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# Improving Goldschmidt Division, Square Root and Square Root Reciprocal 

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#### Abstract

The aim of this paper is to accelerate division, square root and square root reciprocal computations, when Goldschmidt method is used on a pipelined multiplier. This is done by replacing the last iteration by the addition of a correcting term that can be looked up during the early iterations. We describe several variants of the Goldschmidt algorithm assuming 4-cycle pipelined multiplier and discuss obtained number of cycles and error achieved. Extensions to other than 4-cycle multipliers are given.


Key-words: Division, Square root, Square root reciprocal, Convergence division, Computer Arithmetic, Goldschmidt iteration. grant PICS-479, Vers des arithmétiques plus souples et plus précises.

[^0]
## Une amélioration de l'algorithme de Goldschmidt pour la division, la racine carrée et l'inverse de la racine carrée

Résumé : Le but de cet article est l'accélération de la division, et du calcul de racines carrées et d'inverses de racines carrées lorsque la méthode de Goldschmidt est utilisée sur un multiplieur pipe-line. Nous faisons ceci en remplaçant la dernière itération par l'addition d'un terme de correction qui peut être déduit d'une lecture de table effectuée lors des premières itérations. Nous décrivons plusieurs variantes de l'algorithme obtenu en supposant un multiplieur à 4 étages de pipe-line, et donnons pour chaque variante l'erreur obtenue et le nombre de cycles de calcul. Des extensions de ce travail à des multiplieurs dont le nombre d'étages est différent sont présentées.
Mots-clé : Division, Racine carrée, Inverse de la racine carrée, Arithmétique des ordinateurs, Algorithme de Goldschmidt.

## 1 Introduction

Although division is less frequent among the four basic arithmetic operations, a recent study by Oberman and Flynn [7] shows that in a typical numerical program, the time spent performing divisions is approximately the same as the time spent performing additions or multiplications.

This is due to the fact that in most current processors, division is significantly slower than the other operations. Hence, faster implementations of division are desirable.

There are two principal classes of division algorithms. The digit-recurrence methods [4] produce one quotient digit per cycle using residual recurrence which involves (i) redundant additions, (ii) multiplications with a single digit, and (iii) a quotient-digit selection function. The latency and complexity of implementation depends on the radix. The method produces both the quotient which can be easily rounded and the remainder. The iterative, quadratically convergent, methods, such as the Newton-Raphson, the Goldschmidt and series expansion methods (see for instance $[5,6,11]$ ) use multiplications and take advantage of fast multipliers implemented in modern processors. These methods, however, do not produce directly the remainder and rounding requires extra quotient digits. According to [7], roughly twice as many digits of intermediate result are needed as in the final result, unless the iterations are performed using a fused multiply-accumulate operator, that performs computations of the form $a x+b$ with one final rounding only [1].

In this paper, we focus on the latter class of methods. Such methods have been implemented in various microprocessors such as the IBM RS/6000 [12] or the more recent AMD K7 processor [14]. Our goal is to find a way of accelerating the Goldschmidt iteration (G-iteration in the sequel) when implementing it on a pipelined computer. We then extend our work to square root and square root reciprocal calculations.

## 2 Division

### 2.1 Background and G-iteration

Assume two n-bit inputs N and D , that satisfy $1 \leq \mathrm{N}, \mathrm{D}<2$ (i.e., normalized significands of floating-point numbers). We aim at computing $Q=N / D$. The Goldschmidt algorithm consists in finding a sequence $K_{1}, K_{2}, K_{3}, \ldots$ such that $r_{i}=$
$D K_{1} K_{2} \ldots K_{i}$ approaches 1 as $i$ goes to infinity. Hence,

$$
\mathrm{q}_{\mathrm{i}}=\mathrm{NK}_{1} \mathrm{~K}_{2} \ldots \mathrm{~K}_{\mathrm{i}} \rightarrow \mathrm{Q} .
$$

This is done as follows. The first factor $\mathrm{K}_{1}$ may be obtained by table lookup. After that, if $r_{i}=1-\alpha$, we choose $K_{i}=1+\alpha$, which gives $r_{i+1}=1-\alpha^{2}$. To be able to discuss possible alternatives to the basic algorithm, we give in detail the steps used in computing $\mathrm{q}_{4}$.

1. Step 1. Let $D=1 . d_{1} d_{2} \ldots d_{n-1}$, and define $\widehat{D}=1 . d_{1} d_{2} \ldots d_{p}$, where $p \ll n$. Typical values are $n=53$ and $p \approx 10$. Obtain $K_{1}=1 / \widehat{D}$ from a $2^{p} \times p$ table such that

$$
\begin{equation*}
1-2^{-p}<\mathrm{K}_{1} \mathrm{D}<1+2^{-p} \tag{1}
\end{equation*}
$$

Define $\epsilon=1-K_{1} D$. From (1), $|\epsilon|<2^{-p}$. Another solution is to add enough guard bits in the table [2] to get

$$
\begin{equation*}
1-2^{-\mathfrak{p}-0.5}<\mathrm{K}_{1} \mathrm{D}<1+2^{-\mathfrak{p}-0.5} \tag{2}
\end{equation*}
$$

In such a case, $|\epsilon|<2^{-p-0.5}$. We successively compute

- $r_{1}=D K_{1}=1-\epsilon$ (this multiplication will be called mult. 1);
- $\mathrm{q}_{1}=\mathrm{NK}_{1}$ (mult. 2).

2. Step 2. By 2's complementing $r_{1}$, we get $K_{2}=1+\epsilon$. We then compute

- $r_{2}=r_{1} K_{2}=1-\epsilon^{2}$ (mult. 3);
- $\mathrm{q}_{2}=\mathrm{q}_{1} \mathrm{~K}_{2}$ (mult. 4).

Note that

$$
\frac{\mathrm{N}}{\mathrm{D}}=\frac{\mathrm{q}_{2}}{1-\epsilon^{2}}
$$

3. Step 3. By 2 's complementing $r_{2}$, we get $K_{3}=1+\epsilon^{2}$. We then compute

- $r_{3}=r_{2} K_{3}=1-\epsilon^{4}$ (mult. 5);
- $\mathrm{q}_{3}=\mathrm{q}_{2} \mathrm{~K}_{3}$ (mult. 6)
such that

$$
\frac{\mathrm{N}}{\mathrm{D}}=\frac{\mathrm{q}_{3}}{1-\epsilon^{4}}
$$

4. Step 4. By 2's complementing $r_{3}$, we get $\mathrm{K}_{4}=1+\epsilon^{4}$. We then compute $\mathrm{q}_{4}=\mathrm{q}_{3} \mathrm{~K}_{4}$ (mult. 7) such that

$$
\begin{equation*}
\frac{\mathrm{N}}{\mathrm{D}}=\frac{\mathrm{q}_{4}}{1-\epsilon^{8}} \tag{3}
\end{equation*}
$$

For instance, if $p=12$, and if we do not have guard bits in the table that gives $K_{1}$, the previous process gives an approximation $q_{4}$ to $N / D$ that satisfies

$$
\mathrm{q}_{4}<\frac{\mathrm{N}}{\mathrm{D}} \leq \mathrm{q}_{4}\left(1+2^{-96}\right)
$$

If we have guard bits in the table that gives $\mathrm{K}_{1}$, so that (2) is satisfied, we have

$$
\mathrm{q}_{4}<\frac{\mathrm{N}}{\mathrm{D}} \leq \mathrm{q}_{4}\left(1+2^{-8 p-4}\right)
$$

For instance, if $p=9$, the error is bounded by $2^{-76}$. This method has a quadratic convergence: at each step, the number of significant bits of the approximation to the quotient roughly doubles.

Getting correctly rounded results, as required by the IEEE-754 standard [8], may seem less straightforward than with the digit-recurrence methods. And yet, many studies performed during the past recent years $[1,9,15]$ show that with some care this can be done easily, for division as well as for square root. See [1, 12] for more details.

### 2.2 Basic implementation on a pipelined multiplier

In this section, we assume that we use a 4 -cycle $n \times n$ pipelined multiplier. We start counting the cycles when $\mathrm{K}_{1}$ becomes available.

The error committed using this first method is easily obtained from (3): it is around $2^{-8 p}$ (e.g., for $p=12$, it produces around 96 bits of accuracy). This implementation requires 17 cycles. The scheduling of the multiplications in the pipelined multiplier is shown Figure 1. It is worth noticing that we can use the "holes" in the pipeline to interlace independent divisions. By doing that, performing two interlaced divisions requires only 19 cycles (see Figure 1). We can use this method with bipartite tables (see [3]). In such a case, $\mathrm{K}_{1} \approx 1 / \mathrm{D}$ is obtained by looking up two tables with $p$ address bits. One can show

$$
\begin{equation*}
1-2^{-\frac{3 \mathrm{p}}{2}+1}<\mathrm{K}_{1} \mathrm{D}<1+2^{-\frac{3 \mathrm{p}}{2}+1} \tag{4}
\end{equation*}
$$



Figure 1: Schedule of the original G-iteration. It requires 17 cycles to get the final result. It allows interlacing of two independent divisions: it suffices to start multiplication mult. 1 of the second division at cycle 3, mult. 2 at cycle 4, mult. 3 at cycle 7, mult. 4 at cycle 8, mult. 5 at cycle 11, mult. 6 at cycle 12, and mult. 7 at cycle 16. Two interlaced divisions require 19 cycles.

After four conventional iterations, one can get

$$
\mathrm{q}_{4}<\frac{\mathrm{N}}{\mathrm{D}} \leq \mathrm{q}_{4}\left(1+2^{-12 \mathrm{p}+8}\right)
$$

For instance, if $p=9$, this process gives an approximation with error less than $2^{-100}$.

### 2.3 Variant A

As soon as $\epsilon$ becomes available (i.e., in cycle 5), we look-up $\hat{\epsilon}^{4}$ in a table with $p$ address bits, where $\hat{\epsilon}$ is a $p$-bit number, constituted by the bits of $|\epsilon|$ of weight $2^{-p-1}, 2^{-p-2}, \ldots, 2^{-2 p}$. That is, if $\epsilon=0.000 \ldots 0 \epsilon_{\mathfrak{p}+1} \epsilon_{\mathfrak{p}+2} \epsilon_{\mathfrak{p}+3} \epsilon_{\mathfrak{p}+4} \ldots$, then $\widehat{\epsilon}=$ $0.000 \ldots 0 \epsilon_{p+1} \epsilon_{p+2} \ldots \epsilon_{2 \mathfrak{p}}$. Then, instead of successively computing $q_{3}=q_{2}\left(1+\epsilon^{2}\right)$ and $\mathrm{q}_{4}=\mathrm{q}_{3}\left(1+\epsilon^{4}\right)=\mathrm{q}_{2}\left(1+\epsilon^{2}+\epsilon^{4}+\epsilon^{6}\right)$, we compute directly from $\mathrm{q}_{2}$ an approximation $\mathrm{q}_{4}^{\prime}$ to $\mathrm{q}_{4}$ :

$$
\mathrm{q}_{4}^{\prime}=\mathrm{q}_{2}\left(1+\epsilon^{2}+\hat{\epsilon}^{4}\right) .
$$

We now discuss the error in the result produced by this variant. First, neglecting the term in $\epsilon^{6}$ leads to an error around $2^{-6 p}$. Moreover from the expansion

$$
\begin{equation*}
\epsilon^{4}=\left(\hat{\epsilon}+\epsilon_{r}\right)^{4}=\hat{\epsilon}^{4}+4 \epsilon_{r} \hat{\epsilon}^{3}+6 \epsilon_{r}^{2} \hat{\epsilon}^{2}+4 \epsilon_{\mathrm{r}}^{3} \hat{\epsilon}+\epsilon_{\mathrm{r}}^{4} \tag{5}
\end{equation*}
$$

where $\epsilon_{r}=\epsilon-\hat{\epsilon}$ (which gives $\left|\epsilon_{r}\right|<2^{-2 p}$ ), we find that the error committed when replacing $\epsilon^{4}$ by $\hat{\epsilon}^{4}$ is around $2^{-5 p+2}$. For instance, if $p=12$ this variant allows to perform the division in 13 cycles (see Figure 2), with an error around $2^{-58}$. Hence, we save 4 cycles compared to the direct implementation, but at the cost of a poorer accuracy. If we use a bipartite table lookup, the same error $2^{-58}$ is achieved, with $p$ equal to 9 instead of 12 (i.e., with much smaller tables).

### 2.4 Variant B.

To get a better accuracy than with variant A , compute the first error term in (5), that is, $c=4 \epsilon_{r} \hat{\epsilon}^{3}$. This is done by tabulating $\hat{\epsilon}^{3}$ and performing the multiplication $\epsilon_{\mathrm{r}} \times \hat{\epsilon}^{3}$ in the pipelined multiplier. Hence, in this variant we compute a better approximation to $\mathrm{q}_{4}$, that is,

$$
\mathrm{q}_{4}^{\prime \prime}=\mathrm{q}_{2}\left(1+\epsilon^{2}+\hat{\epsilon}^{4}+4 \epsilon_{\mathrm{r}} \hat{\epsilon}^{3}\right) .
$$



Figure 2: Schedule of variant A. Requires 13 cycles, with an accuracy lower than that of the direct implementation. Two interlaced divisions are performed in 15 cycles.

We need one more cycle (for the computation of $c$ ) and one more table (for $\hat{\epsilon}^{3}$ ) than in variant A. However, it is possible to improve this using the following expression for $\mathrm{q}_{4}^{\prime \prime}$ :

$$
\mathrm{q}_{4}^{\prime \prime}=\mathrm{q}_{2}\left(1+\epsilon^{2}+\hat{\epsilon}^{3}\left(4 \epsilon_{\mathrm{r}}+\widehat{\epsilon}\right)\right) .
$$

We now only need one table for $\hat{\epsilon}^{3}$, and one cycle for the computation of $c^{\prime}=$ $\hat{\epsilon}^{3}\left(4 \epsilon_{r}+\widehat{\epsilon}\right)$. The error is about $6 \times 2^{-6 p}$ : for $p=12$, this is less than $2^{-69}$. If we use a bipartite lookup, we get an error $2^{-63}$ with $p=9$. The corresponding schedule is shown Figure 3. On a 4-cycle multiplier, it requires 13 cycles. If we interlace two divisions, the whole calculation requires 17 cycles only. A better performance can be obtained when performing two or three consecutive divisions by the same denominator. This happens, for example, in normalizing 2-D (3-D) vectors. The improvement comes from the fact that the $r_{i}$ 's are the same. Computing $a_{1} / d$ and $a_{2} / d\left(a_{1} / d, a_{2} / d\right.$ and $a_{3} / d$ for 3-D) requires 15 cycles (resp. 16 cycles), whereas first computing $1 / d$ and then multiplying this intermediate result by $a_{1}$ and $a_{2}\left(a_{1}\right.$, $a_{2}$ and $a_{3}$ ) would take 20 cycles (resp. 21 cycles).

### 2.5 Variant C

In Variant B, the values of $\hat{\epsilon}^{3}$ are precomputed and stored in a table with $p$ address bits. If we consider the following formula: $\left(1+\epsilon^{2}+\epsilon^{4}\right)=\left(1+\epsilon^{2}+\epsilon\right)\left(1+\epsilon^{2}-\epsilon\right)$, it is possible to compute ( $1+\epsilon^{2}+\epsilon^{4}$ ) as soon as $\epsilon^{2}$ is known. This technique requires 16 cycles but no table (except the one for $K_{1}$ ) and the error is around $2^{-6 p}$. This variant is probably less interesting than the direct implementation or Variant B. We mention it since it reduces the table requirements.

We present a summary of the properties of these different variants in Table 1.

| Method | number of cycles | bits of accuracy | table size |
| :--- | :---: | :---: | :---: |
| Direct | 17 | 96 | $\mathrm{n} \times 2^{p}$ bits |
| Variant A | 13 | 58 | $2 \mathrm{n} \times 2^{p}$ bits |
| Variant B | 14 | 69 | $2 \mathrm{n} \times 2^{p}$ bits |
| Variant C | 16 | 71 | $\mathrm{n} \times 2^{p}$ bits |

Table 1: Main properties of the proposed variants. The third column gives the amount of memory required including the table used for $\mathrm{K}_{1}$.


Figure 3: Variant B. Mult. $c^{\prime}$ is the computation of $\widehat{\epsilon}^{3}\left(4 \epsilon_{r}+\widehat{\epsilon}\right)$. Mult. $5^{\prime \prime}$ is the final multiplication. It has one more cycle than Variant $A$, but the accuracy is much better. Two interlaced divisions need 17 cycles.

### 2.6 Implementations on multipliers with a different number of cycles

The variants presented so far were illustrated assuming a 4-cycle pipelined multiplier. They can be used on multipliers with less or more than 4 cycles. For instance, let us assume a 2-cycle multiplier. With such a multiplier, the direct iteration (assuming we still wish to compute $\mathrm{q}_{4}$ ) is implemented in 9 cycles. Variant A is implemented in 7 cycles, and variant $B$ is implemented in 8 cycles.

On a 3-cycle multiplier, the direct iteration is implemented in 13 cycles, whereas variant A is implemented in 10 cycles, and variant B in 11 cycles.

Nevertheless, performing two consecutive divisions with multipliers with less than 4 cycles seems less interesting than previously. On a 2 -cycle ( 3 -cycle) multiplier, the direct iteration is implemented in 16 (18) cycles, variant A in 12 (14) cycles, and variant B in 14 (16) cycles.

### 2.7 Implementations with more than four iterations

The same approach is applicable if we want to perform one more iteration of the Goldschmidt algorithm. For example, assume that we add one more step to the algorithm presented in section 2.1. The final result $\mathrm{q}_{5}$ is obtained as

$$
\mathrm{q}_{5}=\mathrm{q}_{4} \times \mathrm{K}_{5}=\mathrm{q}_{4}\left(1+\epsilon^{8}\right)
$$

A direct implementation on a 4-cycle pipelined multiplier requires 21 cycles. However, once $\epsilon$ is known, we can look-up in a table the value $\hat{\epsilon}^{8}$, where $\widehat{\epsilon}$ is the same $p$-bit number as in the previous sections. That value will be used to directly estimate an approximation to $\mathrm{q}_{5}$ from $\mathrm{q}_{3}$. Hence, we can build several variants of the algorithm:

- First variant we compute

$$
\mathrm{q}_{5}^{\prime}=\mathrm{q}_{3}\left(1+\epsilon^{4}+\hat{\epsilon}^{8}\right)
$$

on a 4-cycle multiplier, this requires 17 cycles. The error is less than $2^{-9 p+4}$.

- Second variant we compute

$$
\mathrm{q}_{5}^{\prime \prime}=\mathrm{q}_{3}\left(1+\epsilon^{4}+\hat{\epsilon}^{8}+8 \hat{\epsilon}^{7} \epsilon_{\mathrm{r}}\right)
$$

on a 4 -cycle multiplier, this also requires 17 cycles, and the error is less than $2^{-10 p+6}$. Therefore this variant is more interesting than the previous one.

- Third variant we compute

$$
q_{5}^{\prime \prime \prime}=q_{3}\left(1+\epsilon^{4}+\hat{\epsilon}^{8}+8 \hat{\epsilon}^{7} \epsilon_{r}+28 \hat{\epsilon}^{6} \epsilon_{r}^{2}\right)
$$

on a 4 -cycle multiplier, this requires 18 cycles, and the error is less than $2^{-11 p+7}$.

## 3 Square root and square root reciprocal

### 3.1 Conventional iteration

In this section we will focus on the computation of $1 / \sqrt{x}$ and $\sqrt{x}$ for some real variable $x$. We will start from the generalization of the Goldschmidt method for square-root and square-root reciprocal that was introduced in [11]. An alternative would have been to use Newton-Raphson iteration for $\sqrt{\mathrm{x}}$ :

$$
r_{i+1}=\frac{1}{2} r_{i}\left(3-x r_{i}^{2}\right),
$$

that can be conveniently implemented (as suggested by Schulte and Wires [16]) as:

$$
\begin{aligned}
w_{i} & =r_{i}^{2} \\
d_{i} & =1-w_{i} x \\
r_{i+1} & =r_{i}+r_{i} d_{i} / 2
\end{aligned}
$$

This approach requires three dependent multiplies per iteration, similar to the Goldschmidt method introduced in [11] and cited here.

An interesting discussion on the computation of square roots and square root reciprocals using Newton's method can be found in [10].

Assume we wish to compute $\sqrt{x}$ or $1 / \sqrt{x}$ for $1 \leq x<2$. We shall consider the extension to the binade $2 \leq x<4$ in section 3.3. The first step is similar to what we have done for the division method. Starting from

$$
x=1 . d_{1} d_{2} \ldots d_{n}
$$

we define

$$
\hat{x}=1 . d_{1} d_{2} \ldots d_{p-1}+2^{-p}
$$

so then

$$
|x-\hat{x}| \leq 2^{-p}
$$

where $p \ll n$. From $\hat{x}$ we look-up $n+1$ bit round-to-nearest values of $K_{1}=$ $R N(1 / \hat{x})=.1 d_{2}^{\prime} d_{3}^{\prime} \ldots d_{n+1}^{\prime}$ and $\sqrt{K_{1}}=R N(1 / \sqrt{\hat{x}})=.1 d_{2}^{\prime \prime} d_{3}^{\prime \prime} \ldots d_{n+1}^{\prime \prime}$ in a table with $p-1$ address bits. The trade off of using table lookup of $K_{1}$ rather than computing the square of the table lookup of $\sqrt{\mathrm{K}_{1}}$, saves the latency cost of a dependent multiply while increasing the total table size. Importantly, the tables must be designed so that each table lookup value of $\mathrm{K}_{1}$ corresponds at full target accuracy to the square of the table lookup value of $\sqrt{\mathrm{K}_{1}}$. This requirement precludes the use of bipartite or linear interpolation for constructing both tables to greater initial accuracy as considered for division. Then, the conventional method (assuming we perform 4 iterations) consists in performing the following calculations itemized by groups of independent multiplications depending on results of the previous groups.

1. First group We define the variable $x_{1}$ and a variable $r_{1}$ by the independent multiplications

- $x_{1}=x \times K_{1}$ (called mult. 1 in figure 4)
- $r_{1}=\sqrt{K_{1}}$ if we aim at computing $1 / \sqrt{x}$;
- $r_{1}=x \times \sqrt{K_{1}}$ if we aim at computing $\sqrt{x}$. (mult. $1^{\prime}$ )

These choices are due to the fact that the next iterations compute $\mathrm{r}_{1} / \sqrt{\chi \mathrm{K}_{1}}$.
2. Second group We define $\epsilon_{1}=1-x_{1}$ and compute the independent multiplications:

- $\left(1+\frac{\epsilon_{1}}{2}\right)^{2}=\left(1+\frac{\epsilon_{1}}{2}\right) \times\left(1+\frac{\epsilon_{1}}{2}\right)$
- $r_{2}=\left(1+\frac{\epsilon_{1}}{2}\right) \times r_{1}$

3 Third group We compute

- $x_{2}=\left(1+\frac{\epsilon_{1}}{2}\right)^{2} \times x_{1}$
and we define $\epsilon_{2}$ by $\epsilon_{2}=1-x_{2}$.

4. Fourth group We compute the independent multiplications:

- $\left(1+\frac{\epsilon_{2}}{2}\right)^{2}=\left(1+\frac{\epsilon_{2}}{2}\right) \times\left(1+\frac{\epsilon_{2}}{2}\right)$
- $r_{3}=\left(1+\frac{\epsilon_{2}}{2}\right) \times r_{2}$

5. Fifth group We compute

- $x_{3}=\left(1+\frac{\epsilon_{2}}{2}\right)^{2} \times x_{2}$
and we define $\epsilon_{3}$ by $\epsilon_{3}=1-x_{3}$.

6. Sixth group We compute

- $r_{4}=\left(1+\frac{\epsilon_{2}}{2}\right) \times r_{3}$

Note that these computations would require a number of cycles equal to at least 6 times the multiplier latency in a pipelined multiplier implementation.

The error committed by approximating $\sqrt{x}$ (or $1 / \sqrt{x}$, depending on the initial value we have chosen for $r_{1}$ ) is easily found. Let us define $\epsilon=\epsilon_{1}$. Then recalling $K_{1}$ is rounded to nearest to $n+1$ places,

$$
|\epsilon| \leq\left|1-x K_{1}\right| \leq \frac{|\hat{x}-x|}{\hat{x}}+x 2^{-n-2}<2^{-p}
$$

for $n \geq 2 p$. From $x_{i+1}=\left(1+\frac{\epsilon_{i}}{2}\right)^{2} x_{i}=\left(1+\epsilon_{i}+\frac{1}{4} \epsilon_{i}^{2}\right) x_{i}$ and $\epsilon_{i+1}=1-x_{i+1}$ we easily find

$$
\begin{equation*}
\epsilon_{\mathfrak{i}+1}=\frac{3}{4} \epsilon_{\mathfrak{i}}^{2}+\frac{1}{4} \epsilon_{\mathfrak{i}}^{3} \tag{6}
\end{equation*}
$$

hence

$$
\begin{equation*}
\epsilon_{4} \approx\left(\frac{3}{4}\right)^{7} \epsilon^{8}<2^{-8 p-2} \tag{7}
\end{equation*}
$$

Now, since each time we multiply $x_{i}$ by some factor to get $x_{i+1}$ we multiply $r_{i}$ by the square root of the same factor, we can easily deduce

$$
r_{4}=\sqrt{\frac{x_{4}}{x_{1}}} \times r_{1}
$$

Hence,

$$
r_{4}=\frac{1+\alpha}{\sqrt{x_{1}}} r_{1}
$$

where $|\alpha|<2^{-8 p-3}$. This gives the final result:

- if we compute $1 / \sqrt{x}$ (that is, we have chosen $r_{1}=\sqrt{K_{1}}$ ) then

$$
r_{4}=\frac{1}{\sqrt{x}}(1+\alpha), \text { with }|\alpha|<2^{-8 p-3}
$$

- if we compute $\sqrt{x}$ (that is, we have chosen $r_{1}=x \sqrt{K_{1}}$ ) then

$$
r_{4}=\sqrt{x}(1+\alpha) \text {, with }|\alpha|<2^{-8 p-3} .
$$

### 3.2 Accelerating square root (inverse square root) method

Now, let us try to accelerate the computation by directly deducing an approximation to $r_{4}$ from $r_{2}$. To do that, we first deduce the values of the $\epsilon_{i}$ 's as polynomial identities in $\epsilon$ using (6). We obtain

$$
\begin{aligned}
\epsilon_{2}= & \frac{3}{4} \epsilon^{2}+\frac{1}{4} \epsilon^{3} \\
\epsilon_{3}= & \frac{27}{64} \epsilon^{4}+\frac{9}{32} \epsilon^{5}+\frac{39}{256} \epsilon^{6}+\frac{27}{256} \epsilon^{7} \\
& +\frac{9}{256} \epsilon^{8}+\frac{1}{256} \epsilon^{9} .
\end{aligned}
$$

Using this result, since $\mathrm{r}_{4}=\left(1+\frac{\epsilon_{2}}{2}\right)\left(1+\frac{\epsilon_{3}}{2}\right) \mathrm{r}_{2}$, we can deduce

$$
\begin{equation*}
\mathrm{r}_{4}=\left(1+\frac{\epsilon_{2}}{2}+\phi(\epsilon)\right) \mathrm{r}_{2} \tag{8}
\end{equation*}
$$

where

$$
\begin{aligned}
\phi(y)= & \frac{27}{128} y^{4}+\frac{9}{64} y^{5}+\frac{159}{1024} y^{6}+\frac{135}{1024} y^{7} \\
& +\frac{261}{4096} y^{8}+\frac{1}{32} y^{9}+\frac{27}{2048} y^{10}+\frac{3}{1024} y^{11} \\
& +\frac{1}{4096} y^{12} .
\end{aligned}
$$

This leads to the following solution: once $\epsilon$ is known, we can look-up in a table with $p-1$ address bits the value $\phi(\hat{\epsilon})$, where $\hat{\epsilon}$ (as for the division algorithm) is a $p$-bit number, constituted by the bits of $|\epsilon|$ of weight $2^{-p-1}, 2^{-p-2}, \ldots, 2^{-2 p+1}$ and a terminal unit. That is, if

$$
|\epsilon|=0.000 \ldots 0 \epsilon_{\mathfrak{p}+1} \epsilon_{\mathfrak{p}+2} \epsilon_{\mathfrak{p}+3} \epsilon_{\mathfrak{p}+4} \ldots,|\epsilon|<2^{-p}
$$

then truncating to a midpoint in the $2 p^{\prime}$ th place,

$$
|\hat{\boldsymbol{\epsilon}}|=0.000 \ldots 0 \epsilon_{\mathfrak{p}+1} \epsilon_{\mathfrak{p}+2} \ldots \epsilon_{2 p-1}+2^{-2 p},|\hat{\epsilon}|<2^{-p}
$$

where with $\hat{\epsilon}$ defined to have the same sign as $\epsilon$,

$$
|\epsilon-\hat{\epsilon}| \leq 2^{-2 p}
$$

Then we get the First scheme: We compute

$$
\mathrm{r}_{4}^{\prime}=\left(1+\frac{\epsilon_{2}}{2}+\phi(\hat{\epsilon})\right) .
$$

The error of this first scheme is around $\epsilon_{\mathrm{r}} \phi^{\prime}(\epsilon)$ (where $\epsilon_{\mathrm{r}}=\epsilon-\hat{\epsilon}$, so $\left|\epsilon_{\mathrm{r}}\right| \leq 2^{-2 p}$ ), which is less than $\frac{27}{32} 2^{-5 p}$. This scheme avoids several of the previous dependent multiplies. With a 4 -cycle pipelined multiplier the procedure can be implemented in 16 cycles. We do not discuss this implementation in detail, since the following 2nd scheme is more accurate and implementable in the same number of cycles.

Second scheme: We now assume that $\phi^{\prime}(\widehat{\epsilon})$ is tabulated, and we use the following approximation to $r_{4}$ :

$$
\mathrm{r}_{4}^{\prime \prime}=\left(1+\frac{\epsilon_{2}}{2}+\phi(\hat{\epsilon})+\epsilon_{\mathrm{r}} \phi^{\prime}(\hat{\epsilon})\right) \mathrm{r}_{2} .
$$

The error of the second scheme is around $\frac{\epsilon_{r}^{2}}{2} \phi^{\prime \prime}(\epsilon)$, which is less than $\frac{81}{64} 2^{-6 p}$. For instance, if $p=12$, this gives an error less than $2^{-71}$.

Figure 4 depicts the implementation of the computation of either $1 / \sqrt{x}$ or $\sqrt{x}$ using a 4 -cycle multiplier. Operations $1,2,3$ and 4 correspond to the computations of $x_{1}, r_{2}, x_{2}$ and $r_{4}^{\prime \prime}$. Mult. $1^{\prime}$ is performed only for the computation of $\sqrt{x}$. One can notice that the number of cycles required for both computations is the same since the initialization multiplication $r_{1}=x \sqrt{K_{1}}$ is inserted in the pipeline when computing the square root function.

This method requires 4 tables with $p-1$ address bits each for $K_{1}, \sqrt{K_{1}}, \phi(\hat{\epsilon})$ and $\phi^{\prime}(\hat{\epsilon})$ with total size of $(2 n+2(n-4 p)) \times 2^{p-1}$ bits.

### 3.3 Expanding domain and contracting tables

Depending on whether the exponent of the input value is even or odd, we need to compute the square root and/or square root reciprocal of $x$ or $2 \times x$ to span the domain $[1,4)$. This can be implemented by optionally inserting a multiplication by $\sqrt{2}$ somewhere in the pipeline. This can be done without increasing the latency in view of the one cycle allocated to the addition needed to form $\left(1+\frac{\epsilon}{2}+\phi(\hat{\epsilon})+\right.$ $\left.\epsilon_{r} \phi^{\prime}(\hat{\epsilon})\right)$. In Figure 4, $r_{2}$ is available at cycle 10. Thus we can perform an optional multiplication $\mathrm{r}_{2} \times \sqrt{2}$ from cycle 10 to cycle 13. Another solution is to store tables for both $\sqrt{\mathrm{K}_{1}}$ and $\sqrt{2 \mathrm{~K}_{1}}$. But these tables can be large and avoiding duplication of storage is desirable.


Figure 4: Implementation of $\sqrt{x}$ and $1 / \sqrt{x}$ on a 4-cycle pipelined multiplier. Mult. $1^{\prime}$ is performed only for $\sqrt{x}$; $c$ and $c^{\prime}$ correspond to the computations of $\epsilon^{2}$ and $\epsilon_{\mathrm{r}} \phi^{\prime}(\hat{\epsilon})$.

For the first lookup for $K_{1}$ and $\sqrt{K_{1}}$, we can relax the bound for $\left|\epsilon_{1}\right|$ by storing only a rounding to $p+4$ bits in the output table for $\frac{1}{\sqrt{x}}$. We still obtain $\left|\epsilon_{1}\right|<$ $2^{-p+0.25}$. Importantly, we than should store the exact $2 p+8$ bit square of the output of the first table as the result for $\mathrm{K}_{1}$ to make sure that the table values for $\mathrm{K}_{1}$ and $\sqrt{\mathrm{K}_{1}}$ are correctly related.

The Second scheme requires the computation of $\phi(\hat{\epsilon})+\epsilon_{r} \phi^{\prime}(\hat{\epsilon})$. If we expand $\phi(\hat{\epsilon})+\epsilon_{r} \phi^{\prime}(\hat{\epsilon})$, we may replace the product-sum employing two tables by a single
product employing only one table similar to the procedure in [13].

$$
\begin{equation*}
\phi(\widehat{\epsilon})+\epsilon_{r} \phi^{\prime}(\widehat{\epsilon})=\left(\hat{\epsilon}+4 \epsilon_{r}\right)\left(\frac{27}{128} \widehat{\epsilon}^{3}+\frac{9}{64} \widehat{\epsilon}^{4}+\frac{159}{1024} \widehat{\epsilon}^{5}\right)+\delta \hat{\epsilon}^{4} \epsilon_{r} \tag{9}
\end{equation*}
$$

where $|\delta|<1$.
$\hat{\epsilon}+4 \epsilon_{r}$ can be obtained as follows: we have a 2-bit overlap between $\hat{\epsilon}$ and $4 \epsilon_{r}$ at the $p-1$ th and $p$ th positions. These two bits can be added into $\hat{\epsilon}$ while we do the table lookup for ( $\left.\frac{27}{128} \hat{\epsilon}^{3}+\frac{9}{64} \hat{\epsilon}^{4}+\frac{159}{1024} \hat{\epsilon}^{5}\right)$ using $\hat{\epsilon}$. $\delta \hat{\epsilon}^{4} \epsilon_{r}$ provides the error term. The error is roughly of the same order of magnitude, i.e., $2^{-6(p-0.25)}$. This avoids one lookup table with no extra latency. Note that reducing the precision of the table output of $K_{1}$ and $\sqrt{K_{1}}$ may increase $\hat{\epsilon}$ to slightly more than $2^{-p}$. This could require a $p$ bit index lookup for at least a part of the range of $x$, a small increase compared to the major reduction in table size for $\mathrm{K}_{1}$ and $\sqrt{\mathrm{K}_{1}}$.

## Conclusion

We have presented several variants for implementing division, square roots and square root reciprocals on a 4 -cycle pipelined multiplier. The proposed schemes are based on the Goldschmidt iteration, and require fewer cycles than the original scheme. They also exhibit various trade-offs between computational delay, accuracy, and table size.

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