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Abstract

We present a new tool that generates bounds on the values and the round-off errors of programs using floating point operations. The tool is based on forward error analysis and interval arithmetic. The novelty of our tool is that it produces a formal proof of the bounds that can be checked independently using an automatic proof checker such as Coq and a complete model of floating point arithmetic. For the first time ever, we can easily certify that simple numerical programs such as the ones usually found in real time applications do not overflow and that round-off errors are below acceptable thresholds. Such level of quality should be compulsory on safety critical applications. As our tool is easy to handle, it could also be used for many pieces of software.

Keywords: Round-off error, Overflow, Formal proof, Certification, Safety critical.

Résumé

Nous présentons un nouveau outil capable de générer, à partir de la description d’un programme qui utilise des opérations à virgule flottante, des bornes sur les valeurs des variables et les erreurs d’arrondi commises dans ce programme. L’outil est basé sur l’analyse d’erreur en mode direct et l’arithmétique d’intervalles. La nouveauté de notre outil est qu’il produit une preuve formelle des bornes qui peut être certifiée indépendamment en utilisant un outil automatique de preuve tel que Coq et un modèle complet de l’arithmétique à virgule flottante. Pour la première fois, il est possible de facilement certifier que les programmes numériques simples comme ceux habituellement utilisés dans les applications en temps réel ne commettent pas de dépassement de capacité et que les erreurs d’arrondi sont au-dessous de seuils acceptables. Un tel niveau de qualité devrait être obligatoire sur les applications critiques. Comme notre outil est facile à manipuler, ce niveau devrait également être utilisé pour de nombreux logiciels.

Mots-clés: Erreur d’arrondi, Dépassement de capacité, Preuve formelle, Certification, Logiciels critiques.
Generating formally certified bounds on values and round-off errors*

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1 Introduction

The purpose of this work is to safely qualify the behavior of algorithms concerning the range of all the floating point variables and all the round-off errors. As this work was intended for safety critical applications, we generate a formal proof that is automatically verified by an automatic proof checker such as Coq [13]. However, as our tool is very easy to handle, it can be used for more or less safety critical applications, to provide formally certified bounds on the values and the round-off errors and guarantee that all variables and all round-off errors in a given program are under control.

Deadly and disastrous failures [14, 19, 9] have shown that common professional practices do not guarantee quality of the produced software. Pen-and-paper hand-written formal proofs are not sufficient as errors are commonly printed [17, 29] even in highly regarded publications by honored scientists in well reviewed and cited journals [27, 11]. Results produced by interval libraries, including for example [23, 13], are also not sufficient for safety critical applications. Some hand-written pen-and-paper proofs contain errors and there is no doubt that some highly regarded hand-written programs also contain errors even if they have been well reviewed and produced by honored scientists. In both cases, deriving the conclusions is not sufficient and the process should be certified by an automatic proof checker.

Such certifications have already been used to detect or prevent errors in algorithms and implementations [1]. Few results are available for computer arithmetic [20, 2] and even fewer results consider the effect of round-off errors in floating point arithmetic. Our contacts with leading teams at Intel and NASA LaRC allow us to safely believe that at the time we are reporting on this work, all developments that follow round-off errors are linked to one of the authors of this report [18, 4]. As certifying formal proofs by hand can be extremely tedious and needs specialized training, only carefully chosen applications have been developed so far.

Our tool has been developed to simplify such a task. Given a description of the application, it automatically generates proofs for the various bounds on values and round-off errors. As a result, it can be easily handled by naive users with no special training in Coq or any other formal proof verification tool. Our tool can be linked to many automatic proof checkers. The ones based on higher order logic such as Coq, PVS and HOL are probably the most conservative ones among them; Coq and HOL [10] are probably the safer ones as they rely solely on a very small kernel. Future development of our tool may let users check their proofs on two or three independent proof checkers and get a level of confidence higher than any guarantee provided so far.

It is visible in the following text that some decisions were still open for the implementation of an automatic tool such as ours. We propose educated choices that are useful in regard to common

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practices and to our first application that is used to maintain safety distances between aircrafts. Our main contribution is still our ability to produce forward error analysis proofs based on interval arithmetic that are certified by an automatic proof checker such as Coq as there is no other tool able to do it. Our choices will make a difference when comparable tools will appear in the future.

There might be many reasons for a computing system to fail. The first one might be that some hardware, the operating system or the compiler is not working correctly. Another one might be that the user did not handle the software correctly or that the designer did not understand exactly the intended behavior of the system. Yet another one might be that one programmed structure is not initialized or used correctly, and so on. All these problems are not new with floating point arithmetic and teams around the world are developing tools and methodologies to prevent them.

Failure may also be caused in the substitution of a mathematical question by a finite number of operations. For example when we truncate an infinite series or when we replace the mathematical solution of an equation by a constant number of steps of Newton-Raphson iterations [24, 26]. These errors are also handled elsewhere. Moreover, the initial mathematical question is not necessarily visible in the final compiled program. Indeed it is not always possible to get back to the initial algorithm knowing its implementation only.

Our tool bounds the gap between the behavior of the given program executed on real numbers, with no round-off error, and the behavior of the same program executed on a computer, with round-off errors. Although these bounds are very useful in practice, it is sometimes also interesting to bound the global effect of all errors on the computed values. This is possible with our tool by adding uncertainty to some constants and data of the program. The user is responsible to certify that the modified intervals appropriately account for all truncation and discretization errors of the program compared to the mathematical question. We did so for our aeronautical application and the additional error bounds will soon be formally certified with PVS.

2 Setting up a formal model adapted to automatic proof checkers

In the following, \( x \in \mathbb{R} \) denotes a value \( x \) known with an infinite precision. This is a theoretical number and an algorithm uses an approximated representable value \( \hat{x} \in \mathbb{F} \), with \( \mathbb{F} \) being the set of hardware-representable numbers. We do not mean that \( \hat{x} \) is the rounded value of \( x \), we mean that \( \hat{x} \) is the value that is used and possibly computed by the program in place of \( x \).

A closed mathematical function \( f(x) \) is implemented by a program \( \hat{f}(\hat{x}) \) that takes into account the round-off errors caused by the computation. Discretization and truncation errors have to be handled separately. We use \( \hat{x} \) instead of \( x \) to recall the reader that a program always uses only representable inputs.

The tool has been designed to automatically generate a proof of the range of \( \hat{f}(\hat{x}) \), the difference between the exact value \( f(x) \) and the computed value \( \hat{f}(\hat{x}) \), and so on, when some properties are known on the input variables. For example, under the hypothesis \( \hat{x} \in [1, 4] \), it is easy to prove that \( \sqrt{n} \) lies in the range \([1, 2] \). The tool is here to help certifying proofs on complete applications.

The analysis is done on a low-level description of the application on which arithmetic properties have to be proved. When two variables are summed, our tools do not simply do an addition, it generates proofs about the implemented addition that a computer will effectively execute, for example a floating-point addition on simple precision numbers with respect to the IEEE-754 standard and its default rounding mode as defined in our formal specification. The description of the program must carry enough information in order for the proofs to relate to the real program as implemented and executed.

The tool also takes into account some properties (range, absolute error, relative error) on the input variables. Given the same kind of properties on the output variables, the tool tries to generate a script of proof that states and verifies these properties from the hypotheses on the input variables. Section 2.1 describes one of the mechanisms involved in proving range properties. Section 2.2 is devoted to the proofs of errors, both absolute and relative.
A proof-checker like Coq does not work on the actual proof but on a script of tactics that describes each necessary logic operation to validate the conclusion from the premises. The proof is generated by the automatic proof checker by playing the script. Coq kernel stores the proof as a lambda-terms following Curry-Howard isomorphism. The lambda-terms cannot be read by human beings but it can be double checked by tools independant from Coq kernel and Coq development teams. As it stands now, none of the files manipulated by Coq stores the proof in a human readable format. This is the reason why we always refer to the proof script that is played on the automatic proof checker to print on screen the steps of the proof.

2.1 Bounds on the computed values

Variables may be bounded into explicitly computed ranges certified by an automatic proof checker. For example, a responsible programmer should prove that inputs of all square root operations cannot be negative and that his program never divides by zero. Range checking also let us verify that no result will overflow as floating point algorithms that are not specially built to deal with them will often lead to unanticipated results if one or more operations overflow. Worse, an overflowing operation on non saturated two’s complement fixed point arithmetic will return a wrapped value with the wrong sign.

Ranges are stored as pairs of numbers: a lower bound and an upper bound\(^1\). Such a representation is not optimal and authors may use other ones in the future. It is however sufficient in most cases and as a trade-off, we avoid manipulating complex structures. Indeed, formally checking a proof is a computation intensive operation and using simpler properties (a pair of numbers instead of the whole description of a domain) makes verifications easier.

What is interesting for a proof is the fundamental property of interval arithmetic: if the arguments of an expression are replaced by enclosing intervals, the expression evaluation will lead to an interval that contains the result. Such an inclusion property is exactly what is needed for a proof: given the ranges of the inputs, interval arithmetic will give the range of the output — or more generally an overestimation of this range, but it still is a compatible answer.

Our tool may have handled domains on variables \(x\) but our past experiences on programs lead us to rather bound computed values. The upper and lower bounds will be used in pre- and post-conditions of implemented functions. When a C program contains an assertion \((0.0 \leq x)\), \(x\) refers to the value actually in memory or in a register; consequently the assertion does not apply to \(x\) but to \(\tilde{x}\). When a variable must be proved not to underflow or overflow, we mean to check the property on \(\tilde{x}\), and so on. Most of the properties proved at this stage require knowing the domain of \(\tilde{x}\); the domain of \(x\) may never be needed though it can be estimated with the error bound defined Section 2.2.

Our educated decision to bound the approximated variable rather than the exact quantities has another important consequence: there is no loss if the lower and upper bounds of \(\tilde{x}\) are in \(\mathbb{F}\) as the approximated value \(\tilde{x}\) is an element of \(\mathbb{F}\) by definition and we do not consider intervals of \(\mathbb{R}\) but of \(\mathbb{F}\). The consequence on formal proof checking is substantial, bounds on the exact values could need arbitrarily wide expressions to exactly represent real numbers. On the other hand, bounds on \(\tilde{x}\) are rationals with a very limited precision (for example IEEE standard single or double precision numbers).

Common interval arithmetic considers convex subsets of \(\mathbb{R}\). Consequently, the implementation of the addition operator for a traditional interval arithmetic library is:

\[
[\tilde{a}, \tilde{b}] + [\tilde{c}, \tilde{d}] = [\bigvee(\tilde{a} + \tilde{c}), \bigtriangleup(\tilde{b} + \tilde{d})].
\]

There \(\bigvee\) and \(\bigtriangleup\) denote that operations should be rounded toward \(-\infty\) and \(+\infty\). The bounds may not carry enough precision for \(\tilde{a} + \tilde{c}\) and \(\tilde{b} + \tilde{d}\) to be exactly stored. It is the reason why the two operations are rounded so that the resulting interval encloses all possible values.

\(^1\)We will present in footnotes references to Coq definitions and properties included in annexes. The annexes have been generated through an automatic pretty-printer and they are faithful to the actual scripts automatically validated by Coq. The name of each annex section is the name of the file as it appears in our library. This first definition is given by \texttt{Eint_bound} in the Coq support library.
In our cases, such rounding considerations are not necessary: the hardware-implemented addition $\boxplus$ is considered as a new exact operation with respect to $\mathbb{F}$ and not as the approximation of the addition on the field of the real numbers. That means that if

$$\bar{x} \in [\bar{a}, \bar{b}] \quad \text{and} \quad \bar{y} \in [\bar{c}, \bar{d}] \quad \text{then} \quad x \boxplus y \in [\bar{a} \boxplus \bar{c}, \bar{b} \boxplus \bar{d}]$$

Indeed, the values $\bar{a} \boxplus \bar{c}$ and $\bar{b} \boxplus \bar{d}$ already lie in $\mathbb{F}$, there is no need for an additional rounding. Consequently, since the addition $\boxplus$ is a monotonous function of $\mathbb{F}^2 \rightarrow \mathbb{F}$ whatever the rounding mode, then $\bar{a} \boxplus \bar{c}$ and $\bar{b} \boxplus \bar{d}$ are the bounds.

$$[\bar{a}, \bar{b}] \boxplus [\bar{c}, \bar{d}] = [\bar{a} \boxplus \bar{c}, \bar{b} \boxplus \bar{d}]$$

There is no need to take instead possibly different and non-optimal values $\vee(\bar{a} + \bar{c})$ and $\nabla(\bar{b} + \bar{d})$.

Consequently, the most commonly used interval arithmetic libraries were not appropriate for our project. We chose the Boost parameterizable interval arithmetic library [5]. This is a C++ library that allows to specify the types and the arithmetic to be used instead of being limited to arithmetic on $\mathbb{R}$.

2.2 Bounds on the errors

The other arithmetic property that our tool proves is the difference between the exact result and the computed value. Finite precision of numeric computations is likely to induce a difference between the expected result and the computed value. Such differences will progressively build up and it is necessary for the user to certify that the computed result is meaningful.

Bounds on values and errors are quite similar and we could have decided to handle them at the same time. Each variable would have been represented by four numbers (two ranges) and there would have been new operators inspired by interval arithmetic that would compute on these quartets. Such a representation would have given shorter proof scripts. However there are situations where such a simplicity would be restrictive.

For example, the error range for the expression $x / \sqrt{1 + x^2}$ would be easily computed by this method, but the domain of the result would not. Indeed the result is in the range $[-1, 1]$ for any $x$ [16], and this tight interval may be a precondition for another part of the algorithm. Yet, interval arithmetic is usually not able to yield such a small interval: multiple occurrences of the same variable in the expression leads to an overestimation of the result usually called correlation.

Interval arithmetic is unable to keep track of the relations between sub-expressions and $x / \sqrt{1 + x^2}$ is seen as if it was $x / \sqrt{1 + y^2}$, with $y$ having the same domain as $x$ but not necessarily the same value. Consequently, in this example, if $x$ is in the range $[0, 2]$, $\sqrt{1 + x^2}$ is in the range $[1, \sqrt{5}]$. Standard interval arithmetic will then answer that the whole expression is in $[0, 2] / [1, \sqrt{5}] = [0, 2]$. This interval is no subset of the range $[-1, 1]$ as another part of the algorithm could expect. Standard interval arithmetic cannot be used to validate the domain of this expression and human interaction is needed instead.

Consequently, it may be preferable for bounds and errors proofs to be separated so that if our tool is not able to prove one of them it may still be used for the other one limiting human interactions as much as possible.

2.2.1 Absolute error

The most commonly encountered kind of error representations are absolute and relative errors [7]. Absolute error is useful with fixed point arithmetic since the magnitude of the rounding error does not depend on the magnitude of the computed value. It may also be useful with floating point arithmetic when a subtraction cancels, when the exponents of the results stay close or with subnormal numbers (including zero). Relative errors can be used in all the other cases.

\footnote{This result is given by \texttt{Eplus2Rec bound} for generic floating point numbers.}
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Absolute error$^3$ $\eta_x$ is attached to variable $x$ by equality

$$x = \hat{x} + \eta_x.$$  

We will not define the behavior of all the implemented operators as forward error analysis is very common. As an example, here is the formula giving the absolute error for the addition. $\eta_0$ is the rounding error between the exact result of $\hat{x} + \hat{y}$ and its computed value $\hat{x} \oplus \hat{y}$.

$$z - \hat{z} = (x + y) - (\hat{x} \oplus \hat{y})$$

$$= (x + y) - (\hat{x} + \hat{y} - \eta_0)$$

$$= (x - \hat{x}) + (y - \hat{y}) + \eta_0$$

$$= \eta_x + \eta_y + \eta_0.$$  

We extend the definition to interval bounds. Intervals $A_x$ and $A_y$ are attached to variables $x$ and $y$ and respectively enclose $\eta_x$ and $\eta_y$. Interval $A_0$ encloses all possible values for $\eta_0$. This interval is given by the properties of the underlying arithmetic. For fixed point arithmetic, it is a constant interval. For floating point arithmetic, it depends on the value of $\hat{z}$. Our tool detects automatically Sterbenz’s conditions and proves that $A_0$ is equal to zero if the theorem is verified for all $\hat{x}$ and $\hat{y}$ [28, 3].

$$\eta_z \in A_z \subseteq A_x + A_y + A_0$$

Bounds on values of Section 2.1 were not real numbers and the precision of interval arithmetic was very limited. It is not the case for error intervals and we used common interval arithmetic with rounded overestimations.

2.2.2 Relative error

We only consider relative error$^5$ when $x$ and $\hat{x}$ have the same sign. The most common representation of this error is $\hat{x} = x \times (1 + \epsilon_1)$ but another possible one is $x = \hat{x} \times (1 + \epsilon_2)$ [12].

In order to select one of these two representations, let us consider the case where $\hat{x}$ is one representable number closest to $x$ (there is usually only one $\hat{x}$ value unless $x$ is a mid-point between two representable values). Provided we are sufficiently far from the underflow threshold,

$$-\frac{u}{2} \leq -\frac{u}{2 + u} \leq \epsilon_1 \leq \frac{u}{2 + u} \leq \frac{u}{2} \quad \text{and}$$

$$-\frac{u}{2} \leq -\frac{u}{2(1 + u)} \leq \epsilon_2 \leq \frac{u}{2} \leq \frac{u}{2},$$

where $u$ is the difference between 1 and the next representable floating point number. This bound is important because it is used after each operation to incorporate the additional rounding error.

Now if $\hat{x}$ is the smallest positive normalized number,

$$-\frac{u}{2} \leq -\frac{u}{2 + u} \leq \epsilon_1 \leq \frac{u}{2 - u} \leq \frac{u}{2} \quad \text{and}$$

$$-\frac{u}{2} \leq -\frac{u}{2} \leq \epsilon_2 \leq \frac{u}{2} \leq \frac{u}{2}.$$  

As a conclusion using $x = \hat{x} \times (1 + \epsilon_x)$ gives a range valid on all normalized numbers and consequently the exponent of $\hat{x}$ needs to be checked. Once again, we provide only one

$^3$This definition is given in Coq by \texttt{Einabs}, and by \texttt{Einerror} for symmetric interval.

$^4$This result is given in the general case by \texttt{Eplus_error} and for floating point numbers by \texttt{Eplus_fdec_error} and \texttt{Eplus_exact_fdec_error} depending on whether Sterbenz’s conditions on exponents are verified.

$^5$This definition is given by \texttt{Einrelat}.
example as forward error analysis is very common. For the multiplication \( z = x \times y \), if \( \hat{z} \) is always normalized, the relative error is:

\[
\frac{z}{\hat{z}} = \frac{x \times y}{\hat{x} \hat{y}} = \frac{x \times y}{\hat{x} \times y} (1 + \epsilon_x) = (1 + \epsilon_x)(1 + \epsilon_y)(1 + \epsilon_0).
\]

Once again, we attach intervals \( R_x \) and \( R_y \) to respective variables \( x \) and \( y \) enclosing \( \epsilon_x \) and \( \epsilon_y \). Let interval \( R_0 \) enclose the value of the relative error \( \epsilon_0 \). Its range depends on the properties of the arithmetic and we have seen that in the common cases of floating point arithmetic rounding to nearest, it is \([-u/2, u/2]\) and finally

\[
\epsilon_z \in R_z \subseteq (1 + R_x)(1 + R_y)(1 + R_0) - 1.
\]

All the other cases are handled in our library. By using automatic proof checkers, we guarantee that no case has been forgotten or inappropriately handled.

### 2.2.3 Combining absolute and relative errors

There may be cases where both an interval \( A \) for the absolute error and an interval \( R \) for the relative error can be computed and certified for a given variable \( x \). Since the two intervals are linked by \( \eta_x = \hat{x}\epsilon_x \), if \( I \) is an interval enclosing \( \hat{x} \), then \( A \div I \) is also an interval that encloses the relative error, and \( R \times I \) encloses the absolute error. So starting from \( A \) and \( R \), we can build two sequences of intervals.

\[
\begin{align*}
A_0 &= A & \text{and} & & A_{n+1} &= A_n \cap (R_n \times I) \\
R_0 &= R & \text{and} & & R_{n+1} &= R_n \cap (A_n \div I)
\end{align*}
\]

Such a result would not be useful if a lot of iterations were necessary, since the proof script would become huge. Fortunately, it can be proved that the sequences are stationary after \( A_1 \) and \( R_1 \). So, if both \( A \) and \( R \) are available, possibly tighter intervals can easily be computed.

### 2.3 Formal proof checking

Our tool generates a Coq script of the proof. Such a proof script does not live by itself: it only describes what elementary facts should be applied, it does not provide the proofs of these theorems. Indeed these facts are provided by various libraries: theorems on real numbers arithmetic (the Coq standard library), specification and basic properties of floating point arithmetic [6]. It also depends on a library about interval arithmetic and its floating point extensions we developed so that each generated proof could be formally verified. For example, in order to use interval arithmetic for the addition, this composition property is needed:

\[
\hat{a} \oplus \hat{c} \leq \hat{b} \oplus \hat{d} \quad \text{if} \quad \hat{a} \leq \hat{b} \quad \text{and} \quad \hat{c} \leq \hat{d}.
\]

Non trivial theorems like Sterbenz’s one are also stated [3], extended to floating point intervals in our library, and used by our tool.

On the other hand, our tool is not a part of Coq program and it can also be used with other provers such as HOL Light or PVS with ProofITootnote{http://research.ni.net/~munoz/ProofIT/} package that supports batch proving and proof scripting in PVS. To preserve the independence of our tool, we do not use any technique specific to one prover such as tactics and automatic proof developments. Adapting our tool to other provers

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\[\text{This result is given by } \texttt{Enult\_relat} \text{ in the general case.}\]
will only be a matter of syntax as soon as libraries for computer arithmetic are available on any alternate prover.

We benefit from the fact our tool is not a part of any prover and consequently that it does not need to be proved. For example, when proving the property \( \hat{a} \leq \hat{b} \mp \hat{c} \), the tools acts as an oracle and computes beforehand the sum \( \hat{d} = \hat{b} \mp \hat{c} \) and provides this number \( \hat{d} \) to the proof checker. Consequently, the checker does not have to explicitly compute \( \hat{b} \mp \hat{c} \) in order to prove the property, it only has to verify that \( \hat{d} \) as provided by the tool is the closest\(^8\) (with respect to the rounding mode of \( \mp \)) floating point number to \( \hat{b} + \hat{c} \), and then that \( \hat{a} \leq \hat{d} \). Both operations can be done by the existing Coq floating point library. Using the tool as an oracle avoids having to implement in the checker and prove all the floating point operations with their various rounding.

This solution works fine for the addition and the multiplication as such operations are internal on dyadic numbers and they are provided by the library. Dyadic numbers are pairs \((m, e)\) of \(\mathbb{Z}^2\) interpreted as \(m \times 2^e\). Radix 2 floating point numbers are dyadic numbers and our library relies on these. However mathematical division and square root of dyadic numbers do not generally produce dyadic numbers. Consequently these operators on real numbers cannot be directly used on dyadic numbers and the proofs cannot rely on them. Fortunately these operations can be reverted when rounding to nearest. For example,

\[
\hat{c} = o(\hat{a} \div \hat{b}) \quad \text{iff} \quad \hat{b} \times \hat{c}^- \text{ is closer from } \hat{a} \text{ than } \hat{b} \times \hat{c}^+ \text{ and } \hat{b} \times \hat{c}^+
\]

where \(\hat{c}^-\) and \(\hat{c}^+\) are the floating point numbers respectively just below and just above \(\hat{c}\).

Error bounds are also represented on dyadic numbers from the library. Our decision certainly reduces precision, but the situation is sensible in regard to common practices and it is also fully parameterizable. Moreover, we decided to further limit precision of the error bounds. Since our tool is much faster than any automatic proof checker, it tries to simplify the numbers used in the proof as much as possible. Without any simplification, the length of the error bounds grows continually. So does the time needed to verify each statement of the script as the proof goes on.

For example, we may replace \(m \times 2^e\) by \(\lfloor m \cdot 2^{-k} \rfloor \times 2^{e+k}\) for a given \(k\). The bound is larger but certifying it is faster as \(\lfloor m \cdot 2^{-k} \rfloor\) is a smaller constant. Our tool enlarges all the error ranges as long as they remain smaller than the ones the tool tries to prove. It is an efficient way to reduce the script size and the certification time since the main part of an error is generally its magnitude, not its exact value. For example, certifying the simplified proof was four times faster on the application of Section 3.

### 3 Maintaining safety distances between aircrafts

There is a lot of work done on shifting responsibility for aircraft safety distances from air traffic controllers to pilots themselves. This involves providing pilots with conflict detection systems. One algorithm [22] works in an Euclidean three-dimensional neighborhood and a translation is necessary from geodetic coordinates as given by positioning systems [21] to Euclidean coordinates.

Airplanes are protected by a cylinder (typically 5 miles diameter and 1000 feet height) and a conflict is defined as the overlapping of two protected zones. The above mentioned work has certified that it is equivalent to detect conflicts relative to the intruder. In this case, the intruder has a protected zone of radius 5 miles and height of 2000 feet. The ownership is just a point. A conflict is then defined as the inclusion of the ownership in the protected zone of the intruder. The purpose of conflict detection is to detect such intrusions ahead of time and to propose escape and recovery trajectories.

A lot of approximations are necessary to solve this problem: position and speed of the planes are not exactly known, the use of an Euclidean neighborhood instead of geodetic coordinates creates modeling errors, translation itself incurs modeling and round-off errors. All these errors need to be taken into account so that the safety of the systems is guaranteed. The cylinder is

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\(^8\) This property is verified by the boolean function `evaluate_even_closest` for rounding to nearest and tie-breaking to even.
easily inflated so that fewer and fewer trajectories may cross the initial safety cylinders. Yet, it is life critical to guarantee that the cylinder has been sufficiently inflated to remove any incorrect behavior.

As mentioned, there already is a published algorithm that works in an Euclidean three-dimensional neighborhood. We will describe here a new algorithm responsible of the translation part and use it as an example of safety critical application that can be formally proved thanks to our tool.

The inputs of the Algorithm 1 are the longitudes $\lambda_{1,2}$ and the geodetic latitudes $\phi_{1,2}$ of planes 1 and 2, and their speeds toward north $v_{N1,2}$, toward east $v_{E1,2}$. The outputs are the relative coordinates and the speed of both planes in an Euclidean neighborhood. $t_r$ is a constant. $R_m$ and $R_p$ are Earth radius along a meridian and along a parallel for a given latitude:

$$R_p(\phi) = \frac{a}{1 + (1 - f)^2 \tan^2 \phi}$$

$$R_m(\phi) = \frac{a(1 - e^2)}{(1 - e^2 \sin^2 \phi)^{3/2}}$$

Constants $a$, $e$ and $f$ are described by WGS84\(^6\) and are also used to define the geodetic coordinates in the positioning systems.

**Algorithm 1** Translation from geodetic coordinates to an Euclidean neighborhood

**Input:** $\phi_1$, $\phi_2$, $\lambda_1$, $\lambda_2$, $v_{N1}$, $v_{N2}$, $v_{E1}$, $v_{E2}$

- $\phi_0 \leftarrow (\phi_1 + \phi_2)/2$
- $r_{p0} \leftarrow R_p(\phi_0)$
- $s_{N1} \leftarrow v_{N1}/R_m(\phi_1)$
- $s_{N2} \leftarrow v_{N2}/R_m(\phi_2)$
- $s_x \leftarrow (\lambda_1 - \lambda_2) * r_{p0}$
- $s_y \leftarrow (\phi_1 - \phi_2) * R_m(\phi_0)$
- $v_{x1} \leftarrow v_{E1} * r_{p0}/R_p(\phi_1 + s_{N1} * t_r)$
- $v_{x2} \leftarrow v_{E2} * r_{p0}/R_p(\phi_2 + s_{N2} * t_r)$

These functions use trigonometric functions that must be approximated on computers. Therefore, we decided to approximate directly $R_m$ and $R_p$. This adds more errors as the algorithm uses

\(^6\)http://www.wgs84.com/.
polynomials $\hat{R}_m$ and $\hat{R}_p$ rather than $R_m$ and $R_p$. The program for $\hat{R}_p$ is given Algorithm 2. $\hat{R}_m$ is omitted but it is simple implementation of Horner’s rule. We introduce an uncertainty on the first coefficient of each polynomial to account for the difference between the exact function and the approximated polynomial in our tool.

**Algorithm 2** Definition of $R_p(\phi)$ approximating $R_p(\phi)$

**Input:** $\phi \{ all \ the \ dyadic \ numbers \ are \ floating \ point \ numbers \}$

\[
x \leftarrow 511225 \times 2^{-18} - \phi^2
\]

\[
y \leftarrow 4439091 \cdot 2^{-2} + x \times (9023647 \cdot 2^{-2} + x \times (1386737 \cdot 2^{-6} + x \times (13233647 \cdot 2^{-11} + x \times (-1898597 \cdot 2^{-14} + x \times (-6661427 \cdot 2^{-17}))))
\]

The accuracy of the results depends on the precision of the arithmetic. For this example, we used underlying IEEE-754 simple precision\(^\text{10}\) floating point arithmetic with rounding to nearest (with even tie-breaking) and subnormal numbers are implemented. The speed corresponds to an airliner and the latitudes range from 30°N to 40°N.

Our tool produced a 1650-lines long script. The generated proof guarantees that the absolute error on $s_x$ and $s_y$ is bounded by half a meter, and the error on $v_{x1}$ and $v_{x2}$ is bounded by $5 \times 10^{-4}$ m/s. We conclude that the safety cylinder radius should be increased by 2.60 m for a 300 seconds window.

The 1650 lines of the script can be decomposed between 850 lines of definitions and 800 lines of theorems and proofs. Only these 800 last lines are important with respect to computer arithmetic. The huge number of definitions is caused by the need to describe each operation and each number. For example, when proving a variable is in a given interval, the lower and upper bounds need to be defined separately first.

On the test-machine, the script is verified by Coq in 45 seconds. The verification is a lot slower than the generation itself (a split second). This gap is due to the way Coq handles integer arithmetic: integers are stored as lists of bits and the operations are computed one bit at a time (since it is the way integer arithmetic is proved in Coq standard library).

### 4 Conclusion and Perspectives

The tool that we have just presented is very limited. Yet it is complete enough to deal with our airplane safety distance example. The tool generates a script that can be formally verified by Coq and that proves bounds on the absolute error of the outputs. It can also easily generate a script for the case where the inputs are only known with a given limited accuracy.

Adding functionalities to our tool is time consuming. We have to state and prove the supporting theorems in our library. Today, our tool is limited to floating point arithmetic (rounded to nearest) and absolute error bounds. As there is no difficult lock, more and more functionalities will be incorporated such as fixed point arithmetic and additional arithmetic operations. We may also take into account some DSP specific implementation of floating point arithmetic. Yet it is always possible to state quickly a few unvalidated axioms to incorporate one specific function for setting up a demonstration.

Another perspective is a connexion to Why tool [8]. Why treats C and ML programs and generates proof obligations for automatic proof checkers. As no tool was available for floating point arithmetic, variables were treated as real numbers with no rounding error. We will cooperate with Why developers to extend it with the existing floating point arithmetic.

Finally, our tool only performs simple forward error analysis. It may be linked with other tools such as Fluctuat [25] that is able to propose contracting intervals for loops. Only the certification

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\(^{10} \)The specialization of generic floating point properties to this format is in the Coq module *Efloat\_s764ec*. 
step has to be validated by an automatic proof checker for the result to be fully guaranteed. One may use any tool to provide useful oracles as long as oracles are finally validated by the proofs produced by our tool.

5 Acknowledgment

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References


Module ClosestCalc

Add LoadPath "Float".
Require AllFloat.

Section ClosestCalc.
Variable ndix: Z.
Variable bound: Float.
Variable precision: nat.

Coercion Local FloRadix := (FloR radix).
Hypothesis ndixMoreThanOne := (Zlt (POS xH) ndix).

Local radixNotZero := (Zlt _1 O ? (Zlt _le_weak ? ? ndixMoreThanOne)).
Hypothesis precisionNotZero : ~ precision = O.
Hypothesis precisionMoreThanOne : (lt (1) precision).
Hypothesis pGivesBound : (POS (uNum bound)) = (Zpower_nat radix precision).

Inductive validity: Set :=
| bad : validity
| tie : validity
| good : validity.

Local dp := [x,r;float] [f;float->float]
(Fminus radix x (f (FPred bound radix precision r))).

Local dd := [x,r;float] [f;float->float]
(Fabs (Fminus radix (f r x)));

Local ds := [x,r;float] [f;float->float]
(Fminus radix (f (FSucc bound radix precision r)) x).

Definition evaluate_rounding := [x,r;float] [f;float->float]
let dd := (dd x r f) in
Cases (Fcompare ndix dd (dp x r f)) of
| INFERIEUR => bad
| EQUA => Cases (Fcompare ndix dd (ds x r f)) of
| INFERIEUR => bad
| EQUA => tie
| SUPERIEUR => tie
end

| SUPERIEUR :=
Cases (Fcompare ndix dd (ds x r f)) of
| INFERIEUR => bad
| EQUA => tie
| SUPERIEUR => good
end.

Definition evaluate_closest := [x,r;float]
Cases (evaluate_rounding x r [y;float|y]) of
| bad => false
| _ => true
end.

Lemma Rminus_le_r: (r1,r2;R) "0 ≤ r2 - r1" w "r1 ≤ r2".

Theorem evaluate_closest_correct: (x,r;float)
evaluate_closest x r -> true -> (Fcanonic radix bound r) ->
(Closest bound radix x r).

Definition is_positive_even := [x;positive]
Cases x of
| (xO _) => true
\[ a \Rightarrow \text{false} \]
end.

Definition \textit{is\_float\_even} := \[ x: \text{float} \]
Cases (\texttt{Float \_ x}) of
| \POS y ⇒ (\textit{is\_positive\_even} y)
| \NEG y ⇒ (\textit{is\_positive\_even} y)
| \ZERO ⇒ \text{true}
end.

Lemma \textit{is\_float\_even\_correct} : (\textit{x:float})
(Frac\_to\_n\_radix \_ bound \_ x) → (\textit{is\_float\_even \_ x}) \text{true} → (\textit{F\_\text{even} \_ bound \_ \text{\_ n\_ \_ radix \_ precision \_ x})).

Definition \textit{evaluate\_even\_closest} := \[ x, r: \text{float} \]
Cases (\textit{evaluate\_rounding \_ x \_ r \_ y: \text{float} \_ y}) of
| \text{bad} ⇒ \text{false}
| \text{tie} ⇒ (\textit{is\_float\_even \_ r})
| \_ ⇒ \text{true}
end.

Theorem \textit{evaluate\_even\_closest\_correct} : (\textit{x, r:float})
(\textit{evaluate\_even\_closest \_ x \_ r}) \text{true} → (\textit{F\_\text{even} \_ bound \_ \text{\_ n\_ \_ radix \_ precision \_ x \_ r}) → (\textit{Even\_closest \_ bound \_ \text{\_ n\_ \_ radix \_ precision \_ x \_ r}))

End ClosestCalc.

Module Ebase

Require Reals.

Section Ebase.

Variable base : \textit{Set}.
Variable base\_plus : base → base → base → \textit{Prop}.
Variable base\_minus : base → base → base → \textit{Prop}.
Variable base\_mul : base → base → base → \textit{Prop}.
Variable base\_coe : base → \textit{R}.

Coercion Local base\_2R := base\_coe.

Definition base\_plus\_lu := \[ a, b, c: \text{base} \] \text{\_ (x, y, z:base)}
\text{\_ “a ≤ a” → “b ≤ y” → (base\_plus \_ x \_ y \_ z) → “c ≤ z”}.

Definition base\_plus\_lu := \[ a, b, c: \text{base} \] \text{\_ (x, y, z:base)}
\text{\_ “x ≤ a” → “y ≤ b” → (base\_plus \_ x \_ y \_ z) → “z ≤ e”}.

Definition base\_minus\_l := \[ a, b, c: \text{base} \] \text{\_ (x, y, z:base)}
\text{\_ “a ≤ a” → “y ≤ b” → (base\_minus \_ x \_ y \_ z) → “c ≤ z”}.

Definition base\_minus\_lu := \[ a, b, c: \text{base} \] \text{\_ (x, y, z:base)}
\text{\_ “x ≤ a” → “y ≤ b” → (base\_minus \_ x \_ y \_ z) → “z ≤ e”}.

Definition base\_mul\_lu := \[ a, b, c, d, e, f: \text{base} \] \text{\_ (x, y, z:base)}
\text{\_ “a ≤ a” → “x ≤ c” → “b ≤ y” → “y ≤ d” → (base\_mul \_ x \_ y \_ z) → “e ≤ z ≤ f”}.

Variable error : \textit{Set}.
Variable error\_coe : error → \textit{R}.
Coercion Local error\_2R := error\_coe.

Definition base\_plus\_e := \[ a, b, c, d: \text{base} \] \text{\_ e: error \_ (x, y, z:base)}
\text{\_ “a ≤ a” → “x ≤ c” → “b ≤ y” → “y ≤ d” → (base\_plus \_ x \_ y \_ z) → “(Rabs\_lu \_ (x + y - z)) ≤ e”}.

Definition base\_minus\_e := \[ a, b, c, d: \text{base} \] \text{\_ e: error \_ (x, y, z:base)}
\[a \leq x^i \rightarrow \text{`x`} \leq e^i \rightarrow \text{`y`} \leq d^i \rightarrow (\text{base\_minus x y z}) \rightarrow (\text{Rabsolu (x-y - z)}) \leq e^i.\]

**Definition base\_mul\_e** : = \[a, b, c, d\times\text{base}\] | c: error | (x, y, z: base)

\[a \leq x^i \rightarrow \text{`x`} \leq e^i \rightarrow \text{`y`} \leq d^i \rightarrow (\text{base\_mul x y z}) \rightarrow (\text{Rabsolu (x\times y - z)}) \leq e^i.\]

**Definition Eint\_bound** : = \[d, x: base\] | x: base | \[x: x\leq x\u2032.\]

**Definition Eint\_error** : = \[x: error\] | x: base | \[x: R\] | (\text{Rabsolu (x - x'}) \leq x^i.\]

**Lemma Eint\_error\_trans:**

(me, x: error; x: base; x: R)

(Eint\_error me x x) \rightarrow (\text{Rle me xe}) \rightarrow (\text{Eint\_error xe x' x}).

**Definition Eint** : = \[d, x: base\] | x: error | \[x: R\] | (\text{Rabsolu x} x' x) \rightarrow (\text{Rabsolu x') x' x}).

**Lemma Eplus\_error\_aux:**

(x, y: error) | (x, y: base) | (x, y: R)

(Eint\_error xe x x x) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow

(||\text{Rabsolu} ((x+yy)-(x+y)'||) \leq x+y.\]

**Lemma Eplus\_error:**

(x, y, z, e: error) | (x, y, z: base) | (x, y: R)

(Eint\_error xe x x x) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow (base\_plus x y x z) \rightarrow

\[2x + y + z e \leq x^i \rightarrow (\text{Rabsolu} (x+y - z')) \leq z^e \rightarrow \text{Eint\_error xe z' x''}\].

**Lemma Eminus\_error\_aux:**

(x, y, z, e: error) | (x, y: base) | (x, y: R)

(Eint\_error xe x y z) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow

(||\text{Rabsolu} ((x-yy)-(x-y)'||) \leq x-y.\]

**Lemma Eminus\_error:**

(x, y, z, e: error) | (x, y, z: base) | (x, y: R)

(Eint\_error xe x y z) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow (base\_minus x y y z) \rightarrow

\[2x + y + z e \leq x^i \rightarrow (\text{Rabsolu} (x+y - z')) \leq z^e \rightarrow \text{Eint\_error xe z' x''}\].

**Lemma Emul\_error\_aux:**

(x, y: error) | (x, y, x, y: base) | (x, y: R)

(Eint\_error xe x x x) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow

(||\text{Rabsolu} x \leq x' ||) \leq y' \rightarrow \text{Eint\_error xe ye x' ye'}.\]

**Lemma Emul\_error:**

(x, y: error) | (x, y, x, y: base) | (x, y: R)

(Eint\_error xe x y z) \rightarrow (\text{Eint\_error ye y y y}) \rightarrow (base\_mul x y y z) \rightarrow

\[2x \times y + x \times z + y \times e \leq x^i \rightarrow (\text{Rabsolu} (x\times y - x') \leq z^e \rightarrow \text{Eint\_error xe z' x''}.\]

\[\text{End Ebase.}\]

**Module Ebase2**

Require Import Reals.

Require Import BaseR.

Section Ebase2.
Variable base : Set.
Variable base_plus : base → base → base → Prop.
Variable base_minus : base → base → base → Prop.
Variable base_mult : base → base → base → Prop.
Variable base_coerc : base → R.

Coercion Local base2R := base_coerc.

Definition base_plus (a b c : base) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow (b \leq y)\% R \rightarrow \text{base\_plus } x y z \rightarrow (c \leq z)\% R. \)

Definition base_plus_u (a b c : base) :=
  \( \forall x y z : base, \)
  \( (x \leq a)\% R \rightarrow (y \leq b)\% R \rightarrow \text{base\_plus } x y z \rightarrow (z \leq c)\% R. \)

Definition base_minus (a b c : base) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow (y \leq b)\% R \rightarrow \text{base\_minus } x y z \rightarrow (c \leq z)\% R. \)

Definition base_minus_u (a b c : base) :=
  \( \forall x y z : base, \)
  \( (x \leq a)\% R \rightarrow (b \leq y)\% R \rightarrow \text{base\_minus } x y z \rightarrow (z \leq c)\% R. \)

Definition base_mult (a b c d e f : base) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow \)
  \( (b \leq y)\% R \rightarrow (y \leq d)\% R \rightarrow \text{base\_mult } x y z \rightarrow (e \leq z \leq f)\% R. \)

Variable error : Set.
Variable error_coerc : error → R.

Coercion Local error2R := error_coerc.

Definition base_plus (a b c : base) (e : error) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow \)
  \( (x \leq c)\% R \rightarrow \)
  \( (b \leq y)\% R \rightarrow (y \leq d)\% R \rightarrow \text{base\_plus } x y z \rightarrow (Rabs (x + y - z) \leq e)\% R. \)

Definition base_minus (a b c : base) (e : error) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow \)
  \( (x \leq c)\% R \rightarrow \)
  \( (b \leq y)\% R \rightarrow (y \leq d)\% R \rightarrow \text{base\_minus } x y z \rightarrow (Rabs (x - y - z) \leq e)\% R. \)

Definition base_mult (a b c : base) (e : error) :=
  \( \forall x y z : base, \)
  \( (a \leq x)\% R \rightarrow \)
  \( (x \leq c)\% R \rightarrow \)
  \( (b \leq y)\% R \rightarrow (y \leq d)\% R \rightarrow \text{base\_mult } x y z \rightarrow (Rabs (x \times y - z) \leq e)\% R. \)

Definition Elmt_bound_R (xl xu xz : R) := (xl \leq xz \leq xu)\% R.
Definition Elmt_absol_R (xe xf xx : R) :=
  \( (xe \leq xx - xx \leq xf)\% R. \)

Definition Elmt_mult_R (xe xf xx : R) :=
  \( xe \neq 0)\% R :- R \land (0 < 1 + xe)\% R \land (1 + xe \leq xx / xx \leq 1 + xf)\% R. \)

Definition Elmt_bound (xl xu xx : base) := (Elmt_bound_R xl xu xx).
Definition Elmt_absol (xe xf : error) (xx : base) (xx : R) :=
  \( (\text{Elmt\_absol } R xe xf xx xx). \)

Definition Elmt_mult (xe xf : error) (xx : base) (xx : R) :=
  \( (\text{Elmt\_mult } R xe xf xx xx). \)

Lemma Elmt_absol2absol_R :
∀ (xe xf ye yf xr yr xx yy : R),
\text{Einit\_absol\_R} xe xf xr xx \rightarrow
\text{Einit\_absol\_R} ye yf yr yy \rightarrow
\text{Einit\_absol\_R} (xe + ye)%R (xf + yf)%R (xr + yr)%R (xx + yy)%R.

\text{Lemma Eplus\_absol :}
∀ (xe xf ye yf xe xf ye yf : error) (xr yr xr : base) (xx yy : R),
\text{Einit\_absol} xe xf xr xx \rightarrow \text{Einit\_absol} ye yf yr yy \rightarrow
\text{base\_plus} xr yr xr \rightarrow
(xe \leq xe + ye + we)%R \rightarrow (xf + yf + we \leq sf)%R \rightarrow
\text{Einit\_absol\_R} we wf xr (xr + yr)%R \rightarrow
\text{Einit\_absol} xe xf (xx + yy).

\text{Lemma Emul\_relat\_R :}
∀ (xe xf ye yf xr yr xx yy : R),
\text{Einit\_relat\_R} xe xf xr xx \rightarrow \text{Einit\_relat\_R} ye yf yr yy \rightarrow
\text{Emul\_relat\_R} \left( (1 + xe) \times (1 + ye) - 1 \right)%R \left( (1 + xf) \times (1 + yf) - 1 \right)%R \left( (xe \times ye) \times (xf \times yf) - 1 \right)%R \left( xe \times ye \times fe \right)%R.

\text{Lemma Emul\_absol\_R :}
∀ (xe xf ye yf xe xf xe xf : base) (xx yy : R),
\text{Einit\_absol\_R} xe xf xe xf \rightarrow \text{Einit\_absol\_R} ye yf yr yy \rightarrow
\text{Einit\_bound\_R} xe xf ye yf \rightarrow \text{Einit\_bound\_R} ye yf yr \rightarrow
let (xe,xf) \rightarrow \text{Iplus\_fun} (Iplus\_fun (Imult\_fun (xe,xf) (ye,yf)) (Imult\_fun (ye,yf) (xe,xf))) \rightarrow \text{Imult\_fun (xe,xf) (ye,yf)} \rightarrow
\text{Einit\_absol\_R} xe xf (xx \times yy)%R.

End Ebase2.

\textbf{Module Efloat\_r2ec}

\textbf{Add LoadPath "Float".}
\textbf{Require AllFloat.}
\textbf{Require Ebase.}
\textbf{Require ClosestCalc.}
\textbf{Require IbaseR.}
\textbf{Require Float\_hyp}. 
\textbf{Section Efloat\_r2ec.}

\textbf{Definition radix} := \langle 2 \rangle.

\textbf{Local radixMon ThanOne} := \langle \text{TwoMon} \text{ThanOne} \rangle.
\textbf{Lemma radixNotZero} : '0 < \text{radix}'.

\textbf{Variable precision} : nat.
\textbf{Hypothesis precision\_MonThanOne} : (\langle 1 \rangle \text{ precision}).
\textbf{Lemma precision\_NotZero} : '0 < \text{precision}'.

\textbf{Variable bExp} : nat.
\textbf{Definition bNum} := \langle \text{iter\_nat} \text{ precision} \text{ positive} \ z0 \ zH \rangle.
\textbf{Definition bound} := \langle \text{Bound bNum} \ bExp \rangle.

\textbf{Lemma pGivesBound} : (\text{POS (uNum bound)}) \rightarrow (\text{Zpower\_nat} \langle 2 \rangle \text{ precision}).

\textbf{Record eFloat} : Set := eFloat
value: float;
cFloat.canonic: \{ (Feanonic radix bound value) \}.

Coercion Local cFloat2R := [x:cFloat] (FloR radix (value x)).
Coercion Local float2R := (FloR radix).

Definition EInt_f2 :=( EInt cFloat cFloat2R float float2R).
Definition EInt_f2_bound :=( EInt_bound cFloat cFloat2R).
Definition EInt_f2_error := (EInt_error cFloat cFloat2R float float2R).

Definition EInt_f2_inclusion_helper := [xI,xu,yu,zI,zu :cFloat]
  (andb
   (Fie_bool radix (value xu) (value xu))
   (Fie_bool radix (value yu) (value yu))).

Lemma EInt_f2_inclusion: (xI,xu,yu,yr :cFloat)
  (EInt_f2_bound xu yr) \rightarrow
  (EInt_f2_inclusion_helper xu xu yu yr) \rightarrow
  \(E\)(EInt_f2_bound yu yr).

Definition EInt_f2_union_helper := [a,b :cFloat]
  (Fie_bool radix (FSucc bound radix precision (value a)) (value b)).

Lemma EInt_f2_union: (P : Prop) (aI,ax,ax,ax :cFloat)
  (EInt_f2_union_helper ax ax P) \rightarrow
  \(E\)(EInt_f2_bound ax ax) \rightarrow
  \(E\)(EInt_f2_bound ax ax) \rightarrow P \rightarrow
  \(E\)(EInt_f2_bound ax ax) \rightarrow P \rightarrow P.

Definition Rounded := [a,R;bc :cFloat] (EvenClosest bound radix precision a (value b)).

Definition cFloat_plus := [a,b,c :cFloat]
  (Rounded "a+b c").

Definition cFloat_minus := [a,b,c :cFloat]
  (Rounded "a-b c").

Definition cFloat_mult := [a,b,c :cFloat]
  (Rounded "a\times b c").

Definition cFloat_div := [a,b :cFloat]
  (Rounded "a/b c").

Definition cFloat_error :=
  [a,b,c :cFloat] (Float '1 ' (Zmax (Fexp (value a)) (Fexp (value b))) - '1).

Definition RError := [x,c :cFloat] (Float '1 ' (Fexp (value x)) - '1).

Lemma RError_correct: (x :cFloat; y :R)
  (Rounded y x) \rightarrow \(R\)(Rabsol (y - x)) \leq (RError x).

Lemma RError_range_correct: (x,y :cFloat)
  "x \leq y \rightarrow \(R\)(Rabsol x) \leq \(R\)(Rabsol y) \leq (RError y)".

Lemma RError_range_correct: (x,y :cFloat)
  "x \leq y \rightarrow \(R\)(Rabsol x) \leq \(R\)(Rabsol y) \leq (RError y) \leq e \rightarrow
  \(R\)(RError x) \leq e".

Lemma Rounded_monotone:
  \((x,y) :R (a,b :cFloat) \rightarrow \(R\)(Rabsol x \leq y \rightarrow \(R\)(Rabsol a) \leq \(R\)(Rabsol b) \rightarrow \(R\))a \leq b).

Lemma cFloat_error_RError_b: (x,y :cFloat) \"(RError x) \leq (cFloat_error x y)\".

Lemma cFloat_error_RError_r: (x,y :cFloat) \"(RError y) \leq (cFloat_error x y)\".

Lemma cFloat_error_RError: (x,y :cFloat)
  (EInt_f2_bound x y) \rightarrow \(R\)(RError x) \leq (cFloat_error x y)\".

Definition Eplus_f2ee_bound_helper :=
  [xI,xu,yu,zI,zu :cFloat]
  (andb
   (evaluate_even_closest radix bound precision}
(Fplus radix (value xl) (value yl)) (value zl)
(evaluate_evn_closest radix bound precision
(Fplus radix (value xu) (value yu)) (value zu)).

Definition Eplus_f2ec_error_helper :=
[x, ye, xe:float] [z, zc:Float]
(Flo_bool radix (Fplus radix (Fplus radix xe ye) (cFloat_error d zu)) xe).

Lemma Eplus_f2ec_bound_aux:
(xl,xu,yl,yl,zl,zc:Float) (xt,xt,xt:Float)
(Eint_f2_bound xl xu xt) \rightarrow (Eint_f2_bound yl xu xt)
(cFloat_plus xt yr xt) \rightarrow (cFloat_plus yl yu xt yl xt) \rightarrow (Eint_f2_bound zl zl zl).

Lemma Eplus_f2ec_bound:
(xl,xu,yl,yl,zl,zc:Float) (xt,xt,xt:Float)
(Eint_f2_bound xl xu xt) \rightarrow (Eint_f2_bound yl xu xt)
(cFloat_plus xt yr xt) \rightarrow (Eplus_f2ec_bound_helper xl yl yl zl yl xt) \rightarrow (Eint_f2_bound zl zl zl).

Lemma Eplus_f2ec_error:
(xe, ye, xe:float) (z, zc:Float) (xr, yr, xe:float) (xx, yy:R)
(Eint_f2_error xe xr xx) \rightarrow (Eint_f2_error ye yr yy)
(cFloat_plus xe ye xe) \rightarrow (Eplus_f2ec_error_helper xe ye xe zl zl) \rightarrow (Eint_f2_error xe ye ye xe ye xe y).

Lemma Rminus_le_2: (a,b,c,d:R) "a \leq b^a \rightarrow c \leq d^c \rightarrow "a-c \leq b-c".

Definition Eminus_f2ec_bound_helper :=
[xl,xu,yl,yu,zl,zc:Float]
(andb
(evaluate_evn_closest radix bound precision
(Fminus radix (value xl) (value yu)) (value zl))
(evaluate_evn_closest radix bound precision
(Fminus radix (value xu) (value yl)) (value zu))).

Definition Eminus_f2ec_error_helper := Eplus_f2ec_error_helper.

Lemma Eminus_f2ec_bound_aux:
(xl,xu,yl,yu,zl,zc:Float) (xt,xt,xt:Float)
(Eint_f2_bound xl xu xt) \rightarrow (Eint_f2_bound yl yu xt yu)
(cFloat_minus xt yr xt) \rightarrow (cFloat_minus yl yu zl zl) \rightarrow (cFloat_minus xu yu zl zl) \rightarrow (Eint_f2_bound zl zl zl).

Lemma Eminus_f2ec_bound:
(xl,xu,yl,yu,zl,zc:Float) (xt,xt,xt:Float)
(Eint_f2_bound xl xu xt) \rightarrow (Eint_f2_bound yl yu xt yu)
(cFloat_minus xt yr xt) \rightarrow (Eminus_f2ec_bound_helper xl yl yl zl yl xu) \rightarrow (Eint_f2_bound zl zl zl).

Lemma Eminus_f2ec_error:
(xe, ye, xe:float) (z, zc:Float) (xr, yr, xe:float) (xx, yy:R)
(Eint_f2_error xe xr xx) \rightarrow (Eint_f2_error ye yr yy)
(cFloat_minus xe ye xe) \rightarrow (Eminus_f2ec_error_helper xe ye xe zl zl) \rightarrow (Eint_f2_error xe xe xe xe y).

Definition Eplus_exact_f2ec_bound_helper :=
[xl,xu,yl,yu,zl,zc:Float]
(andb
(Feq_bool radix (Fplus radix (value xl) (value yl)) (value zl))
(Feq_bool radix (Fplus radix (value xu) (value yu)) (value zu)).

Lemma Fplus_exact_Eec_bound:
(\(x\), xu, yu, zuc Float (x, y, zc:Float)
(Hint f2_bound xl xu xr) \(\rightarrow\) (Hint f2_bound yl yu yr) \(\rightarrow\)
(c Float plus xr yr zr) \(\rightarrow\)
(Fplus_exact_Eec_bound_helper xl xu yl yu zl zu) \(\rightarrow\)

Lemma Fplus_exact_Eec_error_aux:
(x, ye:Float (zlu, zuc Float (x, y, zc:Float) (xx, yy:R)
(Hint f2_error xe xr xx) \(\rightarrow\) (Hint f2_error ye yr yy) \(\rightarrow\)
(c Float plus xr yr zr) \(\rightarrow\) (Hint f2_bound xl zu xr) \(\rightarrow\)
(Zle (Fexp (value xl)) (Fexp (value xr))) \(\rightarrow\)
(Zle (Fexp (value xu)) (Fexp (value yr))) \(\rightarrow\)
(Zle (Fexp (value yu)) (Fexp (value yr))) \(\rightarrow\)
(Zle (Fexp (value zl)) (Fexp (value yr))) \(\rightarrow\)
(Hint f2_error (Fplus radix xe ye) zm “xx + yy”).

Definition Fplus_exact_Eec_error_helper :=
let m = (Zmin (Fexp (Fminus radix bound (value xl) (value xu))))
(Fexp (Fminus radix bound (value yl) (value yu))) in
andb
(Feq_bool radix (Fplus radix xe ye) ze)
andb
(Zle_bool (Fexp (value xl)) m)
(Zle_bool (Fexp (value xu)) m)).

Lemma Fplus_exact_Eec_error:
(x, y, zc:Float (xx, yy:R)
(Hint f2_bound xl xu xr xx) \(\rightarrow\) (Hint f2_bound yl yu yr yy) \(\rightarrow\)
(c Float plus xr yr zr) \(\rightarrow\) (Hint f2_bound xl zu xr) \(\rightarrow\)
(Fplus_exact_Eec_error_helper xl xu ye yl yu zl zu xe) \(\rightarrow\)
(Hint f2_error xe ye yl yu ze xl xe zm “xx + yy”).

Definition Eminus_exact_Eec_bound_helper :=
[xl, xu, yl, yu, zl, zu:Float]
andb
(Feq_bool radix (Fminus radix (value xl) (value yu)) (value zl))
(Feq_bool radix (Fminus radix (value xu) (value yl)) (value zu))).

Lemma Eminus_exact_Eec_bound:
(xl, yl, zuc Float (x, y, zc:Float)
(Hint f2_bound xl xu xr) \(\rightarrow\) (Hint f2_bound yl yu yr) \(\rightarrow\)
(c Float minus xr yr zr) \(\rightarrow\)
(Eminus_exact_Eec_bound_helper xl xu yl yu zl zu) \(\rightarrow\)

Axiom cFloat_Ropp: \(\{x:Float (Fcanonical radix bound (Fopp (value x)))\}.\n
Definition cFloat_opp := [c:Float]
(CFloat (Fopp (value y)) (c Float Fopp x)).

Lemma Eminus_exact_Eec_error_aux:
(x, ye:Float (zlu, zuc Float (x, y, zc:Float) (xx, yy:R)
(Hint f2_error xe xr xx) \(\rightarrow\) (Hint f2_error ye yr yy) \(\rightarrow\)
(c Float minus xr yr zr) \(\rightarrow\) (Hint f2_bound xl zu xr) \(\rightarrow\)
(Zle (Fexp (value xl)) (Fexp (value xr))) \(\rightarrow\)
(Zle (Fexp (value xu)) (Fexp (value yr))) \(\rightarrow\)
(Zle (Fexp (value yu)) (Fexp (value yr))) \(\rightarrow\)
Lemma \( \text{Eminus\_exact\_float\_error} \):

\[
(x, y, z) \xrightarrow{\text{Efloat\_exact\_float\_error}} (y, z, x) \xrightarrow{\text{Efloat\_exact\_float\_error}} (z, x, y)
\]

Definition \( \text{Emult\_float\_bound\_helper} \):

\[
\text{Emult\_float\_bound\_helper}(x, y, z) \xrightarrow{\text{Emult\_float\_bound\_helper}} (x, y, z) \xrightarrow{\text{Emult\_float\_bound\_helper}} (z, x, y)
\]

Local \( \text{min\_mult} \):

\[
\text{min\_mult}(x, y, z) \xrightarrow{\text{min\_mult}} (x, y, z) \xrightarrow{\text{min\_mult}} (y, z, x)
\]

Lemma \( \text{Emult\_float\_bound\_helper} \):

\[
\text{Emult\_float\_bound\_helper}(x, y, z) \xrightarrow{\text{Emult\_float\_bound\_helper}} (x, y, z) \xrightarrow{\text{Emult\_float\_bound\_helper}} (z, x, y)
\]

Definition \( \text{Emult\_float\_error\_helper} \):

\[
\text{Emult\_float\_error\_helper}(x, y, z) \xrightarrow{\text{Emult\_float\_error\_helper}} (x, y, z) \xrightarrow{\text{Emult\_float\_error\_helper}} (z, x, y)
\]

Lemma \( \text{Emult\_float\_error\_aux} \):

\[
\text{Emult\_float\_error\_aux}(x, y, z) \xrightarrow{\text{Emult\_float\_error\_aux}} (x, y, z) \xrightarrow{\text{Emult\_float\_error\_aux}} (z, x, y)
\]

Lemma \( \text{Emult\_float\_error\_float} \):

\[
\text{Emult\_float\_error\_float}(x, y, z) \xrightarrow{\text{Emult\_float\_error\_float}} (x, y, z) \xrightarrow{\text{Emult\_float\_error\_float}} (z, x, y)
\]

Axiom \( \text{Emult\_float\_bound} \):
(xl, xu, yl, yu, zl, zuc: Float) (xr, yr, zr: c: Float)
(Eint_f2_bound xl xu zr) → (Eint_f2_bound yl yu yr) →
(c Float_div xr yr zr) →
true → true →
(Eint_f2_bound xl xu zr).

Axiom Ediv_f2ec_error:
(xl, xu, yl, yu, zl, zuc: Float) (d, zuc: Float; ye: float) (d, zuc: Float; ze: float)
(xx, yy: R)
(Eint_f2 xl xu xx yu yy) → (Eint_f2 yl yu xx yy) →
(c Float_div xr yr zr) → (Eint_f2_bound xl xu zr) →
true → true →
(Eint_f2_error ze xz = yz).

End Efloat_r2ec.

Module Efloat_s754ec

Add LoadPath "Float".
Require Reals.
Require AllFloat.
Require Efloat_r2ec.
Require ClosestCalc.
Require Export Float.

Section Efloat_s754ec.

Local precision := 'N:24'.
Local precision: MoreThanOne: (lt (1) precision).
Unfold precision.
Auto with zarith.
Qed.
Local b Exp := 'N:149'.

Definition Eint_s754 := (Eint_f2 precision b Exp).
Definition Eint_s754_bound := (Eint_f2_bound precision b Exp).
Definition Eint_s754_error := (Eint_f2_error precision b Exp).
Definition Eint_s754_union := (Eint_f2_union precision precisionMoreThanOne b Exp).
Definition Eint_s754_inclusion := (Eint_f2_inclusion precision b Exp).

Definition Eplus_s754ec_error := (Eplus_f2ec_error precision precisionMoreThanOne b Exp).
Definition Eplus_s754ec_bound := (Eplus_f2ec_bound precision precisionMoreThanOne b Exp).
Definition Eminus_s754ec_error := (Eminus_f2ec_error precision precisionMoreThanOne b Exp).
Definition Eminus_s754ec_bound := (Eminus_f2ec_bound precision precisionMoreThanOne b Exp).
Definition Emul_s754ec_bound := (Emul_f2ec_bound precision precisionMoreThanOne b Exp).
Definition Emul_s754ec_error := (Emul_f2ec_error precision precisionMoreThanOne b Exp).
Definition Ediv_s754ec_bound := (Ediv_f2ec_bound precision b Exp).
Definition Ediv_s754ec_error := (Ediv_f2ec_error precision b Exp).
Definition Eplus_exact_s754ec_error := (Eplus_exact_f2ec_error precision precisionMoreThanOne b Exp).
Definition Eplus_exact_s754ec_bound := (Eplus_exact_f2ec_bound precision precisionMoreThanOne b Exp).
Definition Eminus_exact_s754ec_error := (Eminus_exact_f2ec_error precision precisionMoreThanOne b Exp).
Definition Eminus_exact_s754ec_bound := (Eminus_exact_f2ec_bound precision precisionMoreThanOne b Exp).

Definition float_s754 := (c Float precision b Exp).
Definition Good_s754 := (Realnum_prec (bound precision b Exp)).
Definition Float_s754 := f (Float p: (Good_s754 f)) (C Float precision b Exp f p).
Definition Plus_s754ec := (c Float_plus precision b Exp).
Definition Minus_s754ec := (c Float_minus precision b Exp).
Definition Mult_s754ec := (c Float_mult precision b Exp).
Definition Div_s754ec := (c Float_div precision b Exp).

End Efloat_s754ec.
Module IbaseR

Require Import Reals.

Section IbaseR.

Definition Iint (zl xu x : R) := (zl ≤ x ≤ xu)%R.

Lemma IplusR :
∀ zl xu yl yu zl xu x y : R,
(zl ≤ zl + yl)%R → (zl + yl ≤ xu)%R →
Iint zl xu x → Iint zl xu (zl + yl) →
Iint zl xu (z + y).

Definition IplusR_func (xi yi : R×R) :=
let (zl, xu) := xi in let (yl, yu) := yi in
((zl + yl)%R, (xu + yu)%R).

Lemma IplusR_func_correct :
∀ zl xu yl yu zl xu x y : R,
Iint zl xu x → Iint zl xu y →
let (zl, xu) := IplusR_func (zl, xu) (yl, yu) in
Iint zl xu (z + y).
intros zl xu yl yu x y lx ly.
simpl.
apply IplusR with zl xu yl yu; auto with real.
Qed.

Lemma IoppR_exact :
∀ zl xu x : R, Iint zl xu x → Iint (- xu) (- zl) (- x).

Lemma IoppR_exact_even :
∀ zl xu x : R, Iint (- xu) (- zl) (- x) → Iint zl xu x.

Lemma Rhe_trans_and : ∀ a b c : R, (a ≤ b ≤ c)%R → (a ≤ c)%R.

Lemma monotony_1p :
∀ a x y : R, (0 ≤ a)%R → (x ≤ y)%R → (a × x ≤ a × y)%R.

Lemma monotony_2p :
∀ a x y : R, (0 ≤ a)%R → (x ≤ y)%R → (x × a ≤ y × a)%R.

Lemma monotony_1n :
∀ a x y : R, (a ≤ 0)%R → (x ≤ y)%R → (a × y ≤ a × x)%R.

Lemma monotony_2n :
∀ a x y : R, (a ≤ 0)%R → (x ≤ y)%R → (y × a ≤ x × a)%R.

Lemma ImulR pp :
∀ zl xu yl yu zl xu x y : R,
(0 ≤ zl)%R → (0 ≤ yl)%R →
(zl ≤ zl × yl)%R → (zl × yl ≤ zu)%R →
Iint zl xu x → Iint yl yu y →
Iint zl xu (z × y).

Lemma ImulR pm :
∀ zl xu yl yu zl xu x y : R,
(0 ≤ zl)%R → (yl ≤ 0)%R → (0 ≤ yu)%R →
(zl ≤ zu × yl)%R → (zu × yl ≤ zu)%R →
Iint zl xu x → Iint yl yu y →
Iint zl xu (z × y).

Lemma ImulR pn :
∀ zl xu yl yu zl xu x y : R,
(0 ≤ zl)%R → (yu ≤ 0)%R →

\[(z_l \leq z_u \times y_l) \% R \rightarrow (z_l \times y_u \leq z_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

Lemma \textit{ImulR}_{mn}:
\[\forall z_l z_u y_l y_u z_l z_u x y : R,\]
\[(z_l \leq 0) \% R \rightarrow \]
\[(0 \leq z_u) \% R \rightarrow \]
\[(0 \leq y_l) \% R \rightarrow \]
\[(z_l \leq z_l \times y_u) \% R \rightarrow \]
\[(z_u \times y_l \leq z_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

Lemma \textit{ImulR}_{mn}:
\[\forall z_l z_u y_l y_u z_l z_u x y : R,\]
\[(z_l \leq 0) \% R \rightarrow (0 \leq z_u) \% R \rightarrow (y_l \leq 0) \% R \rightarrow (0 \leq y_u) \% R \rightarrow \]
\[(z_l \leq z_u \times y_l) \% R \rightarrow (z_l \leq z_l \times y_u) \% R \rightarrow \]
\[(z_u \times y_l \leq z_u) \% R \rightarrow (z_u \times y_u \leq z_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

Lemma \textit{ImulR}_{np}:
\[\forall z_l z_u y_l y_u z_l z_u x y : R,\]
\[(z_u \leq 0) \% R \rightarrow (0 \leq y_l) \% R \rightarrow \]
\[(z_l \leq z_l \times y_u) \% R \rightarrow (z_u \times y_u \leq z_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

Lemma \textit{ImulR}_{mn}:
\[\forall z_l z_u y_l y_u z_l z_u x y : R,\]
\[(z_u \leq 0) \% R \rightarrow (y_l \leq 0) \% R \rightarrow (0 \leq y_u) \% R \rightarrow \]
\[(z_l \leq z_u \times y_l) \% R \rightarrow (z_l \leq z_l \times y_u) \% R \rightarrow \]
\[(z_u \times y_l \leq z_u) \% R \rightarrow (z_u \times y_u \leq z_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

Lemma \textit{ImulR}_{nn}:
\[\forall z_l z_u y_l y_u z_l z_u x y : R,\]
\[(z_u \leq 0) \% R \rightarrow (y_u \leq 0) \% R \rightarrow \]
\[(z_l \leq z_u \times y_u) \% R \rightarrow (z_l \leq z_l \times y_u) \% R \rightarrow \]
\[\text{Int } z_l \text{ xu } x \rightarrow \text{Int } y_l \text{ yu } y \rightarrow \]
\[\text{Int } z_l \text{ zu } (x \times y).\]

**Inductive** \textit{interval} 
\[\text{interval} \text{ (l u : R)} : \text{Set} :=\]
\[\text{Nsign} : (u \leq 0) \% R \rightarrow \text{interval} \text{ l u} \]
\[\text{Msign} : (l \leq 0) \% R \rightarrow (0 \leq u) \% R \rightarrow \text{interval} \text{ l u} \]
\[\text{Psign} : (0 \leq l) \% R \rightarrow \text{interval} \text{ l u}.\]

Lemma \textit{interval} \text{ sign correct} : \forall l u : R, (l \leq u) \% R \rightarrow \text{interval} \text{ sign l u}.

Lemma \textit{Int}_\text{sign} : \forall l u x : R, \text{Int } l u x \rightarrow \text{interval} \text{ sign l u}.

Lemma \textit{Rmin} \text{ trans}:
\[\forall a b c : R, (c \leq \text{Rmin } a b) \% R \rightarrow (c \leq a) \% R \land (c \leq b) \% R.\]

Lemma \textit{Rmax} \text{ trans}:
\[\forall a b c : R, (\text{Rmax } a b \leq c) \% R \rightarrow (a \leq c) \% R \land (b \leq c) \% R.\]

Lemma \textit{Imul} :
∀ zlxu ylyu zlzuxy : R, 
(zl ≤ Rmin(Rmin(zlxyl)(zlxyl))(Rmin(zxuyu)(zxuyu)))%R → 
(Rmax(Rmax(zlxyl)(zlxyl))(Rmax(zxuyu)(zxuyu)))%R → 
Int xl yu x → Int ylu y → Int zl zu (x × y).

Definition ImulR_fun (xi yi : R×R) :=
let (zl,yl) := xi in let (yu,yl) := yi in 
((Rmin(Rmin(zlxyl)(zlxyl))(Rmin(zxuyu)(zxuyu)))%R, 
(Rmax(Rmax(zlxyl)(zlxyl))(Rmax(zxuyu)(zxuyu)))%R).

Lemma ImulR_fun_correct :
∀ zlxu ylyu zlzuxy : R, 
Int xl yu x → Int yl yu y → 
let (zl,yl) := ImulR_fun (zl,yl) in 
Int zl zu (x × y).
intro xl yl yu x y yz yz.
simpl.
apply ImulR with xl yz yu; auto with real.
Qed.

Lemma Rle_Rinv : ∀ x y : R, (0 < x)%R → (x ≤ y)%R → (y ≤ / x)%R.

Lemma Inv_Rp :
∀ yl yu zlzuxy : R, 
(0 < yl)%R → 
Int yl yu y → (zl ≤ / yu)%R → (y l ≤ zu)%R → Int zl zu (/ y).

Lemma Inv_Rn :
∀ yl yu zlzuxy : R, 
(yu < 0)%R → 
Int yl yu y → (zl ≤ / yu)%R → (y l ≤ zu)%R → Int zl zu (/ y).

Lemma InvR :
∀ yl yu zlzuxy : R, 
(yu < 0)%R ∨ (0 < yl)%R → 
Int yl yu y → (zl ≤ / yu)%R → (y l ≤ zu)%R → Int zl zu (/ y).

Lemma InvR_Le :
∀ zlxu ylyu zlzuxy : R, 
(yu < 0)%R ∨ (yl > 0)%R → 
(zl ≤ Rmin(Rmin(zlxyl)(zlxyl))(Rmin(zxuyu)(zxuyu)))%R → 
(Rmax(Rmax(zlxyl)(zlxyl))(Rmax(zxuyu)(zxuyu)))%R → 
Int xl yu x → Int ylu y → 
Int zl zu (x × y).

End IbaseR.