

# SymPy - Python library for symbolic mathematics

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## Contents of this talk:

- Review of the current state:
  - History
  - Different approaches to symbolic manipulation we tried
  - Symbolic limits
  - Integration with SAGE
- Future
  - where to go from here
  - our priorities and principles

- A Python library for symbolic mathematics
- <http://code.google.com/p/sympy/>

Why symbolic mathematics? The same reasons people use Maple/Mathematica, but we want to use it from Python.

```
>>> from sympy import Symbol, limit, sin, oo
>>> x=Symbol("x")
>>> limit(sin(x)/x, x, 0)
1
>>> integrate(x+sinh(x), x)
>>> (1/2)*x**2 + cosh(x)
```

## What SymPy can do

- basics (expansion, complex numbers, differentiation, taylor (laurent) series, substitution, arbitrary precision integers, rationals and floats, pattern matching)
- noncommutative symbols
- limits and some integrals
- polynomials (division, gcd, square free decomposition, groebner bases, factorization)
- symbolic matrices (determinants, LU decomposition...)
- solvers (some algebraic and differential equations)
- 2D geometry module
- plotting (2D and 3D)

Other symbolic manipulation software: GiNaC, Giac, Qalculate, Yacas, Eigenmath, Axiom, PARI, Maxima, SAGE, Singular, Mathematic, Octave, ...

Problems:

- all use their own language (except GiNaC, Giac and SAGE)
- GiNaC and Giac still too complicated (C++), difficult to extend

What we want

- Python library and that's it (no environment, no new language, nothing)
- Rich functionality
- Pure Python (non Python modules could be optional) – works on Linux, Windows, Mac out of the box

Acutally, I didn't tell the full truth, we have one nice thing – isympy:

```
$ bin/isympy
```

```
Python 2.4.4 console for SymPy 0.5.6-hg. These commands work
```

```
>>> from __future__ import division
```

```
>>> from sympy import *
```

```
>>> x, y, z = symbols('xyz')
```

```
>>> k, m, n = symbols('kmn', integer=True)
```

```
In [1]: integrate(ln(x), x)
```

```
Out[1]: -x + x*log(x)
```

# Unicode prettyprinting

```
In [4]: a = Symbol("alpha")
```

```
In [5]: a
```

```
Out[5]:  $\alpha$ 
```

```
In [6]: b = Symbol("beta")
```

```
In [7]: Integral((a+b)**2, a)
```

```
Out[7]:
```

```
|  
|      2  
| (α + β)  dα  
|
```

```
In [8]: Integral((a+b)**2, a).doit()
```

```
Out[8]:
```

```
  3  
α  + α*β  + β*α  
  3
```

Recent changes in isympy:

- pretty printing by default
- use unicode printing if available

- aims to glue together every useful open source mathematics software package and provide a transparent interface to all of them
- <http://www.sagemath.org/>
- More on relationship between SAGE and SymPy later

```
sage: limit(sin(x)/x, x=0)
```

```
1
```

```
sage: integrate(x+sinh(x), x)
```

```
cosh(x) + x^2/2
```

```
In [1]: limit(sin(x)/x, x, 0)
```

```
Out[1]: 1
```

```
In [2]: integrate(x+sinh(x), x)
```

```
Out[2]: (1/2)*x**2 + cosh(x)
```



In 2005, I wanted to use symbolic mathematics in Python

- pyginac used boost-python, very slow compilation (30s per file),
- I wrote swiginac together with Ola Skavhaug in SWIG, it works, but too difficult to extend the GiNaC core behind it
- Is it really that difficult to have a system, that can calculate all I need and still be easy to extend?

Let's reinvent the wheel for the 35th time.

- end of summer 2005: I implemented my first code, mostly translating ideas from GiNaC to Python.
- spring 2006: I discovered the Gruntz algorithm for limits
- end of summer 2006: I implemented limits in SymPy
- February 2007: Fabian Seoane joined and this was the boost to SymPy's development
- Google Summer of Code, SymPy is under the umbrella of Python Software Foundation, the Space Telescope Science Institute and Portland State University

# Contributions

- Fabian: everything, without him, SymPy wouldn't be here
- Mateusz (GSoC): concrete math, symbolic integration, many bugfixes
- Jason (GSoC): geometry, a lot of bugfixes
- Robert (GSoC): polynomials (groebner basis et al.)
- Brian (GSoC): plotting
- Chris (GSoC): linear algebra
- Pearu: new core (10x to 100x speedup)
- Fredrik: fast floating point arithmetics in pure Python (faster than Decimal)
- Jurjen: pretty printing
- Kirill: unicode printing, a lot of bugfixes
- others: bug reports, bug fixes

# Approaches we tried I

GiNaC ".eval()" approach, without their "ex" class:

- classes: Basic, Add, Mul, Pow, Rational, Funcion (sin, cos, exp, log)

Example:

- $x + y + x \rightarrow \text{Add}(\text{Add}(\text{Symbol}("x"), \text{Symbol}("y")), \text{Symbol}("x"))$

```
e = Add(Add(x, y), x)
```

```
e.eval()
```

"e" becomes  $\text{Add}(\text{Mul}(2, x), y)$

Disadvantages:

- User has to call ".eval()" by hand
- Wasteful construction of instances

# Approaches we tried II

Automatic evaluation of ".eval()":

- classes: Basic, Add, Mul, Pow, Rational, Funcion (sin, cos, exp, log)

Example:

- $x + y + x \rightarrow \text{Add}(\text{Add}(\text{Symbol}("x"), \text{Symbol}("y")), \text{Symbol}("x"))$

$e = \text{Add}(\text{Add}(x, y), x)$

"e" becomes  $\text{Add}(\text{Mul}(2, x), y)$  automatically

Disadvantages:

- Wasteful construction of instances

# Approaches we tried III

Not using ".eval()" at all, simplify immediately in "\_\_new\_\_"

- classes: Basic, Add, Mul, Pow, Rational, Funcion (sin, cos, exp, log)

Example:

- $x + y + x \rightarrow \text{Add}(\text{Add}(\text{Symbol}("x"), \text{Symbol}("y")), \text{Symbol}("x"))$

`e = Add(Add(x, y), x)`

"e" becomes `Add(Mul(2, x), y)` immediately, no intermediate classes constructed

# Approaches we tried IV

How to deal with functions:

- Sin, ApplySin, Cos, ApplyCos, ...
  - one class to represent a function (sin)
  - another class to represent "applied" function (sin(x))
  - SAGE way
- sin, cos, ...
  - just one class to represent a function (sin)
  - instance of this class to represent "applied" function (sin(x))
  - SymPy way

We decided to use the second option. Why?

- all logic is in one class, easy to extend and understand
- the less classes, the better

# the Schwarzschild solution in the General Relativity

spherically symmetric metric ( $diag(-e^{\nu(r)}, e^{\lambda(r)}, r^2, r^2 \sin^2 \theta)$ )  $\rightarrow$   
Christoffel symbols  $\rightarrow$  Riemann tensor  $\rightarrow$  Einstein equations  $\rightarrow$   
solver

```
ondra@pc232:~/sympy/examples$ time python relativity.py
```

```
...
```

```
[SKIP]
```

```
...
```

```
-----  
metric:
```

```
-C1 - C2/r 0 0 0  
0 1/(C1 + C2/r) 0 0  
0 0 r**2 0  
0 0 0 r**2*sin(\theta)**2
```

```
real 0m1.092s
```

```
user 0m1.024s
```

```
sys 0m0.068s
```



- Gruntz algorithm
- the algorithm is so simple that everyone should know how it works :)

# Comparability classes

$$L \equiv \lim_{x \rightarrow \infty} \frac{\log |f(x)|}{\log |g(x)|}$$

We define  $<$ ,  $>$ ,  $\sim$ :

- $f > g$  when  $L = \pm\infty$ 
  - $f$  is greater than any power of  $g$
  - $f$  is more rapidly varying than  $g$
  - $f$  goes to  $\infty$  or  $0$  faster than  $g$
- $f < g$  when  $L = 0$ 
  - $f$  is lower than any power of  $g$
  - ...
- $f \sim g$  when  $L \neq 0, \pm\infty$ 
  - both  $f$  and  $g$  are bounded from above and below by suitable integral powers of the other

Examples:

$$2 < x < e^x < e^{x^2} < e^{e^x}$$

$$2 \sim 3 \sim -5$$

$$x \sim x^2 \sim x^3 \sim \frac{1}{x} \sim x^m \sim -x$$

$$e^x \sim e^{-x} \sim e^{2x} \sim e^{x+e^{-x}}$$

$$f(x) \sim \frac{1}{f(x)}$$

# The Gruntz algorithm I

$$f(x) = e^{x+2e^{-x}} - e^x + \frac{1}{x}$$

$$\lim_{x \rightarrow \infty} f(x) = ?$$

Strategy:

- mrv set: the set of most rapidly varying subexpressions
  - $\{e^x, e^{-x}, e^{x+2e^{-x}}\}$
  - the same comparability class
- take an item  $\omega$  converging to 0 at infinity
  - $\omega = e^{-x}$
  - if not present in the mrv set, use the relation  $f(x) \sim \frac{1}{f(x)}$
- rewrite the mrv set using  $\omega$ 
  - $\{\frac{1}{\omega}, \omega, \frac{1}{\omega}e^{2\omega}\}$
- substitute back in  $f(x)$  and expand in  $\omega$ :
  - $f(x) = \frac{1}{x} - \frac{1}{\omega} + \frac{1}{\omega}e^{2\omega} = 2 + \frac{1}{x} + 2\omega + O(\omega^2)$

# The Gruntz algorithm II

Crucial observation:  $\omega$  is from the mrv set, so

$$f(x) = e^{x+2e^{-x}} - e^x + \frac{1}{x} = 2 + \frac{1}{x} + 2\omega + O(\omega^2) \rightarrow 2 + \frac{1}{x}$$

- We iterate until we get just a number, the final limit
- Gruntz proved this always works and converges in his Ph.D. thesis

Generally:

$$f(x) = \underbrace{\dots}_{\infty} + \underbrace{\frac{C_{-2}(x)}{\omega^2}}_{\infty} + \underbrace{\frac{C_{-1}(x)}{\omega}}_{\infty} + C_0(x) + \underbrace{C_1(x)\omega}_0 + \underbrace{O(\omega^2)}_0$$

- we look at the lowest power of  $\omega$
- the limit is one of: 0,  $\lim_{x \rightarrow \infty} C_0(x)$ ,  $\infty$

From SymPy to SAGE:

- using "\_sage\_()" methods:

From SAGE to SymPy:

- using "\_sympy\_()" methods:

Why SymPy in SAGE? Isn't Maxima good enough?

- pure Python
- easily extensible (the main reason I started SymPy), at least we try :)
- small, people can easily use it without SAGE (which is big)
- options are always good

# The question of speed

- Being pure Python has many advantages
- speed is good enough for many purposes
- sympycore project tries to speed SymPy even more
- later, when internals of SymPy settle some more, use C++, C or maybe Cython.

Now what?

- Fix bugs (there are still too many)
- Try to make most of the common tasks easy to do:
  - Playing with defined and undefined functions (  $\text{diff}(f(x), x)$  )
  - most of the integrals, limits, differential/algebraic equations should work
- Collaborate with SAGE, implement only things, that are needed

## Linus

Talk is cheap. Show me the code.

- Have something now, not tomorrow
- Strictly following the Zen of Python ("import this" in Python)
- Every single feature in SymPy must have tests
- Main hg version always needs to pass all tests