

Overview

- Method of moments approach to quadrature formulas
- Adaptive quadrature: recursive quadrature
 - Idea: repeatedly sample intervals of high quadrature error
 - Implementation: recursion (L20.jl)

- $f: \mathbb{R} \to \mathbb{R}$ known through data set (f sample) $\mathcal{D} = \{(x_i, f_i), i = 0, 1, ..., n\}$
- Recall polynomial interpolant technique $f(t) \cong p(t) = \sum_{i=0}^{n} f_i \ell_i(t)$

$$\int_a^b f(t) dt \cong \sum_{i=0}^n \left(\int_a^b \ell_i(t) dt \right) f_i = \sum_{i=0}^n w_i f_i$$

- Alternative approach: impose exact quadrature for members of a basis set
- Set a simple, predefined integration domain, e.g., [0,1], $x_i = ih$, h = 1/n
- Monomial basis set: $\mathcal{M} = \{1, t, t^2, ...\}$ conditions

$$f(t) = 1: \int_0^1 1 \, dt = 1 = \sum_{i=0}^n w_i$$

$$f(t) = t: \int_0^1 t \, dt = \frac{1}{2} = h \sum_{i=0}^n w_i i$$

$$f(t) = t^2: \int_0^1 t^2 \, dt = \frac{1}{3} = h \sum_{i=0}^n w_i i^2$$
....

• Solve above system to find weights w_i

- $f:[a,b] \to \mathbb{R}$ might exhibit regions of slow/rapid variation
- ullet Equidistant sampling over integration domain [a,b] is inefficient
- Adaptive quadrature: place sampling points preferentially to minimize error
- Recursive quadrature:
 - Consider $I_{ab} = \int_a^b f(t) dt$, and let $Q_{ab}(f)$ be some quadrature rule
 - Compare $Q_{\rm ab}(f)$ to $Q_{\rm ac}(f)+Q_{\rm cb}(f)$ with $a\leqslant c\leqslant b$, e.g., c=(a+b)/2

$$e = \frac{|Q_{ac} + Q_{cb} - Q_{ab}|}{|Q_{ab}|}$$

- If error is acceptable, return value, otherwise further subdivide [a,c], [c,b]
- See L20.jl