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ERROR BOUNDS FOR GAUSSIAN QUADRATURE AND WEIGHTED- L^1 POLYNOMIAL APPROXIMATION*

RONALD A. DEVORE† AND L. RIDGWAY SCOTT‡

Abstract. Error bounds for Gaussian quadrature are given in terms of the number of quadrature points and smoothness properties of the function whose integral is being approximated. An intermediate step involves a weighted- L^1 polynomial approximation problem which is treated in a more general context than that specifically required to bound the Gaussian quadrature error.

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1. Introduction. The purpose of this note is to establish error bounds for Gaussian quadrature that reflect the fact that the accuracy is not degraded by certain singularities in the function to be integrated if they occur at the boundaries of the interval of integration rather than in the interior. It is well known that polynomial approximation can achieve greater accuracy at the boundary than in the interior (cf. Timan [5, p. 262]). The basic result to be derived says that the error in N -point Gaussian quadrature approximating the integral of f on $[-1, 1]$ is bounded by

$$(1.1) \quad C_s N^{-s} \int_{-1}^1 |f^{(s)}(x)|(1-x^2)^{s/2} dx$$

for all integers $s \leq 2N$ such that the above integral makes sense. Here, C_s is a constant independent of N and f . The main point of course is that f may have certain singularities at $x = \pm 1$ such that the above integral is finite for some value of s , whereas it would not be finite if the same singularities occurred for $-1 < x < 1$. If f has no singularities in the interval $[-1, 1]$, then an error bound such as (1.1) is not of great interest, as one would expect an exponential rate of convergence in this case (cf. Davis and Rabinowitz [1, p. 239]).

Theorem 3 below gives a result slightly sharper than (1.1) which allows one to interpret (1.1) as being valid for nonintegral s via an interpolation argument. This interpolation argument can be used to determine, for example, the rate of convergence with respect to N for a function with a power law singularity at the boundary (see Remark 4 in § 4). Thus bounds of the type (1.1) appear to be the right way to predict the accuracy of Gaussian quadrature when applied to general classes of singular functions. However, our main motivation for proving (1.1) came from the so-called discrete ordinates method for the transport equation [3]. To bound the error in this method, one must consider the error in Gaussian quadrature applied to a one-parameter family of (singular) functions $f_y(x)$, where the strength of the singularity of f at the boundary (in x) varies with y . Moreover, the singularities are not simple power law singularities, so the estimate (1.1) appeared essential to handle this problem.

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Bounds of the type (1.1) can also be given for composite rules using N points on a mesh that is quadratically graded near ± 1 , provided the basic rule has order of accuracy s (see Remark 5 in § 4). The techniques of proof for the case of composite rules are more elementary than those in the single-interval case. We include this simply for comparison and completeness.

In the following, we shall use interchangeably the words "bound" and "estimate," i.e. we shall refer occasionally to (1.1) as an "error estimate" even though it is not an asymptotic estimate of the actual error, rather simply an "estimate from above."

Our technique of proof is in two parts. Firstly, we analyze a Peano kernel for the quadrature error. This has the effect of establishing the bound (1.1) for $s=1$ and reducing further estimates to a weighted- L^1 approximation problem for polynomials. This part of the analysis is presented in § 2. Secondly, we consider weighted- L^1 approximation by polynomials and prove estimates for the error. This is done in § 3.

Finally, the estimates of §§ 2 and 3 are combined in § 4 to give error bounds for Gaussian quadrature of the type (1.1) for all $s \geq 1$. The calculations in § 2 are done specifically for Gaussian quadrature, although they clearly extend to more general quadrature approximations based on orthogonal polynomials for weighted integrals.

The weighted- L^1 approximation estimates show that, for a rather general class of weights $w > 0$ and sufficiently smooth f , there is a polynomial P of degree at most N so that

$$(1.2) \quad \int_{-1}^1 |f(x) - P(x)| w(x) dx \leq C_s N^{-s} \int_{-1}^1 |f^{(s)}(x)| (1-x)^{s/2} w(x) dx$$

for any s and N such that $N+1 \geq s \geq 1$. In the case $w \equiv 1$ and $s=1$, (1.2) has been established by N. X. Ky [2].

2. Error estimates for Gaussian quadrature. Consider Gaussian quadrature approximation of the form

$$(2.1) \quad \int_{-1}^1 f(x) dx \sim \sum_{j=1}^N \omega_j f(x_j) =: I_N(f)$$

($\{x_j\}$ are the zeros of the Legendre polynomials and $\{\omega_j\}$ are the integrals of the associated Lagrange interpolation polynomials, cf. G. Szegő [4]). The ordering $-1 < x_1 < \dots < x_N < 1$ will be assumed, and we introduce $x_0 := -1$ and $x_{N+1} := 1$. We wish to establish estimates of the error

$$(2.2) \quad e_N(f) := \int_{-1}^1 f(x) dx - I_N(f)$$

in terms of N and properties of f .

Assume $N \geq 2$ (our final estimate in Theorem 1 will hold for $N=1$ as well, as may be easily checked). The Peano kernel theorem allows us to write

$$e_N(f) = \int_{-1}^1 K(t) f'(t) dt,$$

at least for smooth f , where $K(t) = e_N(H_t)$, $|t| \leq 1$, (cf. (3.3) below) and H_t is the Heaviside function

$$(2.3) \quad H_t(x) := \begin{cases} 0, & x < t, \\ 1, & x \geq t. \end{cases}$$

It follows that

$$K(t) = 1 - t - \sum_{x_j > t} \omega_j = \sum_{x_j < t} \omega_j - t - 1.$$

The Chebyshev–Markov–Stieltjes inequality (cf. G. Szegő [4, p. 50]) implies that

$$1 + x_j \leq \sum_{i=1}^j \omega_i \leq 1 + x_{j+1}, \quad j = 1, \dots, N.$$

Therefore, for $j = 1, \dots, N$,

$$x_{j-1} - x_j \leq K(x_j -) \leq 0 \leq K(x_j +) \leq x_{j+1} - x_j.$$

Since K vanishes in each interval $[x_{j-1}, x_j]$ and its slope is one almost everywhere,

$$\max \{|K(x)| : x \in [x_{j-1}, x_j]\} \leq x_j - x_{j-1},$$

for $j = 1, \dots, N+1$. To bound $x_j - x_{j-1}$, recall that (cf. G. Szegő [4, p. 122])

$$(2.4) \quad x_j = -\cos \theta_j \quad \text{where} \quad \frac{(2N+1)\theta_j}{\pi} \in [2j-1, 2j],$$

for $j = 1, \dots, N$. Thus

$$\begin{aligned} x_j - x_{j-1} &= \cos \theta_{j-1} - \cos \theta_j = \int_{\theta_{j-1}}^{\theta_j} \sin \theta \, d\theta \\ &\leq (\theta_j - \theta_{j-1}) \max \{\sin \theta : \theta \in [\theta_{j-1}, \theta_j]\} \\ &\leq \left(\frac{3\pi}{2N}\right) \max \{\sin \theta : \theta \in [\theta_{j-1}, \theta_j]\}. \end{aligned}$$

To bound the latter term, observe that, since $(\sin \theta)/\theta$ is decreasing on $[0, \pi]$, (2.4) implies

$$\sin \theta \leq \left(\frac{\theta}{\theta_{j-1}}\right) \sin \theta_{j-1} \leq \left(\frac{\theta_j}{\theta_{j-1}}\right) \sin \theta_{j-1} \leq 4 \sin \theta_{j-1}$$

for $\theta \in [\theta_{j-1}, \theta_j]$ and $j = 2, \dots, N$. From the symmetry properties of the θ_j 's (cf. [4]), it then follows that

$$\sin \theta \leq 4 \sin \theta_j, \quad \theta \in [\theta_{j-1}, \theta_j], \quad j = 2, \dots, N$$

as well. Thus we have

$$(2.5) \quad \max \{\sin \theta : \theta \in [\theta_{j-1}, \theta_j]\} \leq 4 \min \{\sin \theta : \theta \in [\theta_{j-1}, \theta_j]\}$$

for $j = 2, \dots, N$. Thus we find, for $j = 2, \dots, N$,

$$\begin{aligned} x_j - x_{j-1} &\leq 6\pi \min \{\sin \theta : \theta \in [\theta_{j-1}, \theta_j]\} / N \\ &= 6\pi \min \{\sqrt{1 - \cos^2 \theta} : \theta \in [\theta_{j-1}, \theta_j]\} / N \\ &= 6\pi \min \{\sqrt{1 - x^2} : x \in [x_{j-1}, x_j]\} / N. \end{aligned}$$

Therefore, we have

$$(2.6) \quad |K(x)| \leq \frac{6\pi\sqrt{1-x^2}}{N},$$

at least for x between x_1 and x_N . For $x \in [-1, x_1]$

$$\begin{aligned} |K(x)| &\leq 1+x = \frac{\sqrt{1-x^2}\sqrt{1+x}}{\sqrt{1-x}} \leq \sqrt{1-x^2}\sqrt{1+x_1} \\ &= \sqrt{1-x^2}\sqrt{1-\cos\theta_1} \leq \sqrt{1-x^2} \frac{\theta_1}{\sqrt{2}} \leq \frac{\sqrt{1-x^2}\pi}{\sqrt{2}N}. \end{aligned}$$

A corresponding estimate holds by symmetry for $x \in [x_N, 1]$, so (2.6) is valid for all $x \in [-1, 1]$. Thus we have proved that

$$|e_n(f)| \leq 6\pi N^{-1} \int_{-1}^1 |f'(x)|\sqrt{1-x^2} dx.$$

Since (2.4) implies that

$$\frac{1}{N} \leq \sin\left(\frac{\pi}{2N+1}\right) \leq \sqrt{1-x^2}$$

for $x \in [x_1, x_N]$ and $N \geq 2$, (2.6) implies further that

$$(2.7) \quad |K(x)| \leq 6\pi(1-x^2)$$

for $x \in [x_1, x_N]$. But for $x \in [-1, x_1] \cup [x_N, 1]$, $|K(x)| = 1 - |x|$. Thus (2.7) holds for all $x \in [-1, 1]$, giving the following estimate:

$$|e_N(f)| \leq 6\pi \int_{-1}^1 |f'(x)| \min\left\{\frac{\sqrt{1-x^2}}{N}, 1-x^2\right\} dx.$$

These estimates hold for all $f \in L^1([-1, 1])$ whose weak derivatives are integrable with respect to the weight $1-x^2$, as can be seen by approximating f via smooth functions (cf. the definition of the space Y_w^s given later in § 3). Summarizing the above, we have the following:

THEOREM 1. *Let $e_N(f)$ denote the error in N -point Gaussian quadrature applied to $f \in L^1([-1, 1])$ (see (2.1)–(2.2) for definitions). If the weak derivative, f' , of f is integrable with respect to the weight $1-x^2$, then*

$$|e_N(f)| \leq 6\pi \int_{-1}^1 |f'(x)| \min\left\{\frac{\sqrt{1-x^2}}{N}, 1-x^2\right\} dx.$$

If f' is integrable with respect to $\sqrt{1-x^2}$, then

$$|e_N(f)| \leq \frac{6\pi}{N} \int_{-1}^1 |f'(x)| \sqrt{1-x^2} dx.$$

Estimates involving higher derivatives of f can also be derived by estimating Peano kernels. For example, one may write, for any $1 \leq k \leq 2N$,

$$e_N(f) = \int_{-1}^1 K_k(x) f^{(k)}(x) dx$$

($K_1 = K$ in the previous notation). K_k is a $C^{(k-2)}$ piecewise (k th degree) polynomial (with knots $\{x_i\}$) satisfying $0 = K_k(\pm 1) = \cdots = K_k^{(k-1)}(\pm 1)$ and $K_k^{(k)} = (-1)^k$ between knots. Moreover, $K_k^{(k-1)}(x_j+) - K_k^{(k-1)}(x_j-) = (-1)^{k-1}\omega_j$, $j = 1, \dots, N$. Using these facts together with a special oscillation property of K_2 , it can be shown that

$$|K_2(x)| \leq C \min \left\{ \frac{1-x^2}{N^2}, (1-x^2)^2 \right\}.$$

However, it becomes increasingly difficult to estimate the higher-degree kernels K_k , $k \geq 3$. Thus the following approach proves more fruitful.

Observe that $e_N(f) = e_N(f - P)$ for any $P \in \mathcal{P}_{2N-1}$, where \mathcal{P}_r denotes the set of polynomials of degree not exceeding r . Therefore, Theorem 1 implies that

$$(2.8) \quad |e_N(f)| \leq \inf_{P \in \mathcal{P}_{2N-2}} \int_{-1}^1 |f' - P|(x) w(x) dx,$$

where w is the weight function

$$w(x) = 6\pi \min \left\{ \frac{\sqrt{1-x^2}}{N}, 1-x^2 \right\}.$$

Thus, estimates of $e_N(f)$ are reduced to a weighted- L^1 approximation problem for polynomials. Such problems will be considered in the next section. The results of § 3 on approximation will be combined in § 4 with the results of this section to give higher order estimates of the form

$$(2.9) \quad |e_N(f)| \leq C_s \int_{-1}^1 |f^{(s)}(x)| \min \left\{ \left(\frac{\sqrt{1-x^2}}{N} \right)^s, (1-x^2)^s \right\} dx$$

for arbitrary positive integers s and N such that $N \geq s/2$.

3. Weighted- L^1 approximation. In this section, we shall prove a weighted- L^1 polynomial approximation result of the form

$$(3.1) \quad \inf_{P \in \mathcal{P}_N} \int_{-1}^1 |u - P|(x) w(x) dx \leq C_{s,w} N^{-s} \int_{-1}^1 |u^{(s)}(x)| w(x) (1-x^2)^{s/2} dx,$$

where \mathcal{P}_N denotes the set of polynomials of degree not exceeding N , s is a positive integer, $C_{s,w}$ is a constant independent of N and u , and w is a positive, integrable weight function to be discussed in more detail subsequently. Rather than state our abstract conditions on w and u initially, we shall develop them in the course of deriving (3.1). However, suffice it to say that (3.1) will be proved for a class of weights including the Jacobi weights

$$(3.2) \quad w(x) = (1+x)^\alpha (1-x)^\beta, \quad \alpha, \beta > -1.$$

As a first step, recall the Heaviside function $H_t(x)$ defined in (2.3). If u is sufficiently smooth, we may write

$$(3.3) \quad u(x) = P_1(x) + \frac{1}{(s-1)!} \int_{-1}^1 H_t(x) (x-t)^{s-1} u^{(s)}(t) dt$$

for all $x \in [-1, 1]$, where $P_1 \in \mathcal{P}_{s-1}$. Let $\lambda_t \in \mathcal{P}_{N-s+1}$ be an arbitrary family depending, say, piecewise continuously on t , and define

$$(3.4) \quad P(x) = P_1(x) + \frac{1}{(s-1)!} \int_{-1}^1 \lambda_t(x) (x-t)^{s-1} u^{(s)}(t) dt.$$

Then $P \in \mathcal{P}_N$, and Hölder's inequality and Fubini's theorem imply that

$$(3.5) \quad \int_{-1}^1 |u - P|(x) w(x) dx \leq \frac{1}{(s-1)!} \int_{-1}^1 \left\{ \int_{-1}^1 |\lambda_t - H_t|(x) |x-t|^{s-1} w(x) dx \right\} |u^{(s)}(t)| dt.$$

Thus to prove (3.1), it suffices to construct λ_t in such a way that

$$(3.6) \quad \int_{-1}^1 |\lambda_t - H_t|(x) |x-t|^{s-1} w(x) dx \leq C_s N^{-s} (1-t^2)^{s/2} w(t).$$

For t near ± 1 , this is relatively easy to do under the following assumption:

(A1) There is a constant A_1 such that

$$(i) \quad \int_t^1 w(x) dx \leq A_1(1-t)w(t) \quad \text{for } 0 \leq t \leq 1$$

and

$$(ii) \quad \int_{-1}^t w(x) dx \leq A_1(1+t)w(t) \quad \text{for } -1 \leq t \leq 0.$$

EXAMPLE 1. Assumption (A1) holds for the Jacobi weight $w(x) = (1+x)^\alpha(1-x)^\beta$ provided $\alpha, \beta > -1$.

Proof. To see this, it suffices to verify, say, (i). Then

$$\begin{aligned} \int_t^1 w(x) dx &= \int_t^1 (1+x)^\alpha(1-x)^\beta dx \leq 2^{\max\{\alpha, 0\}} \int_t^1 (1-x)^\beta dx \\ &\leq \frac{2^{\max\{\alpha, 0\}}(1-t)^{1+\beta}}{1+\beta} \leq \frac{2^{|\alpha|}}{1+\beta} (1+t)^\alpha(1-t)^{1+\beta} = A_1(1-t)w(t). \quad \square \end{aligned}$$

LEMMA. Suppose Assumption (A1) holds. Define λ_t by

$$\lambda_t(x) := \begin{cases} 0 & \text{for all } x \in [-1, 1] \quad \text{if } t > 0, \\ 1 & \text{for all } x \in [-1, 1] \quad \text{if } t \leq 0. \end{cases}$$

Then

$$\int_{-1}^1 |\lambda_t - H_t|(x) |x-t|^{s-1} w(x) dx \leq A_1(1-t^2)^s w(t)$$

for all $t \in [-1, 1]$.

Proof. Suppose $t > 0$. From (2.3) and Assumption (A1), we have

$$\begin{aligned} \int_{-1}^1 |\lambda_t - H_t|(x) |x-t|^{s-1} w(x) dx &= \int_t^1 |x-t|^{s-1} w(x) dx \leq (1-t)^{s-1} \int_t^1 w(x) dx \\ &\leq A_1(1-t)^s w(t) \leq A_1(1-t^2)^s w(t). \end{aligned}$$

The case $t \leq 0$ is similar. \square

COROLLARY. Let $\kappa \geq 1$ be arbitrary, and suppose (A1) holds. Then for any $N \geq 1$ and $t \in [-1, 1]$ such that $\sqrt{1-t^2} \leq \kappa/N$, the choices for λ_t given in the lemma satisfy the estimate (3.6) with $C_s = A_1 \kappa^s$.

In view of this corollary, it suffices to assume that $\sqrt{1-t^2} \geq \kappa/N$, where κ can be chosen later at our discretion. We shall subsequently construct, for certain values of t , $\lambda_t = \lambda_{t,r,N} \in \mathcal{P}_{N-2r+1}$ such that

$$(3.7) \quad |H_t - \lambda_t|(x) \leq C_r \left(\frac{\sqrt{1-t^2}}{N} \right)^{2r-1} |x-t|^{-2r+1} \quad \text{for } x \in [-1, 1],$$

where r is any positive integer (to be chosen later depending on w and s). Furthermore, λ_t will be monotone, nondecreasing, with $\lambda_t(-1) = 0$ and $\lambda_t(+1) = 1$. Hence also

$$(3.8) \quad |H_t - \lambda_t|(x) \leq 1 \quad \text{for all } x \in [-1, 1].$$

Assuming these properties of λ_t for some t for the moment, we proceed to prove (3.6) for such t . Let $\delta = \sqrt{1-t^2}/N$. Then

$$(3.9) \quad \begin{aligned} & \int_{-1}^1 |H_t - \lambda_t| |x-t|^{s-1} w(x) dx \\ & \leq \int_{\substack{|x-t| \leq \delta \\ |x| \leq 1}} |x-t|^{s-1} w(x) dx + C_r \int_{\substack{|x-t| \geq \delta \\ |x| \leq 1}} |x-t|^{s-2r} w(x) dx \delta^{2r-1}, \end{aligned}$$

where we used the bounds (3.8) and (3.7) on the first and second integrals, respectively. Thus the estimate (3.6) follows easily from the following two assumptions:

(A2) Let s be a positive integer. Then there exist constants $A_2 < \infty$ and $2 \leq \gamma < \infty$ such that, for all $\delta > 0$ and $t \in [-1, 1]$ satisfying $1-t^2 \geq \gamma\delta$,

$$\int_{|x| \leq \delta} w(x+t) |x|^{s-1} dx \leq A_2 w(t) \delta^s.$$

(A3) There exist $A_3 < \infty$, $2 \leq \gamma < \infty$ and $0 < k_0 < \infty$ such that, for all $\delta > 0$ and $t \in [-1, 1]$ satisfying $1-t^2 \geq \gamma\delta$ and all $k \geq k_0$,

$$\int_{\substack{|x| \geq \delta \\ |t+x| \leq 1}} w(x+t) |x|^{-k} dx \leq A_3 w(t) \delta^{-k+1}.$$

In applying (A2) and (A3) to (3.9), note that, for $\delta = \sqrt{1-t^2}/N$, the condition $1-t^2 \geq \gamma\delta$ is equivalent to $\sqrt{1-t^2} \geq \gamma/N$. This will therefore be satisfied if $\kappa \geq \gamma$, a requirement we now impose on κ . Choosing r such that $2r-s =: k \geq k_0$ thus proves (3.6) for t such that (3.7)–(3.8) hold. Before proving the bounds (3.7) and (3.8), we show that (A2) and (A3) hold for Jacobi weights.

EXAMPLE 2. *Assumption (A2) holds for the Jacobi weights $w(x) = (1+x)^\alpha(1-x)^\beta$ for all $\alpha, \beta \in \mathbb{R}$.*

Proof. Take $\gamma = 3$. By a change of variables,

$$\begin{aligned} \delta^{-s} \int_{|x| \leq \delta} \frac{w(x+t)}{w(t)} |x|^{s-1} dx &= \delta^{-s} \int_{|x| \leq \delta} \left(1 + \frac{x}{1+t}\right)^\alpha \left(1 - \frac{x}{1-t}\right)^\beta |x|^{s-1} dx \\ &= \int_{|y| \leq 1} \left(1 + \left(\frac{\delta}{1+t}\right)y\right)^\alpha \left(1 - \left(\frac{\delta}{1-t}\right)y\right)^\beta |y|^{s-1} dy. \end{aligned}$$

But for $1-t^2 \geq 3\delta$, $\delta/(1 \pm t) \in [0, \frac{2}{3}]$. Thus the integrand is bounded by $3^{\max\{|\alpha|, |\beta|\}}$. \square

EXAMPLE 3. Assumption (A3) holds for the Jacobi weights $w(x) = (1+x)^\alpha(1-x)^\beta$ for all $\alpha, \beta > -1$.

Proof. Take $\gamma = 2$ and $k_0 := 2 + \max\{\alpha, \beta, 0\}$. Then

$$\begin{aligned} & \delta^{k-1} \int_{\substack{|x| \geq \delta \\ |x+t| \leq 1}} \frac{w(x+t)}{w(t)} |x|^{-k} dx \\ &= \delta^{k-1} \left(\int_{\delta}^{1-t} \frac{w(x+t)}{w(t)} x^{-k} dx + \int_{\delta}^{1+t} \frac{w(t-x)}{w(t)} x^{-k} dx \right) \\ &= \delta^{k-1} \left(\int_{\delta}^{1-t} \left(1 + \frac{x}{1+t}\right)^{\alpha} \left(1 - \frac{x}{1-t}\right)^{\beta} x^{-k} dx \right. \\ & \quad \left. + \int_{\delta}^{1+t} \left(1 - \frac{x}{1+t}\right)^{\alpha} \left(1 + \frac{x}{1-t}\right)^{\beta} x^{-k} dx \right) \\ &=: I_{-}(t, \delta, \alpha, \beta, k) + I_{+}(t, \delta, \alpha, \beta, k). \end{aligned}$$

Since $I_{+}(t, \delta, \alpha, \beta, k) = I_{-}(-t, \delta, \beta, \alpha, k)$, it suffices to show that, for all $|t| \leq \sqrt{1-\gamma\delta}$,

$$I_{-}(t, \delta, \alpha, \beta, k) \leq C \quad \text{provided } k \geq 2 + \max\{\alpha, 0\}.$$

By a change of variables, note that

$$I_{-}(t, \delta, \alpha, \beta, k) = \int_1^{(1-t)/\delta} \left(1 + \left(\frac{\delta}{1+t}\right)y\right)^{\alpha} \left(1 - \left(\frac{\delta}{1-t}\right)y\right)^{\beta} y^{-k} dy.$$

Let $M := (1-t)/\delta$. Thus, for $\alpha \geq 0$,

$$\begin{aligned} I_{-} &\leq \int_1^M (1+y)^{\alpha} (1-M^{-1}y)^{\beta} y^{-k} dy \leq 2^{\alpha} \int_1^M (1-M^{-1}y)^{\beta} y^{\alpha-k} dy \\ &\leq 2^{\alpha} \int_1^M (1-M^{-1}y)^{\beta} y^{-2} dy. \end{aligned}$$

In the first estimate above, we used the fact that, for $|t| \leq 1$,

$$(1+t) \geq \frac{1}{2}(1-t)(1+t) \geq \frac{1}{2}\gamma\delta = \delta.$$

For $\alpha < 0$, one has

$$I_{-} \leq \int_1^M (1-M^{-1}y)^{\beta} y^{-k} dy \leq \int_1^M (1-M^{-1}y)^{\beta} y^{-2} dy.$$

A simple calculation shows that, for $M \geq 1$,

$$\int_1^M (1-M^{-1}y)^{\beta} y^{-2} dy \leq C_{\beta} \quad \text{for all } \beta > -1,$$

completing the proof. \square

We now construct λ_t satisfying (3.7)–(3.8). We shall do so by choosing

$$(3.10) \quad \lambda_t(x) := \int_{-1}^x \delta_t(y) dy,$$

where $\delta_t \in \mathcal{P}_{N-2r}$ approximates the Dirac δ -function. In particular, δ_t will be nonnegative and will have integral on $[-1, 1]$ equal to one. Thus λ_t satisfies the claimed monotonicity property as well as having the required values at ± 1 . Hence (3.8) will follow automatically.

To construct δ_n , first note that it suffices to take $\delta_t(x) := \frac{1}{2}$ for all $x \in [-1, 1]$ for $N < 4r$ (recall that we restrict t to satisfying $\sqrt{1-t^2} \geq \gamma/N$ in (3.7)–(3.8)). For $N \geq 4r$, let n be the greatest integer not exceeding $N/2r$. Note that $N/n \leq 3r$. Let T_n denote the Chebyshev polynomial of degree n , and let $\{t_i: 1 \leq i \leq n\}$ be its zeros:

$$T_n(x) := \cos(n(\cos^{-1} x)), \quad t_i := \cos\left(\frac{(i-1/2)\pi}{n}\right) =: \cos \theta_i \quad i = 1, \dots, n.$$

We shall establish (3.7)–(3.8) for $t \in \{t_i: 1 \leq i \leq n\}$. Define

$$(3.11) \quad \delta_{t_i}(x) := c_i \left[\frac{T_n(x)}{(x-t_i)} \right]^{2r},$$

where c_i is chosen so that $\int_{-1}^1 \delta_{t_i}(x) dx = 1$. We now estimate the size of c_i . Observe that $\delta_{t_{n+1-i}}(x) = \delta_{t_i}(-x)$, so that $c_{n+1-i} = c_i$. Thus we may suppose that $i \leq (n+1)/2$. Also, we have $n \geq 2$. Furthermore,

$$|T_n(x)| = |\cos n\theta| \geq \frac{1}{2} \quad \text{for } |n\theta - i\pi| \leq \frac{\pi}{3}.$$

Thus, writing $x = \cos \theta$, we see that for $|n\theta - i\pi| \leq \pi/3$.

$$\begin{aligned} \left| \frac{T_n(x)}{(x-t_i)} \right| &\geq \frac{1}{2} \left| \cos \theta - \cos \theta_i \right|^{-1} \geq \frac{1}{2} \left| \cos\left(\frac{(i+1/3)\pi}{n}\right) - \cos \theta_i \right|^{-1} \\ &\geq \left[\left(\frac{5\pi}{3n} \right) \max \left\{ \sin \theta: \left(\frac{n\theta}{\pi} \right) - i \in \left[-\frac{1}{2}, \frac{1}{3} \right] \right\} \right]^{-1} \geq \left[\left(\frac{5\pi}{n} \right) \sin \theta_i \right]^{-1}, \end{aligned}$$

because $\sin \theta/\theta$ is decreasing on $[0, \pi]$. The measure of the set $\{x = \cos \theta: |n\theta - i\pi| \leq \pi/3\}$ is

$$\cos\left(\frac{(i-1/3)\pi}{n}\right) - \cos\left(\frac{(i+1/3)\pi}{n}\right) \geq \left(\frac{2\pi}{3n}\right) \min \left\{ \sin\left(\frac{(i \pm 1/3)\pi}{n}\right) \right\} \geq \left(\frac{\pi}{3n}\right) \sin \theta_i,$$

since $n \geq 2$ and $\sin \theta_i \leq \min \{\sin((i \pm \frac{1}{3})\pi/n)\}$ unless $i = (n+1)/2$, in which case $\sin \theta_i \leq 2 \min \{\sin((i \pm \frac{1}{3})\pi/n)\}$. Therefore

$$c_i^{-1} = \int_{-1}^1 \left[\frac{T_n(x)}{(x-t_i)} \right]^{2r} dx \geq \left(\left(\frac{5\pi}{n} \right) \sin \theta_i \right)^{-2r} \left(\frac{\pi}{3n} \right) \sin \theta_i.$$

Thus

$$(3.12) \quad c_i \leq 15 \left(\frac{5\pi \sin \theta_i}{n} \right)^{2r-1} = 15 \left(\frac{5\pi \sqrt{1-t_i^2}}{n} \right)^{2r-1} \leq C_r \left(\frac{\sqrt{1-t_i^2}}{N} \right)^{2r-1}$$

To complete the estimate (3.7), observe that

$$|H_{t_i} - \lambda_{t_i}|(x) = c_i \begin{cases} \int_{-1}^x \delta_{t_i}(y) dy, & x \leq t_i, \\ \int_x^1 \delta_{t_i}(y) dy, & x > t_i. \end{cases}$$

Therefore,

$$|H_{t_i} - \lambda_{t_i}|(x) \leq c_i \int_{|x-t_i|}^2 y^{-2r} dy \leq \frac{c_i |x-t_i|^{-2r+1}}{(2r-1)}.$$

Combined with (3.12), this proves (3.7) for $t = t_i$. Hence (3.6) is now verified for $t = t_i$.

For the general case $t_{i+1} < t < t_i$, define $\lambda_t := \lambda_{t_i}$. Note that, since $\sin \theta / \theta$ is decreasing on $[0, \pi]$ (cf. 2.5),

$$\begin{aligned} \frac{\pi \sqrt{1-t^2}}{3n} &\leq \pi \min \{\sin \theta_i, \sin \theta_{i+1}\} / n \leq \int_{\theta_i}^{\theta_{i+1}} \sin \theta d\theta = t_i - t_{i+1} \\ &\leq 3\pi \min \{\sin \theta_i, \sin \theta_{i+1}\} / n \leq \frac{3\pi \sqrt{1-t^2}}{n}. \end{aligned}$$

Therefore,

$$\begin{aligned} \int_{-1}^1 |H_t - \lambda_t|(x) |x-t|^{s-1} w(x) dx &= \int_{-1}^1 |H_t - \lambda_{t_i}|(x) |x-t|^{s-1} w(x) dx \\ &\leq \int_{-1}^1 |H_t - H_{t_i}|(x) |x-t|^{s-1} w(x) dx \\ &\quad + \int_{-1}^1 |H_{t_i} - \lambda_{t_i}|(x) |x-t|^{s-1} w(x) dx. \end{aligned}$$

This first term, via (A2), is bounded by

$$\begin{aligned} \int_t^{t_i} |x-t|^{s-1} w(x) dx &\leq A_2(t_i-t)^s w(t) \leq A_2(t_i-t_{i+1})^s w(t) \\ &\leq A_2 \left(\frac{3\pi \sqrt{1-t^2}}{n} \right)^s w(t) \leq C_{r,s} \left(\frac{\sqrt{1-t^2}}{N} \right)^s w(t). \end{aligned}$$

(The application of Assumption (A2) is valid provided, e.g., $\min \{1-t_i^2, 1-t_{i+1}^2\} \geq \gamma(t_i-t_{i+1})$. The reader may easily check that this holds if the constant κ , mentioned in the corollary and the subsequent discussion, is chosen sufficiently large, depending on r and γ . This observation applies as well to the application of Assumptions (A2) and (A3) in the next set of inequalities.) Since (3.7)–(3.8) hold for t_i , the second term is bounded by ($\delta := \sqrt{1-t^2}/N$)

$$\begin{aligned} C_r \delta^{2r-1} \int_{|x-t| \geq 2|t_i-t_{i+1}|} |x-t_i|^{-2r+1} |x-t|^{s-1} w(x) dx &+ \int_{|x-t| \geq 2|t_i-t_{i+1}|} |x-t|^{s-1} w(x) dx \\ &\leq C_r (2\delta)^{2r-1} \int_{|x-t| \geq 2|t_i-t_{i+1}|} |x-t|^{s-2r} w(x) dx + 2^s A_2 |t_i-t_{i+1}|^s w(t) \\ &\leq C_r (2\delta)^{2r-1} A_3 (2|t_i-t_{i+1}|)^{s-2r+1} w(t) + 2^s A_2 |t_i-t_{i+1}|^s w(t) \\ &\leq \tilde{C}_{r,s} \left(\frac{\sqrt{1-t^2}}{N} \right)^s w(t), \end{aligned}$$

assuming (A2) and (A3) hold. Therefore, (3.6) is now proved for all $t \in [-1, 1]$, and hence the estimate (3.1) is established for smooth u and for weights w satisfying Assumptions (A1)–(A3). In fact, in view of the lemma, one can improve (3.6), and hence (3.1), by replacing the expression $(\sqrt{1-t^2}/N)^s$ by $\min \{(\sqrt{1-t^2}/N)^s, (1-t^2)^s\}$.

To extend the result to more general u , define, for positive integers s ,

$$(3.13) \quad \|u\|_{w,s} = \int_{-1}^1 [|u(x)| + |u^{(s)}(x)|(1-x^2)^s] w(x) dx$$

where $u^{(s)}$ is interpreted as a weak derivative. Define

$$(3.14) \quad Y_w^s := \{u \in L_{\text{loc}}^1(-1, 1) : \|u\|_{w,s} < \infty\}.$$

Assuming (A2)–(A3), we have $w(t) > 0$ for $|t| < 1$ (provided $w \neq 0$). Then Y_w^s is a Banach space having $C^\infty([-1, 1])$ as a dense subspace¹. Using this density, we arrive at the following theorem, which summarizes our results.

THEOREM 2. *Let w be a positive, integrable function on $[-1, 1]$ satisfying Assumptions (A1)–(A3), where s in Assumption (A2) is some positive integer. Let $u \in Y_w^s$ (see (3.13)–(3.14) for the definition). Then for any positive integer $N \geq s-1$,*

$$(3.15) \quad \inf_{P \in \mathcal{P}_N} \int_{-1}^1 |u - P|(x) w(x) dx \leq C_{s,w} \int_{-1}^1 |u^{(s)}(x)| \min \left\{ \left(\frac{\sqrt{1-x^2}}{N} \right)^s, (1-x^2)^s \right\} w(x) dx,$$

where \mathcal{P}_N denotes polynomials of degree not exceeding N and $C_{s,w}$ is a constant independent of N and u .

Remark 1. If $u^{(s)}(x)(1-x^2)^{s/2}$ is integrable on $[-1, 1]$, the above estimate may be simplified to yield

$$\inf_{P \in \mathcal{P}_N} \int_{-1}^1 |u - P|(x) w(x) dx \leq C_{s,w} N^{-s} \int_{-1}^1 |u^{(s)}(x)|(1-x^2)^{s/2} w(x) dx.$$

Remark 2. Assumptions (A2)–(A3) imply, in particular, that $w(t) > 0$ for $|t| < 1$, unless $w \equiv 0$. But such a condition is necessary for an estimate such as (3.15) to hold. To see this, suppose that w is continuous at t and $w(t) = 0$. Choose a sequence $\{u_j\}$ of smooth functions with $u_j^{(s)}$ positive, supported in $\{x+t: |x| \leq 1/j\}$ and $\int_{-1}^1 u_j^{(s)}(x) dx = 1$ (take $j \geq 1/(1-|t|)$ for simplicity). Then u_j converges in L^1 to $u(x) := H_t(x)(x-t)^{s-1}/(s-1)!$ as j tends to infinity. The right-hand side of (3.15) applied to u_j tends to zero. However, the left-hand side certainly will not do so since $u \notin \mathcal{P}_N$ for any N .

4. Higher-order error estimates for Gaussian quadrature. In this section, we combine the results of previous sections to prove an estimate of the form (2.9). From (2.8), we know that

$$|e_N(f)| \leq C \inf_{P \in \mathcal{P}_{2N-2}} \int_{-1}^1 |f' - P|(x) w(x) dx,$$

where $w(x) := \min \{\sqrt{1-x^2}/N, 1-x^2\}$. It is easy to check that, if a collection of weights w_i satisfies (A1), (A2) or (A3), then so does the weight function $\min \{w_i\}$. Thus

¹ *Proof.* Define $u_r(x) := u(rx)$, $0 < r < 1$. Then as $r \rightarrow 1$, $u_r \rightarrow u$ in Y_w^s . Let $u_{r,\varepsilon}$ be obtained from u_r by mollifying: $u_{r,\varepsilon} := u_r * \delta_\varepsilon$ where $\text{supp } \{\delta_\varepsilon\} = [-\varepsilon, \varepsilon]$ and δ_ε has integral one and is smooth. Then $u_{r,\varepsilon}$ is smooth on $[-1, 1]$ for $\varepsilon \leq 1-r$, and $u_{r,\varepsilon} \rightarrow u_r$ in Y_w^s . Now let $r \rightarrow 1$, $\varepsilon := 1-r \rightarrow 0$ to get $u_{r,\varepsilon} \rightarrow u$ in Y_w^s . \square

Theorem 2 applies to the weight w since it is a minimum of two Jacobi weights, yielding

$$\begin{aligned} |e_N(f)| &\leq C_s \int_{-1}^1 |f^{(s)}(x)| \min \left\{ \left(\frac{\sqrt{1-x^2}}{N} \right)^{s-1}, (1-x^2)^{s-1} \right\} w(x) dx \\ &= C_s \int_{-1}^1 |f^{(s)}(x)| \min \left\{ \left(\frac{\sqrt{1-x^2}}{N} \right)^s, (1-x^2)^s \right\} dx. \end{aligned}$$

We summarize this final result as

THEOREM 3. Let $e_N(f)$ denote the error in N -point Gaussian quadrature approximation to the integral of f on $[-1, 1]$ (see (2.1)–(2.2) for definitions). Suppose that $(1-x^2)^s f^{(s)}(x)$ (weak derivative) is integrable on $[-1, 1]$, i.e., $f \in Y_1^s$ where s is any integer such that $1 \leq s \leq 2N$. Then

$$(4.1) \quad |e_N(f)| \leq C_s \int_{-1}^1 |f^{(s)}(x)| \min \left\{ \left(\frac{\sqrt{1-x^2}}{N} \right)^s, (1-x^2)^s \right\} dx,$$

where C_s is independent of N and f .

Remark 3. If $f^{(s)}(x)(1-x^2)^{s/2}$ is integrable on $[-1, 1]$, the above estimate simplifies to

$$|e_N(f)| \leq C_s N^{-s} \int_{-1}^1 |f^{(s)}(x)| (1-x^2)^{s/2} dx.$$

This is the estimate anticipated in (1.1).

Remark 4. Suppose $f(x) = (1-x)^\sigma g(x)$, where $\sigma > -1$ and $g \in C^s([-1, 1])$, where s is the least integer greater than $2\sigma + 2$. Then applying Theorem 3 to f yields, for any $\varepsilon > 0$,

$$\begin{aligned} |e_N(f)| &\leq C \left\{ \int_{-1}^{1-\varepsilon} (1-x)^{\sigma-s} (1-x^2)^{s/2} N^{-s} dx + \int_{1-\varepsilon}^1 (1-x)^{\sigma-s} (1-x^2)^s ds \right\} \\ &\leq C \{ N^{-s} \varepsilon^{\sigma-s/2+1} + \varepsilon^{\sigma+1} \}. \end{aligned}$$

Taking $\varepsilon = N^{-2}$ yields the bound

$$|e_N(f)| \leq CN^{-2\sigma-2}$$

in agreement with the asymptotic estimate of Davis and Rabinowitz [1, (4.6.1.13)].

Remark 5. The weight $\sqrt{1-x^2}$ in the error bounds above arises primarily because the Gaussian quadrature points form a quadratically graded mesh near the boundary points $x = \pm 1$; cf. (2.4). Similar bounds can thus be obtained for composite rules on such a mesh. To be precise, let $y_i = -1 + (i/n)^2$ for $i = 0, 1, \dots, n$, and suppose that $\{w_j, \xi_j: j = 1, \dots, J\}$ is a fixed quadrature rule of order s on $[0, 1]$; i.e., suppose there is some function $K \in L^\infty([0, 1])$ such that for all sufficiently smooth f

$$\int_0^1 f(\xi) d\xi - \sum_{j=1}^J w_j f(\xi_j) = \int_0^1 f^{(s)}(\xi) K(\xi) d\xi,$$

with s some positive integer (we assume $\xi_j \in [0, 1]$ for all j). Define, for $i = 1, \dots, n$,

$$I_i(f) = \Delta y_i \sum_{j=1}^J w_j f(y_{i-1} + \xi_j \Delta y_i)$$

where $\Delta y_i = y_i - y_{i-1}$. For $i = n+1, \dots, 2n$, define $I_i(f)$ by reflection:

$$I_{n+i}(f) = I_{n+1-i}(g), \quad g(x) = f(-x) \quad \text{for } x \in [-1, 0].$$

Then by simple scaling we have

$$\left| \int_{y_{i-1}}^{y_i} f(x) dx - I_i(f) \right| \leq \|K\|_{L^\infty([0,1])} (\Delta y_i)^s \int_{y_{i-1}}^{y_i} |f^{(s)}(x)| dx.$$

Using the fact that, for $2 \leq i \leq n$,

$$\Delta y_i \leq \frac{3(i-1)}{n^2} \leq 3 \min \left\{ \frac{\sqrt{1-y_{i-1}^2}}{n}, 1-y_{i-1}^2 \right\}$$

together with the symmetry of the quadrature rules around $x=0$ thus yields

$$\left| \int_{y_1}^{-y_1} f(x) dx - \sum_{i=2}^{2n-1} I_i(f) \right| \leq 3\|K\|_{L^\infty} \int_{y_1}^{-y_1} |f^{(s)}(x)| \min \{n^{-s}(1-x^2)^{s/2}, (1-x^2)^s\} dx.$$

Provided that the end intervals can be estimated in a similar way, one has a result as in Theorem 3. Specifically, if $\xi_j \neq 0$ for all j (i.e., if $\xi=0$ is *not* a quadrature point), then

$$|K(\xi)| \leq C\xi^s, \quad \xi \in [0, 1],$$

and we easily see that

$$\left| \int_{-1}^{y_1} f(x) dx - I_1(f) \right| \leq C \int_{-1}^{y_1} |f^{(s)}(x)|(1+x)^s dx.$$

Using symmetry again and the definition of y_1 , we thus find

$$(4.2) \quad \left| \int_{-1}^1 f(x) dx - \sum_{i=1}^{2n} I_i(f) \right| \leq C \int_{-1}^1 |f^{(s)}(x)| \min \{n^{-s}(1-x^2)^{s/2}, (1-x^2)^s\} dx$$

for some constant C independent of f and n . Of course, if $\xi=0$ is a quadrature point, a result such as (4.2) cannot hold since the right-hand side is finite for, say, $f(x) = (1+x)^{-1/2}$.

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