# ragni-cas - A Pure Ruby Automatic Differentiation Library for Fast Prototyping of Interfaces

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#### Abstract

This work presents a new Ruby library for symbolic and automatic differentiation, that exposes minimalistic CAS capabilities — i.e. simplifications, substitutions, evaluations, etc. Library aims at rapid prototyping of numerical interfaces and code generation for different target languages, separating mathematical expression from exportation rules — e.g. models from numerical conditioning best practices.

The library is implemented in pure Ruby language and compatible with all Ruby interpreter flavours.

Keywords: CAS, code-generation, Ruby

#### 1. Motivation and significance

Ruby [1] is a purely object-oriented scripting language designed in the mid-1990s by Yukihiro Matsumoto (also known as Matz), internationally standardized since 2012 as ISO/IEC 30170.

With the advent of the *Internet of Things*, a written from scratch version of the *Ruby* interpreter called *mRuby* (*eMbedded Ruby*) [2] has been published on *GitHub* by Matsumoto, in 2014. The new interpreter is a lightweight implementation, aimed at both low power devices and PC, and complies with

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the standard[3]. mRuby has a completely new API, and it is designed to be embedded in complex projects as a front-end interface — e.g. a numerical optimization suite may use mRuby to get problem input definitions.

The Ruby code-base exposes a large set of utilities in core and standard library, that can be furthermore expanded through modules, contained in gems. Even if a high number of gems are deployed and available, there is no module that implements an **automatic symbolic differentiation** (ASD) [4] engine that handles basic computer algebra routines, compatible with all different Ruby interpreters flavours.

Ruby has matured its fame as a web oriented language with Rails, and can efficiently generate code in other languages. An ASD-capable gem is the foundamental step to rapidly develop specific code generators for well known software — e.g. IPOPT [5].

The module described in this work, is a gem implemented in pure *Ruby* code — compatible with all standardized interpreters — that is able to perform symbolic differentiation (SD) and some computer algebra operations [6]. The library aims at:

- be an instrument for rapid development of prototype interface for numerical algorithms and exporting code generated in different target languages;
- generate rapidly descriptions of mathematical models, with easy to implement conditioning rules for numerical issues, changing on request how the code is exported, and how expressions are formulated in the target language;
- separate mathematical expressions from numerical conditioning and workarounds:

• create a complete open-source CAS system for the standard *Ruby* language, as a long-term ambitious impact.

This is not the first gem that tries to implement a CAS. The available computer algebra library for Ruby are:

Rucas [7], Symbolic [8] gems at early stage and with discontinued development status; they offer basic simplification routines. There is no differentiation method, but it is one of the milestones.

Symengine [9] is a wrapper for the C++ library symengine. The backend library is very complete, but it is compatible only with the RVM Ruby interpreter and has several dependencies. At the moment, the SciRuby [10] project reports the gem as broken, and removed it from its codebase. From a direct test, when performing SD of an arbitrary function, the engine always returns nil.

In Section 2 module's container and tree structure is explained in detail and applied to basic CAS tasks. In Section 3 two examples on how to use the library as code generator or as interface are described. In Section 4, the reasons behind the implementation and the long term desired impact are depicted.

### 2. Software description

# 2.1. Software Architecture

ragni-cas is an object oriented ASD gem that supports some computer algebra routines such as simplifications and substitutions. When gem is required, it overloads methods of Fixnum and Float classes, making them compatible with foundamental symbolic classes.

Each symbolic expression (or operation) is the instance of an object, that inherits from a common virtual ancestor: CAS::Op. An operation encapsulates sub-operations recursively, building a linked tree, that is the mathematical equivalent of function composition:

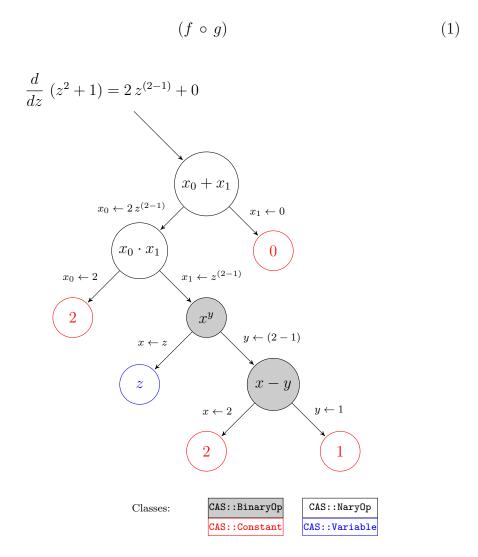


Figure 1: Tree of the expression derived in listing 1

When a new operation is created, it is appended to the tree. The number of branches are determined by the parent container class of the current symbolic function. There are three possible containers. Single argument operations — e.g.  $\sin(\cdot)$  — have as closest parent the CAS::Op class, that links to one sub-tree. Expressions with two arguments — e.g. difference or exponential function — inherit from CAS::BinaryOp, that links to two sub-tree. Operations with arbitrary number of arguments — e.g. sum and product — have as parent the CAS::NaryOp¹, that links to an arbitrary number of sub-tree. Figure 1 contains a graphical representation. The different kind of containers allows to introduce some properties — i.e. associativity and commutativity for sums and multiplications [11]. Each container exposes the sub-tree as instance properties. Containers interfaces and inheritances are shown in Figure 2.

Terminal leafes of the graph are the classes CAS::Constant, CAS::Variable and CAS::Function. The first models a simple numerical value, while the second represents an independent variable, that can be used to perform derivatives and evaluations, and the latter is a prototype of implicit functions. As for now, those leafes exemplify only real scalar expressions, with definition of complex, vectorial and matricial extensions as milestones for the next major release.

SD (CAS::Op#diff) crosses the graph until it reaches ending nodes. A terminal node is the starting point for derivatives accumulation, the mathematical equivalent of the chain rule:

$$(f \circ g)' = (f' \circ g) g' \tag{2}$$

The recursiveness is used also for simplifications (CAS::Op#simplify), substitutions (CAS::Op#subs), evaluations (CAS::Op#call) and code generation.

<sup>&</sup>lt;sup>1</sup>Please note that this container is still at experimental stage

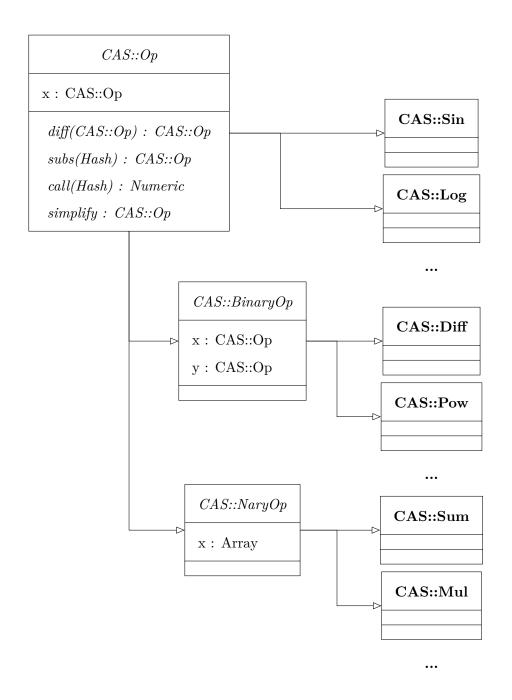


Figure 2: Simplified version of classes interface and inheritance  $\,$ 

## 2.2. Software Functionalities

## 2.2.1. Software installation and prerequisites

No additional dependencies are required. The gem can be installed through rubygems.org provider<sup>2</sup>. Functionalities must be required runtime using the Kernel method: require 'ragni-cas'. All methods and classes are incapsulated in the module CAS.

#### 2.2.2. Basic Functionalities

SD is performed with respect to independent variables (CAS::Variable) through forward accumulation, even for implicit functions. The differentiation is done by the method CAS::Op#diff, having a CAS::Variable as argument:

Listing 1: Differentiation example

Automatic differentiation (AD) is included as plugin and exploits dual numbers [12]. This differentiation strategy is useful in case of complex expressions, when explicit derivative's tree may exceed the call stack depth, that is platform dependent.

Simplifications are not executed automatically, after differentiations. Each node of the tree knows rules for simplify itself, and rules are called recursively, exactly like ASD. Simplifications that require an *heuristic expansion* of the subgraph — i.e. some trigonometric identities — are not defined

<sup>&</sup>lt;sup>2</sup>gem install ragni-cas

for now, but can be easily achieved through **substitutions**:

Listing 2: Simplification example

```
x, y = CAS::vars 'x', 'y'  # creates two variables
f = CAS.log( CAS.sin( y ) )  # symbolic expression
f.subs y: CAS.asin(CAS.exp(x))  # perform substitution
f.simplify  # simplify expression
# => x
```

The tree is numerically **evaluated** when independent variables values are provided in a feed dictionary. The graph is reduced recursively to a single numeric value:

Listing 3: Tree evaluation example

```
x = CAS.vars 'x'  # creates a variable
f = x ** 2 + 1  # define a symbolic expression
f.call x => 2  # evaluate for x = 2
# => 5
```

Symbolic expressions can be used to create comparative expressions, that are stored in special container classes, modeled by the ancestor CAS::Condition — e.g.  $f(\cdot) \geq g(\cdot)$ . This allow the definition of piecewise functions — e.g.  $\max(f(\cdot),g(\cdot))$ .

Listing 4: Expressions and Piecewise functions

```
x, y = CAS.vars 'x', 'y'
f = CAS.declare :f, x
g = CAS.declare :g, x, y
f.greater_equal g
# => (f(x) >= g(x, y))
CAS::max f, g
# => ((f(x) >= g(x, y)) ? f(x) : g(x, y))
```

# 2.2.3. Metaprogramming and Code-Generation

The library is developed explicitly for **metaprogramming** and **generation of code**. Expressions can be exported as source code or used as prototypes for callable *closures* (Proc objects):

Listing 5: Graph evaluation example

```
x = CAS::vars 'x'  # creates a variable
f = CAS::log(CAS::sin(x))  # define a symbolic function

proc = f.as_proc  # exports callable lambda
proc.call 'x' => Math::PI/2
# => 0.0
```

Compiling a closure of a tree is like making its snapshot, thus any further manipulation of the expression do not update the callable object. This drawback is balanced by the faster execution time of a Proc: when a graph needs only to be evaluated in a iterative algorithm, transforming it in a closure reduces the execution time per iteration.

Code generation should be flexible enough to export expressions' trees in a user's target language. Generation methods for common languages are included in specific **plugins**. Users can furthemore expand exporting capabilities by writing specific exportation rules, overriding method for existing plugin, or desining their own exporter:

Listing 6: Example of Ruby code generation plugin

```
# Definition
module CAS
{
    # . . .
    CAS::Variable => Proc.new { "#{name}" }
    CAS::Sin => Proc.new { "Math.sin(#{x.to_ruby})" },
    # . . .
}.each do |cls, prc|
    cls.send(:define_method, :to_ruby, &prc)
end
end

# Usage
x = CAS.vars 'x'
(CAS.sin(x)).to_ruby
# => Math.sin(x)
```

# 3. Illustrative Examples

# 3.1. Code Generation as C Library

TIn this example a model is exported as C library. c-opt plugin implements advanced features such as code optimization and generation of libraries.

The library example implements the model:

$$f(x,y) = x^y + g(x)\log(\sin(x^y)) \tag{3}$$

Expression g(x) belongs to a external object, declared as  $g_{impl}$ , and its interface is described in  $g_{impl}$ . The code is optimized: the intermediate operation  $x^y$  is evaluated once, even if appears twice in our model. The C function that implements our model f(x,y) is declared with the token  $f_{impl}$ . The exporter uses as default type double for variables and function returned values.

Listing 7: Calling optimized-C exporter for library generation

```
require 'ragni-cas/c-opt'

# Model
x, y = CAS.vars :x, :y
g = CAS.declare :g, x

f = x ** y + g * CAS.log(CAS.sin(x ** y))

# Code Generation
g.c_name = 'g_impl'  # g token

CAS::CLib.create "example" do
  include_local "g_impl"  # g header
  implements_as "f_impl", f # token for f
end
```

Library created by class CLib contains the following code:

Listing 8: C Header

Listing 9: C Source

```
// Header file for library: example.c
                                               // Source file for library: example.c
#ifndef example_H
                                               #include "example.h"
#define example H
// Standard Libraries
                                               double f_impl(double x, double y) {
#include <math.h>
                                                 double _{t_0} = pow(x, y);
                                                 double __t_1 = g_impl(x);
// Local Libraries
                                                 double _t_2 = \sin(_t_0);
#include "g_impl"
                                                 double _{-t_3} = \log(_{-t_2});
                                                 double _{t_4} = (_{t_1} + _{t_3});
// Definitions
                                                 double _{t_5} = (_{t_0} + _{t_4});
// Functions
                                                 return __t_5;
double f_impl(double x, double y);
#endif // example_H
                                               // end of example.c
```

The function g(x) models the following operation:

$$g(x) = (\sqrt{x+a} - \sqrt{x}) + \sqrt{\pi + x} \tag{4}$$

and may suffer from catastrophic cancellation [13]. Users can specialize code generation rules for this particular expression, conditioned through rationalization and instead of modifying the model g(x), in listing 10, the rationalization is extended to all differences of square roots <sup>3</sup>. For more insight about \_\_to\_c and \_\_to\_c\_impl please refer to the software manual.

Listing 10: Conditioning in exporting function

```
# Model
a = CAS.declare "PARAM_A"

g = (CAS.sqrt(x + a) — CAS.sqrt(x)) + CAS.sqrt(CAS::Pi + x)

# Particular Code Generation for difference between square roots.
module CAS
```

<sup>3</sup>i.e.: 
$$\sqrt{\phi(\cdot)} - \sqrt{\psi(\cdot)} = \frac{\phi(\cdot) - \psi(\cdot)}{\sqrt{\phi(\cdot)} + \sqrt{\psi(\cdot)}}$$

```
class Diff
   alias :__to_c_impl_old :__to_c_impl

def __to_c_impl(v)
   if @x.is_a? CAS::Sqrt and @y.is_a? CAS::Sqrt
        "(#{@x.x._to_c(v)} + #{@y.x._to_c(v)}) / " +
        "( #{@x.__to_c(v)} + #{@y.__to_c(v)} )"
   else
        self.__to_c_impl_old(v)
   end
  end
end

clib = CAS::CLib.create "g_impl" do
  define "PARAM_A()", 1.0  # Arbitrary value for PARAM_A
  define "M_PI", Math::Pi
  implements_as "g_impl", g
end
```

It should be noted the **separation between the model** — that does not contain conditioning — **and the code generation rule** — that overloads, for this particular case and this particular language, the predefined code generation rule. Obviously, a user can decide to apply directly the conditioning on the model. The result of listing 10 is reported:

Listing 11: g\_impl Header

Listing 12: g\_impl Source

```
// Header file for library: g_impl.c
                                                // Source file for library: g_impl.c
#ifndef g_impl_H
                                               #include "g_impl.h"
#define g_impl_H
                                                double g_impl(double x) {
// Standard Libraries
                                                 double __t_0 = PARAM_A();
#include <math.h>
                                                  double _{t_1} = (x + _{t_0});
                                                  double __t_2 = sqrt(__t_1);
// Local Libraries
                                                  double _{-t_3} = sqrt(x);
                                                  double _{-}t_{4} = (_{-}t_{1} + x) / (_{-}t_{2} +
                                                      __t_3 );
// Definitions
                                                 double _{-t_5} = (M_PI + x);
#define PARAM_A() 1.0
                                                  double __t_6 = sqrt(__t_5);
#define M PI 3.141592653589793
                                                 double _{t_7} = (_{t_4} + _{t_6});
// Functions
                                                  return __t_7;
double g_impl(double x);
#endif // g_impl_H
                                                // end of g_impl.c
```

### 3.2. Using the module as interface

As example, an implementation of an algorithm that extimates the *order* of convergence for trapezoidal integration scheme [14] is provided, using the symbolic differentiation as interface.

Given a function f(x), the trapezoidal rule for primitive estimation in the interval [a, b] is:

$$I_n(a,b) = h\left(\frac{f(a) + f(b)}{2} + \sum_{k=1}^{n-1} f(a+kh)\right)$$
 (5)

with h = (b-a)/n, where n mediates the integration's step size. When exact primitive F(x) is known, approximation error is:

$$E[n] = F(b) - F(a) - I_n(a, b)$$
(6)

This error shows a direct relation:

$$E[n] \propto C \, n^{-p} \tag{7}$$

where p is the convergence order. Using a different value for n, for example  $2\,n$ :

$$\frac{E[n]}{E[2n]} \approx 2^p \quad \to \quad p \approx \log_2\left(\frac{E[n]}{E[2n]}\right)$$
 (8)

Following listings contain the implementation of the described procedure using the described gem and the well known *Python* [15] library *sympy* [16].

Listing 13: Ruby version

Listing 14: Python version

```
require 'ragni cas'
                                              import sympy
                                              import math
def integrate(f, a, b, n)
                                              def integrate(f, a, b, n):
  h = (b - a) / n
                                                 h = (b - a)/n
                                                 x = sympy.symbols('x')
  func = f.as_proc
                                                  func = sympy.lambdify((x), f)
  sum = ((func.call 'x' => a) +
                                                  sums = (func(a) +
        (func.call 'x' => b)) / 2.0
                                                          func(b)) / 2.0
  for i in (1...n)
                                                 for i in range(1, n):
    sum += (func.call 'x' => (a + i*h))
                                                      sums += func(a + i*h)
  return sum * h
                                                  return sums * h
end
def order(f, a, b, n)
                                              def order(f, a, b, n):
  x = CAS.vars 'x'
                                                 x = sympy.symbols('x')
  f_ab = (f.call x => b) -
                                                  f_ab = sympy.Subs(f, (x), (b)).n() \rightarrow
         (f.call x => a)
                                                         sympy.Subs(f, (x), (a)).n()
  df = f.diff(x).simplify
                                                 df = f.diff(x)
  f_1n = integrate(df, a, b, n)
                                                 f_1n = integrate(df, a, b, n)
  f_2n = integrate(df, a, b, 2 * n)
                                                  f_2n = integrate(df, a, b, 2 * n)
  return Math.log(
                                                 return math.log(
    (f_ab — f_1n) /
                                                   (f_ab - f_1n) /
    (f_ab - f_2n),
                                                    (f_ab - f_2n),
  2)
                                                 2)
end
x = CAS.vars 'x'
                                              x = sympy.symbols('x')
f = CAS.arctan x
                                              f = sympy.atan(x)
puts(order f, -1.0, 1.0, 100)
                                              print(order(f, -1.0, 1.0, 100))
# => 1.999999974244451
                                              # => 1.999999974244451
```

#### 4. Impact

There are different complete CAS systems on the market, with complete solutions for analysis of analytical models. But exporting a model, for optimization or any other research activity, requires a lot of work, even with a good CAS software.

This library is a midpoint between a CAS and an AD library. It allows to manipulate expressions while mantaining the complete control on how the code is exported. Each rule is overloaded and applied runtime, without the need of compilation. Each user's model may include the mathematical description, code generation rules and high level logic that should be intrisic to such a rule — e.g. exporting gradients as **patterns** instead of matrices.

Our research group is including ragni-cas in a solver for optimal control problem with indirect methods, as interface for problems' description [17].

As a long term ambitious impact, this library will become a complete CAS for *Ruby* language, filling the empty space reported by *SciRuby* for symbolic math engines. This will require time, and the gem's MIT license allows everyone to contribute to the project.

# 5. Conclusions

This work presents a pure *Ruby* library that implements a minimalistics CAS with automatic and symbolic differentiation that is aimed at code generation and metaprogramming. Although at an early developing stage, the module has promising feature, some of them shown in Section 3. Also, this is the only gem that implements symbolic manipulation for this language.

Language features and lack of dependencies simplify the use of the module as interface, extending model definition capabilities for numerical algorithms. All core functionalities and basic mathematics are defined, with the plan to include more features in next releases. Reopening a class guarantees a *liquid* behaviour, in which users are free to modify core methods and their needs.

Library is published in *rubygems.org* repository and versioned on *github.com*, under MIT license. It can be included easily in projects and in inline interpreter, or installed as a standalone gem.

# Acknowledgements

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# Current code version

Nr.	Code metadata description	Please fill in this column
C1	Current code version	0.0.0
C2	Permanent link to code/repository	github.com/MatteoRagni/cas-rb &
	used for this code version	rubygems.org/gems/ragni-cas
С3	Legal Code License	MIT
C4	Code versioning system used	git (GitHub)
C5	Software code languages, tools, and	Ruby
	services used	
С6	Compilation requirements, operat-	$Ruby \ge 2.x$ , $pry$ for testing console
	ing environments	(optional)
C7	If available Link to developer docu-	rubydoc.info/gems/ragni-cas
	mentation/manual	
С8	Support email for questions	info@ragni.me

Table 1: Code metadata (mandatory)