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# Multiple-Precision Correctly Rounded Gauss-Legendre Quadrature 

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# Multiple-Precision Correctly Rounded Gauss-Legendre Quadrature 

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#### Abstract

Numerical integration is an operation that is frequently available in multiple precision numerical software packages. The different quadrature schemes used are considered well studied but the roundoff errors that result from the computation are often neglected, and the actual accuracy of the results are therefore seldom rigorously proven.

We focus on the Gauss-Legendre quadrature scheme and describe the algorithms needed in our implementation. A thorough error analysis is given as well as experimental error measurements and timings.


Key-words: numerical integration, correct rounding, Gauss-Legendre, multiple precision.

## Intégration de Gauss-Legendre en précision arbitraire avec arrondi correct

Résumé : L'intégration numérique est une opération fréquemment disponible dans les logiciels de calcul numérique en précision arbitraire. Bien que les différentes méthodes d'intégration aient été bien étudiées du point de vue mathématique, l'erreur qui résulte des arrondis lors de l'évaluation est souvent négligée, et la précision finale des calculs n'est pas prouvée.

Nous étudions la méthode d'intégration de Gauss-Legendre et décrivons les algorithmes utilisés dans notre implémentation. Une analyse rigoureuse de l'erreur ainsi que des résultats expérimentaux sont fournis.

Mots-clés : intégration numérique, arrondi correct, Gauss-Legendre, précision arbitraire.

## 1 Introduction

Numerical integration is readily available in most multiple precision numerical computation software (e.g. Pari/GP, MuPAD, Mathematica, Maple, ...). The working precision can usually be changed between computations and the software displays the computed result with desired precision (some software cheat and use a greater internal precision than what is displayed). At this point the user is at a loss to determine how many if any of the digits in the returned result are correct. Let us try a simple example and ask MuPAD 2.5.3, Pari/GP 2.1.6, and Maple 9 for the value of $I=\int_{10^{6}}^{10^{6}+\pi} \sin (\sin (x)) d x$ with 19 digits of precision:

```
>> DIGITS := 19: numeric: :quadrature (sin(sin(x)), \(\left.x=10^{\wedge} 6 . .10^{\wedge} 6+P I\right)\); 1.661291708545107576
```

```
? default(realprecision,19); intnum(x=10^6,10^6+Pi,sin(sin(x)))
```

? default(realprecision,19); intnum(x=10^6,10^6+Pi,sin(sin(x)))
%1 = 1.661291708545990308

```
%1 = 1.661291708545990308
```

> Digits:=19: evalf(Int(sin(sin(x)), $\left.\left.x=10^{\wedge} 6 . .10^{\wedge} 6+P i\right)\right)$;
1.661291708545107059

So we get three different results depending on the software used, which is more than enough. This disappointing observation is easy to explain. Although the semantics of floating-point numbers computations is well defined when those computations are restricted to the four basic operations (thanks to the IEEE754 standard [1]), nothing is guaranteed as soon as computations are composed (think for example of the double rounding problem), or transcendental functions like sin are used. Hence the difficulty of a correctly rounded numerical quadrature operation.

Several approaches were made to surpass these shortcomings when computing integrals. One can mention the use of adaptive quadrature functions (in MuPAD [2]), or dynamic error control (of simple or multiple integrals [3, 4]). Our work differs from these approaches in that we seek a correctly rounded result for which a proven bound on the error is needed. We need to provide more than an estimation and instead give a guaranteed error bound on the final result which takes into account all errors, including the roundoff errors which are the toughest to measure properly. We emphasize the fact that computing the correct rounding of the integral is only possible if its value is not exactly representable by a floating-point number, in which special case the problem is undecidable and we can only provide a result with an absolute error bound as small as desired.

We first describe briefly the Gauss-Legendre integration from a mathematical point of view. A more detailed study of the Gauss family of integration methods with corresponding weight functions can be found in [5]. In this paper, $f:[a, b] \rightarrow \mathbb{R}$ is the $C^{\infty}$ function we want to integrate on a finite domain $[a, b]$ and $n$ is the number of points of the GaussLegendre method. Let

$$
I=\int_{a}^{b} f(x) d x
$$

be the exact value of the integral. We define the inner product of $f$ and $g$ on $[a, b]$ for the admissible weight function $w$ as

$$
<f, g>=\int_{a}^{b} w(x) f(x) g(x) d x
$$

This leads to the definition of a sequence of orthogonal polynomials $\left(p_{i}\right)_{i \geq 0}$ such that:

$$
\begin{gathered}
\forall i \in \mathbb{N}, \operatorname{deg}\left(p_{i}\right)=i \\
\forall(i, j) \in \mathbb{N}^{2},<p_{i}, p_{j}>=0 \quad \text { if } i=j
\end{gathered}
$$

and the leading coefficient $k_{i}$ of $p_{i}$ is positive. For fixed $n>0, p_{n}$ has $n$ distinct roots in $] a, b\left[\right.$ which we name $x_{0}<x_{1}<\ldots<x_{n-1}$. The Gauss quadrature method associated to the weight function $w$ on $[a, b]$ is the interpolatory method at evaluation points $\left(x_{i}\right)_{0 \leq i<n}$ such that

$$
\int_{a}^{b} w(x) p(x) d x=\sum_{i=0}^{n-1} w_{i} p\left(x_{i}\right)
$$

holds for every polynomial $p$ of degree at most $n-1$ (this is enough to define the weights $w_{i}$ although the method will be shown to integrate accurately polynomials of degree at most $2 n-1$ ).

The Gauss-Legendre quadrature method is the Gauss method for the weight function $w=1$. Additionally the Legendre polynomials are usually defined on $[-1,1]$ and normalized such that $P_{n}(1)=1$ and we will follow this custom here.

In section 2 we will describe the algorithms used to compute the Legendre polynomials, the evaluation points and the coefficients of the method. The mathematical error of the method will then be discussed. In section 3 the actual quadrature algorithm is explained and the error analyzed. We follow with experiments and some conclusive remarks.

## 2 Algorithms

In the rest of this paper $P_{n}$ is the Legendre polynomials of degree $n$ defined on $[-1,1]$ as usual. The quadrature method on $[a, b]$ is derived from the quadrature method on $[-1,1]$ from a shifting and scaling in the polynomial. If we name $\left(V_{n}\right)$ the updated family of polynomials on $[a, b]$ we have the simple formulas

$$
V_{n}(u)=P_{n}\left(\frac{2 u-(b+a)}{b-a}\right) \quad \text { and } \quad P_{n}(x)=V_{n}\left(\frac{a+b+x(b-a)}{2}\right) .
$$

When translating the associated quantities (evaluation points and weights) from $[-1,1]$ to the target domain $[a, b]$ the details of the translation are omitted for the sake of simplicity. We will however take into account the error corresponding to this translation.

### 2.1 Legendre Polynomials

Like other orthogonal polynomial sequences, the $\left(P_{n}\right)_{n \geq 0}$ satisfy a recurrence relationship:

$$
\left\{\begin{align*}
P_{0}(X) & =1  \tag{1}\\
P_{1}(X) & =X \\
(n+1) P_{n+1}(X) & =(2 n+1) X P_{n}(X)-n P_{n-1}(X)
\end{align*}\right.
$$

From this we deduce that $P_{n}$ has only monomials of degree the same parity as $n$ and has rational coefficients. We assume that the common denominator is always a power of 2 . Thus $P_{n}$ can be written

$$
P_{n}(X)=\left\{\begin{array}{cc}
2^{-d_{n}} Q_{n}\left(X^{2}\right) & \text { if } n \text { is even } \\
2^{-d_{n}} X Q_{n}\left(X^{2}\right) & \text { otherwise. }
\end{array}\right.
$$

The problem of computing $P_{n}$ is reduced to that of computing $Q_{n}$ and $d_{n}$. The procedure is detailed in Algorithm 1. Details for the case of the constant coefficient have been omitted on line 8 .

```
Algorithm 1 Computation of the Legendre Polynomials
    \(Q_{0} \leftarrow 1\)
    \(Q_{1} \leftarrow 1\)
    \(d_{0} \leftarrow 0\)
    \(d_{1} \leftarrow 0\)
    \(p \leftarrow 0 \quad \triangleright\) holds the parity of the polynomial currently computed
    for \(i \leftarrow 2\) to \(n\) do
        \(\alpha \leftarrow d_{1-p}-d_{p}\)
        \(Q_{p} \leftarrow 2^{\alpha}(i-1) Q_{p}+(2 i-1) X^{p-1} Q_{1-p}\)
        \(r \leftarrow \max \left(\left\{n \in \mathbb{N}\left|2^{n}\right| i\right\}\right)\)
        \(s \leftarrow \max \left(\left\{n \in \mathbb{N}\left|2^{n}\right| Q_{p}\right\}\right)\)
        \(Q_{p} \leftarrow \frac{2^{r-s}}{i} Q_{p}\)
        \(d_{p} \leftarrow d_{1-p}+r-s\)
        \(p \leftarrow 1-p\)
    end for
```


### 2.2 Evaluation points

Computing the roots $\left(x_{i}\right)_{0 \leq i<n}$ of $P_{n}$ reduces to the computation of the roots of $Q_{n}$. Let $m=\left\lfloor\frac{n}{2}\right\rfloor$ and $u_{0}, u_{1}, \ldots u_{m-1}$ be the roots of $Q_{n}$, we have:

$$
\left\{x_{0}, x_{1}, \ldots, x_{n-1}\right\}=\left\{\begin{array}{cc}
\left\{ \pm \sqrt{u_{0}}, \ldots, \pm \sqrt{u_{m-1}}\right\} & \text { if } n \text { is even, } \\
\left\{ \pm \sqrt{u_{0}}, \ldots, \pm \sqrt{u_{m-1}},\right\} \cup\{0\} & \text { otherwise. }
\end{array}\right.
$$

where $0<u_{0}<u_{1}<\ldots<u_{m-1}$. The process of computing the roots of $Q_{n}$ has two steps:

1. root isolation, that is finding $m$ intervals that contain each exactly one root of $Q_{n}$,
2. root refinement.

The root isolation is made using Uspensky's algorithm as described in [6]. The input of the algorithm is $Q_{n}(x)$, and the output is a sequence of $m+1$ intervals of the form $\frac{c_{i}}{2^{l_{i}}}$ where $c_{i}$ and $l_{i}$ are integers and such that $\left[\frac{c_{i}}{2^{l_{i}}}, \frac{c_{i}+1}{2^{l_{i}}}\right]$ contains exactly one root of $Q_{n}$, namely $u_{i}$. At this step, $\log _{2}\left(c_{i}\right)$ bits of $u_{i}$ are known.

The refinement is done using dichotomy and Newton iteration.

### 2.2.1 Dichotomy

Performing a dichotomy to refine each root is a straightforward method but quite slow if done naively. Since $Q_{n}(X)=\sum_{i=0}^{m} a_{i} X^{i}$ has integer coefficients, it is possible to compute the sign of $Q_{n}\left(\frac{c}{2^{i}}\right)$ as $\operatorname{sign}\left(2^{l m} Q_{n}\left(\frac{c}{2^{i}}\right)\right)=\operatorname{sign}\left(\sum_{i=0}^{m} c^{i} 2^{l(m-i)}\right)$ but the cost turns out to be prohibitive as the integers involved are large and only one bit is gained at each step.

Instead we try to decrease the complexity of the computation by using intervals for the coefficients, where the bounds of the intervals are derived from the coefficients by truncating and rounding. Writing the expansion of $Q_{n}$ near the target root $u_{i}$ :

$$
Q_{n}(u)=\left(u-u_{i}\right) Q_{n}^{\prime}\left(u_{i}\right)+\mathcal{O}\left(\left(u-u_{i}\right)^{2}\right)
$$

so it is expected that the required precision used to compute $Q_{n}(u)$ and to successfully tell the sign despite the roundoff errors is approximately the number of current known good bits $-\log _{2}\left|u-u_{i}\right|$, up to a constant factor. If the interval computed for $Q_{n}(u)$ contains 0 , the coefficients of $Q_{n}$ are truncated to a higher precision until we can decide.

The algorithm stops when we have reached the desired accuracy $p$ in number of bits, that is $\log _{2}(c)>p$.

### 2.2.2 Newton iteration

Once the isolating interval for each root is sufficiently small, the faster Newton iteration is used. Since our error analysis requires a guaranteed error bound on the root, we use the interval Newton iteration described in [7].

### 2.3 Weights

The weights $\left(w_{i}\right)_{0 \leq i<n}$ verify

$$
\int_{-1}^{1} p(x) d x=\sum_{i=0}^{n-1} w_{i} p\left(x_{i}\right)
$$

for every polynomial of degree $\leq 2 n-1$ (see section 3.1).
For $i \in[0, n-1]$ we write $L_{i}(x)=\prod_{j \neq i}\left(x-x_{j}\right)$. Notice that $L_{i}(x)=\frac{P_{n}(x)}{\left(x-x_{i}\right) P_{n}^{\prime}\left(x_{i}\right)} . L_{i}^{\prime}$ has
degree $n-2$ so by definition $<L_{i}^{\prime}, P_{n}>=0$.

$$
0=\int_{-1}^{1} P_{n}(x) L_{i}^{\prime}(x) d x=\left[P_{n}(x) L_{i}(x)\right]_{-1}^{1}-\int_{-1}^{1} P_{n}^{\prime}(x) L_{i}(x) d x
$$

$P_{n}^{\prime} L_{i}$ has degree $2 n-1$ so it is integrated exactly by the method:

$$
0=\frac{P_{n}^{2}(1)}{\left(1-x_{i}\right) P_{n}^{\prime}\left(x_{i}\right)}-\frac{P_{n}^{2}(-1)}{\left(-1-x_{i}\right) P_{n}^{\prime}\left(x_{i}\right)}-\sum_{j=0}^{n-1} w_{j} P_{n}^{\prime}\left(x_{j}\right) L_{i}\left(x_{j}\right) .
$$

From Equation (1) we can see that $\left|P_{n}( \pm 1)\right|=1$. Moreover $L_{i}\left(x_{j}\right)=\delta_{i, j}$ (Kronecker delta) so

$$
\begin{gather*}
0=\frac{2}{\left(1-x_{i}^{2}\right) P_{n}^{\prime}\left(x_{i}\right)}-w_{i} P_{n}^{\prime}\left(x_{i}\right) \\
w_{i}=\frac{2}{\left(1-x_{i}^{2}\right) P_{n}^{\prime 2}\left(x_{i}\right)} \tag{2}
\end{gather*}
$$

In Section 2.2 we showed how to compute $x_{i}$ with desired accuracy. Since $P_{n}^{\prime}$ is known exactly we can evaluate $P_{n}^{\prime}\left(x_{i}\right)$ with a dynamic error bound (known as running error in [8]), and from that compute $w_{i}$ with arbitrary accuracy.

## 3 Error bounds

### 3.1 Mathematical error

In this section we prove the bound on the mathematical error made with the Gauss-Legendre quadrature method. A generic proof for any weight function $w$ can be found in [5]. We adapt and recall it here for completeness. Some proofs are omitted and can be found in [9].

Theorem 1. The Gauss-Legendre method on $[a, b]$ with $n$ points is exact for polynomials of degree $\leq 2 n-1$.

PROOF: by definition, the Gauss-Legendre quadrature scheme being of interpolatory type, it is exact of polynomials of degree $\leq n-1$. Let $f$ be a polynomial of degree $\leq 2 n-1$. We write

$$
f=q \cdot P_{n}+r, \quad \text { with } \operatorname{deg}(q) \leq n-1, \operatorname{deg}(r) \leq n-1
$$

Since $P_{n}$ is orthogonal to the set $\mathcal{P}_{n-1}$ of polynomials of degree $\leq n-1$ we have

$$
\int_{-1}^{1} q(x) P_{n}(x) d x=0
$$

and

$$
\int_{-1}^{1} r(x) d x=I(r)
$$

is computed exactly by the method.
Let $E[f]=\int_{-1}^{1} f(x) d x-\sum_{i=0}^{n-1} w_{i} f\left(x_{i}\right)$ be the error of the method for the function $f$. Let $h$ be the polynomial of degree $\leq 2 n-1$ such that

$$
\forall i \in[0, n-1], f\left(x_{i}\right)=h\left(x_{i}\right) \quad \text { and } \quad f^{\prime}\left(x_{i}\right)=h^{\prime}\left(x_{i}\right)
$$

Then the remainder theorem for polynomial interpolation states that

$$
f(x)=h(x)+\frac{f^{(2 n)}(\zeta(x))}{(2 n!)}\left(x-x_{0}\right)^{2}\left(x-x_{1}\right)^{2} \ldots\left(x-x_{n-1}\right)^{2}
$$

for $-1 \leq x \leq 1$ and $a<\zeta(x)<b$. From Theorem (1), $E[h]=0$ so we have

$$
\begin{aligned}
E[f] & =E\left[\frac{f^{(2 n)}(\zeta(x))}{(2 n!)}\left(x-x_{0}\right)^{2}\left(x-x_{1}\right)^{2} \ldots\left(x-x_{n-1}\right)^{2}\right] \\
& =-\int_{-1}^{1} \frac{f^{(2 n)}(\zeta(x))}{(2 n!)} \frac{P_{n}^{2}(x)}{k_{n}^{2}} d x
\end{aligned}
$$

With the mean value theorem there exists $\zeta \in]-1,1[$ such that

$$
E[f]=-\frac{f^{(2 n)}(\zeta)}{k_{n}^{2}(2 n!)} \int_{-1}^{1} P_{n}^{2}(x) d x
$$

By multiplying Equation (1) by $P_{n-1}(x)$ and integrating over $[-1,1]$ we get

$$
(2 n+1) \int_{-1}^{1} x P_{n-1}(x) P_{n}(x) d x=n \int_{-1}^{1} P_{n-1}^{2}(x) d x
$$

Let $v_{n}=\int_{-1}^{1} P_{n}^{2}(x) d x$. By Euclidean division we can write $x P_{n-1}(x)=\frac{k_{n-1}}{k_{n}} P_{n}(x)+R(x)$ where $\operatorname{deg}(R)<n$. So

$$
\begin{aligned}
\frac{(2 n+1) k_{n-1}}{k_{n}} \int_{-1}^{1} P_{n}^{2}(x) d x & =n \int_{-1}^{1} P_{n-1}^{2}(x) d x \\
\frac{(2 n+1) k_{n-1}}{k_{n}} v_{n} & =n v_{n-1} \\
v_{n} & =\left(\frac{n}{2 n+1}\right) \frac{k_{n}}{k_{n-1}} v_{n-1}
\end{aligned}
$$

Again from Equation (1) we get $k_{n}=\frac{2 n-1}{n} k_{n-1}$, so

$$
\begin{aligned}
& v_{n}=\frac{2 n-1}{2 n+1} v_{n-1} \\
& v_{n}=\frac{2}{2 n+1} \\
& k_{n}=\frac{(2 n-1)!}{(n-1)!n!2^{n-1}}
\end{aligned}
$$

Finally the error term is

$$
\begin{aligned}
E[f] & =f^{(2 n)}(\zeta) \frac{2[(n!)(n-1)!]^{2} \cdot 2^{2 n-2}}{((2 n-1)!)^{2}(2 n)!(2 n+1)} \\
& =\frac{2^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} f^{(2 n)}(\zeta)
\end{aligned}
$$

Taking into account the scaling from $[-1,1]$ to $[a, b]$ there is an additional $\left(\frac{2}{b-a}\right)^{n}$ factor in $k_{n}$ and $\frac{b-a}{2}$ in $v_{n}$.
Theorem 2. Let $M$ a bound of $\left|f^{(2 n)}\right|$ on $[a, b]$, then the error of the Gauss-Legendre integration of $f$ on $[a, b]$ with infinite precision is bounded by

$$
\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M
$$

### 3.2 Roundoff errors

```
Algorithm 2 Gauss-Legendre integration
    INPUT: \(\widehat{a}, \widehat{b-a},\left(\widehat{w_{i}}\right), f,\left(\widehat{v_{i}}\right), n \quad \triangleright\) where \(w_{i}\) are the weights and \(v_{i}\) is defined in \(\S 3.2\).
    Output: \(\widehat{I}\).
    for \(i \leftarrow 0\) to \(n-1\) do
        \(t \leftarrow \circ\left((\widehat{b-a}) \cdot \widehat{v_{i}}\right)\)
        \(\widehat{x_{i}} \leftarrow \circ(t+\widehat{a})\)
        \(\widehat{f}_{i} \leftarrow \circ\left(f\left(\widehat{x_{i}}\right)\right)\)
        \(\widehat{y_{i}} \leftarrow \circ\left(\widehat{f_{i}} \cdot \widehat{w_{i}}\right)\)
    end for
    \(\widehat{S} \leftarrow \operatorname{sum}\left(\widehat{y_{i}}, i=0 \ldots n-1\right) \quad \triangleright\) with Demmel and Hida algorithm [10]
    \(\widehat{D} \leftarrow \circ(\widehat{b-a}) / 2\)
    return \(\circ(\widehat{D} \widehat{S})=\widehat{I}\)
```

For the error analysis of Algorithm 2, we need a few useful lemmas concerning the "ulp ${ }^{1}$ calculus", as well as some definitions. The floating-point numbers are represented with radix 2 (this could be generalized for any radix but radix 2 is simpler and is natural on computers). For this section, $p$ is the working precision, and we assume all floatingpoint numbers are normalized, which means in our notations that the exponent range is unbounded.

Definition 1 (Exponent, Ulp). For a non-zero real number $x$ we define $E(x):=1+\left\lfloor\log _{2}|x|\right\rfloor$, such that $2^{E(x)-1} \leq|x|<2^{E(x)}$, and $\operatorname{ulp}(x):=2^{E(x)-p}$.

[^0]For a real $x \neq 0$ and a working precision $p$ we always have $2^{p-1} \operatorname{ulp}(x) \leq|x|<2^{p} u \operatorname{lp}(x)$. If $x$ is a floating-point number, then $\operatorname{ulp}(x)$ is the weight of the least significant bit - zero or not - in the $p$-bit mantissa of $x$. For all real $x, \operatorname{ulp}(x)$ is always greater than zero by definition.

Lemma 1. If $c \neq 0$ and $x \neq 0$ then $c \cdot \operatorname{ulp}(x)<2 \cdot \operatorname{ulp}(c x)$.
Lemma 2. Assuming no underflow (flush to zero) occurs then in all rounding modes for a non zero real $x$ we have: $\operatorname{ulp}(x) \leq \operatorname{ulp}(\circ(x))$, where $\circ(x)$ is the rounding of $x$ in the chosen mode with an unbounded exponent range.

Lemma 3. Let x a non-zero real and $\circ(x)$ its rounding to nearest on $p$ bits. Then $|x| \leq\left(1+2^{-p}\right) \mid \circ$ $(x) \mid$.

Lemma 4. Let $a$ and $b$ be two non-zero floating-point numbers of the same sign and precision $p$ then in all rounding modes

$$
\operatorname{ulp}(a)+\operatorname{ulp}(b) \leq \frac{3}{2} \operatorname{ulp}(\circ(a+b)) .
$$

Lemma 5. For $x$ and $y$ real numbers and using rounding to nearest in precision $p$ we have

$$
|\circ(\circ(x) \circ(y))-x y| \leq\left(\frac{5}{2}+2^{-p}\right) \operatorname{ulp}(\circ(\circ(x) \circ(y))) .
$$

Proof: let $u=\circ(x), v=\circ(y)$ and $z=\circ(u v)$. We can write

$$
\begin{aligned}
& x=u(1+\theta), y=v\left(1+\theta^{\prime}\right), u v=z\left(1+\theta^{\prime \prime}\right) \text { where }|\theta|,\left|\theta^{\prime}\right|,\left|\theta^{\prime \prime}\right| \leq 2^{-p} \\
&|z-x y| \leq \frac{1}{2} \operatorname{ulp}(z)+|u v-x y| \\
& \leq \frac{1}{2} \operatorname{ulp}(z)+|u v|\left(1-(1+\theta)\left(1+\theta^{\prime}\right)\right) \\
& \leq \frac{1}{2} \operatorname{ulp}(z)+|u v|\left(2^{1-p}+2^{-2 p}\right) \\
& \leq \frac{1}{2} \operatorname{ulp}(z)+|z|\left(2^{1-p}+2^{-2 p}\right)\left(1+2^{-p}\right) .
\end{aligned}
$$

With $|z| \leq\left(2^{p}-1\right) \operatorname{ulp}(z)$ we have $|z-x y| \leq \frac{1}{2} \operatorname{ulp}(z)+\left(2+2^{-p}-2^{1-p}-2^{-3 p}\right) \operatorname{ulp}(z)$.
For this section we denote by $\widehat{x}$ the value actually computed (i.e. with all roundoff errors) for a given "exact" value $x$, as would be computed with an infinite precision from the beginning of the algorithm. In order to provide an error bound on the numerical result given by the Gauss-Legendre method, we need to have a step-by-step look into Algorithm 2.

In addition to the parameters of Algorithm 2 we need an upper bound $M$ of $\left|f^{(2 n)}\right|$ on $[a, b] ; p$ is the working precision expressed in the number of bits of the mantissa, $a$ and $b$ given as oracles which returns their values as a rounded to nearest floating-point number of desired precision; $m$ an upper bound of $\left|f^{\prime}\right|$ on $[a, b]$. In the rest of this section we will prove our main theorem:

Theorem 3. Let $\delta_{\widehat{y_{i}}}=\left(\frac{5}{2}+2^{-p}\right) \operatorname{ulp}\left(\widehat{y_{i}}\right)+\left(\frac{17}{4}+23 \cdot 2^{-p-2}+3 \cdot 2^{-2 p-1}\right) m \widehat{w_{i}} u l p\left(\widehat{x}_{i}\right)$, where $\widehat{y_{i}}, \widehat{w_{i}}$ and $\widehat{x_{i}}$ are defined in Algorithm 2. When computing the numerical quadrature of $f$ using Algorithm 2 the total error on the result is bounded by:

$$
E_{\text {total }}=\left(\frac{9}{2}+3 \cdot 2^{-p}\right) \mathrm{ulp}(\widehat{I})+n\left(1+2^{-p}\right) \widehat{D} \cdot \max \left(\delta_{\widehat{y_{i}}}\right)+\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M .
$$

We split the error bound in three terms:

1. the mathematical error $\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M$,
2. the "static" error $E_{\text {stat }}=\left(\frac{9}{2}+3 \cdot 2^{-p}\right)$ ulp $(\widehat{I})$ which comes mostly from the rounding error in the summation,
3. the evaluation error $E_{\text {eval }}=n\left(1+2^{-p}\right) \widehat{D} \cdot \max \left(\delta_{\widehat{y}_{i}}\right)$ which accounts for the rest of the rounding errors and tracks possible cancellations in the summation as well.

Corollary 1. If we assume that $p \geq 2$ we have the simpler formulas:

$$
\left\{\begin{aligned}
\delta_{\widehat{y_{i}}} & =\frac{11}{4} \operatorname{ulp}\left(\widehat{y_{i}}\right)+6 m \widehat{w_{i}} \mathrm{ulp}\left(\widehat{x}_{i}\right) \\
E_{\text {total }} & =\frac{21}{4} \mathrm{ulp}(\widehat{I})+\frac{5 n}{4} \widehat{D} \cdot \max \left(\delta_{\widehat{y_{i}}}\right)+\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M .
\end{aligned}\right.
$$

The algorithm can be analyzed in several steps:

1. the computation of the weights $w_{i}, i \in[0, n-1]$ of the method. They are computed as explained in Section 2 as the rounded to nearest of the exact value:

$$
\left|\widehat{w_{i}}-w_{i}\right| \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{w_{i}}\right) .
$$

2. the computation of $x_{i}$. Let $x_{i}^{\prime}$ be the corresponding evaluation point on $[-1,1]$, and $v_{i}=\frac{1+x_{i}^{\prime}}{2}$. We assume $v_{i}$ is computed as the rounded to nearest of the exact value:

$$
\begin{gathered}
\left|\widehat{v}_{i}-v_{i}\right| \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{v_{i}}\right), \\
\widehat{x}_{i}=\circ\left(\circ\left(\widehat{v_{i}} \cdot(\widehat{b-a})\right)+\widehat{a}\right) .
\end{gathered}
$$

Since $a$ and $b$ are given as oracles, we can assume that $b-a$ as well as $a$ were computed as rounded to nearest of the correct value. The error analysis gives:

$$
\begin{aligned}
\left|\circ\left(\widehat{v_{i}} \cdot \widehat{b-a}\right)-v_{i} \cdot(b-a)\right| & \leq\left[\frac{5}{2}+2^{-p}\right] \operatorname{ulp}\left(\circ\left(\widehat{v_{i}} \cdot \widehat{b-a}\right)\right) \quad[\text { Lemma 5] } \\
\left|\widehat{x_{i}}-x_{i}\right| & \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{x_{i}}\right)+\left[\frac{5}{2}+2^{-p}\right] \operatorname{ulp}\left(\circ\left(\widehat{v_{i}} \cdot \widehat{b-a}\right)\right)+\frac{1}{2} \operatorname{ulp}(\widehat{a}) \\
& \leq\left(\frac{17}{4}+3 \cdot 2^{-p-1}\right) \operatorname{ulp}\left(\widehat{x_{i}}\right) . \quad[\text { Lemma 4] }
\end{aligned}
$$

3. the computation of $f\left(x_{i}\right)$. We assume we have an implementation of $f$ with correct rounding, and we call $f$ which returns the rounding to nearest of the exact value with precision $p$. Such correctly rounded implementations of mathematical functions with arbitrary precision on the result can be found for example in MPFR [11] for non-trivial functions like exp, sin, arctan and numerous others.
With the already estimated error on $\widehat{x}_{i}$ we have:

$$
\left|f\left(\widehat{x}_{i}\right)-f\left(x_{i}\right)\right|=\left|f^{\prime}\left(\theta_{i}\right)\left(\widehat{x}_{i}-x_{i}\right)\right|, \quad \theta_{i} \in\left[\min \left(x_{i}, \widehat{x}_{i}\right), \max \left(x_{i}, \widehat{x}_{i}\right)\right]
$$

and with an upper bound on $f^{\prime}$ we can bound this error absolutely. Let $\widehat{f}_{i}=\circ\left(f\left(\widehat{x}_{i}\right)\right)$ be the floating-point number computed. At this step we now have:

$$
\begin{aligned}
\delta_{\widehat{f}_{i}}=\left|\widehat{f}_{i}-f\left(x_{i}\right)\right| & \leq\left|f^{\prime}\left(\theta_{i}\right)\left(\widehat{x}_{i}-x_{i}\right)\right|+\frac{1}{2} \operatorname{ulp}\left(\widehat{f}_{i}\right) \\
& \leq\left(\frac{17}{4}+3 \cdot 2^{-p-1}\right) m \cdot \operatorname{ulp}\left(\widehat{x}_{i}\right)+\frac{1}{2} \operatorname{ulp}\left(\widehat{f}_{i}\right)
\end{aligned}
$$

4. computation of the $y_{i}=f\left(x_{i}\right) \cdot w_{i}$. The accumulated error so far:

$$
\begin{aligned}
\left|\widehat{y}_{i}-y_{i}\right| & \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{y_{i}}\right)+\left|\widehat{f}_{i} \widehat{w_{i}}-f\left(x_{i}\right) w_{i}\right| \\
& \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{y_{i}}\right)+\widehat{f}_{i}\left|\widehat{w_{i}}-w_{i}\right|+w_{i}\left|\widehat{f}_{i}-f\left(x_{i}\right)\right| \\
& \leq \frac{1}{2} \operatorname{ulp}\left(\widehat{y_{i}}\right)+\frac{1}{2} \widehat{f}_{i} \operatorname{ulp}\left(\widehat{w_{i}}\right)+w_{i} \delta_{\widehat{f_{i}}} \\
& \leq \frac{3}{2} \operatorname{ulp}\left(\widehat{y_{i}}\right)+w_{i}\left[\left(\frac{17}{4}+3 \cdot 2^{-p-1}\right) m \cdot \operatorname{ulp}\left(\widehat{x}_{i}\right)+\frac{1}{2} \operatorname{ulp}\left(\widehat{f}_{i}\right)\right] \quad \text { [Lemmas 1 and 2] } \\
& \leq \frac{3}{2} \operatorname{ulp}\left(\widehat{y_{i}}\right)+\left(1+2^{-p} \left\lvert\, \widehat{w_{i}}\left[\left(\frac{17}{4}+3 \cdot 2^{-p-1}\right) m \cdot \operatorname{ulp}\left(\widehat{x}_{i}\right)+\frac{1}{2} \operatorname{ulp}\left(\widehat{f_{i}}\right)\right] \quad\right.\right. \text { [Lemma 3] } \\
& \leq\left(\frac{5}{2}+2^{-p}\right) \operatorname{ulp}\left(\widehat{y_{i}}\right)+\left(1+2^{-p}\right) m \widehat{w_{i}}\left(\frac{17}{4}+3 \cdot 2^{-p-1}\right) \operatorname{ulp}\left(\widehat{x}_{i}\right) \quad \text { [Lemmas 1 and 2] } \\
& \leq\left(\frac{5}{2}+2^{-p}\right) \operatorname{ulp}\left(\widehat{y}_{i}\right)+\left(\frac{17}{4}+23 \cdot 2^{-p-2}+3 \cdot 2^{-2 p-1}\right) m \widehat{w_{i}} u l p\left(\widehat{x}_{i}\right)=\delta_{\widehat{y_{i}}} .
\end{aligned}
$$

Remark: when bounding the error on $\widehat{x}_{i}, \widehat{f}_{i}$ as well as $\widehat{y}_{i}$, the term with ulp $\left(\widehat{x}_{i}\right)$ vanishes if the error on $\widehat{x}_{i}$ is zero. One can easily show with our assumption that no underflow occurs, and that if $\widehat{x}_{i}=0$ then the error on $\widehat{x}_{i}$ is zero (i.e. $x_{i}=0$ ) and the ill-defined quantity ulp $\left(\widehat{x}_{i}\right)$ vanishes. For the error bound we keep track of only $\max \left(\delta_{\widehat{y}_{i}}\right)$.
5. summation of the $y_{i}{ }^{\prime}$ 's: this is done with Demmel and Hida summation algorithm [10], which guarantees an error of at most 1.5 ulp on the final result. This algorithm uses a larger working precision $p^{\prime} \approx p+\log _{2}(n)$. Let $S=\sum_{i=0}^{n-1} y_{i}$.

$$
|\widehat{S}-S| \leq \frac{3}{2} \operatorname{ulp}(\widehat{S})+n \cdot \max \left(\delta_{\widehat{y_{i}}}\right)
$$

6. multiplication by $\frac{b-a}{2}: I=\frac{b-a}{2} S$. We note $D=\frac{b-a}{2}$ and assume as before that the input $\widehat{b-a}$ was computed as the rounded to nearest of its exact value. Since the division by 2 is exact we have:

$$
\begin{aligned}
|\widehat{D}-D| & \leq \frac{1}{2} \operatorname{ulp}(\widehat{D}) \\
|\widehat{I}-I| & \leq \frac{1}{2} \operatorname{ulp}(\widehat{I})+|\widehat{S} \widehat{D}-S D| \\
& \leq \frac{1}{2} \operatorname{ulp}(\widehat{I})+\frac{1}{2}|\widehat{S}| \operatorname{ulp}(\widehat{D})+D|\widehat{S}-S| \\
& \leq \frac{3}{2} \operatorname{ulp}(\widehat{I})+D\left[\frac{3}{2} \operatorname{ulp}(\widehat{S})+n \cdot \max \left(\delta_{\widehat{y}_{i}}\right)\right] \quad \text { [Lemmas 1 and 2] } \\
& \leq \frac{3}{2} \operatorname{ulp}(\widehat{I})+\left(1+2^{-p}\right) \widehat{D}\left[\frac{3}{2} \operatorname{ulp}(\widehat{S})+n \cdot \max \left(\delta_{\widehat{y}_{i}}\right)\right] \quad \text { [Lemma 3] } \\
& \leq\left(\frac{9}{2}+3 \cdot 2^{-p}\right) \operatorname{ulp}(\widehat{I})+n\left(1+2^{-p}\right) \widehat{D} \cdot \max \left(\delta_{\widehat{y}_{i}}\right) . \quad \text { [Lemmas 1 and 2] }
\end{aligned}
$$

Corollary 2. If we assume furthermore that $f$ does not change sign on $[a, b]$, then we have the following better bound:

$$
\begin{aligned}
E_{\text {total }}= & \left(\frac{9}{2}+3 \cdot 2^{-p}+\left(2+2^{1-p}\right)\left(5+2^{1-p}\right)\left(1+3 \cdot 2^{-p}\right)\right) \mathrm{ulp}(\widehat{I}) \\
& +n m\left(1+2^{-p}\right)\left(\frac{17}{4}+23 \cdot 2^{-p-2}+3 \cdot 2^{-2 p-1}\right) \widehat{D} \max \left(\widehat{w_{i}} u \operatorname{ulp}\left(\widehat{x_{i}}\right)\right) \\
& +\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M .
\end{aligned}
$$

PROOF: Let us assume for example that $f \geq 0$, then we have

$$
\forall i \in[0, n-1], \widehat{y_{i}}=\circ\left(\widehat{w_{i}} \cdot \widehat{f}_{i}\right) \geq 0
$$

so

$$
\operatorname{ulp}\left(\widehat{y_{i}}\right) \leq 2^{1-p} \widehat{y_{i}}
$$

Let $\tilde{S}=\sum_{i=0}^{n-1} \widehat{y}_{i}$, we know that

$$
\begin{aligned}
|\tilde{S}-\widehat{S}| & \leq \frac{3}{2} \operatorname{ulp}(\widehat{S}) \\
\tilde{S} & \leq\left(1+3 \cdot 2^{-p}\right) \widehat{S} \\
L & =\sum_{i=0}^{n-1}\left(\frac{5}{2}+2^{-p}\right) \operatorname{ulp}\left(\widehat{y_{i}}\right) \\
& \leq\left(\frac{5}{2}+2^{-p}\right) 2^{1-p} \sum_{i=0}^{n-1} \widehat{y}_{i} \\
& \leq 2^{1-p}\left(\frac{5}{2}+2^{-p}\right)\left(1+3 \cdot 2^{-p}\right) \widehat{S} \\
& \leq\left(5+2^{1-p}\right)\left(1+3 \cdot 2^{-p}\right) \operatorname{ulp}(\widehat{S})
\end{aligned}
$$

From this we get the following bound on the error on $\widehat{S}$ :
$|\widehat{S}-S| \leq\left(\frac{3}{2}+\left(5+2^{1-p}\right)\left(1+3 \cdot 2^{-p}\right)\right) \operatorname{ulp}(\widehat{S})+n m\left(\frac{17}{4}+23 \cdot 2^{-p-2}+3 \cdot 2^{-2 p-1}\right) \max \left(\widehat{w_{i}} \mathrm{ulp}\left(\widehat{x_{i}}\right)\right)$
and substituting this expression in the bound of $|\widehat{I}-I|$ above yields the announced result. Since $0 \leq \widehat{w_{i}} \leq 1$ and assuming $p \geq 2$ we can write the simpler bound

$$
E_{\text {total }}=30 \mathrm{ulp}(\widehat{I})+8 n m \widehat{D} \max (\operatorname{ulp}(a), \operatorname{ulp}(b))+\frac{(b-a)^{2 n+1}(n!)^{4}}{(2 n+1)[(2 n)!]^{3}} M .
$$

## 4 Experiments

Algorithm 2 was implemented using the MPFR library [11]. In addition to the result of the integration, the program gives an error bound $E_{\text {total }}$ on the computed result split in three terms, as explained in Theorem 3. For our experiments we chose a function and an


Figure 1: The overestimation of the error in bits when computing $\int_{0}^{3} e^{x} d x$ with 113 bits of precision and $n$ evaluation points.
integration domain where the exact value is known, so that we can measure precisely the actual error of the computation (denoted by $E_{\text {meas }}$ ). Figure 1 shows how much pessimistic the error bound is in number of bits, computed as $\log _{2}\left(E_{\text {total }}\right)-\log _{2}\left(E_{\text {meas }}\right)$ when computing the integral $I=\int_{0}^{3} e^{x} d x$ with 113 bits of working precision, the number of evaluation points varying from 2 to 100 .

The error due to roundoffs ( $E_{\text {stat }}+E_{\text {eval }}$ ) is quite stable. This is expected of the GaussLegendre integration scheme where the weights are all positive. With this property the


Figure 2: The roundoff errors $E_{\text {eval }}+E_{\text {stat }}$ and the mathematical error when computing $\int_{0}^{3} e^{x} d x$ with 113 bits of precision and $n$ evaluation points.
integration can benefit from an increase in the number of evaluation points $n$ in order to compute a better result. This is not the case for the Newton-Cotes method for example.

The bound on the total error as given by the algorithm is close to the measured error. In the experimental data we observe a maximum loss of 7 bits of precision on the computed result due to a pessimistic bound. However, this overestimation of the error is noticable for $n \geq 15$ which is the point where the mathematical error becomes less than the roundoff errors. Quite naturally our overestimation is made on the roundoff errors and not on the mathematical error.

## 5 Conclusion

The Gauss-Legendre quadrature scheme provides a robust numerical integration algorithm. In this paper we performed an error analysis of the algorithm that can be used to determine dynamically an error bound on the final result. From this we can easily derive a correctly rounded numerical quadrature algorithm, provided the result is not exactly representable by a floating-point number.

The running time of the algorithm is dominated by the cost of computing the evaluation points and weights (Figure 3). As future work it is planned to improve this time, for example by using precomputed tables of roots optimized for the space.


Figure 3: Roots and weights computation time for 113 bits of precision and $n$ evaluation points. Measurements made on a 3 GHz Pentium 4 processor.

| $p$ | $n_{\text {opt }}$ | Predicted good bits |
| :---: | :---: | :---: |
| 53 | 8 | 47 |
| 113 | 15 | 108 |
| 200 | 22 | 194 |
| 400 | 38 | 395 |
| 1000 | 80 | 995 |

Figure 4: Optimized order for different working precisions $p$ in bits.

Another idea to work on is to consider composition of the integration method and chose automatically the best composition level and $n$ for a given function, integration domain and the required precision with emphasis on the running time. Figure 4 gives the smallest $n$ for which the mathematical error is smaller than the computed bound on the roundoff error $E_{\text {eval }}+E_{\text {stat }}$ in the computation of $\int_{0}^{3} \exp (x) d x$, and gives a first hint in this direction.

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[^0]:    ${ }^{1}$ unit in the last place

