

DEPARTMENT OF INFORMATICS

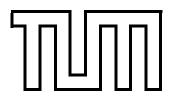
TECHNISCHE UNIVERSITÄT MÜNCHEN

Bachelor's Thesis in Informatics

Multiple Precision Floating Point Arithmetic in Isabelle/HOL

Fabian Hellauer





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Author: Fabian Hellauer

Supervisor: Prof. Tobias Nipkow

Advisor: Fabian Immler Submission date: March 15, 2016



I confirm that this bachelor's thesis is my sources and material used.	own work and I have documented all
München, March 15, 2016	Fabian Hellauer

Abstract

Many problems in geometry or topology systems can only be approximated by large computations with long sequences of operations. To best benefit from the speedup of machine computation, one has to make use of the hardware floating point instructions that are specified by the "Institute of Electrical and Electronics Engineers" (IEEE) 754 standard[1]. However, using the IEEE-floats directly would often create the need for a complicated numerical analysis due to them being affected by round-off. Another one of their properties removes this need however: Alongside the result, the round-off error of an addition or subtraction can be computed. Storing and using it in further operations thus makes the computation error-free. We give a simple data format in Isabelle/HOL [8] that uses this approach to provide fast algorithms for error-free addition, subtraction and multiplication.

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1 Introduction

When attacking computational problems by machine, often times fixed precision numbers have to be used: Arbitrary precision numbers are very slow, mostly because they don't use the hardware floating point operations that are available to modern systems. If the finite set of these machine numbers suffice (or an approximation within their range), the use of hard-wired operations can speed up the computation by a large factor. However, they introduce the problem of round-off, which, when not handled, will affect the output's precision in complex ways.

1.1 Round-off

Round-off occurs when the result of a floating-point operation cannot be represented as a datum of the same format. This happens in particular if the size of exponent and mantissa is fixed, as it is case for floats defined by the IEEE standard for floating-point arithmetic[1]. Round-off also occurs when casting measured or infinitely precise data to such a limited precision format.

Dealing with round-off If round-off affected arithmetic is used in a long sequence of operations, the result will only approximate within a certain range. Correctness proofs for assertions to this range will require a tedious numerical analysis of the algorithms which is very complex to do formally.

Avoiding round-off Another approach is to avoid round-off altogether. However, using an implementation of infinitely precise rationals might severely slow down the code's execution due to them not making use of hardware floating point operations that modern machines provide.

Float expansions If a finite set of numbers with magnitude and precision in the range of IEEE floats suffices however, both the precision and fast execution speed can be preserved: This thesis presents the "float expansion" approach, where the accumulated errors are stored in a list alongside with an approximation for the result of the executed sequence. It provides addition, subtraction and multiplication within the numbers representable in this way (a finite superset to IEEE floats).

1.2 Problem statement

Isabelle[8] already provides the arbitrary precision format real in its Complex-Main library. The widely popular "IEEE-floats" [1] are modelled in an Isabelle theory IEEE-Floating-Point/IEEE [11] provided by the "Archive of Formal Proofs" (AFP).

The task for this bachelor thesis is to use this formalization to present a "multiple precision" float arithmetic in Isabelle/HOL. In several scientific papers this is described as "floating point expansion" [9][5] or "multiple term" [5] strategy. It is an easy way to gain considerable amounts of precision while still using the IEEE floating point specification to enable the widely available acceleration of hardware operations.

1.3 Contributions

We explain different aspects of the "floating point expansion" approach and then provide the data format mpf, which stands for "multiple precision float". It implements error-free addition, subtraction and multiplication within the numbers representable in this format (a finite superset to IEEE floats). We use the formal setting of Isabelle/HOL to specify and prove the algorithms correct, but we make sure all of them can easily be executed by adapting Isabelle's Standard ML (SML) code generation for IEEE-floats.

2 Code Analysis

2.1 IEEE in Isabelle

The IEEE standard for floating point arithmetic (IEEE 754-2008)[1] is already modelled in Lei Yu's AFP entry *IEEE-Floating-Point/IEEE*. The formalization is quite general to accommodate for the many different allowed formats that arise when different precisions and exponent ranges are combined (the decimal formats are omitted). However, a strong precedence for the "binary64" format and the "roundTiesToEven" rounding mode can be observed. The operations using this format and rounding rule (called *float-format* respectively *To-nearest* in the theory) are wrapped into definitions with simpler names, e.g.

definition plus-float :: float \Rightarrow float \Rightarrow float where a + b = Abs-float (fadd float-format To-nearest (Rep-float a) (Rep-float b))

In the case of the format, this is justifiable by "binary64" being widely popular and hardware-implemented on most systems.

In the case of the rounding mode, the IEEE standard defines "roundTiesToEven" to be the default ([1]p. 16).

We also use this format and rounding mode, which enables us to use the code printing defined in the theory *Code-Float* from the same AFP entry. The "roundTiesTo-Even" rule is explicitly required for the TwoSum-properties ([7] sect. 4.3.3) and thus for all the presented algorithms.

2.2 Notation

In theory IEEE, the float operations use the + and - sign via instantiation float

For this thesis, we want the IEEE-operations to use a different symbol, and reserve the + operator for exact number formats. Fortunately, Isabelle provides spare symbols and the **abbreviation** command, which makes sure even the output uses the new notation. We thus state at the beginning of our **theory**:

```
— Use another notation for the possibly inexact IEEE-operations. abbreviation round-affected-plus :: float \Rightarrow float (infix1 \oplus 65) where round-affected-plus a b \equiv a + b
```

abbreviation round-affected-minus :: float \Rightarrow float \Rightarrow float (infixl \ominus 65) where round-affected-minus a $b \equiv a - b$

```
Afterwards,

\mathbf{term} \ a + (b::float)

outputs

a \oplus b :: IEEE.float
```

2.3 Making operations error-free

The core idea is to provide an error-free form of the basic operations between IEEE floats. Since we want the output to be floats as well, and rounding occurs for almost all input values, the only way to do so is to use the round-affected IEEE operation and then computing the error, also represented as floats. For the basic operations, this will turn out to be exactly another float.

2.3.1 Addition

We compute $a \oplus b$ together with y the error value. y will have the sign - or + corresponding to whether $a \oplus b$ is above resp. below the exact mathematical result of a + b. It can be computed by the following sequence, first described by Ole Møller in 1965[6]:

```
definition TwoSum :: float \Rightarrow float \Rightarrow float \times float where TwoSum \ a \ b = (let \ x = a \oplus b; \ b_v = x \ominus a; \ a_v = x \ominus b_v; \ b_r = b \ominus b_v; \ a_r = a \ominus a_v; \ y = a_r \oplus b_r \ in \ (x, \ y))
```

Here, we compute a value y such that a+b=x+y, where $x=a\oplus b$. The following lemma states the latter:

```
lemma TwoSum-correct1: TwoSum a \ b = (x, y) \Longrightarrow x = a \oplus b — x is defined in the first line of TwoSum and not changed thereafter. by (auto simp: TwoSum-def Let-def)
```

The other property needs the preconditions that both the input and the output represent real numbers (as opposed to the special values NaN and $\pm \infty$). This is checked by the predicate Finite. We use the exact arithmetic of real and the conversion Val :: float => real from IEEE-Floating-Point/IEEE.

```
lemma TwoSum-correct2:
fixes a b x y :: float
assumes Finite a
assumes Finite b
```

```
assumes Finite (a \oplus b)
assumes out: (x, y) = TwoSum \ a \ b
shows Val \ a + Val \ b = Val \ x + Val \ y
sorry
```

We assume the lemma by **sorry** for this thesis. Notice that a formal proof using the theorem prover Coq[2] is available online[3]

2.3.2 Subtraction

```
For a-b, we can compute:

definition TwoDiff:: float \Rightarrow float \Rightarrow float \times float where

TwoDiff \ a \ b = TwoSum \ a \ (float-neg \ b)
```

To drop the additional negation step, we could instead perform:

```
TwoDiff' \ a \ b = (let \ x = a \ominus b; \ b_v = x \ominus a; \ a_v = x \oplus b_v; \ b_r = b_v \ominus b; \ a_r = a \ominus a_v; \ y = a_r \oplus b_r \ in \ (x, y))
```

according to Shewchuk[10]. Note that we still need a + on the right side of the equation:

```
lemma TwoDiff\text{-}correct2:
fixes a\ b\ x\ y:: float
assumes Finite\ a
assumes Finite\ b
assumes Finite\ (a\ominus b)
assumes out: (x,\ y) = TwoDiff\ a\ b
shows Val\ a - Val\ b = Val\ x + Val\ y
```

To not further increase the amount of unproven lemmas, we drop this optimization for this thesis.

2.4 Code Analysis

The new data format is designed to implement the idea of storing all the errors as an unevaluated sum. It is defined as follows:

```
— Define the "Multiple Precision Float" type-synonym mpf = float \times float list fun approx :: mpf \Rightarrow float where approx (a, es) = a
```

```
fun errors :: mpf \Rightarrow float \ list \ \mathbf{where} errors (a, es) = es
```

where the tuple of a *float* and a *float list* should be seen together as a non-empty float list, ordered by decreasing magnitude. The approximation *approx* is just stored separately to avoid having to check for an empty list on access. Its quality depends on the executed algorithms (proofs about it can be found at [10]).

The mpf's represented value is the infinite precise sum of all its components:

```
fun Val-mpf :: mpf \Rightarrow real where Val-mpf (a, es) = Val \ a + listsum (map \ Val \ es)
```

Note that multiple mpfs can represent the same value. Many of these are invalid if we enforce the "non-overlapping" property proposed by Shewchuk[10] on the float list. However, since it needs to read out the bit representation of the IEEE float, there is no easy way to check this condition using the AFP-formalization. We could instead decrease the number of valid representations by not allowing zero components in the list's tail:

```
fun valid :: mpf \Rightarrow bool where
valid (a, es) = (case \ Iszero \ a \ of
True \Rightarrow es = [] \mid
False \Rightarrow Finite \ a \land list-all \ (\lambda f. \ Isdenormal \ f \lor Isnormal \ f) \ es)
```

where λf . Isdenormal $f \vee Isnormal f$ returns False for zero-floats:

```
lemma Iszero fl \Longrightarrow \neg(\lambda f. \ Isdenormal \ f \lor Isnormal \ f) \ fl

using float-distinct

by (metis Isnormal-def Iszero-def is-normal-def is-zero-def order-less-irreft)
```

Since zero components don't contribute to the mpf's value, omitting them is an easy way to save storage by decreasing the list size. The problem with this property is that the algorithms don't preserve the constraint by default. As Shewchuk puts it:

"A complicating characteristic of all the algorithms for manipulating expansions is that there may be spurious zero components scattered throughout the output expansions, even if no zeros were present in the input expansions."

As he shows by an example, they even occur in the middle of output lists that provably have all non-zero-components sorted. He also states:

"Unfortunately, accounting for these zero components could complicate the correctness proofs significantly." [10]

In other words: We could modify the algorithms to drop the zero component onthe-fly, but the extra branch would drastically increase the proof size. We instead settle for an even weaker property:

```
fun Finite-mpf :: mpf \Rightarrow bool where
  Finite-mpf (a, es) \longleftrightarrow Finite a \land list-all Finite es
```

Using this property and the well-known TwoSum-correct2 property described above ($\llbracket Finite ?a; Finite ?b; Finite (?a \oplus ?b); (?x, ?y) = TwoSum ?a ?b \rrbracket \Longrightarrow$ Val ?a + Val ?b = Val ?x + Val ?y, unproven in this thesis), we will be able to

```
prove computations error-free in the next section.
Here is a proof that valid implies Finite:
lemma valid-finite: valid (a, es) \Longrightarrow Finite-mpf(a, es)
 apply (simp split: bool.splits)
 using float-cases-finite float-distinct apply fastforce
 by (metis (no-types, lifting) Ball-set Finite-def)
definition safe-TwoSum \ a \ b =
 (let \ r = TwoSum \ a \ b \ in
   if Finite (fst r) \wedge Finite (snd r)
   then Some r
   else None)
definition safe-TwoDiff a b =
 (let \ r = TwoDiff \ a \ b \ in
   if Finite (fst r) \wedge Finite (snd r)
   then Some r
   else None)
lemma safe-TwoSum-finite:
 assumes safe-TwoSum\ a\ b = Some\ (s,\ e)
 shows safe-TwoSum-finite1: Finites
 and safe-TwoSum-finite2: Finite e
 using assms
 by (auto simp: safe-TwoSum-def Let-def split: split-if-asm)
lemma safe-TwoSum-correct1:
 safe-TwoSum a b = Some (x, y) \Longrightarrow x = a \oplus b
 by (auto simp: safe-TwoSum-def Let-def TwoSum-correct1 split: split-if-asm)
lemma safe-TwoSum-correct2:
 fixes a \ b \ x \ y :: float
 assumes Finite a Finite b Finite (a \oplus b)
 assumes out: safe-TwoSum\ a\ b = Some\ (x, y)
 shows Val \ a + Val \ b = Val \ x + Val \ y
 using assms
by (auto intro!: TwoSum-correct2 simp: safe-TwoSum-def Let-def split: split-if-asm)
```

```
definition IsZero-mpf\ mpf \longleftrightarrow Iszero\ (approx\ mpf) \land errors\ mpf = []
lemma float-distinct-10: \neg (Isnormal f \land Iszero f)
  by (auto simp add: float-defs is-normal-def is-zero-def)
lemma valid-no-zero-components: valid (a, es) \Longrightarrow list-all (\lambda f. \neg Iszero f) es
  apply (simp split: bool.splits)
  apply (induction es)
  using float-distinct(9) float-distinct-10
  apply auto
  done
lemma rec-val: Val-mpf (a, e \# es) = Val \ a + Val-mpf \ (e, es)
lemma rec-finite: Finite-mpf (a, e \# es) \longleftrightarrow Finite a \land Finite-mpf (e, es)
  by simp
fun safe-grow-mpf-rec :: mpf \Rightarrow float \Rightarrow mpf \ option \ \mathbf{where}
  safe-grow-mpf-rec(a, []) <math>f =
      (x, y) \leftarrow safe\text{-}TwoSum \ f \ a;
      Some (x, [y])
  safe-grow-mpf-rec~(a,~e~\#~es)~f=
    do \{
      (a', es') \leftarrow safe\text{-}grow\text{-}mpf\text{-}rec\ (e, es)\ f;
      (x, y) \leftarrow safe\text{-}TwoSum \ a' \ a;
      Some (x, y \# es')
```

At this point, we could implement the zero removal explained before, by modifying the last lines of the blocks:

```
fun safe-grow-mpf-rec-no-0 :: mpf \Rightarrow float \Rightarrow mpf option where safe-grow-mpf-rec-no-0 (a, []) f = do \{ (x, y) \leftarrow safe-TwoSum f a; if Iszero y then <math>Some (x, []) else Some (x, [y]) \} | safe-grow-mpf-rec-no-0 (a, e \# es) f = do \{ (a', es') \leftarrow safe-grow-mpf-rec-no-0 (e, es) f; (x, y) \leftarrow safe-TwoSum a' a; if Iszero y then Some (x, es') else Some (x, y \# es') \}
```

However, we don't pursue this idea further due to the problems mentioned there.

We rename the induction cases:

 $\textbf{lemmas} \ safe-grow-mpf-induct = safe-grow-mpf-rec. induct [case-names \ no-error \ in-between]$

```
lemma preserve-finite:
 assumes safe-grow-mpf-rec mpf x = Some r
 assumes Finite x Finite-mpf mpf
 shows Finite-mpf r
using assms
proof (induction mpf x arbitrary: r rule: safe-grow-mpf-induct)
  The base case is the case where the mpf is a single float with an empty error-list:
case (no\text{-}error\ a\ f)
— We apply the definition of safe-grow-mpf-rec:
from no-error.prems(1) have do \{(x, y) \leftarrow safe-TwoSum f a; Some (x, [y])\} = Some r
   unfolding safe-grow-mpf-rec.simps(1).
  Since we required the result to be some value, we can give it a name:
 then obtain x y where xy: safe-TwoSum f a = Some (x, y) and r: r = (x, [y])
   by (auto simp: bind-eq-Some-conv)
— and then delegate to the corresponding property of safe-TwoSum:
 moreover from safe-TwoSum-finite[OF xy]
   have Finite x Finite y.
 ultimately show ?case
   by simp
\mathbf{next}
case (in-between a e es f r-full)
 note in-between.prems(1)[simplified, unfolded bind-eq-Some-conv, simplified]
 then obtain l \ r where goal1: safe-grow-mpf-rec (e, es) \ f = Some \ (l, r)
   and r1: do \{(x, y) \leftarrow safe\text{-}TwoSum \ l \ a; Some \ (x, y \# r)\} = Some \ r\text{-}full
    by blast
 then obtain l2 r2 where l2: safe-TwoSum l a = Some (l2, r2) and
    r2: (l2, r2 \# r) = r\text{-full}
    using r1[unfolded bind-eq-Some-conv, simplified] by auto
 from r2 have ?case = Finite-mpf (l2, r2 \# r) by simp
 moreover have Finite 12
   using safe-TwoSum-finite1[OF l2].
 moreover have Finite r2
   using safe-TwoSum-finite2[OF l2].
 moreover from in-between.IH[OF qoal1 in-between.prems(2)] have list-all Finite r
   using in-between.prems(3) by auto
 ultimately
   show ?case
   by simp
qed
```

Notice that the "assignments" (\leftarrow) in a Monad like mpf option can also be written using the \gg -operator (bind) and λ -notation. This is also what Isabelle's state panel will output, e.g.

```
do \{(x, y) \leftarrow safe\text{-}TwoSum \ l \ a; \ Some \ (x, y \# r)\}
```

becomes

```
safe-TwoSum l \ a \gg (\lambda(x, y). \ Some \ (x, y \# r))
etc. We perform the next proof using this style:
lemma preserve-val:
 assumes safe-grow-mpf-rec mpf x = Some r
 assumes Finite x Finite-mpf mpf
 shows Val\text{-}mpf \ r = Val\text{-}mpf \ mpf + Val \ x
using assms
proof (induction mpf x arbitrary: r rule: safe-grow-mpf-induct)
case (no\text{-}error\ a\ f)
 from no-error.prems(1) have safe-TwoSum f a \gg (\lambda(x, y). Some (x, [y]) = Some r
   unfolding safe-grow-mpf-rec.simps(1).
 then obtain x y where xy: safe-TwoSum f a = Some (x, y) and r: r = (x, [y])
   by (auto simp: bind-eq-Some-conv)
 from safe-TwoSum-finite1 [OF xy]
 have Finite x.
 from no-error have an: Finite a by simp
 show ?case
   using safe-TwoSum-correct2[OF \langle Finite f \rangle \ an - xy] \langle Finite x \rangle
     safe-TwoSum-correct1[OF xy]
   by (auto simp: r split: prod.split)
next
case (in-between a e es f r-full)
 note in-between.prems(1)[simplified, unfolded bind-eq-Some-conv, simplified]
 then obtain l r where goal1: safe-grow-mpf-rec (e, es) f = Some (l, r)
   and r1: safe-TwoSum l a \gg (\lambda(x, y)). Some (x, y \# r) = Some r-full
     by blast
 then obtain l2 r2 where l2: safe-TwoSum l a = Some (l2, r2) and
    r2: (l2, r2 \# r) = r-full
    using r1 [unfolded bind-eq-Some-conv, simplified] by auto
 then have Val\text{-}mpf \ r\text{-}full = Val\text{-}mpf \ (l2, r2 \# r) by simp
 also have ... = Val \ l2 + Val-mpf \ (r2, r)
   by (simp add: rec-val)
 also have ... = Val \ l2 + Val \ r2 + listsum(map \ Val \ r)
   by simp
 also have \dots = Val \ l + Val \ a + listsum(map \ Val \ r)
   proof -
     from in-between.prems have Finite l
      using goal1 preserve-finite by auto
     moreover have Finite a
      using in-between.prems(3) by simp
     moreover have Finite(l + a)
      using l2 safe-TwoSum-correct1 safe-TwoSum-finite1 by auto
     moreover have Val\ l + Val\ a = Val\ l2 + Val\ r2
      using safe-TwoSum-correct2[OF\ calculation\ l2].
     ultimately show ?thesis
      by simp
   qed
 finally show ?case
```

```
\begin{array}{c} \textbf{using} \ \textit{in-between goal1 rec-finite} \ \textbf{by} \ \textit{auto} \\ \textbf{qed} \end{array}
```

Note that the proof tactics for *preserve-finite* and *preserve-val* are very similar (identical up to the **obtain** commands in both induction cases). They could be combined by stating

```
shows preserve-finite: Finite-mpf r
and preserve-val: Val-mpf r = Val-mpf mpf + Val x
```

in a single lemma with the same assumptions. However, to actually remove redundancy in the proofs, both goals would have to be combined again via

```
unfolding atomize-conj
```

Since the second result depends partly on the first one, many fact names (or alternatively: large HOL predicates combining unrelated facts) would accumulate during the proof. To maintain readability, we stick to the two-proof-solution. Instead of stating both goals in one lemma, we collect the results afterwards:

```
\begin{array}{l} \textbf{lemmas} \ safe\text{-}grow\text{-}mpf\text{-}correct = \\ preserve\text{-}finite \\ preserve\text{-}val \end{array}
```

2.4.1 Tail recursive version

We can also implement grow-mpf in a tail-recursive way. For simplicity, we drop the overflow-check via the *option* monad for now.

```
fun grow\text{-}mpf\text{-}it :: float \ list \Rightarrow float \ list \Rightarrow mpf where grow\text{-}mpf\text{-}it \ [] \ f\ hs = (f,\ hs) \ | grow\text{-}mpf\text{-}it \ (e \ \# \ es) \ f\ hs = (let (x,\ y) = TwoSum\ f\ e in\ grow\text{-}mpf\text{-}it \ es\ x\ (y\ \# \ hs))
```

This transformation was comparably easy because grow-mpf only needs one linear pass as the graphic shows.

```
fun grow-mpf-tr :: mpf \Rightarrow float \Rightarrow mpf where grow-mpf-tr (a, es) f = (let (a', es') = grow-mpf-it (rev es) f []; (x, y) = TwoSum a' a in <math>(x, y \# es')
```

An interesting realisation is that the list recursion can instead be implemented using fold:

```
fun grow-mpf-step :: float \Rightarrow mpf \Rightarrow mpf where grow-mpf-step f (a, es) = (let (x, y) = TwoSum \ a \ f in (x, y \# es))
```

```
fun grow-by-fold :: mpf \Rightarrow float \Rightarrow mpf where
 grow-by-fold (a, es) f = foldr grow-mpf-step (a \# es) (f, [])
```

We prepare an equivalence proof for grow-mpf-tr and grow-by-fold by providing some lemmas about grow-mpf-tr and op @.

```
lemma grow-it-append-accumulator:
 grow-mpf-it as f (hs @ hs') = (let
    (a, es) = grow-mpf-it \ as f \ hs
   in (a, es @ hs'))
 apply (induction as arbitrary: f hs hs')
 apply simp-all
 apply (metis (no-types, lifting) Cons-eq-appendI case-prod-beta)
 done
lemma grow-it-append-expansion:
 grow-mpf-it (as @ es) fhs = (let
   (a', es') = grow-mpf-it \ as f \ hs
 in grow-mpf-it es a' es')
 apply (induction as arbitrary: f hs)
 by (simp-all add: prod.case-eq-if)
Its effect on the wrapper grow-mpf-tr is
lemma grow-append-rev:
 grow-mpf-tr(a, es @ es') f = (let
   (a'', es'') = grow-mpf-it (rev es') f [];
   (a', es') = grow-mpf-it (rev es) a'' es'';
   (x, y) = TwoSum a' a
   in (x, y \# es'))
   by (simp add: case-prod-beta grow-it-append-expansion)
In case of an increase by a singleton, this can be simplified:
lemma grow-snoc-rev:
 (grow-mpf-tr\ (a,\ es\ @\ [h])\ f)=(let
    (x, y) = TwoSum f h;
```

```
(a', es') = grow-mpf-it (rev es) x [y];
   (x', y') = TwoSum a' a
 in (x', y' \# es'))
unfolding grow-append-rev[of a es [h] f]
apply simp
by (simp add: split-def)
```

The right part of the equation can also be written using grow-mpf-tr:

```
lemma gm\text{-}snoc1: (grow\text{-}mpf\text{-}tr\ (a,\ es\ @\ [h])\ f)=(let
       (x, y) = TwoSum f h;
       (a', es') = grow-mpf-tr(a, es) x
      in (a', es' @ [y]))
    by (induction es arbitrary: a) (simp-all add: case-prod-beta grow-it-append-expansion)
```

We expect compilers that optimize tail recursion to also optimize *foldr*. Thus, it is no longer necessary to make the functions tail recursive if they can be expressed by fold.

2.5 Further Operations

With

```
fun mpf-neg :: mpf ⇒ mpf where
  mpf-neg (a, es) = (float-neg a, map float-neg es)

lemma valid-zero-mpf:
  shows valid Plus-zero-mpf
  and valid Minus-zero-mpf
by (simp-all add: Plus-zero-mpf-def Minus-zero-mpf-def float-zero1 float-zero2)
```

One way to inspect which computations will be performed is to define a test mpf with dummy values and then use Isabelle's simplifier to apply the methods simplifications to the desired point:

```
definition a_4 = undefined
definition a_3 = undefined
definition a_2 = undefined
definition a_1 = undefined
definition a_0 = undefined
definition test-mpf = (a_4, [a_3, a_2, a_1])
definition output = grow-by-fold\ test-mpf\ a_0
```

To make the simplifier apply the definition, we need to state a lemma:

```
lemma P output unfolding output-def test-mpf-def grow-by-fold.simps
— We can now use various proof methods to get a neatly arranged output:
apply (clarsimp split: prod.splits) oops
```

where P is an undefined dummy predicate. At the last step, the output is as follows:

To demonstrate the TwoSum sequence carried out by grow-mpf, we use the following graphic:

```
value approx output delivers
```

Plus- $zero \oplus One \oplus undefined \oplus undefined \oplus undefined \oplus undefined \oplus undefined <math>\oplus undefined$:: IEEE.float

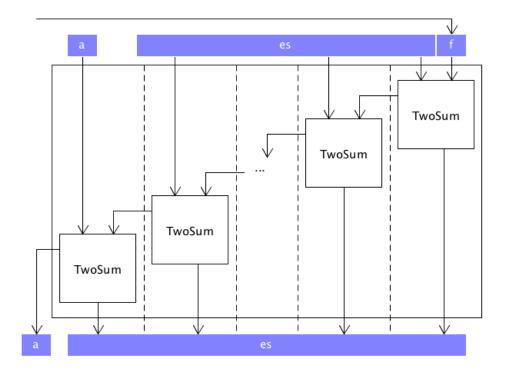


Figure 2.1: The float f is added to the mpf (a, es). TwoSum is represented by a box where the larger value, x, is output to the left side and the smaller one, y, to the bottom. On the top, the function call is passed on. The returned mpf (bottom) is built from right to left.

3 Code generation

3.1 Use of SML floats

To enable computation for hardware floats, **theory** Code-Float[11] provides the built-in operators of the target language:

```
code-printing constant op / :: float \Rightarrow float \Rightarrow float \rightarrow (SML) Real.' / ((-), (-)) and (OCaml) Pervasives.('/.) declare divide-float-def [code del]
```

The other operations are defined analogously.

Even ML's comparisons can be used (ML's bool is already defined as translation for HOL's bool in the Main theory HOL):

```
code-printing constant Orderings.less :: float \Rightarrow float \Rightarrow bool \rightarrow (SML) Real. < ((-), (-)) and (<math>OCaml) Pervasives. (<) declare less-eq-float-def[code\ del]
```

3.2 Printing Floats

If we decide that an unchecked code module is safe enough for us, we can use the format Float.float[4] from Isabelle's HOL-library.

To enable the conversion from *IEEE.float* to *Float.float* in the generated code, we first insert the possibility to produce them from integers:

```
definition float-of-int i = Float (real-of-int i)
context includes integer.lifting begin
lift-definition float-of-integer::integer \Rightarrow float is float-of-int.
end
lemma float-of-int[code]:
  float-of-int i = float-of-integer (integer-of-int i)
  by (simp add: float-of-integer-def)

code-printing
  constant float-of-integer :: integer \Rightarrow float \rightharpoonup (SML) Real.fromInt
declare [[code drop: float-of-integer]]

Then, the conversion is possible:

— convert hardware floats to Float.float for an exact representation
code-printing
code-module ToManExp \rightharpoonup (SML)
```

3 Code generation

```
\forall fun\ tomanexp\ x =
  let
    val \{man = m, exp = e\} = Real.toManExp x;
   val \ p = Math.pow \ (2.0, 53.0);
   val \ ms = m * p;
   val \ mi = Real.floor \ ms;
    val\ ei = op\ Int.-(e, 53);
  in (mi, ei)
  end
consts tomanexp::float \Rightarrow integer * integer
code-printing constant tomanexp :: float \Rightarrow integer * integer \rightarrow
  (SML) tomanexp
definition toFloat::float \Rightarrow Float.float where
  toFloat \ x = (let \ (m, e) = tomanexp \ x \ in \ Float \ Float \ (int-of-integer \ m) \ (int-of-integer \ m)
e))
We can now define a test list:
definition list :: float list where
 - Note that floats with magnitude < 1 can only be defined via op div:
  list = [
   float-of-int 43,
   float-of-int 34538,
   float-of-int 3 / float-of-int 44,
   float-of-int \theta,
   float-of-int 0,
   float-of-int (-348976754389282980)]
To use the ML operators, we have to insert the transformation to a term and back:
instantiation float::term-of
definition term\text{-}of::float \Rightarrow term where term\text{-}of \ x = undefined
instance ..
end
code-printing
code-module FromManExp 
ightharpoonup (SML)
  \langle fun\ from manexp\ m\ e=Real.from ManExp\ \{man=Real.from LargeInt\ m,\ exp=e\} \rangle
consts from man exp::integer \Rightarrow integer \Rightarrow float
code-printing constant from manexp :: integer \Rightarrow integer \Rightarrow float \rightarrow
  (SML) frommanexp
definition of-Float::Float.float \Rightarrow float where
 of	ext{-}Float\ x = from manexp\ (integer	ext{-}of	ext{-}int\ (Float.mantissa\ x))\ (integer	ext{-}of	ext{-}int\ (Float.exponent)
lemma [code]: term\text{-}of\text{-}class.term\text{-}of (x::float) \equiv
  Code	ext{-}Evaluation. App
```

```
(Code-Evaluation.termify of-Float)
(term-of-class.term-of (normfloat (toFloat x)))
by (rule term-of-anything)

We can now print the list without an error:
value list

produces
[of-Float 43, of-Float (Float.Float 17269 1),
of-Float (Float.Float 1228254443828317 (- 54)),
of-Float (Float.Float 0 0), of-Float (Float.Float 0 0),
of-Float (Float.Float (- 5452761787332547) 6)]
:: float list

which is an error-free representation.

abbreviation toNF :: float \Rightarrow Float.float where
toNF \equiv normfloat \ o \ toFloat
```

4 Results and conclusion

The IEEE 754 floats were already modelled in Isabelle. Using this formalization, this work provides an easy way to use them for fast and error-free addition and subtraction. For these operations, we translated algorithms from the literature and adapted them for our purposes.

4.1 Impact

We give a more practice-oriented analysis of Shewchuk's algorithms and offer explanations for challenges that can arise when implementing them. We also give ideas and solutions for verifying them in a functional setting. Based on the existing formalization of IEEE-floats, we then specified a data format to provide an easy access to these algorithms. This means that users have a new option for a number format to perform verified computations using fast and error-free addition and subtraction. As results of our thorough testing of generated code, an error in polyML's float handling has been detected and removed. This means the code generated using the AFP-theory IEEE-Floating-Point/Code-Float has now a clearer semantics.

4.2 Future Work

A correctness proof for the *TwoSum* method needs to be converted to Isabelle's IEEE754 formalization. This will then also enable proofs for Shewchuk's "nonoverlapping" property, which, when implemented, allows more assertions about float expansions to be formally verified, e.g. about the maximum length of a valid *mpf*, or the quality of the approximation stored in the first component. Another improvement could be made by adapting code generation for IEEE-floats to support more of Isabelle's target languages. This will make our arithmetic library more flexible for use in languages than SML. However, the correct behaviour of floats in the language should be ensured beforehand, to avoid getting wrong results when using the generated code.

4 Results and conclusion

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