LEMA: Towards a Language for Reliable Arithmetic


N° 7258
April 2010

Algorithms, Certification, and Cryptography
LEMA: Towards a Language for Reliable Arithmetic

Vincent Lefèvre*, Philippe Théveny†, Florent de Dinechin‡, Claude-Pierre Jeannerod§, Christophe Mouilleron¶, David Pfannholzer∥, Nathalie Revol∗∗

Abstract: Generating certified and efficient numerical codes requires information ranging from the mathematical level to the representation of numbers. Even though the mathematical semantics can be expressed using the content part of MathML, this language does not encompass the implementation on computers. Indeed various arithmetics may be involved, like floating-point or fixed-point, in fixed precision or arbitrary precision, and current tools cannot handle all of these.

Therefore we propose in this paper LEMA (Langage pour les Expressions Mathématiques Annotées), a descriptive language based on MathML with additional expressiveness. LEMA will be used during the automatic generation of certified numerical codes. Such a generation process typically involves several steps, and LEMA would thus act as a glue to represent and store the information at every stage.

First, we specify in the language the characteristics of the arithmetic as described in the IEEE 754 floating-point standard: formats, exceptions, rounding modes. This can be generalized to other arithmetics. Then, we use annotations to attach a specific arithmetic context to an expression tree. Finally, considering the evaluation of the expression in this context allows us to deduce several properties on the result, like being exact or being an exception. Other useful properties include numerical ranges and error bounds.

Key-words: design, languages, performance, reliability, verification, annotated mathematical expressions, XML, MathML, certified numerical properties, formal proof, computer arithmetic, floating-point arithmetic
LEMA: vers un langage pour une arithmétique fiable

Résumé : La génération de codes numériques certifiés et efficaces demande de l’information allant du point de vue mathématique à la représentation des nombres. Bien que la sémantique mathématique puisse être exprimée en utilisant la partie “contenu” de MathML, ce langage ne couvre pas l’implémentation sur les ordinateurs. En effet, diverses arithmétiques peuvent être impliquées, comme la virgule flottante ou la virgule fixe, en précision fixe ou arbitraire, et les outils actuels ne les supportent pas toutes.

C’est pour cela que nous proposons dans ce papier, LEMA (Langage pour les Expressions Mathématiques Annotées), un langage descriptif basé sur MathML avec plus d’expressivité. LEMA sera utilisé dans la génération automatique de codes numériques certifiés. Un tel processus nécessite typiquement plusieurs étapes, et LEMA jouerait ainsi le rôle d’une glu pour représenter et stocker les informations à chaque étape.

Tout d’abord, nous spécifions dans le langage, les caractéristiques de l’arithmétique comme cela est décrit dans la norme IEEE 754 pour la virgule flottante: formats, exceptions, modes d’arrondi. Ceci peut être généralisé à d’autres arithmétiques. Ensuite, nous utilisons des annotations pour attacher un contexte arithmétique spécifique à un arbre d’expressions. Finalement, considérer l’évaluation de l’expression dans ce contexte nous permet de déduire des propriétés sur le résultat, comme le fait d’être exact ou de produire une exception, ou de déterminer une plage de valeurs englobant le résultat et des bornes d’erreur.

Mots-clés : design, langages, performance, fiabilité, vérification, expressions mathématiques annotées, XML, MathML, propriétés numériques certifiées, preuve formelle, arithmétique des ordinateurs, arithmétique en virgule flottante
1 Introduction

A major problem with numerical applications (from libraries to user code) is the control of accuracy with acceptable performance. While some applications are more focused on performance, for other ones, accuracy of the results is crucial. For instance, the requirement can be:

- a guaranteed error bound, provided by the user as an input;
- correct rounding to some given target format, that is, the result obtained as if all internal computations were carried out in infinite precision before the final rounding;
- an exact result, when it is known to be exactly representable in some format.

In addition to accuracy, the range of the computed values needs to be considered too, in order to either avoid overflows and underflows, or control what happens if such exceptions occur. Arithmetics for which this kind of properties can be guaranteed are referred to as reliable arithmetics.

Even applications focused on performance need meaningful results, which one may want to certify. This is not always possible in a reasonable time, though, since errors tend to compensate. Values (either final or internal results) can be expressed in various arithmetics, which can be mixed even by a common application for, say, performance reasons. The most common ones are (see [1] and [22] for comprehensive descriptions):

- integers, either bounded (in a fixed format) or in arbitrary precision;
- floating point in some radix $\beta$ (in general, $\beta = 2$ or 10) and precision $p$ (fixed or arbitrary): a number $x$ has the form $x = s \cdot m \cdot \beta^e$, where $s = \pm 1$ is the sign, $m = x_0.x_1x_2\ldots x_{p-1}$ (with $0 \leq x_i \leq \beta - 1$) is the significand, and the integer $e$ is the exponent (whose range can be bounded or not);
- fixed point, which is quite similar to integer arithmetic.

But one can also construct derived arithmetics, such as double-double arithmetic [21, 9], where each number $x$ is represented by a pair of double-precision floating-point numbers $(x_{hi}, x_{lo})$ and interpreted as $x = x_{hi} + x_{lo}$ (exactly).

The variety of architectures or inputs implies the support of some previously mentioned arithmetics, which in turn yields many variants in the implementation of a single algorithm. In order to avoid tedious work for similar solutions, we seek to generate efficient code automatically.

However, speed is not our main concern. We also want to use the same description of inputs and of the chosen arithmetic so as to prove the validity and accuracy of the implementation with formal methods. Indeed, a traditional solution is to validate the programs with hand-written proofs. In the domain of numerical applications, such proofs often require many tedious calculations, and in practice, they often contain errors (e.g., in corner cases), making them globally incorrect. Moreover if the problem slightly changes (due to a different hypothesis, architecture, input...), the whole proof may need to be redone or manually checked.

To satisfy these needs, we present here a new descriptive language called LEMA.

The outline of this paper is as follows. We start in Section 2 by presenting several established software environments. We show there why they do not allow us to achieve directly the goals mentioned above, and how the single language LEMA could act as a glue between some of them. After describing the LEMA language requirements in Section 3, we argue our decision to use XML as a basis for LEMA in Section 4. Then Section 5 illustrates with the help of examples the use we make of the tree nature of XML and its extensibility. In particular, we show how mathematical expressions are annotated with their evaluation in a given floating-point context and with associated proofs.
2 Related Work

2.1 Existing Environments to Develop Numerical Codes

Various environments already exist for either generating efficient numerical codes or analyzing/proving the numerical properties of given programs. But, like any software, such tools have their own limitations. One can cite:

- **Spiral**\(^1\) This project develops a program generation system capable of producing C or Verilog code implementing digital signal processing algorithms in (complex or real) floating-point arithmetic, like DFTs \(^23, 24\), and since recently, some non-linear transforms as well \(^14\). Given some architectural features (like, typically, various forms of parallelism: SMP, SIMD, etc.), the Spiral system automatically optimizes codes for efficiency. However, although overflows are avoided in some specific applications as described in \(^8\), the generation process does not seem to implement error-analysis techniques.

- **Fluctuat**\(^2\) This software is a static analyzer devoted to the study of the propagation of rounding errors inherent to floating-point computations \(^25\). Given a (possibly very large) numerical code, Fluctuat can detect automatically catastrophic losses of accuracy and identify the corresponding faulty variable(s). However, although the results are validated using affine arithmetic \(^2, 7\), formal proofs are not produced; also, since Fluctuat does not handle arbitrary precision and since it is not freely available, its extension is an issue.

- **Why**\(^3\) This platform for code certification is based on the idea that properties of some existing code can be proven by annotating the code through comments. Then these comments are turned into different proof obligations. Afterwards, one can deal with each proof obligation using an appropriate tool. Frama-C\(^4\), a C source static analyzer, has a plug-in named Jessie which converts C code with comments written in the ANSI/ISO C Specification Language\(^5\) (ACSL) into Why code, thus allowing to reuse all the functionalities of the Why platform.

The main limitation of this approach lies in its restriction to one particular programming language. Frama-C for instance only works on preprocessed C code and will not yield a proof for some extended class of C implementations. Moreover, working on a specific code has two drawbacks: First, useful information could have been lost when the developer has implemented his algorithm; Second, it is conflicting with our idea to go from a mathematical description of a problem to code generation.

2.2 Tools Commonly Used for Specific Tasks

On the other hand, to assist the design of fast and provably-accurate implementations of mathematical functions, the following tools can be extremely useful. However, as we shall see, each of them is used to perform very specific tasks, so that the programmer usually has to combine them in order to get a full design flow:

- **Maple**\(^6\) Symbolic computation can be useful to simplify symbolic expressions or to obtain mathematical properties like the monotonicity of a given function.

---

\(^1\) http://spiral.net/
\(^2\) http://www-list.cea.fr/labos/gb/LSL/fluctuat/index.html
\(^3\) http://why.lri.fr/
as well as \(^10, 11\).
\(^4\) http://frama-c.com/
\(^5\) http://frama-c.com/download/acsl_1.4.pdf
\(^6\) http://www.maplesoft.com/Products/Maple/

RR n° 7258
• **Sollya**[7] The implementation of mathematical functions is often performed by means of polynomial approximation. Sollya is able to compute such an approximation. In addition, one can ask for a guaranteed upper bound on the supremum norm of the difference between a function and its polynomial approximant, thus allowing to control the error committed at this approximation step.

• **Gappa**[8] “a tool intended to help verifying and formally proving properties on numerical programs dealing with floating-point or fixed-point arithmetic.”

• **CGPE**[9] This tool returns some efficient evaluation schemes for univariate and bivariate polynomials, optimized for a specific target architecture. Even though the code latency is the main focus, one can ask that the error due to the evaluation order does not exceed a given bound.

• **Tools to search for “worst cases”**. [17] Such tools target specifically the correct rounding of mathematical functions in a fixed precision, mainly double precision. From the input data (precision, function, tested interval, etc.), these tools generate C code that performs the search. The most important part related to code generation is here a hierarchical approximation of a high-degree polynomial by degree-2 polynomials on consecutive sub-intervals, using a modulo-1 fixed-point arithmetic (provided by the mpn layer of GMP), where each variable has its own precision (which must be a multiple of 32 or 64 bits, depending on the target). Even though dynamical error analysis is done to automatically determine the precision of each variable, all the proofs have been done manually. Some optimizations for current architectures or for specific functions would need to redo most of the error analysis and modify a large part of the tools. This problem was actually at the origin of LEMA, and its need to support some exotic arithmetics.

• **GNU MPFR**[10] This library offers floating-point arithmetic in arbitrary precision with correct rounding. Here, arbitrary precision means that the precision is a parameter of each function call. Note that this library underlies the computations of the aforementioned software.

Most of these tools have been routinely used to produce and validate significant parts of the codes of mathematical software libraries like CRlibm [5] and FLIP [12].

### 2.3 Tool Integration Through LEMA

When implementing an application with certified numerical properties, one can use in turn some of the tools that have been previously presented. Note that they do not act on the same type of data: the information needed by a proof assistant may not be relevant to a general computer algebra system (CAS), and a polynomial specialized tool may not be able to deal with abstract mathematical properties. As there certainly is a natural sequence of calling these tools from the mathematical level to the hardware one, we could have inserted small translators between each application in order to automate the whole process. But some steps would also need data of the specification not produced by its predecessor. For instance, the hardware instruction set and the characteristics of each useful instruction are needed when generating efficient code but not before. Indeed, it is unlikely that a CAS would process them. Conversely, knowledge of

---

[8] http://gappa.gforge.inria.fr/ as well as [20], [5], [4].
the abstract level may be relevant everywhen: the mathematical properties often help to choose
the most efficient algorithm at the implementation stage. For this reason, we want to gather
information produced by all the tools and to store them along with the specifications of the
problem.

Figure 1 shows how we intend to address the issue of certified numerical code generation
using LEMA: The user provides a description of his problem written in LEMA; Then interaction
with the previously mentioned tools allows to enrich this description; Finally, when enough
information has been gathered, the user can proceed with the generation of code possibly with
proof.

This can be C code, possibly with function calls to some libraries, such as GNU MPFR if
multiple precision is needed. Other targets are possible, such as the GCC middle-end.
3 Requirements Analysis

We analyze in this section the requirements of a language that must be sufficiently rich to describe numerical algorithms without loss of information.

We could have reused the annotation languages (like ACSL) that have been presented in the previous section but they are quite general and have only limited support for floating-point arithmetic. Instead of extending one of them, we propose a single language for all data. Such data are either inputs provided by the user, or properties generated by some tools (e.g., by a tool doing value range propagation or from error analysis), or some non-input properties provided by the user because of the difficulty of finding them in an automatic way. In the latter case, a property can be proven later with our tools. The language should hold the status of these data: how they have been added, whether they need to be checked by some prover, whether and how they have been checked, and so on.

A consequence of integrating all kind of information in one place is that this language must mix different domains: mere mathematics, floating-point arithmetic, and hardware possibilities. In each domain, we want to express as simply as possible some basic and high level properties.

In the mathematical domain, it shall be possible to express any kind of mathematical expression or relation that a computer algebraic system can understand. The way expressions are written must preserve their intrinsic parallelism, that is their independence with respect to an order of evaluation. This is another reason to exclude annotations in a procedural language.

In the floating-point domain, we want to express simple properties like exactness in a computation, error bound of an evaluation, or the range of a floating-point function, and there shall be possible to bind such properties to their proofs. To state initial specifications, we need a higher level of expressiveness to describe

- various arithmetics, including arbitrary precision, but also particular arithmetics such as fixed point modulo 1 in arbitrary precision (needed to generalize the hierarchical approximations of a function by polynomials, as described in [17]);
- the rounding mode and its properties;
- the measure of error as a function of the exact value and its approximation;
- how standards like IEEE 754-2008 [15] and C99 [16] define the behavior of functions on special values;
- formal proofs, possibly with properties declared as hypotheses (or axioms) if it is too difficult or impossible to prove them with current tools.

In the hardware description domain, the useful information ranges from which integer and floating-point formats are available on the target platform, to a more complex level for the description of the instruction set characteristics in term of execution time and parallelism capabilities.

And most of all, the language must provide innate extensibility, so that we can create it in a minimal form and enhance it progressively in response to our needs.

4 Rationale for XML and MathML

The first thing we need to express is numerical expressions. For the moment, let us ignore problems related to rounding. The problem that needs to be solved is generally expressed in

RR n° 7258
a mathematical form, and we want to remain close to this form. This is better for the understanding of the computations and for the proof. For this reason, imperative style such as in C or FORTRAN with variables that can be reused in control structures such as loops was rejected. A functional style is highly preferable. We considered using XML as a basis for our language, since

- it allows us to define the features we want, including clean extensibility (e.g., via namespaces);
- it natively expresses trees, thus perfectly fits to the functional style;
- the use of decorations permits the representation of properties parametrized by the arithmetic, without duplication (see Figure 2 for an example);
- it already has various validating parsers, making this solution rather robust (validation could be seen as the first step of a proof: we ensure that what the user writes has an unambiguous meaning, and the various tools working on this language will always get a sane input).

Then the question was: how do we write numerical expressions, that are in fact mathematical expressions? We could define our own XML elements for that, but since specification work had already been done with the XML-based language MathML, we chose to reuse it. As a bonus, this allows to interoperate with other tools that understand MathML, though this was not the primary motivation.

5 Annotating MathML Expressions With Evaluations

In this section, we present how we have extended the MathML language in order to support floating-point evaluations.

As a simple example, let us evaluate the interval

\[ [0.17, 10714811169606510337534739638811517442326528] \]
in single precision (that is in the binary32 format defined in the IEEE 754-2008 standard [15]) with rounding to nearest mode and even-rounding rule for the halfway cases. The exact mathematical interval we choose can be written in pure MathML as follows:

```xml
<math xmlns="http://www.w3.org/1998/Math/MathML">
  <interval>
    <cn id="left" type="real">0.17</cn>
    <cn id="right" type="integer">
      10714811169606510337534739638811517442326528
    </cn>
  </interval>
</math>
```

Here, both endpoints are numbers for briefness, but they could be more sophisticated expressions as well, like polynomials or rational functions. Using a tool like Gappa, we can produce certified evaluations in any floating-point context and then include this information in the previous document. From a lexical point of view, what we want to provide in XML is

- means to bind evaluated values to the exact value from which they stem,
- means to record floating-point properties of these evaluations,
- means to attach certificates to evaluations.

From an efficiency point of view, we want each expression to be evaluated only once in a given floating-point context. Consequently, evaluated values have to be easy to find knowing the exact value node, and vice versa. Additionally, evaluated values must be recorded in such a way that they can be sent to external tools with minimal manipulation.

### 5.1 Numbers in MathML

First, let us review the support of numbers defined in the content markup section of MathML 3.0 (see §4.2.1 “Numbers <cn>” in [18]). As we can see in the above example, the content number element <cn> encodes numbers and it specifies their kind with the `type` attribute. In strict MathML, the `type` attribute may only take four values: `integer`, `real`, `double`, and `hexdouble`, the last two ones being dedicated to floating-point numbers in double precision. This restriction to a unique precision prevents us to use them as a general means for floating-point number encoding. Numbers of the `real` type are written as “an optional sign (+ or −) followed by a string of digits possibly separated into an integer and a fractional part by a decimal point. Some examples are 0.3, 1, and −31.56.” The drawback with this type is the absence of exponent, which hinders its use for numbers whose magnitude is either tiny or huge. In non-strict MathML, floating-point numbers can be encoded with the `e-notation` type. For instance, `<cn type="e-notation">12.3<sep/>3</cn>` represents $12.3 \times 10^3$. Note also that in non-strict MathML a `base` attribute can be used to specify the base in which the text content of the `cn` element should be interpreted. This `e-notation` type allows writing floating-point numbers in arbitrary precision but it requires a conversion before the number can be read by non MathML-aware tools.

Furthermore, the MathML vocabulary can be enriched using symbols defined in OpenMath content dictionaries and, indeed, the content dictionary `bigfloat1.cd` of [5] extends the representation of numbers to more general floating-point formats. For example, the following sample encodes the floating-point value $12.625 = 101.2^{-3}$:
Again, a number encoded this way has to be converted before being sent to external tools.

5.2 Floating-Point Numbers in LEMA

To address this problem, we choose a floating-point number representation stricter than the one specified in §5.12.3 “External hexadecimal-significand character sequences representing finite numbers” in [15], so that it is common to all tools we are considering: first, the mandatory sign, followed by the ‘0x’ prefix, and the hexadecimal integral significand with digits in lower case; second, the binary exponent in the following form: the exponent indicator ‘p’, the mandatory exponent sign, and at last the exponent value written in decimal. For instance, the right endpoint value in our example

\[10714811169606510337534739638811517442326528\]

is written in this format as

\[+0x7bp+136\].

This allows us to represent binary floating-point of any precision. Since such an encoding is not available in MathML, we define in a new namespace the special attribute lema:type of the cn element. The values of lema:type are custom floating-point types like “Binary32” for a floating-point number belonging to the set defined by the binary32 format of the IEEE 754-2008 standard [15] and it indicates that the text content of cn should be interpreted as a number written in the form described above.

The same LEMA namespace is also used for any other floating-point properties not already defined in MathML. We define some additional attributes for the cn element: the floating-point format and the rounding mode are specified in separate attributes called lema:type and lema:rounding, respectively. This helps filter the document for a particular precision or rounding mode. Moreover, each of the values lema:type and lema:rounding is used as a suffix to the value of the initial id attribute in the evaluated number id, so that the connection between the initial value and its evaluation is immediate to a human reader. This entails some data duplication but consistency is easy to check. Finally, the boolean attributes lema:exact and lema:overflow indicate, respectively, if the evaluation is exact\(^{11}\) and if it overflows. The example may be rewritten as follows:

```xml
<cn id="right_Binary32_Nearest"
    lema:type="Binary32"
    lema:rounding="Nearest"
    lema:exact="true"
    lema:overflow="true">+0x7bp+136</cn>
```

\(^{11}\)The notion of exactness is defined regardless of the range of the exponent.
5.3 Annotating with Floating-Point Evaluations

Whereas we had to extend the MathML number encoding, we can reuse the mechanism provided by MathML to annotate elements with application specific information: the pair `<semantics>`, `<annotation-xml>` of elements (see §4.2.8 “Attribution via semantics” in [13]).

The `semantics` element is a container whose first child is the expression being annotated and whose other children are the annotations, each annotation being enclosed in an `annotation` or `annotation-xml` element. By this means, it is possible to provide several alternative presentations of the expression or to change its mathematical meaning with additional information. We use this annotation system to attach an evaluated value to the exact value from which it is derived. This evaluated value can be seen as an interpretation of the mathematical value in a given floating-point context. The `encoding` attribute of an `annotation` element indicates the data format of its text contents; for our needs we use the `application/lema-evaluation+xml` value, thus following the OpenMath example which uses `application/openmath+xml` and the recommendations of RFC 3023 for XML Media Types [23]. Therefore, our example with the evaluations can be written as follows:

```xml
<math xmlns="http://www.w3.org/1998/Math/MathML"
     xmlns:lema="http://www.ens-lyon.fr/LIP/Arenaire/lema">
  <interval>
    <semantics>
      <cn id="left" type="real">0.17</cn>
      <annotation-xml lema:type="Binary32_Nearest"
                       encoding="application/lema-evaluation+xml">
        <cn id="left_Binary32_Nearest" lema:type="Binary32"
            lema:rounding="Nearest"
            lema:exact="false">+0xae147bp-26</cn>
      </annotation-xml>
    </semantics>
    <semantics>
      <cn id="right" type="integer">1071481169606510337534739638811517442326528</cn>
      <annotation-xml lema:type="Binary32_Nearest"
                       encoding="application/lema-evaluation+xml">
        <cn id="right_Binary32_Nearest" lema:type="Binary32"
            lema:rounding="Nearest"
            lema:exact="true"
            lema:overflow="true">+0x7bp+136</cn>
      </annotation-xml>
    </semantics>
  </interval>
</math>
```

As a `semantics` element admits several `annotation-xml` children, we can attach to a single number many evaluations in different floating-point contexts. Contexts are differentiated with the `lema:type` attribute of the `annotation-xml` element, which eases the retrieval of a given evaluation by just browsing among siblings of the exact number node. Thanks to this proximity,
the converse operation, that is, finding the exact value knowing the evaluated value node, is simple too.

5.4 Annotating Evaluations with Certificates

Furthermore, we would like evaluated values and properties such as exactness to be certified, typically by using Gappa. In the following example, we present an excerpt where such a Gappa proof is embedded after the evaluated value using the ability of the annotation-xml element to contain application-specific elements:

```xml
<semantics>
  <cn id="right" type="integer">
    10714811169606510337534739638811517442326528
  </cn>
  <annotation-xml lema:type="Binary32_Nearest" encoding="application/lema-evaluation+xml">
    <cn id="right_Binary32_Nearest" lema:type="Binary32" lema:rounding="Nearest" lema:exact="true" lema:overflow="true">+0x7bp+136</cn>
    <lema:proof href="right_Binary32_Nearest" type="gappa">
        <![CDATA[
          @rndn = float< 24, -126, ne >;
          MaxFloat = 0xf.fffffp+124;
          right = 10714811169606510337534739638811517442326528;
          right_Binary32_Nearest = +0x7bp+136;

          {
            right_Binary32_Nearest - rndn(right) in [0, 0]
            \ right_Binary32_Nearest - right in [0, 0]
            \ right_Binary32_Nearest - MaxFloat >= 0
          }
        ]]>}
    </lema:proof>
  </annotation-xml>
</semantics>
```

Here we have introduced a new lema:proof element as a container for the script proving the evaluation whose identifier is referenced by the href attribute. Several proofs for the same evaluation but written in different tool languages can be embedded at this point, and the type attribute differentiates them. Here, the CDATA content is a script that Gappa can interpret. A Gappa script is composed of two or three sections: The first one is where symbols are defined; The second one, written between curly brackets, is a logical formula to be proven; The last one, which is optional, is a series of hints like rewriting rules or bisection directives, and is aimed at helping the Gappa engine (see [19] for further information).

In the above example, the first four lines in the CDATA part belong to the definition section: the symbol rndn defines both the rounding mode and the floating-point format, while MaxFloat corresponds to the maximal number that can be represented in the binary32 format. The values
of the id attributes are reused to define symbols for the corresponding numbers, making it easier to map them to their counterpart in the XML document.

The lines between curly brackets form the logical formula, which is a conjunction of statements. The first line states that the rounded value truly is +0x7bp+136, the second one, that the exact value of the right endpoint is represented without rounding error, and the last one that it actually overflows.

From this script, Gappa can also derive formal proofs in Coq and HOL Light, which could be embedded next to it. But as the formal proofs are mere certificates that are not used thereafter and that should not be corrupted by subsequent transformations, it is preferred to save them in external files, using the src attribute value to record their URI, as in:

```xml
<semantics>
  <cn id="left">0.17</cn>

  <annotation-xml lema:type="Binary32_Nearest" encoding="application/lema-evaluation+xml">
    <cn id="left_Binary32_Nearest" lema:rounding="Nearest" lema:exact="false">
      +0xae147bp-26
    </cn>
    <lema:proof href="left_Binary32_Nearest" type="gappa" src="left_Binary32_Nearest.gappa"/>
    <lema:proof href="left_Binary32_Nearest" type="coq" src="left_Binary32_Nearest.v"/>
  </annotation-xml>
</semantics>
```

This shortens somewhat the XML document, as Coq proofs are very lengthy. In addition, if all proofs are saved in a single directory, it is easy to check them all by sending them in a row to the proof checker.

6 Conclusion and Perspectives

In this paper, we have presented LEMA, our extension of MathML that suits our needs for code generation and formal proofs of numerical codes. We have exemplified the extension to floating-point arithmetic: representation of floating-point numbers and creation of the lema:type attribute, incorporation of their evaluation in floating-point arithmetic through annotations, expressions of floating-point properties through attributes and links to their proof.

Beyond floating-point evaluation, LEMA should express other arithmetic properties like error bounds and value ranges, special values (infinity, signed zero, NaN), and arithmetics and their associated formats. This is currently under development. At a higher level, specifications, such as a description of hardware capabilities or a function specification, can be reused. Thus, we intend to set up a database mechanism. Meanwhile, we are elaborating an XML schema to validate LEMA documents.

The LEMA language is developed simultaneously with a library that enables to integrate various tools, as shown in Figure 1. Linking with the tools other than Gappa is work in progress.
7 Acknowledgement

This work is supported by the ANR project EVA-Flo.

References


