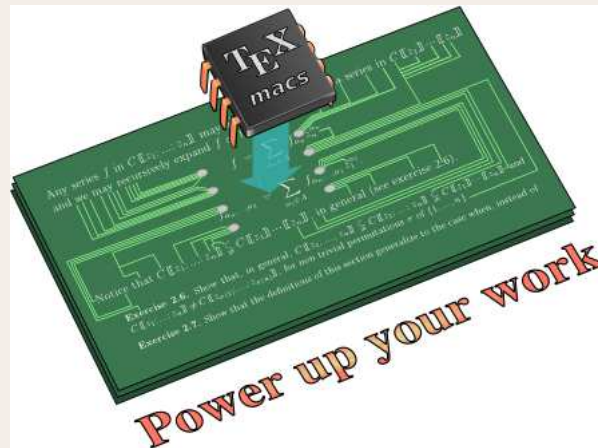


# Effective real numbers in MMLIB

by Joris van der Hoeven



Presentation with **GNU TEX**macs ([www.texmacs.org](http://www.texmacs.org))



# Effective real numbers in MMLIB



<http://www.mathemagix.org/mmxweb/web/welcome-mml.en.html>

**Mmx**  $\gg$   $x: \text{Real} = \sin \sin \text{real } 2;$

**Mmx**  $\gg$   $x$

$7.891 \cdot 10^{-1}$

Real

**Mmx**  $\gg$   $\text{approximate}(x, 1.0e-35);$

$7.8907234357288836143140304248688412 \cdot 10^{-1}$

Interval

**Mmx**  $\gg$   $M: \text{Matrix Real} = \begin{pmatrix} x & x+2 \\ 2-x^2 & \cos x \end{pmatrix};$

**Mmx**  $\gg$   $M;$

$\begin{bmatrix} 7.891 \cdot 10^{-1} & 2.789 \\ 1.377 & 7.045 \cdot 10^{-1} \end{bmatrix}$

Matrix(Real)

**Mmx**  $\gg$   $M^{20};$

$\begin{bmatrix} 2.284 \cdot 10^8 & 3.181 \cdot 10^8 \\ 1.571 \cdot 10^8 & 2.188 \cdot 10^8 \end{bmatrix}$

Matrix(Real)

**Mmx**  $\gg$   $\exp(x + \exp(-\text{real } 100)) - \exp(x);$

$8.189 \cdot 10^{-44}$

Real

**Mmx**  $\gg$   $\exp(x + \exp(-\text{real } 1000)) - \exp(x);$

0

Real

Mmx >>



# Effective analytic functions in MMLIB



**Mmx** >>  $z$ : Analytic == analytic(0, 1);

**Mmx** >> exp( $z$ );

$1.000 + 1.000 z + 5.000 10^{-1} z^2 + 1.667 10^{-1} z^3 + 4.167 10^{-2} z^4 + 8.333 10^{-3} z^5 + 1.389 10^{-3} z^6 + 1.984 10^{-4} z^7 + 2.480 10^{-5} z^8 + 2.756 10^{-6} z^9 + O(z^{10})$  Analytic

**Mmx** >> exp( $z$ )[int 20];

$4.110 10^{-19}$  Complex

**Mmx** >>  $\ell$ : Analytic == log(1 -  $z$ );

**Mmx** >> radius( $\ell$ );

$9.99937726184725761359 10^{-1}$  Floating

**Mmx** >> evaluate( $\ell$ , complex(1/2));

$-6.931 10^{-1}$  Complex

**Mmx** >> continue( $\ell$ , complex(1/2));

$-6.931 10^{-1} - 2.000 z - 2.000 z^2 - 2.667 z^3 - 4.000 z^4 - 6.400 z^5 - 1.067 10^1 z^6 - 1.829 10^1 z^7 - 3.200 10^1 z^8 - 5.689 10^1 z^9 + O(z^{10})$  Analytic

**Mmx** >> continue( $\ell$ , turn(complex(1)));

$6.283 i - 1.000 z - 5.000 10^{-1} z^2 - 3.333 10^{-1} z^3 - 2.500 10^{-1} z^4 - 2.000 10^{-1} z^5 - 1.667 10^{-1} z^6 - 1.429 10^{-1} z^7 - 1.250 10^{-1} z^8 - 1.111 10^{-1} z^9 + O(z^{10})$  Analytic





# Solving differential equations



$$f'' = (z^2 + 1) f' + e^z f; \quad f(0) = 1, f'(0) = 1 + 2i.$$

**Mmx**  $\gg$  `f: Analytic == solve_lde((z^2 + 1, exp(z)), (complex(1), complex(1, 2)));`

**Mmx**  $\gg$  `f;`

$1.000 + (1.000 + 2.000 i) z + (1.000 + 1.000 i) z^2 + (6.667 \cdot 10^{-1} + 1.000 i) z^3 + (5.417 \cdot 10^{-1} + 5.833 \cdot 10^{-1} i) z^4 + (3.500 \cdot 10^{-1} + 4.833 \cdot 10^{-1} i) z^5 + (2.278 \cdot 10^{-1} + 2.750 \cdot 10^{-1} i) z^6 + (1.343 \cdot 10^{-1} + 1.742 \cdot 10^{-1} i) z^7 + (7.882 \cdot 10^{-2} + 9.767 \cdot 10^{-2} i) z^8 + (4.361 \cdot 10^{-2} + 5.572 \cdot 10^{-2} i) z^9 + O(z^{10})$  Analytic

**Mmx**  $\gg$  `u: Complex == evaluate(f, complex(1/10));`

**Mmx**  $\gg$  `u;`

$1.111 + 2.111 \cdot 10^{-1} i$  Complex

**Mmx**  $\gg$  `approximate(u, 1.0e-81);`

$1.11072457537794457102292725574830566357052541308848196626687047567517902828392834 + 2.1106346012282867466007605052618438398248510727864880851400427655460836641117663 \cdot 10^{-1} i$   
Complexify(Interval)

**Mmx**  $\gg$



# Definition of effective real numbers



- $\tilde{x} \in \mathbb{D} = \mathbb{Z} 2^{\mathbb{Z}}$  is an  $\varepsilon$ -approximation of  $x \in \mathbb{R}$  if  $|\tilde{x} - x| < \varepsilon$ .
- *Approximation algorithm* for  $x$ : computes  $\varepsilon \mapsto \varepsilon$ -approximation of  $x$ .
- *Effective real number*:  $x \in \mathbb{R}$  which admits an approximation algorithm.
- *Complexity* of  $x$ : time needed to compute a  $2^{-l}$ -approximation.
- No zero-test for effective real numbers.
- References: Bishop and Bridges, Blanck, Müller, vdH, etc.



# Implementation by layers



## **Fast arithmetic.**

Fast computations on mantissas (Karatsuba, FFT, Brent, Chudnovsky<sup>2</sup>, VdH, etc.).

Implementations: GMP, CLN, ...

## **Low-level validated arithmetic.**

Correct rounding or well-specified bounds on the error.

Implementations: MPFR, ...

## **High-level validated arithmetic.**

Template types for intervals, balls, Lipschitz balls, etc.

Implementations: MPFI, IRRAM, MMXLIB, ...

## **High-level interface.**

Given a required precision for the result, automatically find precisions for all intermediate computations.

Implementations: IRRAM, MMXLIB, ...





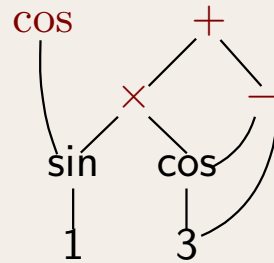
# Modeling computations with effective real numbers



- Example: addition

```
class add_real_rep: public real_rep {
    real x, y;
    add_real_rep (const real& x2, const real& y2):
        x (x2), y (y2) {}
    dyadic approximate (const dyadic& eps) {
        return x->approximate (eps/2) + y->approximate (eps/2); }
};
```

- Model sets of effective real numbers by acyclic graphs:



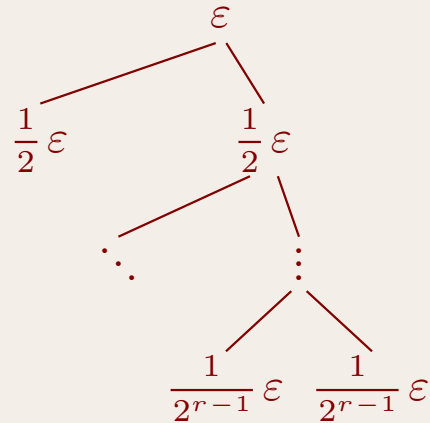
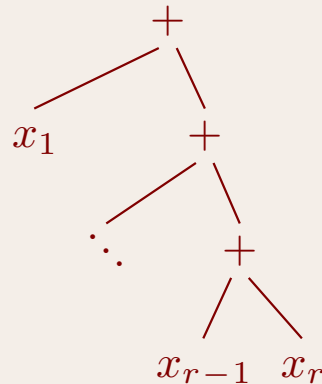
- Computations stored in memory → don't use classical numerical algorithms.



# A priori error estimates



- Distribute tolerance  $\varepsilon$  a priori over nodes of  $n$ -ary operations ( $n > 1$ ).
- Can be bad in case of badly nested expressions:



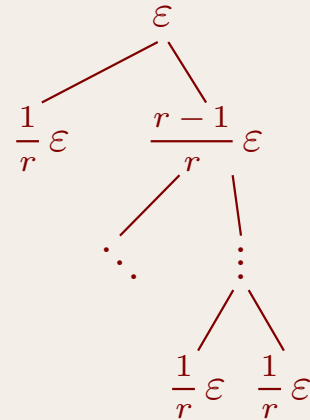
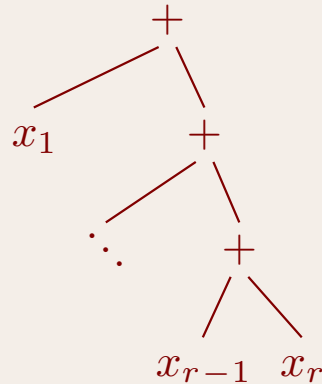
- Possible loss of  $\log w$  bits of precision; still unacceptable.



# A priori error estimates



- Distribute tolerance  $\varepsilon$  a priori over nodes of  $n$ -ary operations ( $n > 1$ ).
- Balanced error estimates: redistribute as a function of *weight*:



- Possible loss of  $\log w$  bits of precision; still unacceptable.



## *A posteriori* error estimates



- Perform whole computation using interval arithmetic.

While result not precise enough:

Double precision and redo entire computation.

- First improvement:

For each instance of `real_rep`, keep best current approximation in memory.

- Second improvement:

Don't double precision, but estimated computation time.



# Model for complexity analysis



## Global approximation problem

**Input:** an acyclic graph  $G$  with for each node  $\alpha \in G$ :

- A real function  $f_\alpha$  from the library.  
Induces by induction a real number  $x_\alpha = f_\alpha(x_{\alpha[1]}, \dots, x_{\alpha[|\alpha|]})$ .
- A tolerance  $\varepsilon_\alpha \in \mathbb{D}^>$ .

**Output:** for each node an interval  $\mathbf{x}_\alpha \ni x_\alpha$  with

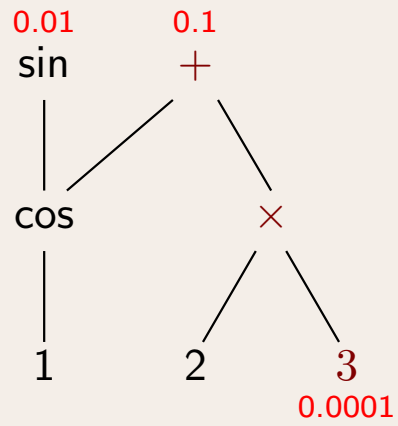
- $r_{\mathbf{x}_\alpha} < \varepsilon_\alpha$ .
- $\mathbf{x}_\alpha \supseteq f_\alpha(\mathbf{x}_{\alpha[1]}, \dots, \mathbf{x}_{\alpha[|\alpha|]})$ .

## Drawbacks

- Does not model incremental computations.
- No dependency of computations on intermediate results.

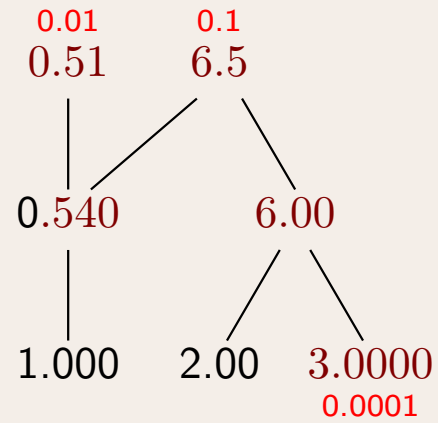


# Example





# Example





## Doubling computation time approach

- Final complexity vs. total complexity  $t^{\text{fin}} \leq t \leq (\log_2 t^{\text{fin}}) t^{\text{fin}}$ .
- Final complexity vs. optimal complexity  $t^{\text{opt}} \leq t^{\text{fin}} \leq 2 s t^{\text{opt}}$ .

## Faster approaches in “the rigid case”

- Rigid dag: for each  $x_\alpha = f_\alpha(x_{\alpha[1]}, \dots, x_{\alpha[|\alpha|]})$ , the  $\partial x_\alpha / \partial x_{\alpha[i]}$  are known with a fixed relative precision (say  $1/2$ ).
- A sufficient precision can be determined quite sharply beforehand (IRRAM).
- Using backward error bounds and balancing, the precisions can be adapted locally in a quasi-optimal way.