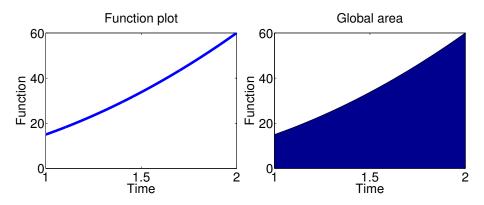
• globally: compute the area under the curve directly

$$\int_{t_0}^T g(t) dt \approx \int_{t_0}^T P_n(t) dt$$

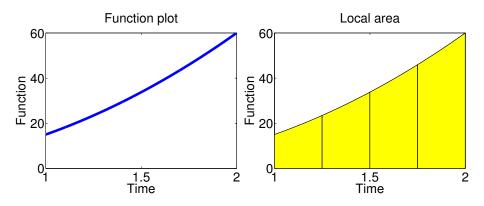
• locally: compute the area under curve by cutting it into pieces (composite rule)

$$\int_t^{t+h} g(t) dt \approx \sum_{i=0}^{N-1} \int_{t_i}^{t_i+h} P(t) dt$$

Global integration



Local integration



Let us choose *n* pairwise distinct points t_i in time interval $I = [t_0, T]$, and let us compute the value of function $g_i := g(t_i)$ in these points. Then the polynomial can be found by global interpolation

$$g(t_i) = P(t_i) = \sum_{j=0}^n a_j t_i^j$$

The last relation is equivalent to the process of solving

$$Qa = w$$

in which

$$Q = [t_i^j], \quad a = [a_j], \quad w = [g(t_i)]$$

Polynomial interpolation

The matrix Q is of the form

$$Q = \begin{pmatrix} 1 & t_0 & \dots & t_0^n \\ 1 & t_1 & \dots & t_1^n \\ \vdots & \vdots & & \vdots \\ 1 & t_n & \dots & t_n^n \end{pmatrix}$$

and hence is regular. This further means that one may compute the vector of unknown coefficients a by solving

$$Qa = w$$

To solve the previous equation one may use one of the iterative methods introduced to you before.

Once the coefficients a_i are known the integration becomes simple

$$\int_{t_0}^T g(t) dt \approx \int_{t_0}^T P(t) dt = \int_{t_0}^T \sum_{j=0}^n a_j t_j^j dt$$

Before we give general expression for the previous interval, let us first focus on specific polynomiall approximations:

- piecewise constant interpolation
- linear interpolation
- quadtratic interpolation

Global piecewise constant interpolation

To compute the integral

$$\int_{t_0}^T g(t) dt$$

we approximate the function g by a zero order polynomial

 $P_1(t)=a_0$

such that for given point $(t_i, g_i := g(t_i))$ the following interpolation condition holds

$$P_1(t_i) = g(t_i) \Rightarrow a_0 = g(t_i)$$

Global piecewise constant interpolation

This then leads to

$$\int_{t_0}^T g(t) dt = \int_{t_0}^T P(t) dt = \int_{t_0}^T a_0 dt = g_i (T - t_0)$$

Hence,

$$\int_{t_0}^T g(t) dt pprox g(t_i) (T-t_0)$$

In other words, one computes the integral as area of rectangle.

Global piecewise constant interpolation

Depending on the choice of $g(t_i)$ one may distinguish

- left sum: $g_i = g(t_0) \Rightarrow \int_{t_0}^T g(t) dt \approx g_i(T t_0)$
- right sum: $g_i = g(T) \Rightarrow \int_{t_0}^T g(t) dt \approx g_i(T t_0)$

• lower sum:
$$g_i = \min_{t \in I} g(t) \Rightarrow \int_{t_0}^T g(t) dt \approx g_i(T - t_0)$$

• upper sum:
$$g_i = \max_{t \in I} g(t) \Rightarrow \int_{t_0}^T g(t) dt \approx g_i(T - t_0)$$

• mid-point rule:
$$\int_{t_0}^T g(t) dt \approx 0.5(g(t_0) + g(T))(T - t_0)$$

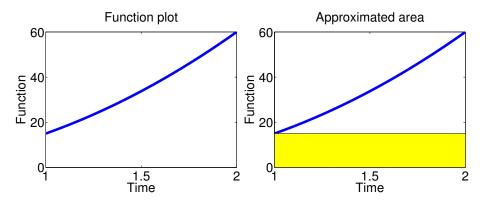
Pros:

• easy to perfom numerical integration

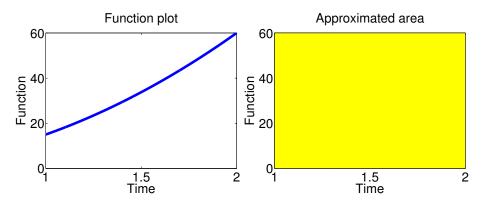
Cons:

not very much accurate

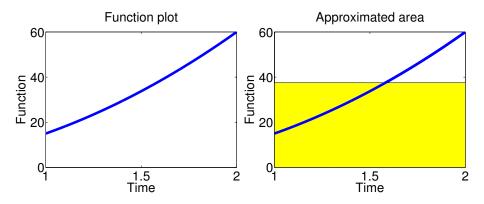
Left (also lower) integration rule



Right (also upper) integration rule



Mid-point integration rule



Exercise: Global interpolation

Problem: Compute the integral of $g(t) = 15t^2$ on interval [1,2].

Answer: having

$$g(t_0) = 15, \quad g(t_1) = 15 \cdot 2^2 = 60$$

approximation of integral by left (here also lower) rule becomes

$$\int_{t_0}^{T} g(t) dt \approx g(t_0) (T - t_0) = 15 \cdot (2 - 1) = 15$$

The correct value is

$$\int_{1}^{2} 15t^{2}dt = 15\frac{t^{3}}{3}|_{1}^{2} = \frac{15}{3}(2^{3} - 1^{3}) = 5 \cdot 7 = 35$$



Thus, global interpolation using this rule is not correct. The right/upper rule gives

$$\int_{t_0}^{T} g(t) dt \approx g(T_0)(T-t_0) = 60 \cdot (2-1) = 60$$

and leads to overestimation. Finally, the mid-point rule



$$\int_{t_0}^{T} g(t) dt \approx 0.5(g(t_0) + g(T))(T - t_0) = 37.5 \cdot (2 - 1) = 37.5$$

is also not correct. However, it is the closest to the solution.

Local piecewise constant interpolation

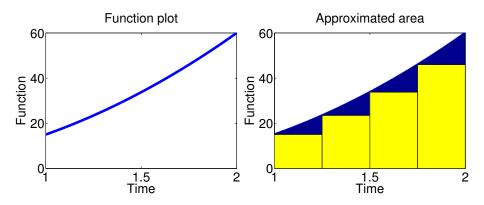
To increase accuracy one may apply the local piecewise interpolation. That is to divide the interval $[t_0, T]$ into N subintervals such that the integral reads

$$\int_{t_0}^T g(t) dt pprox \sum_{i=0}^{N-1} \int_{t_i}^{t_i+h} g(t) dt$$

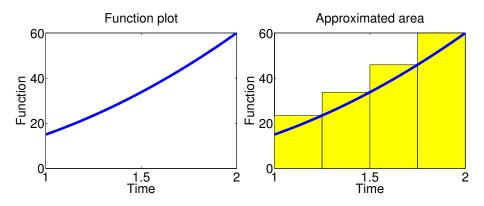
Then in each subinterval apply interpolation by

- left rule
- right rule
- Iower sum
- upper sum
- mid-point rule

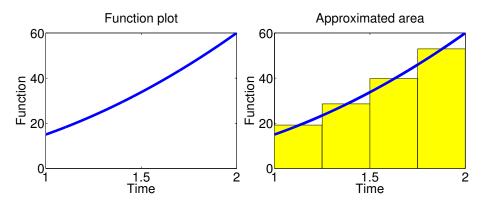
Local left (here also lower) rule



Local right (here also upper) rule



Local midpoint rule



Exercise: Local interpolation

Problem: Compute the integral of $g(t) = 15t^2$ on interval [1, 2] using composite left rule.

Answer: Divide the interval into N small subintervals and take $h = \frac{2-1}{N}$. Then compute the values of function g(t) in points $t_i = t_0 + hi$, $i = 0, 1, \dots, N$. Let us take, for example, N = 2, then h = 0.5 and the points are $t_0 = 1$, $t_1 = 1.5$, $t_2 = 2$. The values of function in those points are:



$$g(t_0) = 15$$
, $g(t_1) = 15 \cdot 1.5^2 = 33.75$, $g(t_2) = 15 \cdot 2^2 = 60$

Left rule reads:

$$\int_1^2 g(t)dt = \sum_{i=0}^{N-1} g(t_i)h$$

Exercise: Local interpolation

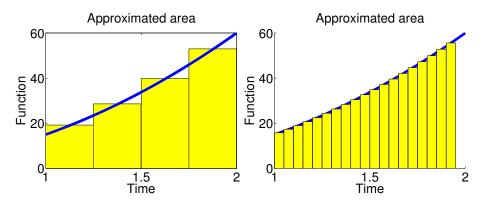
In our case N = 2 and hence

$$\int_{1}^{2} g(t)dt = (15 + 33.75) \cdot 0.5 = 24.375$$

This value is closer to the correct one compared to the global rule. Precision will improve with the increase of number N. On the other side, with the increase of N the number of function evaluations $g(t_i)$ will also increase (and hence the computation time).



Local midpoint rule:coarser vs finer



To compute the integral

$$\int_{t_0}^T g(t) dt$$

we may approximate the function g by a linear polynomial

$$P_1(t) = a_0 + a_1 t$$

such that for given two points $(t_i, g_i := g(t_i))$ and $(t_{i+1}, g_{i+1} := g(t_{i+1}))$ in I the following interpolation conditions hold

$$P_1(t_i) = g(t_i), \quad P_1(t_{i+1}) = g(t_{i+1})$$

The last relation represents the system of equations

$$\begin{pmatrix} 1 & t_i \\ 1 & t_{i+1} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = \begin{pmatrix} g(t_i) \\ g(t_{i+1}) \end{pmatrix} = \begin{pmatrix} g_i \\ g_{i+1} \end{pmatrix}$$

After solving one obtains

$$a_1=\frac{g_{i+1}-g_i}{t_i-t_{i+1}}$$

and

$$a_0 = g_i - rac{g_{i+1} - g_i}{t_i - t_{i+1}} t_i$$

Now substitute back the coefficients into polynomial:

$$P_1(t) = a_1 t + a_0$$

$$P_1(t) = \frac{g_{i+1} - g_i}{t_i - t_{i+1}} t + g_i - \frac{g_{i+1} - g_i}{t_i - t_{i+1}} t_i$$

$$= \frac{(g_{i+1} - g_i)t - g_i t_{i+1} - g_{i+1}t_i}{t_i - t_{i+1}}$$

$$P_1(t) = \frac{t - t_{i+1}}{t_i - t_{i+1}} g_i + \frac{t - t_i}{t_{i+1} - t_i} g_{i+1}$$

Note that in

$$P_1(t) = rac{t-t_{i+1}}{t_i-t_{i+1}}g_i + rac{t-t_i}{t_{i+1}-t_i}g_{i+1}$$

we have polynomials

$$\ell_i := rac{t-t_{i+1}}{t_i-t_{i+1}}, \quad \ell_{i+1} := rac{t-t_i}{t_{i+1}-t_i}g_{i+1}$$

These are known as **Lagrange polynomials**. Hence, our polynomial $P_1(t)$ is actually Lagrange expansion:

$$g(t) \approx P_1(t) = \ell_i g_i + \ell_{i+1} g_{i+1}$$

Once we know this, the integration can be performed.

Properties of ℓ_i

We have

$$\ell_i(t_k) := \delta_{ik} := \left\{ egin{array}{cc} 1, & k=i \ 0, & k
eq i \end{array}
ight.,$$

where δ_{ik} is the so-called Kronecker-product.

Integration reads

$$\int_{t_0}^T P_1(t) dt = \int_{t_0}^T \ell_i g_i dt + \int_{t_0}^T \ell_{i+1} g_{i+1} dt$$

$$\int_{t_0}^T P_1(t)dt = g_i \int_{t_0}^T \ell_i dt + g_{i+1} \int_{t_0}^T \ell_{i+1} dt$$
$$\int_{t_0}^T P_1(t)dt = g_i w_i + g_{i+1} w_{i+1}$$

where

$$w_i := \int_{t_0}^T \ell_i dt, \quad w_{i+1} = \int_{t_0}^T \ell_{i+1} dt$$

Weights in last relation do not depend on the function g(t). They read

$$w_0 := \int_{t_0}^T \frac{t-T}{t_0-T} dt, \quad w_1 = \int_{t_0}^T \frac{t-t_0}{T-t_0} dt$$

However, they do depend on the interval $I = [t_0, T]$, which prevents us from precomputing them. To resolve this issue, one may transform interval $[t_0, T]$ to [0, 1]. To achieve this use the transformation formula

$$\tau := \frac{t - t_0}{T - t_0}$$

which for $t = t_0$ gives 0 and for t = T gives 1.

From the transformation formula one has

$$d\tau = \frac{1}{T - t_0} dt$$

as well as

$$t=\tau(T-t_0)+t_0$$

This leads to

$$w_0 := \int_{t_0}^T \frac{t - t_0}{T - t_0} dt = \int_0^1 \frac{\tau(T - t_0) + t_0 - t_0}{T - t_0} (T - t_0) d\tau$$

i.e.

$$w_0 := (T-t_0) \int_0^1 \tau d\tau$$

in last relation

$$w_0 := (T-t_0) \int_0^1 \tau d\tau$$

integral $\int_0^1 \tau d\tau$ can be computed analytically. Let us denote its value as α_0 such that

$$w_0 := (T - t_0) \int_0^1 \tau d\tau = (T - t_0) \alpha_0$$

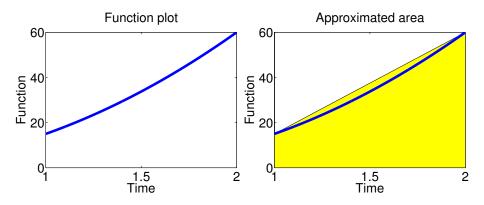
holds. Note that α_0 is the weigth which can be stored in a table. in a similar manner one may compute

$$w_1 = \int_{t_0}^T \frac{t-t_0}{T-t_0} dt$$

Once the weights are found, the final formula for linear numerical integration reads

$$\int_{t_0}^{T} g(t) dt = \int_{t_0}^{T} P_1(t) dt = g_0 w_0 + g_1 w_1$$
$$= (T - t_0) \sum_{i=0}^{1} \alpha_i g_i = 0.5(T - t_0)(g(t_0) + g(T)).$$

Hence, the integral is approximated by area of trapozoid. Thus, this numerical integration is named as **trapezoidal rule**.



Exercise: Global interpolation

Compute the integral of $f(x) = 15x^2$ on interval [1, 2]. Hence, $t_0 = 1, t_1 = 2$.

$$\int_{1}^{2}g(t)dt=\int_{1}^{2}15t^{2}=rac{1}{2}(t_{1}-t_{0})[g(t_{0})+g(t_{1})]$$

$$\int_{1}^{2} 15t^{2} dt = \frac{1}{2}(2-1)[15 \cdot 1^{2} + 15 \cdot 2^{2}] = 0.5 \cdot 75 = 37.5$$

True value:

$$\int_{1}^{2} 15t^{2}dt = 15\frac{t^{3}}{3}|_{1}^{2} = \frac{15}{3}(2^{3} - 1^{3}) = 5 \cdot 7 = 35$$



Local linear interpolation

In similar manner as before the global linear interpolation can be translated to the local linear interpolation by using cummulative rule

$$\int_{t_0}^T g(t) dt \approx \sum_{i=0}^{N-1} \int_{t_i}^{t_i+h} g(t) dt$$

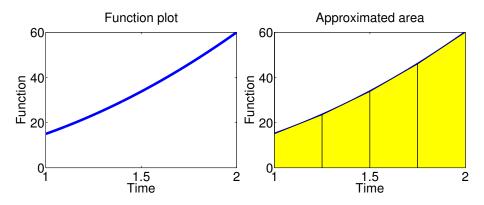
and in each subinterval interpolating function g(t) by a linear polynomial

$$\int_{t_i}^{t_i+h} P_1(t) dt = \int_{t_i}^{t_i+h} \ell_i g_i dt + \int_{t_i}^{t_i+h} \ell_{i+1} g_{i+1} dt$$

Finally,

$$\int_{t_0}^{T} g(t) dt \approx \sum_{i=0}^{N-1} 0.5h(g_i + g_{i+1})$$

Local linear interpolation



Exercise: Local interpolation

Compute the integral of $g(t) = 15t^2$ on interval [1,2] by composite rule. Hence, let us take N = 2. Then, $t_0 = 1, t_1 =$ $1.5, t_2 = 2.$

$$\int_{1}^{2} g(t)dt = \int_{1}^{1.5} 15t^{2} + \int_{1.5}^{2} 15t^{2}$$

$$= \frac{1}{2}(t_{1} - t_{0})[g(t_{0}) + g(t_{1})] + \frac{1}{2}(t_{2} - t_{1})[g(t_{2}) + g(t_{1})]$$

$$\int_{1}^{2} 15t^{2}dt = \frac{1}{2}(1.5 - 1)[15 \cdot 1^{2} + 15 \cdot (1.5)^{2}]$$

$$+ \frac{1}{2}(2 - 1.5)[15 \cdot (1.5)^{2} + 15 \cdot (2)^{2}] = 12.1875 + 23.4375 = 35.625$$

2

To compute the integral

$$\int_{t_0}^T g(t) dt$$

we approximate the function g by a quadratic polynomial

$$g(t)\approx P_2(t)=a_2t^2+a_1t+a_0$$

where

$$P_2(t_i) = g_i$$

 $P_2(t_{i+1}) = g_{i+1}$
 $P_2(t_{i+2}) = g_{i+2}$

Here, points t_i are equidistant in the interval $[t_0, T]$.

Having that $t_{i+1} = \frac{t_i + t_{i+2}}{2}$ and after solving previous system for a_2, a_1 and a_0 one obtains

$$a_2 = \frac{(2g_i - 4g_{i+1} + 2g_{i+2})}{(t_i^2 - 2t_i t_{i+2} + t_{i+2}^2)}$$

$$egin{aligned} &a_1=-rac{(g_it_i-4g_{i+1}t_i+3g_it_{i+2}+3g_{i+2}t_i-4g_{i+1}t_{i+2}+g_{i+2}t_{i+2})}{(t_i^2-2t_it_{i+2}+t_{i+2}^2)}\ &a_0=rac{(g_it_{i+2}^2+g_{i+2}t_i^2+g_it_it_{i+2}-4g_{i+1}t_it_{i+2}+g_{i+2}t_it_{i+2})}{(t_i^2-2t_it_{i+2}+t_{i+2}^2)} \end{aligned}$$

After substitution in $P_2(t)$ one obtains:

$$egin{aligned} P_2(t) &= rac{(t-t_{i+1})(t-t_{i+2})}{(t_i-t_{i+1})(t_i-t_{i+2})}g_i + rac{(t-t_i)(t-t_{i+2})}{(t_{i+1}-t_i)(t_{i+1}-t_{i+2})}g_{i+1} \ &+ rac{(t-t_i)(t-t_{i+2})}{(t_{i+2}-t_i)(t_{i+2}-t_{i+1})}g_{i+2}. \end{aligned}$$

in which the quadratic Lagrange polynomials read

$$\ell_{i} = \frac{(t - t_{i+1})(t - t_{i+2})}{(t_{i} - t_{i+1})(t_{i} - t_{i+2})}, \quad \ell_{i+1} = \frac{(t - t_{i})(t - t_{i+2})}{(t_{i+1} - t_{i})(t_{i+1} - t_{i+2})}$$
$$\ell_{i+2} = \frac{(t - t_{i})(t - t_{i+2})}{(t_{i+2} - t_{i})(t_{i+2} - t_{i+1})}$$

Hence,

$$P_2(t) = \ell_i g_i + \ell_{i+1} g_{i+1} + \ell_{i+2} g_{i+2}$$

which after integration leads to

$$\int_{t_0}^T P_2(t) dt = g_i \int_{t_0}^T \ell_i dt + g_{i+1} \int_{t_0}^T \ell_{i+1} dt + g_{i+2} \int_{t_0}^T \ell_{i+2} dt$$

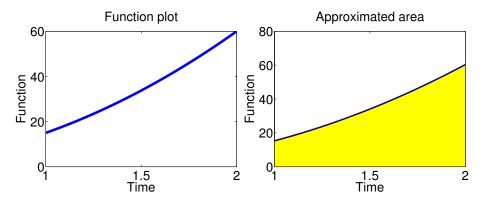
If we employ transformation of the interval $[t_0, T]$ into [0, 1] in a similar manner as for linear interpolation one obtains

$$\int_{t_0}^{T} P_2(t) dt = (T - t_0) \sum_{i=0}^{n} g(t_i) \underbrace{\int_{0}^{1} \prod_{j=0 \atop j \neq i}^{n} \frac{nt - j}{i - j}}_{=:\alpha_i} dt$$

After simple mathematical operations one finally obtains:

$$\int_{t_0}^T g(t) dt = \int_{t_0}^T P_2(t) dt = \frac{T - t_0}{6} (g_i + 4g_{i+1} + g_{i+2})$$

which is known as Simpson's rule.



Exercise: Global interpolation

Compute the integral of $g(t) = 15t^2$ on interval [1,2]. Hence, $t_0 = 1, t_1 = 1.5, t_2 = 2$.

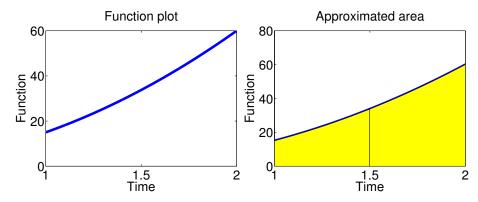
$$\int_{1}^{2} g(t) dt = \frac{T - t_0}{6} (g_i + 4g_{i+1} + g_{i+2})$$

$$\int_{1}^{2} g(t) dt = \frac{1}{6} (15(1)^{2} + 4 \cdot 15 \cdot (1.5)^{2} + 15 \cdot (2)^{2} = 35$$

This shows us that the global quadratic interpolation is exact for the function of polynomial order 2. Namely, this method is exact for function of polynomial order 3 or less.



Quadratic interpolation applied on each subinterval h of interval l is called local. And can be applied in a similar manner as previous interpolations.



Exercise: Local interpolation

Compute the integral of $g(t) = 15t^2$ on interval [1,2] by composite rule. Hence, N = 2 and $t_0 = 1$, $t_1.25$ $t_2 = 1.5$, $t_3 = 1.75$ $t_4 = 2$.

$$\int_{1}^{2} g(t)dt = \int_{t_{0}}^{t_{2}} g(t)dt + \int_{t_{2}} t_{3}g(t)dt$$
$$\int_{1}^{2} g(t)dt = \frac{0.5}{6}(15(1)^{2} + 4 \cdot 15 \cdot (1.25)^{2} + 15 \cdot (1.5)^{2} + \frac{0.5}{6}(15(1.5)^{2} + 4 \cdot 15 \cdot (1.75)^{2} + 15 \cdot (2)^{2})$$
$$= 35$$



Hence, the same conclusion as for global interpolation.

Let us define the function

$$\phi(t) = g(t) - P_n(t) - \frac{g(t_p) - P_n(t_p)}{v(t_p)}v(t)$$

in which g(t) is the function we would like to integrate (belongs to $C^{n+1}(I)^a$) and $P_n(t)$ is the *n*-th order interpolation polynomial (through t_i , i = 0, ..., n interpolation points). Argument t_p is some chosen (fixed) point in interval of definition I, not neccessary the interpolation point. Here,

$$v(t) = \prod_{i=0}^{n} (t - t_i), v(t_p) = \prod_{i=0}^{n} (t_p - t_i)$$

 ^{a}g is in the interval I n + 1-times continuously differentiable

The function

$$\phi(t) = g(t) - P_n(t) - \frac{g(t_p) - P_n(t_p)}{v(t_p)}v(t)$$

has the following properties

•
$$\phi(t) = 0$$
 when $t = t_i$ or $t = t_p$, i.e. it has $(n+2)$ roots

•
$$\phi'(t)$$
 has $n+1$ roots

• $\phi^{(n+1)}(t)$ has only one root denoted by ξ , i.e. $\phi^{(n+1)}(\xi) = 0$

This allows us to write

$$\phi^{(n+1)}(t) = \frac{d^{(n+1)}}{dt^{(n+1)}} (g(t) - P_n(t) - \frac{g(t_p) - P_n(t_p)}{v(t_p)} v(t)$$

= $g^{(n+1)}(t) - P_n^{(n+1)}(t) - \frac{g(t_p) - P_n(t_p)}{v(t_p)} v^{(n+1)}(t)$
= $g^{(n+1)}(t) - P_n^{(n+1)}(t) - \frac{g(t_p) - P_n(t_p)}{v(t_p)} (n+1)!$

Furthermore,

$$\phi^{(n+1)}(\xi) = 0 \Rightarrow g^{(n+1)}(\xi) - P_n^{(n+1)}(\xi) - \frac{g(t_p) - P_n(t_p)}{v(t_p)}(n+1)! = 0$$

Here note that $P_n^{(n+1)}(\xi) = 0$ since P_n is polynomial of order n

Thus,

$$g^{(n+1)}(\xi) - rac{g(t_p) - P_n(t_p)}{v(t_p)}(n+1)! = 0$$

from which one obtains the estimate of the error

$$g(t_{P}) - P_{n}(t_{P}) = rac{1}{(n+1)!}g^{(n+1)}(\xi)v(t_{P}) = rac{1}{(n+1)!}g^{(n+1)}(\xi)\prod_{i=0}^{n}(t_{P}-t_{i})$$

in point t_p of interval.

Theorem

Let $g \in C^{n+1}(I)$, then there exists for every t a $\xi \in I_s$ (smallest interval which contains points) such that

$$g(t) - P_n(t) = \frac{1}{(n+1)!}g^{(n+1)}(\xi)\prod_{j=0}^n(t-t_j)$$

holds. Since ξ is unknown, the error is estimated by

$$|g(t) - P_n(t)| \leq \frac{1}{(n+1)!} \max_{\xi \in I_s} |g^{(n+1)}(\xi)| \left| \prod_{j=0}^n (t-t_j) \right|.$$

Error of numerical integration

Aim: Find an error estimator for

$${\sf E}_n^I(g):=\int_{t_0}^T g(t)dt \ -\sum_{j=0}^n lpha_j g(t_j)$$

Theorem: Let $H := \frac{T-t_0}{n}$ and g be sufficiently often differentiable in I. Then there exists a constant c_n independent of the interval I such that

$$|E_n'(g)| \leq rac{c_n}{(n+1)!} \max_{\xi \in I} \left| g^{(n+1)}(\xi)
ight|, \quad c_n := |\int_{t_0}^T \prod_{i=0}^n (t-t_i)|$$

holds. If n is even, then there exists a constant d_n independent of the interval I such that

$$|E_n^I(g)| \leq rac{d_n}{(n+2)!} \max_{\xi \in I} \left| g^{(n+2)}(\xi)
ight|, \quad d_n := |\int_{t_0}^T t \prod_{i=0}^n (t-t_i)|$$

holds.

Exercise: error

Before we have used linear global interpolation to compute the value of intergral of a function $f(x) = 15t^2$ on interval [1,2]. We got the following result $\begin{array}{c} exact \\ 35 \\ 37.5 \end{array}$ Let us now compute the error using error estimate

$$|E_n^l(g)| \leq rac{c_n}{(n+1)!} \max_{\xi \in I} \left| g^{(n+1)}(\xi)
ight|, \quad c_n := |\int_{t_0}^T \prod_{i=0}^n (t-t_i)|$$



in which n + 1 = 2.

Exercise: error

Having

$$g^{'} = 30t, g^{''} = 30$$

one concludes that

$$\max_{\xi\in I} \left| g^{(n+1)}(\xi) \right| = 30$$

Furthermore,

$$\int_{t_0}^T \prod_{i=0}^n (t-t_i) = \int_0^1 (t-0)(t-1)dt = \frac{t^3}{3} - \frac{t^2}{2} = -\frac{1}{6}$$

Thus,

$$|E'_n(g)| \le \frac{rac{1}{6}}{(2)!}30 = 2.5$$



Let us compute the error of linear global interpolation to compute the value of intergral of a function f(x) = 15t on interval [1, 2]. by applying linear integration one gets the following result $\frac{45}{2}$ $\frac{45}{2}$ Let us now compute the error using error estimate

$$|E_n'(g)| \leq rac{c_n}{(n+1)!} \max_{\xi \in I} \left| g^{(n+1)}(\xi) \right|, \quad c_n := |\int_{t_0}^T \prod_{i=0}^n (t-t_i)|$$



in which n + 1 = 2.

Exercise: error

Having

$$g^{'} = 15, g^{''} = 0$$

one concludes that

$$\max_{\xi\in I} \left|g^{(n+1)}(\xi)\right| = 0$$

and hence

$$|E_n'(g)| \leq 0$$

Thus, our integration is exact for linear polynomials (as well as for constant).



So far we have learned that given any set of n points t_i over the time interval I we may build Newton-Cotes quadrature rule

w_i, t_i

which can be used to exactly integrate polynomials of order n - 1 or less by using the formula

$$\int_{t_0}^{T} P(t) \ dt = \sum_{i=0}^{n-1} P(t_i) w_i$$

The question is: can we do better than that?

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Yes, we can

- Newton-Cotes formula
 - one chooses *n* points *t_i* in which the value of function will be evaluated (uniformly distributed)
 - weigths are computed according to given points t_i
 - accurate for polynomials of order n-1 given n points
- improvement: Gauss quadrature
 - vary both *n* points *t_i* and *n* weights *w_i* (both are unknown)
 - make the integration formula exact for polynomials of order 2n-1

The Gauss quadrature has for a goal to vary the placements t_i such that the integration is more accurate. In general Gauss formula approximates:

$$\int_{-1}^{1} g(t) dt \approx \int_{-1}^{1} P(t) dt = \sum_{i=0}^{n} g(t_i) w_i, \quad w_i = \int_{-1}^{1} \ell_i(t) dt$$

in the same way as Newton-Cotes. But, t_i are unknown and have to be found. Additionally to them are also unknown the coefficients of a polynomial P(t).

1 point Gauss quadrature

Let us have 1 point formula in which t_i and hence w_i are unknown. Having two unkowns, one has to build the system of two equations. This can be achieved by letting the formula to exactly integrate constant and linear polynomial. One does not loose on generality by letting

$$\int_{-1}^{1} 1dt = w_i g(t_i) = w_i \cdot 1.$$
$$\int_{-1}^{1} tdt = w_i g(t_i) = w_i \cdot t_i.$$

Having that $\int_{-1}^{1} 1 dt = 2$ and $\int_{-1}^{1} t dt = 0$, the last system reduces to

$$t_i w_i = 0$$

$$w_i = 2$$

Hence, $t_i = 0$, $w_i = 2$ is 1 point Gauss rule which exactly integrates constant and linear polynomials over [-1, 1]. Do not forget that in Newton-Cotes formula we needed two points, i.e. two function evaluations.

2 point- Gauss quadrature

Let us have 2 point formula in which t_{i-1} , t_i and hence w_{i-1} , w_i are unknown. Having four unkowns, one has to build the system of four equations. This can be achieved by letting the formula to exactly integrate polynomial up to third order. One does not loose on generality by letting

$$\int_{-1}^{1} 1 dt = w_{i-1} + w_i$$
$$\int_{-1}^{1} t dt = t_{i-1} w_{i-1} + t_i w_i$$
$$\int_{-1}^{1} t^2 dt = t_{i-1}^2 w_{i-1} + t_i^2 w_i$$
$$\int_{-1}^{1} t^3 dt = t_{i-1}^3 w_{i-1} + t_i^3 w_i$$

This leads to a system of equatons whose solution is

$$t_{i-1}=-\sqrt{rac{1}{3}},$$
 $w_{i-1}=1$ $t_i=\sqrt{rac{1}{3}},$ $w_i=1$

Hence, the two point Gauss-quadrature rule approximates exactly the polynomials of order 3. Newton Cotes formula needed 4 evaluations.

Generalisation

We need generalisation of 1-point and 2-pont rule. For this, note that 1-point rule

$$t_0 = 0, w_0 = 2$$

is placed in $t_0 = 0$ which is the root of a first order polynomial q(t) = t. Two point rule

$$t_0 = -\sqrt{\frac{1}{3}}, w_0 = 1, \quad t_1 = \sqrt{\frac{1}{3}}, w_1 = 1$$

is placed at roots of second order polynomial $q(t) = 0.5(3t^2 - 1)$. These two polynomials are known as **Legendre polynomials**.

What about n - point rule?

Generalisation

Let us express 2n-1 order polynomial P(t) over the *n*-th order Lagrange polynomial q(t) whose roots are the integration points t_i such that

$$\mathsf{P}(t) = q(t)\phi(t) + r(t)$$

holds. Here, $\phi(t)$ and reminder r(t) are at most of order n-1 or less. By applying integration rule

$$\int_{-1}^{1} g(t) dt \approx \int_{-1}^{1} P(t) dt = \int_{-1}^{1} (q(t)\phi(t) + r(t)) dt \Rightarrow \int_{-1}^{1} g(t) dt = \int_{-1}^{1} r(t) dt$$

one may conclude that the last expression reduces only to $\int_{-1}^{1} r(t)dt$ having that the polynomial q(t) is equal to zero in integration points (its own roots) t_i . Since r(t) is of most n-1 order, the last integration is exact given n points.

In a similar manner as for Newton-Cotes formula, one may derive the error estimate:

$$\left| {E_n^{[a,b]} (g)}
ight| \le (b-a)^{2n+1} rac{(n!)^4}{[(2n)!]^3 (2n+1)} \max_{\xi \in [a,b]} \left| {g^{(2n)} (\xi)}
ight|$$

Note that Gauß-formulas can be also made composite in a similar manner as Newton-Cotes formula.

Another interval

So we know how to compute

$$\int_{-1}^{1} g(t) dt$$

What to do if

$$\int_a^b g(t)dt = ?$$

Then one may apply the transformation formula

$$\int_a^b g(t)dt = \int_{-1}^1 f(p)dp$$

in which

$$t = qp + s, dt = qdp$$

such that p = -1 for t = a and p = 1 for t = b.

Another interval

By solving one obtains that

$$q=\frac{b-a}{2}, \quad s=\frac{a+b}{2}$$

Hence,

$$\int_{a}^{b} g(t)dt = \int_{-1}^{1} g(\frac{b-a}{2}p + \frac{a+b}{2})\frac{b-a}{2}dp = \int_{-1}^{1} f(p)dp$$

and now you are ready to use the quadrature formula on $\left[-1,1\right].$

Note that Gauss-Legendre formula is the most often used, but it is not the only one. Integration points can also be selected as roots of other kind of polynomials. If the integral

 $\int_a^b g(t)dt$

can be represented as

$$\int_a^b g(t)dt = \int_a^b v(t)f(t)dt = \sum_{i=0}^{n-1} w_i f(t_i)$$

in which v(t) is the weight, then integration rule will depend on the type of weight v(t) as given in the following slide.

Other Gauss quadratures

Some of Gauss quadrature rules:

Weight $v(t)$	Interval (a, b)	Polynomial
1	[-1, 1]	Lagrange
e^{-t^2}	$(-\infty,\infty)$	Hermite
e^{-t}	$(0,\infty)$	Laguerre
$\sqrt{1-t^2}$	[-1, 1]	Chebishev (I kind)