

Ch. 3 ~~Key~~ Runge-Kutta Methods

Key Points:

Quadrature (Newton-Cotes, Gaussian, ...)

weights, nodes

orthogonal polynomials (Jacobi, Legendre, Chebyshev, ...)

Runge-Kutta schemes.

- RK matrix
- RK weights
- RK nodes
- stages

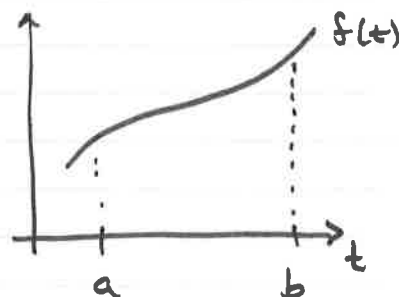
Can we tie this back to ideas
used in multistep case even though
these will be single-step methods

(i.e. interpolation used there for $\int p(t) dt$?)

Numerical Quadrature (see also Heath, Ch. 8)

We've talked a little about ways to approximate an integral

$$I(f) \equiv \int_a^b f(t) dt$$



"left endpoint"

$$\int_a^b f(t) dt \approx (b-a) \cdot f(a)$$

"right endpoint"

$$\int_a^b f(t) dt \approx (b-a) \cdot f(b)$$

midpoint rule

$$\int_a^b f(t) dt \approx (b-a) f\left(\frac{a+b}{2}\right)$$

trapezoid rule

$$\int_a^b f(t) dt \approx \frac{(b-a)}{2} [f(a) + f(b)]$$

These are exact for the case where $f(t) = \text{polynomial of degree } 0$ (i.e. a constant function)

...
but midpoint is exact also for a polynomial of degree 1.

This is exact if the function is linear. (polynomial of degree one).

What about more general methods?

Here is an example — Simpson's Method

Approximate the integral

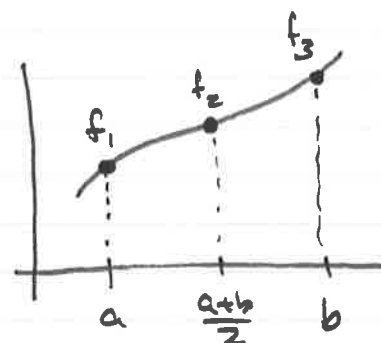
$$I(f) = \int_a^b f(t) dt$$

using ~~three~~ three equally-spaced points

~~using~~ (3-pt. quadrature rule)

so

$$\int_a^b f(t) dt \approx w_1 f_1 + w_2 f_2 + w_3 f_3$$



$$f_1 = f(a)$$

$$f_2 = f\left(\frac{a+b}{2}\right)$$

$$f_3 = f(b)$$

where the weights w_1, w_2 and w_3 are chosen so that the quadrature rule integrates the first 3 polynomial basis functions exactly. We'll use monomial basis functions $(1, t, t^2)$ in this example.

$$f(t) = 1: \quad \int_a^b 1 dt = (b-a) = w_1 \cdot 1 + w_2 \cdot 1 + w_3 \cdot 1$$

$$f(t) = t: \quad \int_a^b t dt = \left. \frac{1}{2} t^2 \right|_a^b = \frac{1}{2} (b^2 - a^2) = w_1 a + w_2 \left(\frac{a+b}{2} \right) + w_3 b$$

$$f(t) = t^2: \quad \int_a^b t^2 dt = \left. \frac{1}{3} t^3 \right|_a^b = \frac{1}{3} (b^3 - a^3) = w_1 a^2 + w_2 \left(\frac{a+b}{2} \right)^2 + w_3 b^2$$

This leads to the linear system

$$\begin{bmatrix} 1 & 1 & 1 \\ a & \frac{a+b}{2} & b \\ a^2 & \left(\frac{a+b}{2}\right)^2 & b^2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} = \begin{bmatrix} b-a \\ \frac{1}{2}(b^2-a^2) \\ \frac{1}{3}(b^3-a^3) \end{bmatrix}$$

Solving for the weights gives

$$w_1 = \frac{1}{6}(b-a) \quad w_2 = \frac{2}{3}(b-a) \quad w_3 = \frac{1}{6}(b-a)$$

So

$$\int_a^b f(t) dt = (b-a) \left[\frac{1}{6} f(a) + \frac{2}{3} f\left(\frac{a+b}{2}\right) + \frac{1}{6} f(b) \right]$$

This is Simpson's Quadrature Rule.

(order 3)

For discussion of degree/order see
P. (49.2) in notes.

Consider the ~~most general~~ numerical quadrature case ~~where we~~ ~~approximate the integral~~ where we approximate the integral

$$I(f) \equiv \int_a^b f(t) dt$$

by the n -point quadrature rule

$$I(f) \approx Q_n(f) = \sum_{i=1}^n w_i f(t_i)$$

where $a \leq t_1 < t_2 < \dots < t_n \leq b$. The points t_i are called the nodes and the coefficients w_i are called the weights.

Plan: choose the nodes + weights for optimal performance (accuracy, computational cost, stability, ...)

Consider first the case where the nodes are specified (possibly equally-spaced but not necessarily).

Given $t_1, t_2, \dots, t_n$ find $w_1, w_2, \dots, w_n$ so that the quadrature rule integrates the first n basis functions exactly. Again, we'll use ~~the~~ the monomial basis functions here

$$1, t, t^2, \dots, t^{n-1}$$

(As we shall note below, this gives a quadrature rule of order n .)

$$\int_a^b 1 dt = b-a = w_1 + w_2 + \dots + w_n$$

$$\int_a^b t dt = \frac{b^2 - a^2}{2} = w_1 t_1 + w_2 t_2 + \dots + w_n t_n$$

$$\vdots$$

$$\int_a^b t^{n-1} dt = \frac{b^n - a^n}{n} = w_1 t_1^{n-1} + \dots + w_n t_n^{n-1}$$

This leads to the linear system

$$\begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ t_1 & t_2 & t_3 & \dots & t_n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ t_1^{n-1} & t_2^{n-1} & t_3^{n-1} & \dots & t_n^{n-1} \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{bmatrix} = \begin{bmatrix} b-a \\ \frac{1}{2}(b^2-a^2) \\ \vdots \\ \frac{1}{n}(b^n-a^n) \end{bmatrix}$$

fix in notes...

Vandermonde Matrix

- this nonsingular (although possibly not well-conditioned if n is large) can be solved to uniquely determine the weights $w_1, w_2, \dots, w_n$.

An alternative way to view this same problem is ~~is~~ in terms of the Lagrangian basis functions. Basically, what we just did was to fit ~~the~~ ^{the function at} ~~the~~ ⁿ points to a polynomial of degree $n-1$ and then integrated that polynomial.

So in terms of the Lagrangian basis functions

$$l_j(t) = \frac{\prod_{k=1, k \neq j}^n (t - t_k)}{\prod_{k=1, k \neq j}^n (t_j - t_k)} \quad j=1, 2, \dots, n$$

So

$$f(t) \approx p(t) = f_1 l_1(t) + f_2 l_2(t) + \dots + f_n l_n(t)$$

Note: $l_j(t_i) = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases} \quad f_j = f(t_j)$

So

$$\int_a^b f(t) dt \approx \int_a^b p(t) dt = \sum_{j=1}^n f_j \underbrace{\int_a^b l_j(t) dt}_{w_j}$$

That is,

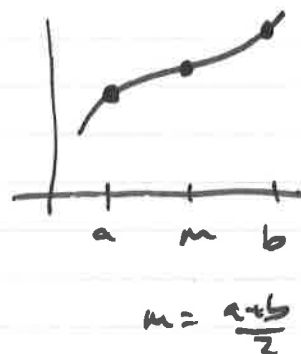
$$w_j = \int_a^b l_j(t) dt \quad j=1, 2, \dots, n$$

EX (Simpson's Rule)

$$l_1(t) = \frac{(t-m)(t-b)}{(a-m)(a-b)}$$

$$l_2(t) = \frac{(t-a)(t-b)}{(m-a)(m-b)}$$

$$l_3(t) = \frac{(t-a)(t-m)}{(b-a)(b-m)}$$



... after some effort ...

$$W_1 = \int_a^b l_1(t) dt = \frac{1}{6}(b-a)$$

$$W_2 = \int_a^b l_2(t) dt = \frac{2}{3}(b-a)$$

$$W_3 = \int_a^b l_3(t) dt = \frac{1}{6}(b-a)$$

so

$$\int_a^b f(t) dt \approx W_1 f_1 + W_2 f_2 + W_3 f_3.$$

Error in Quadrature Rules

see also
(Heath, p. 344)

$$I(f) = \int_a^b f(t) dt$$

$$E(f) \equiv I(f) - Q_n(f)$$

$$Q_n(f) = \sum_{i=1}^n w_i f(t_i)$$

~~Suppose the quadrature rule is exact~~ Suppose the quadrature rule is exact

for a polynomial of degree $n-1$.

That is,

$$Q_n(f) = I(p_{n-1})$$

Note: $Q_n(f) = Q_n(p_{n-1})$
since p_{n-1} interpolates f

Note that later we'll
make more specific
error estimates for
midpoint, Simpson,
trap. & in some cases
get improved estimates

Then

$$|E(f)| = |I(f) - Q_n(f)|$$

$$= |I(f) - I(p_{n-1})|$$

$$= |I(f - p_{n-1})|$$

$$\leq (b-a) \|f - p_{n-1}\|_{\sup}$$

$$\leq \frac{(b-a)h^n}{4n} \|f^{(n)}\|_{\sup}$$

$$\leq \frac{1}{4} h^{n+1} \|f^{(n)}\|_{\sup}$$

see notes p. (48.1)

$f^{(n)}$ = n th deriv. of f .

$$\begin{cases} h = \max h_i \\ h_i = t_{i+1} - t_i \end{cases}$$

$$h = \frac{b-a}{n}$$

Def: A quadrature rule is ~~order n~~ if it is
exact for every polynomial of degree $n-1$ but not
exact for a polynomial of degree n

Isaacs
p. 33/34

Note: (p. 344)

Heath defines the degree of a quadrature rule..

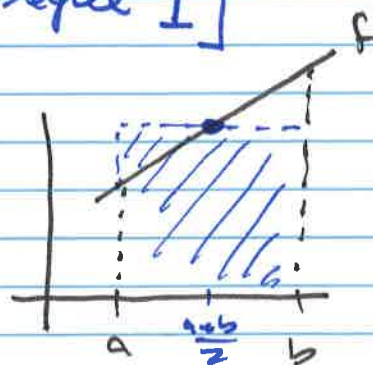
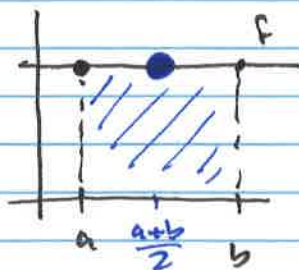
Def: A quadrature rule is said to be of degree d if it is exact for every polynomial of degree d but is not exact for some polynomial of degree $d+1$.

(i.e. ~~the~~ the largest positive integer n such that the formula is exact for t^k

- An n -pt. interpolatory quadrature rule is ~~is~~ of ~~degree~~ $k=0,1,\dots,n$ degree at least $n-1$.

Observe:

- The Midpoint Rule is a 1-pt. interpolatory quadrature rule. Note that it gets both constant functions (polynomial of degree 0) and linear functions (polynomial of degree 1) exact [Degree 1]



- The Trapezoid Rule is a 2-pt. interpolatory quadrature rule. It gets constant and linear functions exact. It does not get quadratic functions exact. [Degree 1]

Iserles refers to these as Order 2. ["Order" = "Degree" + 1]

(see Heath, p. 324)

✓
48.1

Note: If a polynomial of degree $n-1$, $P_{n-1}(t)$, interpolates a sufficiently smooth function $f(t)$ at $t = t_1, t_2, \dots, t_n$ distinct points, then

$$f(t) - P_{n-1}(t) = \frac{f^{(n)}(\theta)}{n!} (t-t_1)(t-t_2)\dots(t-t_n)$$

where $\theta \in [t_1, t_n]$.

If $|f^{(n)}(t)| < M$ for $t \in [t_1, t_n]$

and $h = \max \{t_{i+1} - t_i \mid i=1, 2, \dots, n-1\}$ then

$$\|f(t) - P_{n-1}(t)\|_{\sup} \leq \frac{M h^n}{4n}$$

$$\|\cdot\|_{\sup} = \sup \{|\cdot| \mid t \in [t_1, t_n]\} \quad \text{least upper bound}$$

See also Iserles result p. 33 (written here for weighting function $w(t)=1$).

$$\left| \int_a^b f(t) dt - \sum_{i=1}^n w_i f(t_i) \right| \leq C \max_{a \leq t \leq b} |f^{(p)}(t)|$$

(proved using
Peano kernel theorem)

constant > 0
indep. of f

↑
order "p" quadrature
rule.

~~constant > 0~~

Summarizing so far ... (Iserles Lemma 3.1)

Lemma 3.1

Given any distinct set of nodes $t_1, t_2, \dots, t_n$, it is possible to find a set of weights $w_1, w_2, \dots, w_n$ such that

$$\int_a^b f(t) dt \approx \sum_{i=1}^n w_i f(t_i)$$

is of order $p \geq n$. (so degree $\geq n-1$)

Proof: this uses the construction we've already examined using basis functions $(1, t, t^2, \dots, t^{n-1})$, Vandermonde matrix, etc.

Remark:

- Again recall the example of the midpoint rule that is an interpolatory quadrature rule (1-pt) (i.e. $n=1$) but has order 2 (i.e. $p=2, n=1$) (degree 1)

$$p \geq n.$$

- Simpson's Rule, a 3-pt interpolatory quadrature rule, (i.e. $n=3$) has order 4 (degree 3) (i.e. $p=4, n=3$)
- Trapezoid Rule, a 2-pt interpolatory quadrature rule, (i.e. $n=2$) has order 2 (so here $p=2, n=2$)

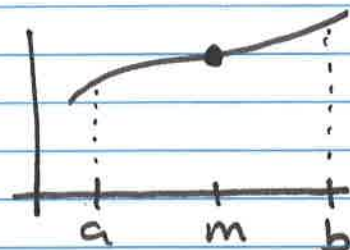
Comment: In general, an n -pt N-C quadrature method with $n \equiv \text{odd}$ has ~~degree~~ degree one higher than the interpolating polynomial.

(49.1)

Error in Midpoint Rule

$$I(f) = \int_a^b f(t) dt$$

$$Q_{\text{MID}}(f) = (b-a) f(m)$$



$$m = \frac{a+b}{2}$$

$$\underbrace{\hspace{2cm}}_{\equiv h}$$

$$E(f) = I(f) - Q_{\text{MID}}(f)$$

$$= \int_a^b f(t) dt - (b-a) f(m)$$

Newton-Cotes
(given equally spaced points)

Let us expand $f(t)$ in a Taylor Series about $t=m$

$$f(t) = f(m) + f'(m)(t-m) + \frac{1}{2} f''(m)(t-m)^2 + \frac{1}{3!} f'''(m)(t-m)^3 + \frac{1}{4!} f^{(4)}(\theta)(t-m)^4$$

$$\int_a^b f(t) dt = f(m)(b-a) + f'(m) \int_a^b (t-m) dt + \frac{1}{2} f''(m) \int_a^b (t-m)^2 dt + \frac{1}{3!} f'''(m) \int_a^b (t-m)^3 dt + \frac{1}{4!} f^{(4)}(\theta) \int_a^b (t-m)^4 dt$$

$$= f(m)(b-a) + \frac{1}{2} f''(m) \left[\frac{1}{3} (b-m)^3 - \frac{1}{3} (a-m)^3 \right] + \frac{f^{(4)}(\theta)}{4!} \left[\frac{1}{5} (b-m)^5 - \frac{1}{5} (a-m)^5 \right]$$

$$= f(m)(b-a) + \frac{1}{2} f''(m) \left[\frac{1}{3} h^3 - \frac{1}{3} (-h)^3 \right] + O(h^5)$$

$$= f(m)(b-a) + \frac{1}{3} h^3 f''(m) + O(h^5)$$

So

$$E(f) = I(f) - Q_{\text{MID}}(f) \\ = \frac{1}{3} h^3 f''(m) + O(h^5)$$

this formula suggests that $E=0$ (method is exact) for constant and linear functions (for which $f''=0$).

interpolating polynomial is degree 0

(ie. midpoint $\rightarrow$ degree 1, order 2)

- A similar calculation can be made (see Burden + Faires) to show for Simpson's rule

$$E(f) = I(f) - Q_{\text{SIMP}}(f) \approx h^5 f^{(4)}(\theta)$$

Simpson (3-pt quad.) is exact for polynomials of degree 0, 1, 2, 3.

interpolating polynomial is degree 2

(ie. Simpson's Rule $\rightarrow$ degree 3, order 4)

- Similarly, Trapezoid Rule has (Burden + Faires)

$$E(f) = I(f) - Q_{\text{TRAP}}(f) \approx h^3 f''(\theta)$$

interpolating polynomial is degree 1

(degree 1, order 2)

(compare conditioning?) 50

Stability in Quadrature Rules

$$\text{let } Q_n(f) = \sum_{i=1}^n w_i f(t_i)$$

be some quadrature rule (n-pt.)

If there is a perturbation in the function — call it $\hat{f}$

$$|Q_n(\hat{f}) - Q_n(f)| = \left| \sum_{i=1}^n w_i \hat{f}(t_i) - \sum_{i=1}^n w_i f(t_i) \right|$$

$$= \left| \sum_{i=1}^n w_i (\hat{f}(t_i) - f(t_i)) \right|$$

$$\leq \sum_{i=1}^n |w_i| |\hat{f}(t_i) - f(t_i)|$$

triangle inequality

$$\leq \left(\sum_{i=1}^n |w_i| \right) \|\hat{f} - f\|_{\sup}$$

For stability (i.e. output relatively insensitive to input)

we hope for $\left(\sum_{i=1}^n |w_i| \right)$ to not be large.

Note: if w_i 's are all positive then $\sum_{i=1}^n |w_i| = \sum_{i=1}^n w_i$

and we know from the Vandermonde matrix (row 1 eq.)

that $\sum_{i=1}^n w_i = (b-a)$

↗ This is as good as you can expect

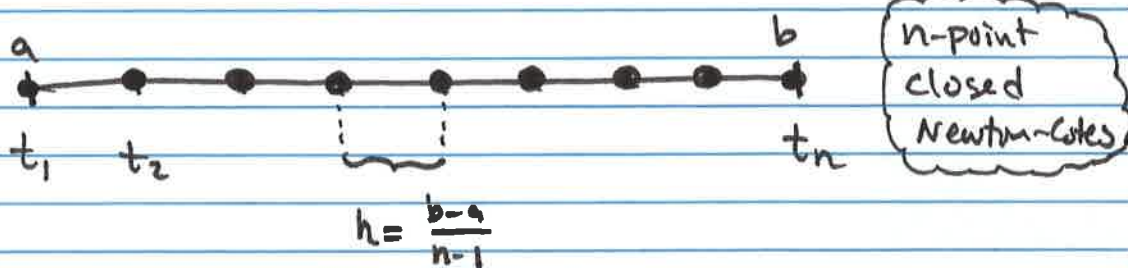
a) this is the condition# for integration (see Heath, p. 341)

If the w_i 's have different signs then the quantity $\sum_{i=1}^n |w_i|$ could be ~~be~~ considerably larger than $b-a$.

Types of Quadrature

• Newton-Cotes Quadrature:

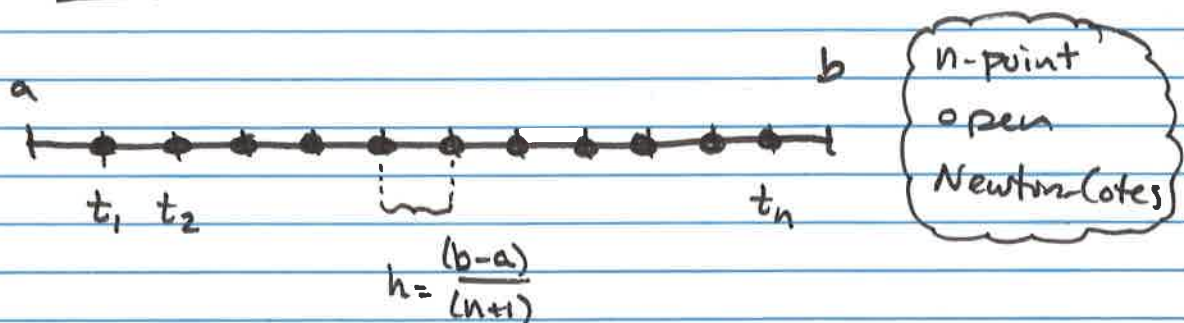
- here the nodes are equally-spaced on the interval $[a, b]$. These can be "closed"



e.g. Trapezoid - 2pt. closed N-C.

Simpson - 3pt. closed N-C

or "open"



e.g. Midpoint Rule - 1pt. open N-C

Further comments on Newton-Cotes

- polynomial "wiggle" becomes a problem as n increases
- in fact, according to Heath (p. 350)

$$\sum_{i=1}^n |w_i| \rightarrow \infty \text{ as } n \rightarrow \infty$$

indicating loss of stability.

- from an error point of view odd ~~is~~ # of points (n) seems better than even.

↳ clue: maybe a careful choice of nodes can make things better!

Some of these issues can be addressed using quadrature rules involving

- orthogonal polynomials
- Gaussian quadrature

Orthogonal Polynomials

Suppose p and g are polynomial functions. Define the inner product of p and g with respect to a weight function $w(t)$ (nonnegative)

$$\langle p, g \rangle = \int_a^b w(t) p(t) g(t) dt$$

with $\left| \int_a^b t^j w(t) dt \right| < \infty$ for $j = 0, 1, 2, \dots$ and $\int_a^b w(t) dt > 0$.

p and g are orthogonal if $\langle p, g \rangle = 0$

with respect
to the
weighting
function $w(t)$

A set of polynomials $\{p_j\}$ is orthonormal if

$$\langle p_i, p_j \rangle = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$

EX

Legendre Polynomials $w(t) = 1$ on $[-1, 1]$.

$$P_0(t) = 1$$

$$P_1(t) = t$$

$$P_2(t) = \frac{1}{2}(3t^2 - 1)$$

$$P_3(t) = \frac{1}{2}(5t^3 - 3t)$$

$\vdots$

Recurrence

$$(k+1)P_{k+1}(t) = (2k+1)tP_k(t) - kP_{k-1}(t)$$

inner product $\langle p, g \rangle = \int_a^b p(t)g(t)w(t) dt$

SS.1
OP.1

Orthogonal Polynomials

$(P_0(t), P_1(t), P_2(t), P_3(t), \dots)$

$[a, b]$

EX Legendre ... $[-1, 1]$

$w(t) = 1, P_k(1) = 1$

$P_0(t) = A = 1$

normalization
(note: not orthogonal by previous def.)

~~$P_0 = A$~~

$P_1 = At + B$ 1: $0 = \int_{-1}^1 (At+B) \cdot 1 \cdot 1 dt = \left. \frac{1}{2} At^2 + Bt \right|_{-1}^1 = \left(\frac{1}{2} A + B \right) - \left(-\frac{1}{2} A + B \right) = 2B$ $B=0$

$P_1(t) = At$

$P_1(1) = 1 \Rightarrow A = 1$

$P_1(t) = t$

$P_2 = At^2 + Bt + C$ 1: $0 = \int_{-1}^1 (At^2 + Bt + C) \cdot 1 dt = \left. \frac{1}{3} At^3 + \frac{1}{2} Bt^2 + Ct \right|_{-1}^1 = \left(\frac{1}{3} A + \frac{1}{2} B + C \right) - \left(-\frac{1}{3} A + \frac{1}{2} B - C \right)$
 $0 = \frac{2}{3} A + 2C$

t: $0 = \int_{-1}^1 (At^2 + Bt + C) \cdot t dt = \left. \frac{1}{4} At^4 + \frac{1}{3} Bt^3 + \frac{1}{2} Ct^2 \right|_{-1}^1 = \frac{2}{3} B$

~~$B = 0$~~

$B = 0$

$P_2(1) = 1 = A + C$

$A = -3C$

$C = -\frac{1}{2}$

$A = \frac{3}{2}$

$P_2(t) = \frac{3}{2} t^2 - \frac{1}{2} = \frac{1}{2} (3t^2 - 1) = P_2(t)$

So in general, we can generate ~~these~~ $P_k(t)$

* by requiring
$$0 = \int_{-1}^1 t^j P_k(t) dt \quad \text{for } j=0, 1, 2, \dots, k-1$$

and
$$P_k(1) = 1$$

P_k = degree k polynomial.
(i.e. P_k is orthogonal w.r.t weight function $w(t)=1$ to all lower degree polynomials)

... but there is also a recurrence relation

*

$$(k+1) P_{k+1}(t) = (2k+1)t P_k(t) - k P_{k-1}(t)$$

so for example since we already have

$$P_0(t) = 1$$

$$P_1(t) = t$$

$$P_2(t) = \frac{1}{2}(3t^2 - 1) \quad \left[\text{note: } 2P_2(t) = 3tP_1(t) - P_0(t) \right]$$

$$= 3t^2 - 1$$

$$3 \cdot P_3(t) = 5t P_2(t) - 2P_1(t)$$

$$= 5t \left(\frac{1}{2}(3t^2 - 1) \right) - 2t$$

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

$$P_3(t) = \frac{5}{2}t^3 - \frac{3}{2}t$$

more generally...

(with $p_m \neq 0$)

Def: of $\mathbb{P}_m$

really this should be interpreted as polynomials of degree less than or equal to m .

53.2

we say $p_m \in \mathbb{P}_m$

($\mathbb{P}_m$ = set of all real polynomials of degree $\leq m$)

is an m^{th} orthogonal polynomial

see Iserles p. 34/35

with respect to the weight function $w(t)$ if

$$\langle p_m, \hat{p} \rangle = 0$$

"monic" means leading coefficient = 1.

for every $\hat{p} \in \mathbb{P}_{m-1}$

(note $\mathbb{P}_m \neq \mathbb{P}_{m-1}$)

if $\mathbb{P}_m$ is $\mathbb{P}_{m-1}$ then $\mathbb{P}_m = \mathbb{P}_{m-1}$

That is, p_m is orthogonal to all lower degree polynomials.

(it follows easily...)

$$at^{n-1}$$

$$at^{n-1} + bt^{n-2}$$

$$at^{n-1} + bt^{n-2} + ct^{n-3} \dots$$

EX

Note: $\langle P_2, P_j \rangle = 0$ on $[-1, 1]$ w.r.t. $w(t) = 1$ $j = 0, 1$.

$$\int_{-1}^1 \frac{1}{2}(3t^2 - 1) \cdot t \, dt = \int_{-1}^1 \underbrace{\left(\frac{3}{2}t^3 - \frac{1}{2}t\right)}_{\text{odd function}} \, dt = 0 \quad \checkmark$$

as noted on previous pages

and

$$\int_{-1}^1 \frac{1}{2}(3t^2 - 1) \cdot 1 \, dt = \frac{1}{2} \left(t^3 - t \right) \Big|_{-1}^1 = \frac{1}{2}(0 - 0) = 0 \quad \checkmark$$

See comment on uniqueness of monic ~~poly~~ orthogonal polynomials

(Iserles, p. 35)

EX Chebyshev

$$w(t) = \frac{1}{(1-t^2)^{1/2}} \quad \text{on } [-1, 1]$$

$$T_0(t) = 1$$

$$T_1(t) = t$$

$$T_2(t) = 2t^2 - 1$$

$$T_3(t) = 4t^3 - 3t$$

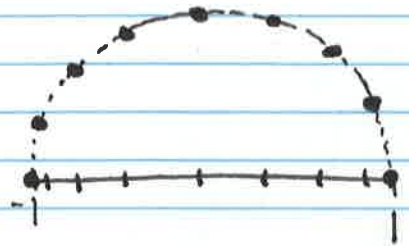
⋮

Recurrence

$$T_{k+1}(t) = 2t T_k(t) - T_{k-1}(t)$$

also... $T_k(t) = \cos(k \arccos(t))$

zeros at... $t_i = \cos\left(\frac{(2i-1)\pi}{2k}\right) \quad i=1, 2, \dots, k$



— the zeros tend to bunch up near the ends -1 and 1 .

How are these related to Gaussian Quadrature?

Gaussian Quadrature

Basic idea: Previously, we started with a given set of nodes t_j and then selected the weights w_j in order to maximize the number of polynomials (maximize the degree) that we could represent exactly. If we used an n -pt quadrature rule we were, in general, able to get a quadrature method of degree $n-1$ (maybe n in certain cases).

Here, for Gaussian Quadrature, we shall use the nodes as unknowns as well and ~~require that~~ use the requirements that additional polynomial degrees be exactly represented to determine the weights + nodes.

EX (Heath, p.351)

$$I(f) = \int_{-1}^1 f(t) dt \approx \underbrace{w_1 f(t_1) + w_2 f(t_2)}_{\text{2-pt quadrature rule.}}$$

treat w_1, w_2 AND t_1, t_2 as "to be determined"

(using $w(t) = 1$)

Regime

$$p(t) = 1$$

$$\int_{-1}^1 1 dt = w_1 f(t_1) + w_2 f(t_2)$$

$$2 = w_1 \cdot 1 + w_2 \cdot 1$$

$$p(t) = t$$

$$\int_{-1}^1 t dt = w_1 f(t_1) + w_2 f(t_2)$$

$$0 = w_1 t_1 + w_2 t_2$$

$$p(t) = t^2$$

$$\int_{-1}^1 t^2 dt = w_1 f(t_1) + w_2 f(t_2)$$

$$\frac{2}{3} = w_1 t_1^2 + w_2 t_2^2$$

$$p(t) = t^3$$

$$\int_{-1}^1 t^3 dt = w_1 f(t_1) + w_2 f(t_2)$$

$$0 = w_1 t_1^3 + w_2 t_2^3$$

Solve these
nonlinear
equations
for w_1, w_2
 t_1, t_2

$$\Rightarrow \begin{aligned} t_1 &= -\frac{1}{\sqrt{3}} & w_1 &= 1 \\ t_2 &= \frac{1}{\sqrt{3}} & w_2 &= 1 \end{aligned}$$

$n=2$ pts.

i.e. $2 \cdot n - 1$

In general, this method is degree 3 (order 4)

since $1, t, t^2, t^3$ are ~~exactly~~ integrated exactly.

Before connecting orthogonal polynomials to Gaussian Quadrature, let's go back and generalize slightly our definition of quadrature ... introducing a weight function $w(t)$

$$(*) \quad \int_a^b w(t) f(t) dt \approx \sum_{i=1}^n w_i f(t_i)$$

where

$$0 < \int_a^b w(t) dt < \infty, \quad \left| \int_a^b t^j w(t) dt \right| < \infty \quad j=1,2,\dots$$

Here's the main Gaussian Quadrature Theorem

[but see also
Lemma 3.2 (p.36)
for background]

Thm 3.3

^{nth}
(orthogonal polynomial)

Let $t_1, t_2, \dots, t_n$ be the zeros of $\forall p_n \in \mathbb{P}_n$

(set of all polynomials of degree n), and let $w_1, w_2, \dots, w_n$

be the solution to the Vandermonde system

$$\sum_{j=1}^n w_j t_j^m = \int_a^b t^m w(t) dt \quad m=0,1,2,\dots,n-1$$

$w(t) = \text{given}$

(requiring the quadrature rule be exact for polynomials $1, t, t^2, \dots, t^{n-1}$).

Then

(i) The quadrature method (*) is of order $2n$ ("degree" $2n-1$)

(ii) No other quadrature (n-pt.) can exceed this order.

Proof: of (i)

Idea: show that the coefficients (weights) w_j and the nodes t_j are such that ~~the~~ the quadrature rule

$$Q_n(f) = \sum_{j=1}^n w_j f(t_j)$$

is exact for polynomials $\hat{p} \in \mathbb{P}_{2n-1}$ (all polynomials of degree $2n-1$ or less)

$$\left(\text{i.e. } \int_a^b \hat{p}(t) w(t) dt = \sum_{j=1}^n w_j \hat{p}(t_j) \right) \quad \text{(not just } \mathbb{P}_{n-1} \text{)}$$

Let $\hat{p} \in \mathbb{P}_{2n-1}$. Note that $\hat{p}$ can be

expressed in terms of

$$\hat{p} = p_n g + r \quad \begin{array}{l} \swarrow \text{"quotient"} \\ \nwarrow \text{"remainder"} \end{array} \quad \left[\frac{\hat{p}}{p_n} = g + \frac{r}{p_n} \right]$$

where p_n is an ^{nth} orthogonal polynomial of degree n

and g, r are polynomials of degree $n-1$ or less. Consider

$\deg(g) \leq n-1$
 $\deg(r) \leq n-1$

$$\begin{aligned} \int_a^b \hat{p}(t) w(t) dt &= \int_a^b p_n(t) g(t) w(t) dt + \int_a^b r(t) w(t) dt \\ &= \cancel{\langle p_n, g \rangle} + \int_a^b r(t) w(t) dt \end{aligned}$$

since p_n is an n^{th} orthogonal polynomial and $g \in \mathbb{P}_{n-1}$

Also

$$\sum_{j=1}^n w_j \hat{p}(t_j) = \sum_{j=1}^n w_j \cancel{p_n(t_j)} g(t_j) + \sum_{j=1}^n w_j r(t_j) = \sum_{j=1}^n w_j r(t_j)$$

$p_n(t_j) = 0$ by definition of t_j .

We have already seen (Lemma 3.1), given nodes $t_1, t_2, \dots, t_n$ it is possible to choose weights w_j ($j=1, \dots, n$) such that if $r \in \mathbb{P}_{n-1}$

$$\int_a^b w(t) r(t) dt = \sum_{j=1}^n w_j r(t_j)$$

So, choose the weights w_j based on ~~these~~ these conditions.

Then

$$\int_a^b w(t) \hat{p}(t) dt = \int_a^b w(t) r(t) dt = \sum_{j=1}^n w_j r(t_j) = \sum_{j=1}^n w_j \hat{p}(t_j)$$

That is, our choice of t_j and w_j allow our quadrature scheme to integrate exactly any polynomial of degree $2n-1$. (i.e. it works for $\hat{p} \in \mathbb{P}_{2n-1}$). Therefore, the order of ~~the~~ quadrature method (*) is $2n$ (degree $2n-1$).

$\Rightarrow$ So with a carefully selected set of nodes we basically double degree of the quadrature rule.

How does this not apply if $\hat{p} \in \mathbb{P}_{2n-1}$?

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Proof of (ii)

Suppose the method integrates $\hat{p} \in \mathbb{P}_{2n}$ exactly.

In particular, consider the polynomial

$$\hat{p}(t) = \prod_{i=1}^n (t - t_i)^2 \geq 0.$$

(so $\hat{p}(t_i) = 0$ and $\hat{p}(t) \geq 0$). Note

$$\sum_{j=1}^n w_j \hat{p}(t_j) = 0$$

but, recalling that $w(t)$ is non-negative

$$\int_a^b w(t) \hat{p}(t) dt > 0$$

(note, this integral could be zero only if $w(t) = \begin{cases} 0 & t \neq t_i \\ \neq 0 & \text{elsewhere} \end{cases}$

but this function — indicator function — would have $\int_a^b w(t) dt = 0$.

Therefore, it is not possible that

$$\int_a^b w(t) \hat{p}(t) dt = \sum_{j=1}^n w_j \hat{p}(t_j)$$

so the method does not integrate $\hat{p} \in \mathbb{P}_{2n}$ exactly.

Therefore, the quadrature rule is order $2n$ (degree $2n-1$).

answer.

if $\hat{p} = \prod_{i=1}^{n-1} (t - t_i)^2 \geq 0$
for example
 $\sum_{j=1}^n w_j \hat{p}(t_j) \neq 0$
since $\hat{p}(t_n) \neq 0$
so the argument does not apply

A generalization of this Thm is...

Thm 3.4 (Iserles, p.37)

Let $r \in \mathbb{P}_n$ obey the orthogonality conditions

$$\langle r, \hat{p} \rangle = 0 \quad \text{for every } \hat{p} \in \mathbb{P}_{m-1}$$

and

$$\langle r, t^m \rangle \neq 0 \quad \text{for some } m \in \{0, 1, 2, \dots, n\}$$

Let $t_1, t_2, \dots, t_n$ be the zeros of the polynomial r .

Choose $w_1, w_2, \dots, w_n$ as in

$$\sum_{j=1}^n w_j t_j^m = \int_a^b t^m w(t) dt \quad m = 0, 1, 2, \dots, n-1$$

Then the quadrature formula

$$\int_a^b w(t) f(t) dt \approx \sum_{j=1}^n w_j f(t_j)$$

has order $p = n+m$ (degree $n+m-1$).

Comments: • $m \in \{0, 1, 2, \dots, n\}$ so $m \leq n$.

and the order $p = n+m \leq 2n$

- r does not satisfy "full" orthogonality conditions
... only a partial set of conditions
- this gives a quadrature scheme with an order reduced from the ~~the~~ original Gaussian quad. rule,