The Formalization of Syntax-Based Mathematical Algorithms Using Quotation and Evaluation*

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Abstract. Algorithms like those for differentiating functional expressions manipulate the syntactic structure of mathematical expressions in a mathematically meaningful way. A formalization of such an algorithm should include a specification of its computational behavior, a specification of its mathematical meaning, and a mechanism for applying the algorithm to actual expressions. Achieving these goals requires the ability to integrate reasoning about the syntax of the expressions with reasoning about what the expressions mean. A syntax framework is a mathematical structure that is an abstract model for a syntax reasoning system. It contains a mapping of expressions to syntactic values that represent the syntactic structures of the expressions; a language for reasoning about syntactic values; a quotation mechanism to refer to the syntactic value of an expression; and an evaluation mechanism to refer to the value of the expression represented by a syntactic value. We present and compare two approaches, based on instances of a syntax framework, to formalize a syntax-based mathematical algorithm in a formal theory $T$. In the first approach the syntactic values for the expressions manipulated by the algorithm are members of an inductive type in $T$, but quotation and evaluation are functions defined in the metatheory of $T$. In the second approach every expression in $T$ is represented by a syntactic value, and quotation and evaluation are operators in $T$ itself.

1 Introduction

A great many of the algorithms employed in mathematics work by manipulating the syntactic structure of mathematical expressions in a mathematically meaningful way. Here are some examples:

1. Arithmetic operations applied to numerals.

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2. Operations such as factorization applied to polynomials.
3. Simplification of algebraic expressions.
4. Operations such as transposition performed on matrices.
5. Symbolic differentiation and antidifferentiation of expressions with variables.

The study and application of these kinds of algorithms is called *symbolic computation*. For centuries symbolic computation was performed almost entirely using pencil and paper. However, today symbolic computation can be performed by computer, and algorithms that manipulate mathematical expressions are the main fare of computer algebra systems.

In this paper we are interested in the problem of how to formalize syntax-based mathematical algorithms. These algorithms manipulate members of a formal language in a computer algebra system, but their behavior and meaning are usually not formally expressed in a computer algebra system. However, we want to use these algorithms in formal theories and formally understand what they do. We are interested in employing existing external implementations of these algorithms in formal theories as well as implementing these algorithms directly in formal theories.

As an illustration, consider an algorithm, say named *RatPlus*, that adds rational number numerals, which are represented in memory in some suitable way. (An important issue, that we will not address, is how the numerals are represented to optimize the efficiency of *RatPlus*.) For example, if the numerals $\frac{2}{5}$ and $\frac{3}{8}$ are given to *RatPlus* as input, the numeral $\frac{31}{40}$ is returned by *RatPlus* as output. What would we need to do to use *RatPlus* to add rational numbers in a formal theory $T$ and be confident that the results are correct? First, we would have to introduce values in $T$ to represent rational number numerals as syntactic structures, and then define a binary operator $O$ over these values that has the same input-output relation as *RatPlus*. Second, we would have to prove in $T$ that, if $O(a, b) = c$, then the sum of the rational numbers represented by $a$ and $b$ is the rational number represented by $c$. And third, we would have to devise a mechanism for using the definition of $O$ to add rational numbers in $T$.

The second task is the most challenging. The operator $O$, like *RatPlus*, manipulates numerals as syntactic structures. To state and then prove that these manipulations are mathematically meaningful requires the ability to express the interplay of how the numerals are manipulated and what the manipulations mean with respect to rational numbers. This is a formidable task in a traditional logic in which there is no mechanism for directly referring to the syntax of the expressions in the logic. We need to reason about a rational number numeral $\frac{2}{5}$ both as a syntactic structure that can be deconstructed into the integer numerals 2 and 5 and as an expression that denotes the rational number 2/5.

Let us try to make the problem of how to formalize syntax-based mathematical algorithms like *RatPlus* more precise. Let $T$ be a theory in a traditional logic like first-order logic or simple type theory, and let $A$ be an algorithm that manipulates certain expressions of $T$. To formalize $A$ in $T$ we need to do three things:
1. Define an operator $O_A$ in $T$ that represents $A$: Introduce values in $T$ that represent the expressions manipulated by $A$. Introduce an operator $O_A$ in $T$ that maps the values that represent the input expressions taken by $A$ to the values that represent the output expressions produced by $A$. Write a sentence named $\text{CompBehavior}$ in $T$ that specifies the computational behavior of $O_A$ to be the same as that of $A$. That is, if $A$ takes an input expression $e$ and produces an output expression $e'$, then $\text{CompBehavior}$ should say that $O_A$ maps the value that represents $e$ to the value that represents $e'$.

2. Prove in $T$ that $O_A$ is mathematically correct: Write a sentence named $\text{MathMeaning}$ in $T$ that specifies the mathematical meaning of $O_A$ to be the same as that of $A$. That is, if the value of an input expression $e$ given to $A$ is related to the value of the corresponding output expression $e'$ produced by $A$ in a particular way, then $\text{MathMeaning}$ should say that the value of the expression representing $e$ should be related to the value of the expression representing $e'$ in the same way. Finally, prove $\text{MathMeaning}$ from $\text{CompBehavior}$ in $T$.

3. Devise a mechanism for using $O_A$ in $T$: An application $O_A(a_1, \ldots, a_n)$ of $O_A$ to the values $a_1, \ldots, a_n$ can be computed in $T$ by instantiating $\text{CompBehavior}$ with $a_1, \ldots, a_n$ and then simplifying the resulting formula to obtain the value of $O_A(a_1, \ldots, a_n)$. For the sake of convenience or efficiency, we might want to use $A$ itself to compute $O_A(a_1, \ldots, a_n)$. We will know that results produced by $A$ are correct provided $A$ and $O_A$ have the same computational behavior.

If we believe that $A$ works correctly and we are happy to do our computations with $A$ outside of $T$, we can skip the writing of $\text{CompBehavior}$ and use $\text{MathMeaning}$ as an axiom that asserts $A$ has the mathematical meaning specified by $\text{MathMeaning}$ for $O_A$. The idea of treating specifications of external algorithms as axioms is a key component of the notion of a biform theory [2,6].

So to use $A$ in $T$ we need to formalize $A$ in $T$, and to do this, we need a system that integrates reasoning about the syntax of the expressions with reasoning about what the expressions mean. A syntax framework [9] is a mathematical structure that is an abstract model for a syntax reasoning system. It contains a mapping of expressions to syntactic values that represent the syntactic structures of the expressions; a language for reasoning about syntactic values; a quotation mechanism to refer to the syntactic value of an expression; and an evaluation mechanism to refer to the value of the expression represented by a syntactic value. A syntax framework provides the tools needed to reason about the interplay of syntax and semantics. It is just what we need to formalize syntax-based mathematical algorithms.

Reflection is a technique to embed reasoning about a reasoning system (i.e., metareasoning) in the reasoning system itself. Reflection has been employed in logic [13], theorem proving [12], and programming [5]. Since metareasoning very often involves the syntactic manipulation of expressions, a syntax framework is a natural subcomponent of a reflection mechanism.

This paper attacks the problem of formalizing a syntax-based mathematical algorithm $A$ in a formal theory $T$ using syntax frameworks. Two approaches
are presented and compared. The first approach is local in nature. It employs a syntax framework in which there are syntactic values only for the expressions manipulated by $A$. The second approach is global in nature. It employs a syntax framework in which there are syntactic values for all the expressions of $T$. We will see that these two approaches have contrasting strengths and weaknesses. The local approach offers an incomplete solution at a low cost, while the global approach offers a complete solution at a high cost.

The two approaches will be illustrated using the familiar example of polynomial differentiation. In particular, we will discuss how the two approaches can be employed to formalize an algorithm that differentiates expressions with variables that denote real-valued polynomial functions. We will show that algorithms like differentiation that manipulate expressions with variables are more challenging to formalize than algorithms like symbolic arithmetic that manipulate numerals without variables.

The following is the outline of the paper. The next section, Section 2, presents the paper’s principal example, polynomial differentiation. The notion of a syntax framework is defined in Section 3. Sections 4 and 5 present the local and global approaches to formalizing syntax-based mathematical algorithms. And the paper concludes with Section 6.

2 Example: Polynomial Differentiation

We examine in this section the problem of how to formalize a symbolic differentiation algorithm and then prove that the algorithm actually computes derivatives. We start by defining what a derivative is.

Let $f : \mathbb{R} \to \mathbb{R}$ be a function over the real numbers and $a \in \mathbb{R}$. The derivative of $f$ at $a$, written $\text{deriv}(f, a)$, is

$$
\lim_{h \to 0} \frac{f(a + h) - f(a)}{h}
$$

if this limit exists. The derivative of $f$, written $\text{deriv}(f)$, is the function

$$
\lambda x : \mathbb{R} . \text{deriv}(f, x).
$$

Notice that we are using the traditional definition of a derivative in which a derivative of a function is defined pointwise.

Differentiation is in general the process of finding derivatives which ultimately reduces to finding limits. Symbolic differentiation is the process of mechanically transforming an expression with variables that represents a function over the real numbers into an expression with variables that represents the derivative of the function. For example, the result of symbolically differentiating the expression $\sin(x^2)$ which represents the function $\lambda x : \mathbb{R} . \sin(x^2)$ is the expression $2 \cdot x \cdot \cos(x^2)$ which represents the function $\lambda x : \mathbb{R} . 2 \cdot x \cdot \cos(x^2)$. Symbolic differentiation is performed by applying certain differentiation rules and simplification rules to a starting expression until no rule is applicable.
Let us look at how symbolic differentiation works on polynomials. A polynomial is an expression constructed from real-valued constants and variables by applying addition, subtraction, multiplication, and natural number exponentiation. For example, \( x \cdot (x^2 + y) \) is a polynomial. The symbolic differentiation of polynomials is performed using the following well-known differentiation rules:

**Constant Rule**
\[
\frac{d}{dx}(c) = 0 \quad \text{where } c \text{ is a constant or a variable different from } x.
\]

**Variable Rule**
\[
\frac{d}{dx}(x) = 1.
\]

**Sum and Difference Rule**
\[
\frac{d}{dx}(u \pm v) = \frac{d}{dx}(u) \pm \frac{d}{dx}(v).
\]

**Product Rule**
\[
\frac{d}{dx}(u \cdot v) = \frac{d}{dx}(u) \cdot v + u \cdot \frac{d}{dx}(v).
\]

**Power Rule**
\[
\frac{d}{dx}(u^n) = \begin{cases} 
0 & \text{if } n = 0, \\
 n \cdot u^{n-1} \cdot \frac{d}{dx}(u) & \text{if } n > 0.
\end{cases}
\]

Written using traditional Leibniz notation, the rules specify how symbolic differentiation is performed with respect to the variable \( x \). The symbols \( u \) and \( v \) range over polynomials that may contain \( x \) as well as other variables, and the symbol \( n \) ranges over natural numbers. Notice that these rules are not meaning preserving in the usual way; for example, the rule \( \frac{d}{dx}(c) = 0 \) is not meaning preserving if we view \( c \) as a value and not as an expression.

Let \( \text{PolyDiff} \) be the algorithm that, given a polynomial \( u \) and variable \( x \), applies the five differentiation rules above to the starting expression \( \frac{d}{dx}(u) \) until there are no longer any expressions starting with \( \frac{d}{dx} \) and then simplifies the resulting expression using the rules \( 0 + u = u + 0 = 0 \) and \( 1 \cdot u = u \cdot 1 = u \) and collecting like terms. Applied to \( x \cdot (x^2 + y) \), \( \text{PolyDiff} \) would perform the following steps:

\[
\frac{d}{dx}(x \cdot (x^2 + y)) = \frac{d}{dx}(x) \cdot (x^2 + y) + x \cdot \frac{d}{dx}(x^2 + y) \\
= 1 \cdot (x^2 + y) + x \cdot \left( \frac{d}{dx}(x^2) + \frac{d}{dx}(y) \right) \\
= 1 \cdot (x^2 + y) + x \cdot \left( 2 \cdot x^1 \cdot \frac{d}{dx}(x) + 0 \right) \\
= 1 \cdot (x^2 + y) + x \cdot (2 \cdot x^1 \cdot 1 + 0) \\
= 3 \cdot x^2 + y
\]
Line (1) is by the Product Rule; (2) is by the Variable and Sum and Difference Rules; (3) is by the Power and Constant Rules; (4) is by the Variable Rule; and (5) is by the simplification rules. Thus, given the function

\[ f = \lambda x : \mathbb{R}. x \cdot (x^2 + y), \]

using PolyDiff we are able to obtain the derivative

\[ \lambda x : \mathbb{R}. 3 \cdot x^2 + y \]

of \( f \) via mechanical manipulation of the expression \( x \cdot (x^2 + y) \).

Algorithms similar to PolyDiff are commonly employed in informal mathematics. In fact, they are learned and applied by every calculus student. They should be as available and useful in formal mathematics as they are in informal mathematics. We thus need to formalize them as described in the Introduction.

The main objective of this paper is to show how syntax-based mathematical algorithms can be formalized using PolyDiff as example. We will begin by making the task of formalizing PolyDiff precise.

Let a theory be a pair \( T = (L, \Gamma) \) where \( L \) is a formal language and \( \Gamma \) is a set of sentences in \( L \) that serve as the axioms of the theory. Define \( T_R = (L_R, \Gamma_R) \) to be a theory of the real numbers in (many-sorted) simple type theory. We assume that \( L_R \) is a set of expressions over a signature that includes a type \( \mathbb{R} \) of the real numbers, constants for each natural number, and constants for addition, subtraction, multiplication, natural number exponentiation, and the unary and binary \( \text{deriv} \) operators defined above. We assume that \( \Gamma_R \) contains the axioms of a complete ordered field as well as the definitions of all the defined constants in \( L_R \) (see [8] for further details).

Let \( L_{\text{var}} \subseteq L_R \) be the set of variables of type \( \mathbb{R} \) and \( L_{\text{poly}} \subseteq L_R \) be the set of expressions constructed from members of \( L_{\text{var}} \), constants of type \( \mathbb{R} \), addition, subtraction, multiplication, and natural number exponentiation. Finally, assume that \( \text{PolyDiff} : L_{\text{poly}} \times L_{\text{var}} \rightarrow L_{\text{poly}} \) is the algorithm described in the previous section adapted to operate on expressions of \( L_R \).

Thus to formalize PolyDiff we need to:

1. Define an operator \( O_{\text{pd}} \) in \( T_R \) that represents PolyDiff.
2. Prove in \( T_R \) that \( O_{\text{pd}} \) is mathematically correct.
3. Devise a mechanism for using \( O_{\text{pd}} \) in \( T \).

Formalizing PolyDiff should be much easier than formalizing differentiation algorithms for larger sets of expressions that include, for example, rational expressions and transcendental functions. Polynomial functions are total (i.e., they are defined at all points on the real line) and their derivatives are also total. As a result, issues of undefinedness do not arise when specifying the mathematical meaning of PolyDiff.

However, functions more general than polynomial functions as well as their derivatives may be undefined at some points. Thus using a differentiation algorithm to compute the derivative of one of these more general functions requires
care in determining the precise domain of the derivative. For example, differentiating the rational expression $x/x$ using the well-known Quotient Rule yields the expression 0, but the derivative of $\lambda x : \mathbb{R} . x/x$ is not $\lambda x : \mathbb{R} . 0$. The derivative is actually the partial function

$$\lambda x : \mathbb{R} . \text{if } x \neq 0 \text{ then } 0 \text{ else } \bot.$$ 

We restrict our attention to differentiating polynomial functions so that we can focus on reasoning about syntax without being concerned about issues of undefinedness.

### 3 Syntax Frameworks

A syntax framework [9] is a mathematical structure that is intended to be an abstract model of a system for reasoning about the syntax of an interpreted language (i.e., a formal language with a semantics). It will take several definitions from [9] to present this structure.

**Definition 1 (Interpreted Language).** An interpreted language is a triple $I = (L, D_{\text{sem}}, V_{\text{sem}})$ where:

1. $L$ is a formal language, i.e., a set of expressions.
2. $D_{\text{sem}}$ is a nonempty domain (set) of semantic values.
3. $V_{\text{sem}} : L \rightarrow D_{\text{sem}}$ is a total function, called a semantic valuation function, that assigns each expression $e \in L$ a semantic value $V_{\text{sem}}(e) \in D_{\text{sem}}$.  

A syntax representation of a formal language is an assignment of syntactic values to the expressions of the language:

**Definition 2 (Syntax Representation).** Let $L$ be a formal language. A syntax representation of $L$ is a pair $R = (D_{\text{syn}}, V_{\text{syn}})$ where:

1. $D_{\text{syn}}$ is a nonempty domain (set) of syntactic values. Each member of $D_{\text{syn}}$ represents a syntactic structure.
2. $V_{\text{syn}} : L \rightarrow D_{\text{syn}}$ is an injective, total function, called a syntactic valuation function, that assigns each expression $e \in L$ a syntactic value $V_{\text{syn}}(e) \in D_{\text{syn}}$ such that $V_{\text{syn}}(e)$ represents the syntactic structure of $e$.  

A syntax language for a syntax representation is a language of expressions that denote syntactic values in the syntax representation:

**Definition 3 (Syntax Language).** Let $R = (D_{\text{syn}}, V_{\text{syn}})$ be a syntax representation of a formal language $L_{\text{obj}}$. A syntax language for $R$ is a pair $(L_{\text{syn}}, I)$ where:

\footnote{No distinction is made between how expressions are constructed in this definition as well as in subsequent definitions. In particular, expressions constructed by binding variables are not treated in any special way.}

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[9] No distinction is made between how expressions are constructed in this definition as well as in subsequent definitions. In particular, expressions constructed by binding variables are not treated in any special way.
1. \( I = (L, D_{\text{sem}}, V_{\text{sem}}) \) in an interpreted language.
2. \( L_{\text{obj}} \subseteq L \), \( L_{\text{syn}} \subseteq L \), and \( D_{\text{syn}} \subseteq D_{\text{sem}} \).
3. \( V_{\text{sem}} \) restricted to \( L_{\text{syn}} \) is a total function \( V'_{\text{sem}} : L_{\text{syn}} \to D_{\text{syn}} \).

Finally, we are now ready to define a syntax framework:

**Definition 4 (Syntax Framework in an Interpreted Language).**

Let \( I = (L, D_{\text{sem}}, V_{\text{sem}}) \) be an interpreted language and \( L_{\text{obj}} \) be a sublanguage of \( L \). A syntax framework for \((L_{\text{obj}}, I)\) is a tuple \( F = (D_{\text{syn}}, V_{\text{syn}}, L_{\text{syn}}, Q, E) \) where:

1. \( R = (D_{\text{syn}}, V_{\text{syn}}) \) is a syntax representation of \( L_{\text{obj}} \).
2. \( (L_{\text{syn}}, I) \) is a syntax language for \( R \).
3. \( Q : L_{\text{obj}} \to L_{\text{syn}} \) is an injective, total function, called a quotation function, such that:

   **Quotation Axiom.** For all \( e \in L_{\text{obj}} \),
   \[
   V_{\text{sem}}(Q(e)) = V_{\text{syn}}(e).
   \]

4. \( E : L_{\text{syn}} \to L_{\text{obj}} \) is a (possibly partial) function, called an evaluation function, such that:

   **Evaluation Axiom.** For all \( e \in L_{\text{syn}} \),
   \[
   V_{\text{sem}}(E(e)) = V_{\text{sem}}(V_{\text{syn}}^{-1}(V_{\text{sem}}(e)))
   \quad \text{whenever } E(e) \text{ is defined.} \]

A syntax framework is depicted in Figure 1. For \( e \in L_{\text{obj}} \), \( Q(e) \) is called the quotation of \( e \). \( Q(e) \) denotes a value in \( D_{\text{syn}} \) that represents the syntactic structure of \( e \). For \( e \in L_{\text{syn}} \), \( E(e) \) is called the evaluation of \( e \). If it is defined, \( E(e) \) denotes the same value in \( D_{\text{sem}} \) that the expression represented by the value of \( e \) denotes. Since there will usually be different \( e_1, e_2 \in L_{\text{syn}} \) that denote the same syntactic value, \( E \) will usually not be injective. \( Q \) and \( E \) correspond to the quote and eval operators in Lisp and other languages.

Common examples of syntax frameworks are based on representing the syntax of expressions by Gödel numbers, strings, and members of an inductive type. Programming languages that support metaprogramming — such as Lisp, F# [10], MetaML [18], MetaOCaml [15], reFLect [11], and Template Haskell [16] — are instances of a syntax framework if mutable variables are disallowed. See [9] for these and other examples of syntax frameworks.

The notion of a syntax framework can be easily lifted from an interpreted language to an interpreted theory. This is the version of a syntax framework that we will use in this paper.

**Definition 5 (Model).** Let \( T = (L, \Gamma) \) be a theory. A model of \( T \) is a pair \( M = (D_{\text{sem}}^M, V_{\text{sem}}^M) \) such that \( D_{\text{sem}}^M \) is a nonempty set of semantic values that includes the truth values \( \top \) (true) and \( \bot \) (false) and \( V_{\text{sem}}^M : L \to D_{\text{sem}}^M \) is a total function such that, for all sentences \( A \in \Gamma \), \( V_{\text{sem}}^M(A) = \top. \) □
Definition 6 (Interpreted Theory). An interpreted theory is a pair \( I = (T, \mathcal{M}) \) where \( T \) is a theory and \( \mathcal{M} \) is a set of models of \( T \). (If \( T = (L, \Gamma) \), \((L, D^M_{\text{sem}}, V^M_{\text{sem}})\) is obviously an interpreted language for each \( M \in \mathcal{M} \).) \( \Box \)

Definition 7 (Syntax Framework in an Interpreted Theory).
Let \( I = (T, \mathcal{M}) \) be an interpreted theory where \( T = (L, \Gamma) \) and \( L_{\text{obj}} \subseteq L \). A syntax framework for \((L_{\text{obj}}, I)\) is a triple \( F = (L_{\text{syn}}, Q, E) \) where:

1. \( L_{\text{syn}} \subseteq L \).
2. \( Q : L_{\text{obj}} \rightarrow L_{\text{syn}} \) is an injective, total function.
3. \( E : L_{\text{syn}} \rightarrow L_{\text{obj}} \) is a (possibly partial) function.
4. For all \( M = (D^M_{\text{sem}}, V^M_{\text{sem}}) \in \mathcal{M} \), \( F^M = (D^M_{\text{syn}}, V^M_{\text{syn}}, L_{\text{syn}}, Q, E) \) is a syntax framework for \((L_{\text{obj}}, (L, D^M_{\text{sem}}, V^M_{\text{sem}}))\) where \( D^M_{\text{syn}} \) is the range of \( V^M_{\text{sem}} \) restricted to \( L_{\text{syn}} \) and \( V^M_{\text{syn}} = V^M_{\text{sem}} \circ Q \).

Let \( I = (L, D, V) \) be an interpreted language, \( L_{\text{obj}} \subseteq L \), and \( F = (D_{\text{syn}}, V_{\text{syn}}, L_{\text{syn}}, Q, E) \) be a syntax framework for \((L_{\text{obj}}, I)\). \( F \) has built-in quotation if there is an operator (which we will denote as \text{quote} \) such that, for all \( e \in L_{\text{obj}} \), \( Q(e) \) is the syntactic result of applying the operator to \( e \) (which we will denote as \text{quote}(e)) \( \). \( F \) has built-in evaluation if there is an operator (which we will denote as \text{eval} \) such that, for all \( e \in L_{\text{syn}} \), \( E(e) \) is the syntactic result of applying the operator to \( e \) (which we will denote as \text{eval}(e)) \) whenever \( E(e) \) is defined. There are similar definitions as those above when \( F \) is a syntax framework in an interpreted theory.

A syntax framework \( F \) for \((L_{\text{obj}}, I)\), where \( I \) is either an interpreted language or an interpreted theory, is replete if \( L_{\text{obj}} = L \) and \( F \) has both built-in quotation and evaluation. If \( F \) is replete, it has the facility to reason about the syntax.

Fig. 1. A Syntax Framework
of all of \( L \) within \( L \) itself. Examples of a replete syntax framework are rare. The programming language Lisp with a simplified semantics is the best known example of a replete syntax framework \[^9\]. T. Å. Mogensen’s self-interpretation of lambda calculus \[^4\] and the logic Chiron \[^7\], derived from classical NBG set theory, are two other examples of replete syntax frameworks \[^9\].

4 Local Approach

In order to formalize PolyDiff in \( T_R \) we need the ability to reason about the polynomials in \( L_{\text{poly}} \) as syntactic structures (i.e., as syntax trees). This can be achieved by constructing a syntax framework for \( (L_{\text{poly}}, I_R') \) where \( I_R' = \langle T_R', M' \rangle \) is an interpreted theory such that \( T_R' \) is a conservative extension of \( T_R \). Since we seek to reason about just the syntax of \( L_{\text{poly}} \) instead of a larger language, we call this the local approach.

The construction of the syntax framework requires the following steps:

1. Define in \( T_R \) an inductive type whose members are the syntax trees of the polynomials in \( L_{\text{poly}} \). The inductive type should include a new type symbol \( S \) and appropriate constants for constructing and deconstructing expressions of type \( S \). Let \( L_{\text{syn}} \) be the set of expressions of type \( S \). For example, if \( x + 3 \) is a polynomial in \( L_{\text{poly}} \), then an expression like \( \text{plus}(\text{var}(s_x), \text{con}(s_3)) \) could be the expression in \( L_{\text{syn}} \) that denotes the syntax tree of \( x + 3 \). Next add an unspecified “binary” constant \( O_{\text{pd}} \) of type \( S \rightarrow (S \rightarrow S) \) to \( L_R \) (that is intended to represent \( \text{PolyDiff} \)). Let \( T_R' = \langle L_R', I_R' \rangle \) be the resulting extension of \( T_R \). \( T_R' \) is clearly a conservative extension of \( T_R \).

2. In the metatheory of \( T_R' \) define an injective, total function \( Q : L_{\text{poly}} \rightarrow L_{\text{syn}} \) such that, for each polynomial \( u \in L_{\text{poly}} \), \( Q(u) \) is an expression \( e \) that denotes the syntax tree of \( u \). For example, \( Q(x + 3) \) could be \( \text{plus}(\text{var}(s_x), \text{con}(s_3)) \).

3. In the metatheory of \( T_R' \) define an injective, total mapping \( E : L_{\text{syn}} \rightarrow L_{\text{poly}} \) such that, for each expression \( e \in L_{\text{syn}} \), \( E(e) \) is the polynomial whose syntax tree is denoted by \( e \). For example, \( E(\text{plus}(\text{var}(s_x), \text{con}(s_3))) \) would be \( x + 3 \).

Let \( (L_{\text{poly}}, I_R') \) where \( I_R' = \langle T_R', M' \rangle \) and \( M' \) is the set of standard models of \( T_R' \) in simple type theory (see \[^8\]). It is easy to check that \( F = \langle L_{\text{syn}}, Q, E \rangle \) is a syntax framework for \( (L_{\text{poly}}, I_R') \). Notice that \( E \) is the left inverse of \( Q \) and hence the law of disquotation holds: For all \( u \in L_{\text{poly}} \), \( E(Q(u)) = u \).

We are now ready to formalize PolyDiff in \( T_R' \). First, we need to define an operator in \( T_R' \) to represent \( \text{PolyDiff} \). We will use \( O_{\text{pd}} \) for this purpose. We write a sentence \textbf{CompBehavior}

\[
\lambda a, b : S . \text{is-var}(b) \Rightarrow B(a, b, O_{\text{pd}}(a))(b)
\]

in \( T_R' \) where, for all \( u \in L_{\text{poly}} \) and \( x \in L_{\text{var}} \),

\[
B(Q(u), Q(x), O_{\text{pd}}(Q(u))(Q(x)))
\]

holds iff

\[
\text{PolyDiff}(u, x) = E(O_{\text{pd}}(Q(u))(Q(x)))).
\]
That is, we specify the computational behavior of $O_{pd}$ to be the same as that of PolyDiff.

Second, we need to prove that $O_{pd}$ is mathematically correct. We write the sentence MathMeaning for all $u \in L_{poly}$,

$$\text{deriv}(\lambda x: \mathbb{R}. u) = \lambda x: \mathbb{R}. E(O_{pd}(Q(u))(Q(x)))$$

in the metatheory of $T'_R$ that says $O_{pd}$ computes derivatives (with respect to the variable $x$). That is, we specify the mathematical meaning of $O_{pd}$ to be the same as that of deriv. And then we prove in $T'_R$ that MathMeaning follows from CompBehavior. The proof requires showing that the value of $E(O_{pd}(Q(u))(Q(x)))$ equals deriv($\lambda x: \mathbb{R}. u$, $x$), the derivative of $\lambda x: \mathbb{R}. u$ at $x$, which is

$$\lim_{h \to 0} \frac{(\lambda x: \mathbb{R}. u)(x + h) - (\lambda x: \mathbb{R}. u)(x)}{h}.$$ 

The details of the proof are found in any good calculus textbook such as [17].

Third, we need to show how PolyDiff can be used to compute the derivative of a function $\lambda x: \mathbb{R}. u$ in $T'_R$. There are two ways. The first way is to instantiate the sentence CompBehavior with $Q(u)$ and $Q(x)$ and then simplify the resulting expression (e.g., by beta-reduction). The second way is to replace $E(O_{pd}(Q(u))(Q(x)))$ in MathMeaning with the result of applying PolyDiff to $u$ and $x$. The first way requires that PolyDiff is implemented in $T'_R$ as $O_{pd}$. The second way does not require that PolyDiff is implemented in $T'_R$, but only that its meaning is specified in $T'_R$.

The local approach is commonly used to reason about the syntax of expressions in a formal theory. It embodies a deep embedding [1] of the object language (e.g., $L_{poly}$) into the underlying formal language (e.g., $L_R$). The local approach to reason about syntax can be employed in almost any proof assistant in which it is possible to define an inductive type (e.g., see [11,20]).

The local approach has both strengths and weaknesses. These are the strengths of the local approach:

1. **Indirect Reasoning about the syntax of $L_{poly}$ in the Theory.** In $T'_R$ using $L_{syn}$, we can indirectly reason about the syntax of the polynomials in $L_{poly}$. This thus enables us to specify the computational behavior of PolyDiff via $O_{pd}$.

2. **Direct Reasoning about the syntax of $L_{poly}$ in the Metatheory.** In the metatheory of $T'_R$ using the formula

$$\text{for all } u \in L_{poly}, x \in L_{var}, \text{PolyDiff}(u, x) = E(O_{pd}(Q(u))(Q(x))),$$

we can directly reason about the syntax of the polynomials in $L_{poly}$ and specify the mathematical meaning of PolyDiff.

And these are the weaknesses:

1. **Syntax Problem.** We cannot refer in $T'_R$ to the syntax of polynomials. Also the variable $x$ is free in $x + 3$ but not in $Q(x + 3) = \text{plus}(\text{var}(s_2), \text{con}(s_1))$. As a result, $Q$ and $E$ cannot be defined in $T'_R$ and thus PolyDiff cannot be fully formalized in $T'_R$. In short, we can reason about syntax in $T'_R$ but not about the interplay of syntax and semantics in $T'_R$. 


2. **Coverage Problem.** The syntax framework \( F \) can only be used for reasoning about the syntax of polynomials. It cannot be used for reasoning, for example, about rational expressions. To do that a new syntax framework must be constructed.

3. **Extension Problem.** \( L_{\text{poly}}, L_{\text{syn}}, Q, \) and \( E \) must be extended each time a new constant of type \( R \) is defined in \( T'_R \).

In summary, the local approach only gives us indirect access to the syntax of polynomials and must be modified to cover new or enlarged contexts. If \( L_{\text{obj}} \) (which is \( L_{\text{poly}} \) in our example) does not contain variables, then we can define \( E \) to be a total operator in the theory. (If the theory is over a traditional logic, we will still not be able to define \( Q \) in the theory.) This variant of the local approach is used, for example, in the Agda reflection mechanism [19].

5 Global Approach

The **global approach** described in this section utilizes a replete syntax framework. Assume that we have modified \( T_R \) and simple type theory so that there is a replete syntax framework \( F = (L_{\text{syn}}, Q, E) \) for \( (L_R, I_R) \) where \( I_R = (T_R, \mathcal{M}) \) and \( \mathcal{M} \) is the set of standard models of \( T_R \) in the modified simple type theory. Let us also assume that \( L_{\text{syn}} \) is the set of expressions of type \( S \) and \( L_R \) includes a constant \( O_{pd} \) of type \( S \to (S \to S) \). By virtue of \( F \) being replete, \( F \) embodies a deep embedding of \( L_R \) into itself.

As far as we know, no one has ever worked out the details of how to modify simple type theory so that it admits built-in quotation and evaluation for the full language of a theory. However, we have shown how NBG set theory can be modified to admit built-in quotation and evaluation for its entire language [7]. Simple type theory can be modified in a similar way. We plan to present a version of simple type theory with a replete syntax framework in a future paper.

We can formalize \( \text{PolyDiff} \) in \( T_R \) as follows. We will write \( \text{quote}(e) \) and \( \text{eval}(e) \) as \( \lceil e \rceil \) and \( \llbracket e \rrbracket \), respectively. First, we define the operator \( O_{pd} \) in \( T_R \) to represent \( \text{PolyDiff} \). We write a sentence \( \text{CompBehavior} \)

\[
\lambda a, b : S . \text{is-poly}(a) \land \text{is-var}(b) \Rightarrow B(a, b, O_{pd}(a)(b))
\]

in \( T_R \) where, for all \( u \in L_{\text{poly}} \) and \( x \in L_{\text{var}} \),

\[
B(\lceil u \rceil, \llbracket x \rrbracket, O_{pd}(\lceil u \rceil)(\llbracket x \rrbracket))
\]

holds iff

\[
\text{PolyDiff}(u, x) = \llbracket O_{pd}(\lceil u \rceil)(\llbracket x \rrbracket) \rrbracket.
\]

That is, we specify the computational behavior of \( O_{pd} \) to be the same as that of \( \text{PolyDiff} \).

Second, we prove in \( T_R \) that \( O_{pd} \) is mathematically correct. We write the sentence \( \text{MathMeaning} \)

\[
\forall a : S . \text{is-poly}(a) \Rightarrow \text{deriv}(\lambda x : R . \llbracket a \rrbracket) = \lambda x : R . \llbracket O_{pd}(a)(\llbracket x \rrbracket) \rrbracket
\]
in $T_R$ that says $O_{pd}$ computes derivatives (with respect to the variable $x$). That is, we specify the mathematical meaning of $O_{pd}$ to be the same as that of $\text{deriv}$. And then we prove in $T'_R$ that $\text{MathMeaning}$ follows from $\text{CompBehavior}$.

Third, we use $\text{PolyDiff}$ to compute the derivative of a function $\lambda x : \mathbb{R} . u$ in $T'_R$ in either of the two ways described for the local approach.

The strengths of the global approach are:

1. **Direct Reasoning about the syntax of $L_{\text{poly}}$ in the Theory.** In $T_R$ using $L_{\text{syn}}$, $\text{quote}$, and $\text{eval}$, we can directly reason about the syntax of the polynomials in $L_{\text{poly}}$. As a result, we can formalize $\text{PolyDiff}$ in $T_R$ as described in the Introduction.

2. **Direct Reasoning about the syntax of $L_R$ in the Theory.** In $T_R$ using $L_{\text{syn}}$, $\text{quote}$, and $\text{eval}$, we can directly reason about the syntax of the expressions in the entire language $L_R$. As a result, the syntax framework $F$ can cover all current and future syntax reasoning needs. Moreover, we can express such things as syntactic side conditions, formula schemas, and substitution for a variable directly in $T_R$ (see [7] for details).

In short, not only does the global approach enable us to formalize $\text{PolyDiff}$ in $T_R$, it provides us with the facility to move syntax-based reasoning from the metatheory of $T_R$ to $T_R$ itself. This seems to be a wonderful result that solves the problem of formalizing syntax-based mathematical algorithms. Unfortunately, the global approach has the following serious weaknesses that temper the enthusiasm one might have for its strengths:

1. **Evaluation Problem.**

   Claim: $\text{eval}$ cannot be defined on all expressions in $L_R$.

   Proof: Suppose $\text{eval}$ is indeed total. $T_R$ is sufficiently expressive, in the sense of Gödel’s incomplete theorem, to apply the the diagonalization lemma [3], to obtain a formula $\text{LIAR}$ such that

   $$\text{LIAR} = \lnot \llbracket \text{LIAR} \rrbracket \Rightarrow \neg \llbracket \text{LIAR} \rrbracket.$$

   Then

   $$\llbracket \text{LIAR} \rrbracket = \llbracket \lnot \llbracket \text{LIAR} \rrbracket \rrbracket = \neg \llbracket \text{LIAR} \rrbracket,$$

   which is a contradiction. □

   This means that the liar paradox limits the use of $\text{eval}$ and, in particular, the law of disquotation does not hold universally, i.e., there are expressions $e$ in $L_R$ such that $\llbracket \lnot e \rrbracket \neq e$.

2. **Variable Problem.** The variable $x$ is not free in the expression $\lnot x + 3 \Rightarrow$ (or in any quotation). However, $x$ is free in $\llbracket x + 3 \rrbracket$ because $\llbracket x + 3 \rrbracket = x + 3$. If the value of the variable $e$ is $\lnot x + 3 \Rightarrow$, then both $e$ and $x$ are free in $\llbracket e \rrbracket$ because $\llbracket e \rrbracket = \llbracket x + 3 \rrbracket = x + 3$.

   This example shows that the notions of a free variable, substitution for a variable, etc. are significantly more complex when expressions contain $\text{eval}$.
3. **Extension Problem.** We can define $L_{\text{con}} \subseteq L_{\text{syn}}$ in $T_R$ as the language of expressions denoting the syntactic values of constants in $L_R$.

Claim: Assume the set of constants in $L$ is finite and $T'_R = (L'_R, \Gamma'_R)$ is an extension of $T_R$ such that there is a constant in $L'_R$ but not in $L_R$. Then $T'_R$ is not a conservative extension of $T_R$.

Proof: Let $\{c_1, \ldots, c_n\}$ be the set of constants in $L$. Then

$$L_{\text{con}} = \{\llbracket c_1 \rrbracket, \ldots, \llbracket c_n \rrbracket\}$$

is valid in $T_R$ but not in $T'_R$. $\Box$.

This shows that in the global approach the development of a theory via definitions requires that the notion of a conservative extension be weakened.

4. **Interpretation Problem.**

Let $T = (L, \Gamma)$ and $T' = (L', \Gamma')$ be two theories in a simple type theory that has been modified to admit built-in quotation and evaluation for the entire language of a theory.

Claim: Let $\Phi$ be an interpretation of $T$ in $T'$ such that $\Phi$ is a homomorphism with respect to the logical operators of the underlying logic. Then $\Phi$ must be injective on the constants of $L$.

Proof: Assume that $\Phi$ is not injective on constants. Then there are two different constants $a, b$ such that $\Phi(a) = \Phi(b)$. $\llbracket a \rrbracket \neq \llbracket b \rrbracket$ is valid in $T$. Hence

$$\Phi(\llbracket a \rrbracket \neq \llbracket b \rrbracket) = \llbracket \Phi(a) \rrbracket \neq \llbracket \Phi(b) \rrbracket$$

since $\Phi$ is a homomorphism, and the latter inequality must be valid in $T'$ since $\Phi$ is an interpretation (which maps valid formulas of $T$ to valid formulas of $T'$). However, since $Q$ is injective, our hypothesis $\Phi(a) = \Phi(b)$ implies $\llbracket \Phi(a) \rrbracket = \llbracket \Phi(b) \rrbracket$, which is a contradiction. $\Box$

This shows that the use of interpretations is more cumbersome in a logic that admits quotation than one that does not.

6 Conclusion

Syntax-based mathematical algorithms are employed throughout mathematics and are one of the main offerings of computer algebra systems. They are difficult, however, to formalize since they manipulate the syntactic structure of expressions in mathematically meaningful ways. We have presented two approaches to formalizing syntax-based mathematical algorithms in a formal theory, one called the **local approach** and the other the **global approach**. Both are based on the notion of a **syntax framework** which provides a foundation for integrating reasoning about the syntax of expressions with reasoning about what the expressions mean. Syntax frameworks include a syntax representation, a syntax language for reasoning about the representation, and quotation and evaluation mechanisms. Common syntax reasoning systems are instances of a syntax framework.
The local approach and close variants are commonly used for formalizing syntax-based mathematical algorithms. Its major strength is that it provides the means to formally reason about the syntactic structure of expressions, while its major weakness is that the mathematical meaning of a syntax-based mathematical algorithm cannot be expressed in the formal theory. Another weakness is that an application of the local approach cannot be easily extended to cover new or enlarged contexts.

The global approach enables one to reason in a formal theory $T$ directly about the syntactic structure of the expressions in $T$ as well as about the interplay of syntax and semantics in $T$. As a result, it is possible to fully formalize syntax-based algorithms like PolyDiff and move syntax-based reasoning, like the use of syntactic side conditions, from the metatheory of $T$ to $T$ itself. Unfortunately, these highly desirable results come with a high cost: Significant change must be made to the underlying logic as illustrated by the Evaluation, Variable, Extension, and Interpretation Problems given in the previous section.

One of the main goals of the MathScheme project [2], led by J. Carette and the author, is to see if the global approach can be used as a basis to integrate axiomatic and algorithmic mathematics. The logic Chiron [7] demonstrates that it is possible to modify a traditional logic to support the global approach. Although we have begun an implementation of Chiron, it remains an open question whether a logic modified in this way can be effectively implemented. As part of the MathScheme project, we are now pursuing this problem as well as developing the techniques needed to employ the global approach.

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