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The 2Sum and Fast2Sum algorithms are important building blocks in numerical computing. They are used (implicitely or explicitely) in many *compensated* algorithms (such as compensated summation or compensated polynomial evaluation). They are also used for manipulating floating-point *expansions*. We show that these algorithms are much more robust than it is usually believed: The returned result makes sense even when the rounding function is not round-to-nearest, and they are almost immune to overflow.

Additional Key Words and Phrases: Floating-point, error-free transformation, rounding errors, faithful rounding, 2Sum, Fast2Sum

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1. MOTIVATIONS

One easily shows that, provided that no overflow occurred, the error of a rounded-tonearest floating-point addition or subtraction is exactly representable by a floatingpoint number. The 2Sum [Knuth 1998] and Fast2Sum [Dekker 1971] algorithms make it possible to compute that error, under someconditions that will be noted below. That error can therefore be used later in a calculation: This is the underlying idea behind *compensated* algorithms. This has allowed for the development of various techniques, such as very accurate (compensated) summation [Kahan 1965; Neumaier 1974; Rump et al. 2008a, 2008b; Demmel and Nguyen 2013] or dot products, accurate polynomial evaluation [Graillat et al. 2009], efficient manipulation of floating-point expansions [Priest 1991; Shewchuk 1997; Hida et al. 2001] (floating-point expansions represent real numbers as unevaluated sums of floating-point numbers; algorithms for adding and multiplying these expansions make much use of 2Sum and Fast2Sum), and so on. However, these techniques suffer from some limitations:

—As noticed, among others, by Boldo and Daumas [2003], when the rounding function differs from round-to-nearest, the error of floating-point addition/subtraction may

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not be exactly representable—although this is often the case, see Lemma 2.4 below. And yet, rounding functions such as round towards $\pm\infty$ and round towards zero (called "directed roundings") are very useful. They allow one to get certain lower and/or upper bounds on the exact result of a computation and to easily implement interval arithmetic or stochastic arithmetic. Also, on many processors, changing the rounding mode is rather costly (it may require flushing the arithmetic pipeline): Hence, someone who required directed roundings in previous parts of one's program may be reluctant to switch to round-to-nearest before using algorithms 2Sum and Fast2Sum. Therefore, with directed roundings, even if we cannot always obtain the "exact" error of floating-point addition, it would still be useful to obtain a value close to that error. This problem was partly dealt with by Demmel and Nguyen [2013] and later by Graillat et al. [2015] for the Fast2Sum algorithm and by Martin-Dorel et al. [2013] in the case of "double roundings." We aim at tackling this issue in a more general context, and we wish to study the behaviour of 2Sum and Fast2Sum just assuming "general" rounding functions (see Definition 2.1 below). There already exist some kind of error-free transformations for summation with faithful rounding (see Priest [1992]). But these algorithms are costly, and our work shows that they are not necessary to achieve a good accuracy in many compensated algorithms: The simple and well-known Fast2Sum and 2Sum algorithms will frequently suffice.

—In the literature, these algorithms are usually considered as returning a correct result provided that no underflow or overflow occurs. The case of underflow can be handled fairly intuitively, using a property mentioned by Hauser [1996]—see below. The case of overflow is more problematic; the central question is as follows: *Can we have a "spurious" overflow?*, that is, can we have situations where the initial addition does not overflow, and yet one of the arithmetic operations performed in the 2Sum or Fast2Sum algorithm overflows? We will see in the following that such a situation almost never arises: This gives more confidence on many compensated algorithms and on algorithms that use floating-point expansions for "middle-precision" (e.g., around 100 digits) computations.

The rest of the article is organized as follows. In Section 2, we introduce some notation, definitions, and preliminary remarks used. The accuracy of Fast2Sum with no overflow is analyzed in Section 3 while the accuracy of 2Sum is dealt with in Section 4. Section 5 is devoted to show that Fast2Sum is immune to overflow and Section 6 to show that 2Sum is almost immune to overflow.

2. NOTATION, DEFINITIONS, AND PRELIMINARY REMARKS

Throughout this article, we assume a radix-2, precision-p, floating-point (FP) arithmetic of extremal exponents e_{\min} and e_{\max} . We also assume that subnormal numbers are available. We denote by Ω the largest representable floating-point number:

$$\Omega = (2 - 2^{1-p}) \cdot 2^{e_{\max}}$$

The floating-point predecessor of an FP number x will be noted pred(x). Its successor will be noted succ(x). If x is a real number, satisfying $2^k \le |x| < 2^{k+1}$, where k is an integer, then we define ulp(x) as follows:

$$ulp(x) = 2^{\max(k, e_{\min}) - p + 1}.$$

When an arithmetic operation τ is performed, with input FP operands a and b, what is actually computed is $\circ(a\tau b)$, where \circ is a rounding function. The IEEE-754 Standard for Floating-point arithmetic defines five rounding functions (round towards $+\infty$ —denoted RU below—, round towards $-\infty$ —denoted RD below—, round towards zero, round to nearest ties to even, and round to nearest ties to infinity). The two

round-to-nearest functions will be denoted RN in the following (the choice of the tie-breaking rule is not important here). We say that the FP number y is a *faithful* rounding of the real number x if $y \in {\text{RD}(x), \text{RU}(x)}$.

The rounding functions considered in this article satisfy the following definition (introduced by Kulisch [1971] under the name of *optimal rounding*).

Definition 2.1 (Rounding Function—"Optimal Rounding" in Kulisch [1971]). Let F_p be the set of the precision-p binary floating-point numbers. Function \circ from \mathbb{R} to F_p is a rounding function if

$$\begin{split} & - \forall x \in F_p, \circ(x) = x; \\ & - \forall (x, y) \in \mathbb{R}^2, x \leq y \Rightarrow \circ(x) \leq \circ(y). \end{split}$$

Remark 2.2. If \circ is a rounding function, then for any $x, \circ(x) \in \{\text{RD}(x), \text{RU}(x)\}$, where RD and RU are the rounds-towards $-\infty$ and round-towards $+\infty$ rounding functions.

The Fast2Sum algorithm was first introduced by Dekker [1971]. It allows one to compute the error of a (rounded to nearest) floating-point addition. That algorithm is

ALGORITHM 1: Conventional Fast2Sum Algorithm.

 $s \leftarrow \text{RN}(a+b)$ $z \leftarrow \text{RN}(s-a)$ $t \leftarrow \text{RN}(b-z)$

The conventional 2Sum algorithm, from Knuth [1998] and [Møller 1965], is

ALGORITHM 2: Conventional 2Sum Algorithm.

 $\begin{array}{ll} (1) & s \leftarrow \mathrm{RN}(a+b) \\ (2) & a' \leftarrow \mathrm{RN}(s-b) \\ (3) & b' \leftarrow \mathrm{RN}(s-a') \\ (4) & \delta_a \leftarrow \mathrm{RN}(a-a') \\ (5) & \delta_b \leftarrow \mathrm{RN}(b-b') \\ (6) & t \leftarrow \mathrm{RN}(\delta_a+\delta_b) \end{array}$

We know that, in the absence of overflow, if the radix β of the floating-point system being used is less than or equal to 3, and if the floating-point exponents e_a and e_b of a and b satisfy $e_a \ge e_b$, then the values s and t returned by Algorithm 1 satisfy s + t = a + b, that is, t is the error of the floating-point addition $s \leftarrow \text{RN}(a + b)$. Depending on the environment, testing the exponents of a and b may prove difficult. However, if $|a| \ge |b|$, then $e_a \ge e_b$. Algorithm 2 gives the same results as Algorithm 1 but without any requirement on β or on the exponents of a and b: It works in all cases provided that no overflow occurs. Due to the large penalty of a wrong branch prediction on modern architectures, if we do not have preliminary information on the respective orders of magnitude of a and b, calling the 6-operation 2Sum algorithm (Algorithm 2) is, in general, more efficient than comparing |a| and |b|, swapping them if needed, and calling the 3-operation algorithm Fast2Sum (Algorithm 1).

Algorithms 1 and 2 allow one to compute the error of a floating-point addition, provided that this addition was performed using a round-to-nearest rounding function. The computed error can be re-injected later in a calculation to compensate for it. This makes these "error-free transformations" very useful. However, when a rounding function different from round-to-nearest is used, the error of a floating-point addition is not always equal to a floating-point number. For instance, Muller et al. [2010],

in a radix-2 and precision-*p* arithmetic, assume rounding toward $-\infty$, if a = 1 and $b = -2^{-3p}$, then

$$s = \text{RD}(a+b) = 0.\underbrace{111111\cdots 11}_{p}$$

= 1-2^{-p},

and

which cannot be exactly represented with precision p (it would require precision 2p).

Therefore, with rounding functions different from RN, it is important to know what Algorithms 1 and 2 (or, rather, a modified version, with different rounding functions, of these algorithms) will return, to know if they are still of any use.

This issue was already dealt with by Martin-Dorel et al. [2013] in the restricted case where the rounding function is round to nearest with a possible "double rounding."¹ Demmel and Nguyen show that if $4ulp(a) \le |b| \le a$, then Algorithm 1 returns the error of the floating-point addition of a and b when directed rounding functions are used.

Graillat et al. [2015] give an error bound on the value returned by Algorithm 1 when directed rounding functions are used. We will improve on their bound, showing that the algorithm always returns the best possible result, namely a floating-point number t closer to the error of the floating-point addition of a and b than any other floating-point number. We will perform a similar analysis with the 2Sum algorithm.

There is another issue with these two algorithms. One can rather easily convince oneself that they are immune to underflow. The main reason for this is that, as shown by Hauser [1996], if the sum a + b of two floating-point numbers is below the underflow threshold, then that sum is a floating-point number, which implies that it is computed exactly, with any rounding function (it can be viewed as a consequence of Lemma 2.4 below). It is, however, much more difficult to know if these algorithms are, at least for some restricted input domains, immune to *overflow*. More precisely, if the first operation (namely the floating-point addition of a and b) does not overflow, can one of the following operations overflow?

The goal of this article is to deal with these two issues and to show that Fast2Sum and 2Sum (Algorithms 1 and 2) are much more robust than it is in general believed: For any combination of rounding functions (we can even have a different rounding function at each step of the algorithm) they are immune to overflow (except for a very limited number of "extreme" cases that are easy to detect), and they always produce a very accurate estimate of the error of the floating-point addition a + b. The algorithms that we will analyze are the following:

ALGORITHM 3: Fast2Sum With Faithful Roundings: \circ_1 , \circ_2 , \circ_3 are Rounding Functions (SEE Definition 2.4).

<i>s <</i>	$-\circ_1(a+b)$
$z \leftarrow$	$-\circ_2(s-a)$
$t \leftarrow$	$-\circ_3(b-z)$

We will make much use of the following result, due to Sterbenz [1974] (see, for instance, Hauser [1996] or Muller et al. [2010] for a proof).

¹This happens when the result of an operation is first rounded to a wider floating-point format before being rounded to the destination format.

ALGORITHM 4: 2Sum With Faithful Roundings: \circ_i , for i = 1, ..., 6, are Rounding Functions (See Definition 2.1).

 $(1) s \leftarrow \circ_1(a+b)$ $(2) a' \leftarrow \circ_2(s-b)$ $(3) b' \leftarrow \circ_3(s-a')$ $(4) \delta_a \leftarrow \circ_4(a-a')$ $(5) \delta_b \leftarrow \circ_5(b-b')$ $(6) t \leftarrow \circ_6(\delta_a+\delta_b)$

LEMMA 2.3 (STERBENZ). In a radix- β floating-point system with subnormal numbers available, if x and y are finite floating-point numbers such that

$$\frac{y}{2} \le x \le 2y,$$

then x - y is a floating-point number.

Lemma 2.4 below is common computer arithmetic folklore. We give a proof of it for the sake of completeness.

LEMMA 2.4. Let a and b be two binary FP numbers of respective exponents e_a and e_b . Let $s \in \{\text{RD}(a + b), \text{RU}(a + b)\}$. If the exponent e_s of s is less than or equal to $\min(e_a, e_b)$, then s = a + b exactly.

PROOF. First, *a* and *b* are multiples of 2^{e_a-p+1} and 2^{e_b-p+1} , respectively. Since $e_s \leq \min(e_a, e_b)$, the number a + b is an integer multiple of 2^{e_s-p+1} . Hence:

—the largest multiple of 2^{e_s-p+1} less than or equal to a + b is a + b itself, and —the smallest multiple of 2^{e_s-p+1} larger than or equal to a + b is a + b itself.

Therefore RD(a + b) = a + b, and RU(a + b) = a + b. Hence s = a + b. \Box

The following lemma allows one to understand the behavior of the first two lines of Fast2Sum.

LEMMA 2.5. Let a and b be two binary FP numbers, with $e_a \ge e_b$. Let $s \in \{\text{RD}(a + b), \text{RU}(a + b)\}$. The number s - a is a floating-point number (which implies that it will be computed exactly, with any rounding function).

Notice that Lemma 2.5 only holds in radix 2. With floating-point systems of higher radices, we can build counter-examples. For instance, in radix 3 with p = 4 and $\circ = RU$, if $a = 1002_3 = 29_{10}$ and $b = 2222_3 = 80_{10}$, then $s = RU(a + b) = 11010_3 = 111_{10}$, so $s - a = 10001_3 = 82_{10}$ is not exactly representable with precision 4.

PROOF. We have $a = M_a \cdot 2^{e_a - p + 1}$ and $b = M_b \cdot 2^{e_b - p + 1}$, with $|M_a|, |M_b| \le 2^p - 1$. Without loss of generality, we assume $M_a \ge 0$. Let M_s and e_s be the integral significand and the exponent of s, respectively. Since $|s| \le 2 \max\{|a|, |b|\}$, we have $e_s \le e_a + 1$.

(1) If $e_s = e_a + 1$, then

$$M_{s} \in \left\{ \left\lfloor \frac{M_{a}}{2} + \frac{M_{b}}{2^{1+(e_{a}-e_{b})}} \right\rfloor, \left\lceil \frac{M_{a}}{2} + \frac{M_{b}}{2^{1+(e_{a}-e_{b})}} \right\rceil \right\}.$$
 (1)

Defining $\mu = 2M_s - M_a$, from Equation (1), we obtain

$$\frac{M_b}{2^{e_a-e_b}}-2<\mu<\frac{M_b}{2^{e_a-e_b}}+2,$$

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which implies $|\mu| \leq |M_b| + 1 \leq 2^p$. An immediate consequence is that $s - a = \mu \cdot 2^{e_a - p + 1}$ is a floating-point number.

(2) If $e_s \leq e_a$, then first notice that if $e_s \leq e_b$, then s = a + b exactly by Lemma 2.4 so s - a = b is a floating-point number. Therefore, we need only to focus on the case $e_s > e_b$. In that case,

$$s \in \left\{ \left\lfloor 2^{e_a - e_s} M_a + 2^{e_b - e_s} M_b \right\rfloor \cdot 2^{e_s - p + 1}, \left\lceil 2^{e_a - e_s} M_a + 2^{e_b - e_s} M_b \right\rceil \cdot 2^{e_s - p + 1} \right\}$$

 $\mathbf{S0}$

$$\left(2^{e_b-e_s}M_b-1
ight)\cdot 2^{e_s-p+1} < s-a < \left(2^{e_b-e_s}M_b+1
ight)\cdot 2^{e_s-p+1}$$

Hence |s - a| is of the form $K \cdot 2^{e_s - p + 1}$, with

$$|K| \le rac{|M_b|}{2} + 1 < 2^p,$$

which implies that it is a floating-point number. \Box

The following lemma shows that even when the rounding function is not round-tonearest, the error of a floating-point addition will very frequently be exactly representable by a floating-point number.

LEMMA 2.6. Let a and b be binary, precision-p, floating-point numbers. Let $s \in \{\text{RD}(a+b), \text{RU}(a+b)\}$. If the difference $|e_a - e_b|$ of the exponents of a and b does not exceed p - 1, then s - (a + b) is a binary, precision-p, floating-point number.

PROOF. Without loss of generality, we assume $|a| \ge |b|$, and $e_a - e_b \le p - 1$. The numbers a and b are multiple of $2^{e_b - p + 1}$, therefore a + b and s are multiple of $2^{e_b - p + 1}$ too. Therefore, there exists an integer Z such that

$$(a+b) - s = Z \cdot 2^{e_b - p + 1}.$$
(2)

Let e_s be the FP exponent of s. Since |s - (a + b)| < ulp(s), we have $|(a + b) - s| < 2^{e_s - p + 1}$. Since $|b| \le |a|$, we have $|s| \le 2|a|$, which implies $e_s \le e_a + 1$. Therefore,

$$|(a+b) - s| < 2^{e_a - p + 2} \le 2^{e_b + 1}.$$
(3)

By combining Equations (2) and (3), we deduce that $|Z| \le 2^p - 1$, and therefore (a+b)-s is a FP number. \Box

3. ACCURACY OF FAST2SUM IN THE ABSENCE OF OVERFLOW

Let us first deal with the Fast2Sum algorithm with arbitrary rounding functions (Algorithm 3).

THEOREM 3.1. If no overflow occurs, and $e_a \ge e_b$, then the values s and t returned by Algorithm 3 satisfy

$$t = \circ_3((a+b) - s),$$

that is, t is a faithful rounding of the error of the FP addition $s \leftarrow \circ_1(a+b)$.

Notice that if we combine this theorem with Lemma 2.6, we deduce that if the difference of the exponents of *a* and *b* does not exceed p - 1 (which will occur in many practical cases), then *t* is exactly (a + b) - s.

PROOF. Lemma 2.5 above implies that s-a is a floating-point number. Hence, z = s-a, so

$$t = \circ_3(b-z) = \circ_3((a+b)-s). \quad \Box$$

4:6

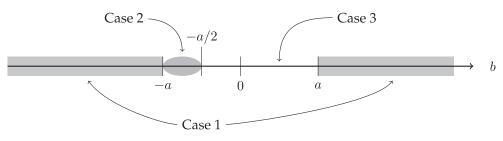


Fig. 1. Various cases discussed in the proof.

4. ACCURACY OF 2SUM IN THE ABSENCE OF OVERFLOW

We now consider the 2Sum algorithm with arbitrary rounding functions (Algorithm 4). Contrarily to what happens in the previous section with Algorithm 3, we do not always obtain a final value t equal to a faithful rounding of (a + b) - s. Consider the following example, in binary32/single-precision arithmetic (p = 24):

 $\begin{array}{l} -a = 3076485 \cdot 2^{-21}, \, b = -6130317 \cdot 2^{-49}; \\ -\circ_1 = \circ_2 = \circ_5 = \mathrm{RU}, \, \circ_3 = \circ_4 = \circ_6 = \mathrm{RD}. \end{array}$

We successively obtain

 $\begin{array}{l} s &= a = 3076485 \cdot 2^{-21}; \\ a' &= 12305941 \cdot 2^{-23}; \\ b' &= -2^{-23}; \\ \delta_a &= -2^{-23}; \\ \delta_b &= 15244637 \cdot 2^{-47}; \\ t &= -1532579 \cdot 2^{-47}, \end{array}$

and, since (a + b) - s = b is a floating-point number, with any rounding function \circ , $\circ((a + b) - s) = b$ differs from *t*. However, $(a + b - s) - t = -2^{-49}$, so *t* remains a very good approximation to (a + b) - s. As we are going to see, this is always true. More precisely, we prove together the following two results. The first one (Theorem 4.1) is the main result of this section. The second one (Lemma 4.2) is needed in the proof of Theorem 6.2.

THEOREM 4.1. If $p \ge 4$ and no overflow occurs, then the values s and t returned by Algorithm 3 satisfy

$$t = (a+b) - s + \alpha,$$

with $|\alpha| < 2^{-p+1} \cdot \text{ulp}(a+b) \le 2^{-p+1} \cdot \text{ulp}(s)$. Furthermore, if the floating-point exponents e_s and e_b of s and b satisfy $e_s - e_b \le p - 1$, then t is a faithful rounding of (a+b) - s.

LEMMA 4.2. If $p \ge 4$ and no overflow occurs in lines (1) to (5) of Algorithm 4, then the variables δ_a and δ_b computed at lines (4) and (5) satisfy

$$|\delta_a + \delta_b| \le \text{ulp}(a+b).$$

PROOF. We prove together Theorem 4.1 and Lemma 4.2. This means the case split and intermediate results are the same, but they do not rely on one another. Without loss of generality, we assume $a \ge 0$. Figure 1 illustrates the various cases that are discussed in the proof. Case 1 ($|b| \ge a$) just uses Theorem 3.1 (we can exhibit Algorithm 3, hidden in the lines of Algorithm 4). Case 2 (|b| < a and $|s| \le |b|$) is a simple application of Sterbenz' Lemma (Lemma 2.3), and Case 3 (|b| < a and |s| > |b|) requires more effort. Notice that in Cases 1 and 2, *t* will always be a faithful rounding of a + b - s. Hence,

the part in the theorem that is specific to the case $e_s - e_b \le p - 1$ does not need to be addressed in these first two cases.

Case 1 – If $|\mathbf{b}| \ge \mathbf{a}$, then lines (1), (2), and (4) of Algorithm 4 are equivalent to Fast2Sum(b,a). Therefore, from Theorem 3.1, we have a' = s - b and $\delta_a = \circ_4(a + b - s)$, hence $|\delta_a| \le ulp(a + b)$. An immediate consequence of a' = s - b is b' = b and $\delta_b = 0$. From this, we find

$$t = \circ_4(a+b-s),$$

which proves the result of Theorem 4.1, and $|\delta_a + \delta_b| \leq ulp(a + b)$, so the result of Lemma 4.2 holds.

Case 2 – If $|\mathbf{b}| < \mathbf{a}$ and $|\mathbf{s}| \le |\mathbf{b}|$ (which is equivalent to saying that $-a < b \le -a/2$), then by Sterbenz Lemma, s = a + b. An immediate consequence is a' = a, b' = b, $\delta_a = \delta_b = 0$ (so, obviously, the result of Lemma 4.2 holds), t = 0. Hence t = (a + b) - s (so the result of Theorem 4.1 holds).

Case 3 – If $|\mathbf{b}| < \mathbf{a}$ and $|\mathbf{s}| > |\mathbf{b}|$ (which is equivalent to saying that -a/2 < b < a), notice that we have s > 0. Let $u = 2^{1-p}$ (i.e., u is the rounding unit for directed roundings). We have

$$s = (a+b) \cdot (1+\epsilon_1); \text{with}|\epsilon_1| \le u;$$

$$a' = (s-b) \cdot (1+\epsilon_2); \text{with}|\epsilon_2| \le u.$$

Thus $a' = (a + a\epsilon_1 + b\epsilon_1) \cdot (1 + \epsilon_2)$. Since |b| < a, $a\epsilon_1 + b\epsilon_1$ can be written $2a\epsilon_3$, with $|\epsilon_3| \le u$. Therefore

$$a' = a \cdot (1 + \eta),$$

with $|\eta| \le 3u + 2u^2$. As soon as $p \ge 4$, we have $|\eta| < 1/2$, so $a' \ge 0$ and $a/2 \le a' \le 2a$. Therefore, Sterbenz Lemma applies to line (4) of Algorithm 4, and

$$\delta_a = a - a'. \tag{4}$$

Also, since s > |b|, lines (2) and (3) of Algorithm 4 are equivalent to the first two lines of Fast2Sum(s, -b), so

$$b' = s - a' \tag{5}$$

and

$$\delta_b = \circ_5 (a' - (s - b)). \tag{6}$$

Notice that, from Lemma 2.6, as soon as the exponents e_s and e_b of s and b satisfy $e_s - e_b \le p - 1$, Equation (6) implies $\delta_b = a' - (s - b)$, which, combined with Equation (4), gives $t = \circ_6(a + b - s)$, so Theorem 4.1 holds. Also, in that case, $\delta_a + \delta_b = (a + b) - s$, so Lemma 4.2 holds. Hence, *let us now assume that* $e_s - e_b \ge p$. Notice that this implies

$$|b| < 2^{e_b+1} \le 2^{e_s-p+1} = ulp(s).$$

Hence,

 $a' \in {\text{succ}(s), s, \text{pred}(s), \text{pred}(\text{pred}(s))}.$

Notice that the case a' = pred(pred(s)) can occur only when *s* is a power of 2, as seen in Figures 2 and 3.

(1) If $\mathbf{a}' = \mathbf{s}$, then b' = 0. It follows that $\delta_b = b$ and $\delta_a = a - s$, for which we deduce $t = \circ_6(\delta_a + \delta_b) = \circ_6(a + b - s)$, so Theorem 4.1 holds, and $|\delta_a + \delta_b| = |a + b - s| < ulp(a + b)$, so Lemma 4.2 holds.

(2) If $a' \neq s$, then

$$a' = s - \sigma \cdot \operatorname{ulp}(s)$$
, with $\sigma \in \{-1, 1/2, 1\}$,

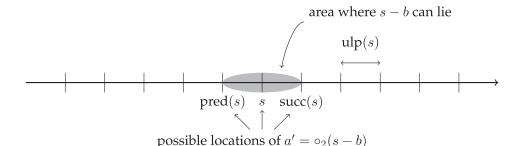
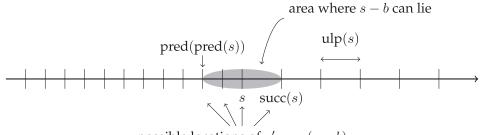


Fig. 2. General case: *s* is not a power of 2.



possible locations of $a' = \circ_2(s-b)$

Fig. 3. Special case: *s* is a power of 2.

(the case $\sigma = 1/2$ can occur only when *s* is a power of 2), and we have

$$b' = \sigma \cdot ulp(s)$$

$$\delta_a = a - s + \sigma \cdot ulp(s)$$

$$\delta_b = \circ_5(b - \sigma \cdot ulp(s)).$$

We know that |b| < ulp(s). Furthermore, *b* has the same sign as σ . Therefore

—either $|b| \ge |\sigma|/2 \cdot \text{ulp}(s)$, in which case Sterbenz Lemma implies $\delta_b = b - \sigma \cdot \text{ulp}(s)$, so $\delta_a + \delta_b = a + b - s$. This has two consequences: First, $|\delta_a + \delta_b| = |a + b - s| < \text{ulp}(a + b)$, so Lemma 4.2 holds, and, second, $t = \circ_6(a + b - s)$, and therefore Theorem 4.1 holds;

—or $|b| < |\sigma|/2 \cdot ulp(s)$, in which case

$$|b - \sigma \cdot ulp(s)| < |\sigma| \cdot ulp(s)$$

(unless b = 0 but that case is straightforwardly handled), so (since $|\sigma| \cdot ulp(s)$ is a power of 2)

$$|\delta_b - (b - \sigma \cdot \mathbf{ulp}(s))| < \frac{|\sigma|}{2} \mathbf{ulp}(\mathbf{ulp}(s)) = |\sigma| \cdot 2^{-p} \mathbf{ulp}(s)$$

(since ulp(s) is a power of 2). An immediate consequence is

$$|(\delta_a + \delta_b) - (a + b - s)| < |\sigma| \cdot 2^{-p} ulp(s).$$

$$\tag{7}$$

Since we already know that |(a + b) - s| < ulp(a + b), we deduce

$$|\delta_a + \delta_b| < \operatorname{ulp}(a+b) + |\sigma| \cdot 2^{-p} \operatorname{ulp}(s).$$
(8)

Let us try to slightly improve on the bound (8). First, from $|b| \le ulp(s)$, one easily deduces a > s/2 (otherwise, we would have $a + b \le s/2 + ulp(s)$, which would

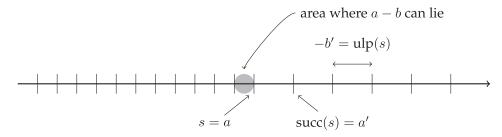


Fig. 4. Last case: *s* is a power of 2, a + b < s, and $\sigma = -1$.

imply $s = o_1(a + b) \le o_1(s/2 + ulp(s)) = s/2 + ulp(s))$. Hence δ_a is a multiple of $\frac{1}{2}ulp(s)$. Also, ulp(a + b) is equal to ulp(s) or $\frac{1}{2}ulp(s)$.

Finally, $|b| < |\sigma|/2 \cdot \text{ulp}(s)$ and $b' = \sigma \cdot \text{ulp}(s)$ imply $|b - b'| > \frac{|\sigma|}{2} \text{ulp}(s)$, so $|\delta_b| \ge \frac{|\sigma|}{2} \text{ulp}(s)$, which implies that δ_b is a multiple of $|\sigma| \cdot 2^{-p} \text{ulp}(s)$. All this implies that $\delta_a + \delta_b$ is a multiple of $|\sigma| \cdot 2^{-p} \text{ulp}(s)$. Hence, from Equation (8), we deduce

$$|\delta_a + \delta_b| \le \operatorname{ulp}(a+b)$$

First, this proves Lemma 4.2. Furthermore, since ulp(a + b) is a power of 2, we obtain

$$|\circ_6 (\delta_a + \delta_b) - (\delta_a + \delta_b)| \le \frac{1}{2} \operatorname{ulp}(\operatorname{ulp}(a + b)) = 2^{-p} \operatorname{ulp}(a + b).$$

Combined with Equation (7), this gives

$$|t - (a + b - s)| < 2^{-p} \cdot (\operatorname{ulp}(a + b) + |\sigma| \cdot \operatorname{ulp}(s)).$$
(9)

This already gives $|t - (a + b - s)| < 2^{-p+1} \cdot ulp(s)$. Let us now try to express a bound on |t - (a + b - s)| as a function of ulp(a + b) only. We have four cases to consider

- (a) if s is not a power of 2, or if $a + b \ge s$, then ulp(a + b) = ulp(s), which gives $|t (a + b s)| < 2^{-p+1} \cdot ulp(a + b)$, so Theorem 4.1 holds;
- (b) if s is a power of 2 and a + b < s and $\sigma = 1/2$, then $ulp(a + b) = 1/2 \cdot ulp(s)$, and Equation (9) implies $|t (a + b s)| < 2^{-p+1} \cdot ulp(a + b)$, so Theorem 4.1 holds;
- (c) the case when *s* is a power of 2, a + b < s, and $\sigma = 1$ is impossible: We assumed $|b| < |\sigma|/2 \cdot \text{ulp}(s) = 1/2 \cdot \text{ulp}(s)$, which implies $s b \ge s 1/2 \cdot \text{ulp}(s) = \text{pred}(s)$, which implies $a' = \circ_2(s b) \ge \text{pred}(s)$, which is not compatible with the assumption $\sigma = 1$, since $a' = s \sigma \text{ulp}(s)$;
- (d) if *s* is a power of 2, a + b < s, and $\sigma = -1$, then we have the following relations (see Figure 4): $a' = \operatorname{succ}(s) = s + \operatorname{ulp}(s)$, $b' = -\operatorname{ulp}(s)$, and $-1/2 \cdot \operatorname{ulp}(s) < b < 0$. We deduce that $a > \operatorname{pred}(s) = s - \frac{1}{2}\operatorname{ulp}(s)$ (otherwise, we would have $a + b < \operatorname{pred}(s)$, which would imply $s = \circ_1(a + b) \leq \operatorname{pred}(s)$). Similarly, we have $a < \operatorname{succ}(s) = s + \operatorname{ulp}(s)$ (otherwise, we would have $a + b \geq \operatorname{succ}(s) - \frac{1}{2}\operatorname{ulp}(s) > s$). Therefore a = s, from which we immediately deduce $\delta_a = -\operatorname{ulp}(s)$ and $\delta_b = \circ_5(b + \operatorname{ulp}(s))$. Now, δ_a and δ_b have opposite signs, and

$$\frac{1}{2}\mathbf{ulp}(s) < b + \mathbf{ulp}(s) < \mathbf{ulp}(s),$$

(notice that since ulp(s) is a power of 2, this implies $ulp(b + ulp(s)) \le 2^{-p}ulp(s)$) from which we deduce

$$\frac{|\delta_a|}{2} = \frac{1}{2} \operatorname{ulp}(s) \le \circ_5(b + \operatorname{ulp}(s)) = \delta_b \le |\delta_a| = \operatorname{ulp}(s),$$

hence we can apply Sterbenz lemma to the addition of δ_a and δ_b , which gives

$$t = \circ_6(\delta_a + \delta_b) = \delta_a + \delta_b$$

= $-ulp(s) + \circ_5(b + ulp(s))$
= $b + \eta$
= $a + b - s + \eta$,

(since a = s), with $|\eta| < ulp(b + ulp(s)) \le 2^{-p} \cdot ulp(s) = 2^{-p+1}ulp(a + b)$, hence Theorem 4.1 holds. \Box

5. FAST2SUM IS IMMUNE TO OVERFLOW

Let us now consider Algorithm 3, with $e_a \ge e_b$, where e_a and e_b are the floating-point exponents of a and b, and let us assume that no overflow occurred in the first line $(s \leftarrow \circ_1(a+b))$. Without loss of generality, we can assume a > 0. Let us first deal with the second line of the algorithm $(z \leftarrow \circ_2(s-a))$.

We have $s = a + b + \epsilon$, with $|\epsilon| < ulp(a + b) \le 2ulp(a)$. Hence $s - a = b + \epsilon$. Therefore, if the computation of $z = o_2(s - a)$ overflows, it means that either $b < -\Omega + 2ulp(a)$ or $b > \Omega - 2ulp(a)$.

The second case is impossible: If $b > \Omega - 2ulp(a) \ge \Omega - 2ulp(\Omega)$, then (since $e_a \ge e_b$, which here implies $e_a = e_b = e_{\max}$) $a + b \ge \Omega - 2ulp(\Omega) + 2^{e_{\max}} = 3 \cdot 2^{e_{\max}} - 3 \cdot 2^{e_{\max}-p+1}$, which implies that a + b overflows. Let us consider the first case. In that case, we have $b < -\Omega + 2ulp(a) \le -\Omega + 2ulp(\Omega)$ and (since $e_a \ge e_b$ which implies here $e_a = e_{\max}$), $\Omega/2 < 2^{e_{\max}} \le a \le \Omega$), in the first operation we are in the conditions of Sterbenz Lemma, so s = a + b, which implies z = b: In that case the computation of z does not overflow.

Hence, in all cases, the second line of Algorithm 3 cannot overflow. Let us now deal with the last line $(t \leftarrow \circ_3(b-z))$. We know from Lemma 2.5 that z = s - a, so b - z = a + b - s. The computation of t can overflow only if $|b - z| > \Omega$, but this is impossible since

$$|b-z| = |(a+b)-s| < ulp(s) < |s|.$$

We immediately deduce

THEOREM 5.1. Assume that we perform Algorithm 3 with input values a and b whose exponents satisfy $e_a \ge e_b$. If the computation of s (first line of the algorithm) does not overflow, then the other lines of the algorithm cannot overflow.

6. 2SUM IS ALMOST IMMUNE TO OVERFLOW

The overflow analysis of Algorithm 4 will be significantly more difficult. Our main result is Theorem 6.2 below. To make its proof simpler, we first prove the following result.

LEMMA 6.1. If there are no overflows at lines (1) to (5) of Algorithm 4, then there cannot be an overflow at line (6).

PROOF. From Lemma 4.2, we know that $|\delta_a + \delta_b| \leq ulp(a+b)$. Since no overflow occurs at line (1), a + b is in the representable range, so $ulp(a+b) \leq 2^{-p+1}|a+b|$ is much below the overflow threshold. Hence line (6) of Algorithm 4 (namely, $t \leftarrow \circ_6(\delta_a + \delta_b)$) cannot overflow. \Box

THEOREM 6.2. If the first input value a of Algorithm 4 satisfies $|a| < \Omega$ and if there is no overflow at line (1) of the algorithm, then there will be no overflow at lines (2) to (6).

PROOF. Without loss of generality, we assume a > 0 and $b \neq 0$. Assume that no overflow occurred in the first line ($s \leftarrow \circ_1(a+b)$).

(1) If b > 0

The monotonicity of the rounding functions implies the following: (i) $s \ge b$, so $a' \ge \circ_2(0) = 0$, and (ii) $a' \le \circ_2(s) = s$. Therefore

$$0 \le a' \le s,\tag{10}$$

which implies that there is no overflow at line (2) of the algorithm. Now, Equation (10) implies $0 \le s - a' \le s$, so

$$0 \le b' \le s. \tag{11}$$

As a consequence, there is no overflow at line (3) of the algorithm.

Now, since a > 0 and $a' \ge 0$, we deduce $|a - a'| \le \max\{a, a'\}$, and hence line (4) cannot overflow.

Similarly, since b > 0 and $b' \ge 0$, we obtain $|b - b'| \le \max\{b, b'\}$, and hence line (5) cannot overflow.

Lemma 6.1 implies that line (6) cannot overflow.

(2) If b < 0

Notice that there cannot be an overflow at line (1): |a + b| (hence |s|) is less than or equal to max $\{|a|, |b|\}$.

(a) **if** -b < a, then a + b - ulp(a + b) < s < a + b + ulp(a + b), so

$$a + b - \operatorname{ulp}(a) < s < a + b + \operatorname{ulp}(a)$$
(12)

(since |a + b| < a, which implies $ulp(a + b) \le ulp(a)$). We therefore deduce

$$a - \operatorname{ulp}(a) < s - b < a + \operatorname{ulp}(a).$$

Therefore, unless $a = \Omega$, there will be no overflow at line (2) of the algorithm, and $a' = \circ_2(s - b)$ will satisfy

$$a - \operatorname{ulp}(a) \le a' \le a + \operatorname{ulp}(a).$$
(13)

(this is deduced using the monotonicity of the rounding function \circ_2 and the fact that a - ulp(a) and a + ulp(a) are floating-point numbers). We now assume $a \neq \Omega$ (which, with our assumption -b < a, implies $-b < pred(\Omega)$), that is, since *b* is a floating-point number,

$$|b| = -b \le \operatorname{pred}(\operatorname{pred}(\Omega)). \tag{14}$$

From Equations (12) and (13), we find

$$b - 2ulp(a) < s - a' < b + 2ulp(a),$$
 (15)

This, along with Equation (14) and $ulp(a) \le ulp(\Omega)$ implies that line (3) of the algorithm cannot overflow. Notice that 0 < -b < a implies

$$|b \pm 2\mathfrak{ulp}(a)| < a + 2\mathfrak{ulp}(a). \tag{16}$$

It also implies that a cannot be the smallest nonzero subnormal floating-point number $2^{e_{\min}-p+1}$. Hence $a \ge 2^{e_{\min}-p+2}$, so $a \ge 2ulp(a)$. This and Equation (16) give $|b \pm 2ulp(a)| \le 2a$ so $ulp(b \pm 2ulp(a)) \le 2ulp(a)$. Combined with Equation (15), this gives

$$b - 4ulp(a) \le b' \le b + 4ulp(a).$$
⁽¹⁷⁾

Now, from Equations (13) and (17), we deduce $|a - a'| \le ulp(a)$ and $|b - b'| \le 4ulp(a)$, so lines (4) and (5) of the algorithm cannot overflow. Lemma 6.1 implies that line (6) cannot overflow.

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(b) **if** $-\mathbf{b} \geq \mathbf{a}$. First, notice that the case $a \geq -b/2$ is easily handled, since Sterbenz Lemma applied to line (1) of Algorithm 4 implies s = a + b, so a = a', b = b', and $\delta_a = \delta_b = t = 0$. Hence we only need to focus on the case $\mathbf{a} < -\mathbf{b}/2$. From $0 \leq a < -b/2$, we deduce $b \leq a + b < b/2$, which implies²

$$b \le s \le b/2. \tag{18}$$

The consequence of Equation (18) is twofold. First, we immediately deduce $0 \le s - b \le -b/2$, so line (2) of Algorithm 4 cannot overflow, and, second, the Sterbenz Lemma can be applied to line (2) of Algorithm 4, so a' = s - b. It follows that b' = b and line (3) cannot overflow. Therefore a - a' = a + b - s, so |a-a'| < ulp(a+b), hence line (4) cannot overflow. We finally have $\delta_b = b - b' = 0$ and $t = \delta_a$: Lines (5) and (6) cannot overflow. \Box

Notice that condition $|a| < \Omega$ is necessary. Assume all rounding functions are RN (with ties-to-even tie-breaking rule). The choice $a = \Omega$ and $b = -(3/2) \cdot ulp(\Omega)$ will give no overflow at line (1) and an overflow at line (2).

Conclusion

We have shown that, in binary floating-point arithmetic, the 2Sum and Fast2Sum algorithms are more "robust" than it is usually believed: Even when the error of the initial floating-point addition is not exactly representable, they return a very good approximation to that error. Also, they are almost totally immune to overflow: The only case where a "spurious" overflow may occur is with 2Sum, when the absolute value of operand a is equal to the largest floating-point number.

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²Unless b/2 is not a floating-point number: this can happen only of b is subnormal, and in that case, with $0 \le a < -b/2$, overflow is of course impossible.

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