

# Examples of Interp. Rules

Trapezoidal Rule (an example of the first special case):

$$T(f) \equiv \int_a^b P_1(x) dx,$$

where  $x_0 = a$  and  $x_1 = b$ . We then have,

$$P_1(x) = l_0(x)f_0 + l_1(x)f_1 = \frac{x - x_1}{x_0 - x_1} f_0 + \frac{x - x_0}{x_1 - x_0} f_1.$$

Therefore we have

$$\begin{aligned} T(f) &= \int_a^b \frac{x - b}{a - b} dx f(a) + \int_a^b \frac{x - a}{b - a} dx f(b), \\ &= \left( \frac{b - a}{2} \right) f(a) + \left( \frac{b - a}{2} \right) f(b) = \left( \frac{b - a}{2} \right) [f(a) + f(b)]. \end{aligned}$$



# Examples of Interp. Rules

We also have that  $\Pi_1(x) = (x - a)(x - b)$  is negative for  $x \in [a, b]$  and  $\int_a^b \Pi_1(x) dx = -\frac{(b-a)^3}{6}$ . We therefore have satisfied the conditions of the first special case and this implies,

$$T(f) = \left(\frac{b-a}{2}\right)[f(a) + f(b)], \quad E^T(f) = \frac{-f''(\eta)}{12}(b-a)^3.$$

**Simpsons Rule** (an example of the second special case):

$$S(f) \equiv \int_a^b P_2(x) dx,$$

with  $x_0 = a, x_1 = \frac{a+b}{2}, x_2 = b$ .



# Simpsons Rule

Exercise: Using

$$P_2(x) = l_0(x)f(a) + l_1(x)f\left(\frac{a+b}{2}\right) + l_2(x)f(b),$$

where

$$l_0(x) = \frac{(x - \frac{a+b}{2})(x - b)}{(a - \frac{a+b}{2})(a - b)}, \quad l_1(x) = \frac{(x - a)(x - b)}{(\frac{a+b}{2} - a)(\frac{a+b}{2} - b)},$$

$$l_2(x) = \frac{(x - a)(x - \frac{a+b}{2})}{(b - a)(b - \frac{a+b}{2})}.$$

Simplify and verify (after some tedious algebra) that,

$$\begin{aligned} S(f) &= \left[ \int_a^b l_0(x) dx \right] f(a) + \left[ \int_a^b l_1(x) dx \right] f\left(\frac{a+b}{2}\right) + \left[ \int_a^b l_2(x) dx \right] f(b), \\ &\vdots \\ &= \left( \frac{b-a}{6} \right) \left[ f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right]. \end{aligned}$$



# Simpsons Rule (cont)

Note that for  $x \in [a, b]$ ,  $\Pi_2(x)$  is antisymmetric about  $\frac{a+b}{2}$  and this implies  $\int_a^b \Pi_2(x) dx = 0$ . Furthermore by choosing  $x_3 = \frac{a+b}{2}$  we have

$$\Pi_3(x) = (x - a)\left(x - \frac{a + b}{2}\right)^2(x - b),$$

is of one sign and this implies,

$$E^S(f) = I(f) - S(f) = \frac{1}{4!} f^4(\eta) \int_a^b \Pi_3(x) dx.$$

But  $\int_a^b \Pi_3(x) dx = -\frac{4}{15} \left(\frac{b-a}{2}\right)^5$  so we have,

$$S(f) = \left(\frac{b-a}{6}\right) \left[ f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right]$$

$$E^S(f) = \frac{-f^4(\eta)}{90} \left(\frac{b-a}{2}\right)^5$$



# Gaussian Quadrature

Recall that the error for interp. rules satisfies,

$E(f) = \int_a^b f[x_0 x_1 \cdots x_n x] \Pi_n(x) dx$ , and if  $\int_a^b \Pi_n(x) dx = 0$  we have,

$$E(f) = \int_a^b f[x_0 x_1 \cdots x_{n+1} x] \Pi_{n+1}(x) dx,$$

for any  $x_{n+1}$ . Now if  $\int_a^b \Pi_{n+1}(x) dx = 0$  as well we can show similarly,

$$E(f) = \int_a^b f[x_0, x_1, \cdots x_{n+2}, x] \Pi_{n+2}(x) dx.$$

In general if we let  $q_0(x) \equiv 1$  and  $q_i(x) \equiv (x - x_{n+1}) \cdots (x - x_{n+i})$  for  $i = 1, 2, \cdots (m - 1)$ . We can then show that if  $\int_a^b \Pi_n(x) q_i(x) dx = 0$ , for  $i = 0, 1, \cdots (m - 1)$  then,

$$E(f) = \int_a^b f[x_0 x_1 \cdots x_{n+m} x] \Pi_{n+m}(x) dx.$$



# Gaussian Quadrature (cont)

The key idea of GQ is to choose the interpolation points,  $(x_0, x_1, \dots, x_n)$  such that  $\int_a^b \Pi_n(x)q(x)dx = 0$  for all polynomials,  $q(x)$ , of degree at most  $n$ . In particular for the choice  $q(x) = q_i(x)$  for  $i = 0, 1, \dots, n$  we have  $\int_a^b \Pi_n(x)q_i(x)dx = 0$  and,

$$E(f) = \int_a^b f[x_0x_1 \cdots x_{2n+1}x]\Pi_{2n+1}(x)dx.$$

To ensure that  $\Pi_{2n+1}(x)$  is of one sign for  $x \in [a, b]$  we can choose  $x_{n+i+1} = x_i$  for  $i = 0, 1, \dots, n$  and we then have  $\Pi_{2n+1}(x) = \Pi_n^2(x)$ ,

$$E(f) = f[x_0x_1 \cdots x_{2n+1}\xi] \int_a^b \Pi_n^2(x)dx = \frac{1}{(2n+2)!} f^{2n+2}(\eta) \int_a^b \Pi_n^2(x)dx.$$

Note that these rules will be exact for all polynomials of degree at most  $2n + 1$  since  $f^{2n+2}(\eta) \equiv 0$ .



# GQ – Orthogonal Polynomials

How do we choose the  $x_i$ 's to ensure that  $\int_a^b \Pi_n(x)q(x)dx = 0$  for all polynomials,  $q(x)$  of degree at most  $n$  ? This question leads to the study of orthogonal polynomials.

- Definition: The set of polynomials  $\{r_0(x), r_1(x), \dots, r_k(x)\}$  is orthogonal on  $[-1, 1]$  iff the following two conditions are satisfied:
  - $\int_{-1}^1 r_i(x)r_j(x)dx = 0$ , for  $i \neq j$ ,
  - The degree of  $r_i(x)$  is  $i$  for  $i = 0, 1, \dots, k$ .



# Properties

- Properties of orthogonal polynomials:
  - Any polynomial  $q_s(x)$  of degree  $s \leq k$  can be expressed as.

$$q_s(x) = \sum_{j=0}^s c_j r_j(x).$$

- $r_k(x)$  is orthogonal to all polynomials of degree less than  $k$ . That is,  $\int_{-1}^1 r_k(x)q_s(x)dx = 0$  for  $s < k$ . (This follows from the previous property.)
- $r_k(x)$  has  $k$  simple zeros all in the interval  $[-1, 1]$ .



# Proof (last property)

For  $r_k(x)$ , let  $\{\mu_1, \mu_2, \dots, \mu_m\}$  be the set of points in  $[-1, 1]$  where  $r_k(x)$  changes sign. It is clear that each  $\mu_j$  is a zero of  $r_k(x)$  and all simple zeros of  $r_k(x)$  in  $[-1, 1]$  must be in this set. We then have  $m \leq k$  as the maximum number of zeros of a polynomial of degree  $k$  is  $k$ . Assume  $m < k$ . We then have,

$$\hat{q}_m(x) \equiv \prod_{i=1}^m (x - \mu_i),$$

is a polynomial of degree  $m < k$  that changes sign at each  $\mu_i$  and,

$$\int_{-1}^1 \hat{q}_m(x) r_k(x) dx = 0.$$

But  $\hat{q}_m(x)$  and  $r_k(x)$  have the same sign for all  $x$  in  $[-1, 1]$  (they change sign at the same locations). This implies a contradiction (the integrand is of one sign but the integral is zero)– our assumption must be false. We must therefore have  $m = k$ .



# 3-Term Recurrence

The  $r_k(x)$  also satisfy,

$$r_{s+1}(x) = a_s(x - b_s)r_s(x) - c_sr_{s-1}(x),$$

for  $s = 1, 2, \dots, k$ , where the  $a_s$  are normalization constants,  $r_{-1}(x) = 0$ , and if  $t_s = \int_{-1}^1 r_s^2(x)dx$  then,

$$b_s = \frac{1}{t_s} \int_{-1}^1 xr_s^2(x)dx, \quad c_s = \frac{a_s t_s}{a_{s-1} t_{s-1}}.$$

For example, we obtain the classical Legendre polynomials if we normalise so  $r_s(-1) = 1$ . This leads to,

$$a_s = \frac{2s+1}{s+1}, \quad b_s = 0, \quad c_s = \frac{s}{s+1}.$$



# Orthogonal Polys on $[a, b]$

- To transform orthogonal polynomials defined on  $[-1, 1]$  to  $[a, b]$  consider the linear mapping from  $[-1, 1] \rightarrow [a, b]$  defined by  $x = \frac{b-a}{2}y + \frac{a+b}{2}$ . The inverse mapping is  $y = \frac{1}{b-a}[2x - b - a]$  and from calculus we know,

$$\int_a^b g(x)dx = \left(\frac{b-a}{2}\right) \int_{-1}^1 g\left(\frac{b-a}{2}y + \frac{a+b}{2}\right)dy.$$

This relationship, combined with the properties of Legendre polynomials give a prescription for the choice of the  $x_i$ 's for GQ:

For  $i = 0, 1, \dots, n$ , set  $y_i$  to the  $i^{th}$  zero of the Legendre Polynomial,  $r_{n+1}(y)$ . With this choice we note that  $\prod_{j=0}^n (y - y_j) = K r_{n+1}(y)$  for some  $K \neq 0$ .



# Choice of the $x_i$ 's for GQ

Then with the choice  $x_i = \frac{b-a}{2}y_i + \frac{b+a}{2}$  we have,

$$\begin{aligned}\Pi_n(x) &= \Pi_n\left(\frac{b-a}{2}y + \frac{a+b}{2}\right) = \prod_{j=0}^n \left(\frac{b-a}{2}y + \frac{a+b}{2} - x_j\right) \\ &= \prod_{j=0}^n \left(\frac{b-a}{2}y + \frac{a+b}{2} - \left(\frac{b-a}{2}y_j + \frac{a+b}{2}\right)\right) \\ &= \prod_{j=0}^n \left[\frac{b-a}{2}(y - y_j)\right] = \left(\frac{b-a}{2}\right)^{n+1} \prod_{j=0}^n (y - y_j) \\ &= \left(\frac{b-a}{2}\right)^{n+1} K r_{n+1}(y).\end{aligned}$$



# Choice of the $x_i$ 's (cont)

Therefore for any polynomial,  $q(x)$  of degree at most  $n$ ,

$$\begin{aligned} & \int_a^b \Pi_n(x)q(x)dx \\ &= \left(\frac{b-a}{2}\right) \int_{-1}^1 \Pi_n(x)q\left(\frac{b-a}{2}y + \frac{b+a}{2}\right)dy, \\ &= \left(\frac{b-a}{2}\right) \int_{-1}^1 \Pi_n(x)\hat{q}(y)dy, \\ &= \left(\frac{b-a}{2}\right)^{n+2} K \int_{-1}^1 r_{n+1}(y)\hat{q}(y)dy = 0. \end{aligned}$$

since  $\hat{q}(y)$  is a polynomial of degree at most  $n$ .

That is with the  $x_i$ 's chosen as the 'transformed zeros' of the Legendre polynomial,  $r_{n+1}(y)$ , we have the property we need.



# Composite Quadrature Rules

Approximating the integrand with a PP leads to the class of Composite Rules. Let  $a = x_0 < \cdots < x_M = b$  and  $S(x)$  be a PP approximation to  $f(x)$  (defined on this mesh). We can then use  $\int_a^b S(x)dx$  as the approximation to  $I(f) = \int_a^b f(x)dx$ . Recall that  $S(x) \equiv p_{i,n}(x)$  for  $x \in [x_{i-1}, x_i]$   $i = 1, \cdots, M$ . From calculus we have,

$$\int_a^b S(x)dx = \sum_{i=1}^M \int_{x_{i-1}}^{x_i} S(x)dx = \sum_{i=1}^M \int_{x_{i-1}}^{x_i} p_{i,n}(x)dx,$$

–A sum of basic interpolatory rules.

If we use equally spaced  $x_i$ 's and low degree interpolation we obtain familiar rules.



# Error Estimates for GQ

Let  $G_n(f) = \sum_{i=0}^n \omega_i f(x_i)$  denote the  $(n + 1)$  – point Gaussian quadrature rule.

- We have shown,

$$I(f) - G_n(f) = O(b - a)^{2n+3}, \text{ as } (b - a) \rightarrow 0.$$

- The rules  $G_{n+1}, G_{n+2}, \dots$ , are more accurate (as  $(b - a) \rightarrow 0$ ) so we could use,

$$\widehat{EST}_{G_n} \equiv G_{n+k}(f) - G_n(f) = E_{G_n} + O(b - a)^{2(n+k)+3}.$$

- The rules  $G_{n+k}$  and  $G_n$  have at most one common interpolation point so the computation of this error estimate more than doubles the cost ( $2n + k + 2$  integrand evaluations).



# Error Est for GQ (cont)

- An alternative (to forming an error estimate based on  $G_{n+k}$ ) is to use the integrand evaluations already available (for the computation of  $G_n(f)$ ) and introduce only the minimum number of extra evaluations required to obtain an effective error estimate.
- This approach leads to a class of quadrature rules called Kronrod quadrature rules,  $K_{n+k}(f)$ . The error estimate for  $G_n(f)$ , is then  $K_{n+k}(f) - G_n(f)$ , where  $K_{n+k}(f)$  is more accurate and less expensive to compute than is  $G_{n+k}(f)$ . Kronrod proposed a particularly effective class of such rules where  $k = n + 1$ ,

$$K_{2n+1}(f) \equiv \sum_{i=0}^n a_i f(x_i) + \sum_{j=0}^{n+1} b_j f(y_j),$$



$$K_{2n+1}(f) \equiv \sum_{i=0}^n a_i f(x_i) + \sum_{j=0}^{n+1} b_j f(y_j)$$

- The  $x_i$ 's are the interpolation points associated with  $G_n(f)$ , and the  $y_j$ 's are the extra interpolation points necessary to define an accurate approximation to  $I(f)$ . Kronrod derived these weights (the  $a_i$ 's and the  $b_j$ 's) and the extra interpolation points  $(y_0, y_1, \dots, y_n)$  so that the resulting rule is order  $3n + 3$ .
- The resulting error estimate is then,

$$EST_{G_n} \equiv K_{2n+1}(f) - G_n(f),$$

with an associated cost of  $2n + 3$  integrand evaluations and an order of accuracy of  $O((b - a)^{3n+4})$ .

- These Gauss-Kronrod pairs of rules can be the basis for composite quadrature rules and adaptive methods. These methods are widely used and implemented in numerical libraries.



# 2D Quadrature

- Consider the problem of approximating integrals in two dimensions,

$$I(f) = \int \int_D f(x, y) dx dy,$$

This problem is more complicated than the one dimensional case since  $D$  can take many forms.

- One can develop the analogs of Gaussian rules or interpolatory rules but the weights and nodes will depend on the region  $D$ . Such rules can be determined and tabulated for simple regions such as rectangles, triangles and circles.
- An arbitrary region must then be transformed onto one of these simple regions before the rule can be used. Such a transformation will generally be nonlinear and may introduce an approximation error as well.



# 2D Quadrature (cont)

- An alternative approach is to apply a ‘product rule’, where one reduces the  $2D$ -integral to a sequence of two  $1D$ -integrals:

$$\int_a^b \int_{\alpha(y)}^{\beta(y)} f(x, y) dx dy = \int_a^b g(y) dy,$$

where

$$g(y) \equiv \int_{\alpha(y)}^{\beta(y)} f(x, y) dx.$$

- Note that  $g(y)$  is a  $1D$ -integral with upper and lower bounds depending on  $y$ .
- In this case  $g(y)$  is approximated, for a fixed value of  $y$ , by a standard method (for example,  $\approx \sum_{j=0}^M \omega_j f(x_j, y)$ ), and  $\int_a^b g(y) dy$  is also approximated by a standard method.



# 2D Quadrature (cont)

That is,

$$\begin{aligned}\int_a^b \int_{\alpha(y)}^{\beta(y)} f(x, y) dx dy &= \int_a^b g(y) dy \approx \sum_{r=0}^{M'} \hat{\omega}_r g(y_r), \\ &\approx \sum_{r=0}^{M'} \hat{\omega}_r \left( \sum_{j=0}^M \omega_j f(x_j, y_r) \right), \\ &= \sum_{r=0}^{M'} \sum_{j=0}^M (\hat{\omega}_r \omega_j) f(x_j, y_r).\end{aligned}$$

Note that error estimates for product rules are not easy to develop since the function  $g(y) \approx \sum_{j=0}^M \omega_j f(x_j, y)$  will not be a 'smooth' function of  $y$  unless  $M$  and the  $x_j$ 's are fixed (which is unlikely since  $\alpha(y)$  and  $\beta(y)$  are not fixed). In particular this 'inner rule' cannot be adaptive.

