

Verified Compilation and Optimization of Floating-Point Programs in CakeML

Heiko Becker, Robert Rabe, Eva Darulova, Magnus O. Myreen,
Zachary Tatlock, Ramana Kumar, Yong Kiam Tan, Anthony Fox



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Floating-Point Arithmetic in Unverified & Verified Compilers

Floating-Point Arithmetic in Unverified & Verified Compilers

COMPCERT

- IEEE-754 arithmetic
- no performance optimizations
- full correctness proof

Floating-Point Arithmetic in Unverified & Verified Compilers



- IEEE-754 arithmetic
- fast-math optimizations
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Floating-Point Arithmetic in Unverified & Verified Compilers



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- no floating-point support

COMPCERT

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Floating-Point Arithmetic in Unverified & Verified Compilers



LLVM

- IEEE-754 arithmetic
- fast-math optimizations
- no correctness guarantees

in this talk



- IEEE-754 arithmetic
- fast-math-style optimizations
- correctness & accuracy proofs

COMPCERT

- IEEE-754 arithmetic
- no performance optimizations
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Fast-Math-Style Optimizations in Compilers



LLVM

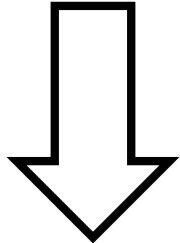
$$x * (x * (x * x)) \rightarrow (x * x) * (x * x)$$

Fast-Math-Style Optimizations in Compilers



LLVM

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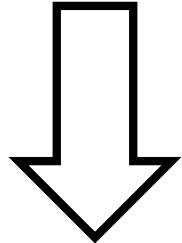
changes bit-level result

Fast-Math-Style Optimizations in Compilers



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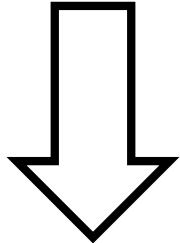
preserve IEEE-754 floating-point arithmetic

Fast-Math-Style Optimizations in Compilers



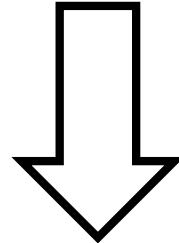
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changes bit-level result

preserve IEEE-754 floating-point arithmetic



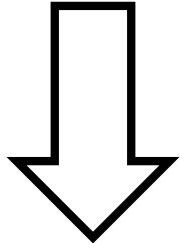
requires bit-level accuracy

Fast-Math-Style Optimizations in Compilers



LLVM

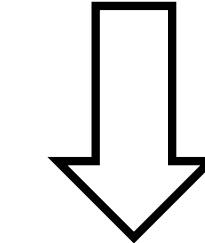
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changes bit-level result



preserve IEEE-754 floating-point arithmetic



requires bit-level accuracy

Fast-Math-Style Optimizations in Compilers

$x * (x * (x *$



COMPCERT

KEML

Implementation of ML

Icing: Supporting Fast-math Style Optimizations in a Verified Compiler

Heiko Becker¹, Eva Darulova¹, Magnus O. Myreen², and Zachary Tatlock^{3*}

¹ MPI-SWS, Saarland Informatics Campus (SIC), {hbecker,eva}@mpi-sws.org

² Chalmers University of Technology, myreen@chalmers.se

³ University of Washington, ztatlock@cs.washington.edu

change

Abstract. Verified compilers like CompCert and CakeML offer increasingly sophisticated optimizations. However, their deterministic source

CAV'19

fast arithmetic



accuracy

Icing: Supporting Fast-math Style Optimizations in a Verified Compiler [CAV'19]

verified floating-point
optimizations

Icing: Supporting Fast-math Style Optimizations in a Verified Compiler [CAV'19]

proof-of-concept optimizer

verified floating-point
optimizations

Icing: Supporting Fast-math Style Optimizations in a Verified Compiler [CAV'19]

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fine-grained control

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Icing: Supporting Fast-math Style Optimizations in a Verified Compiler [CAV'19]

non-deterministic semantics

proof-of-concept optimizer

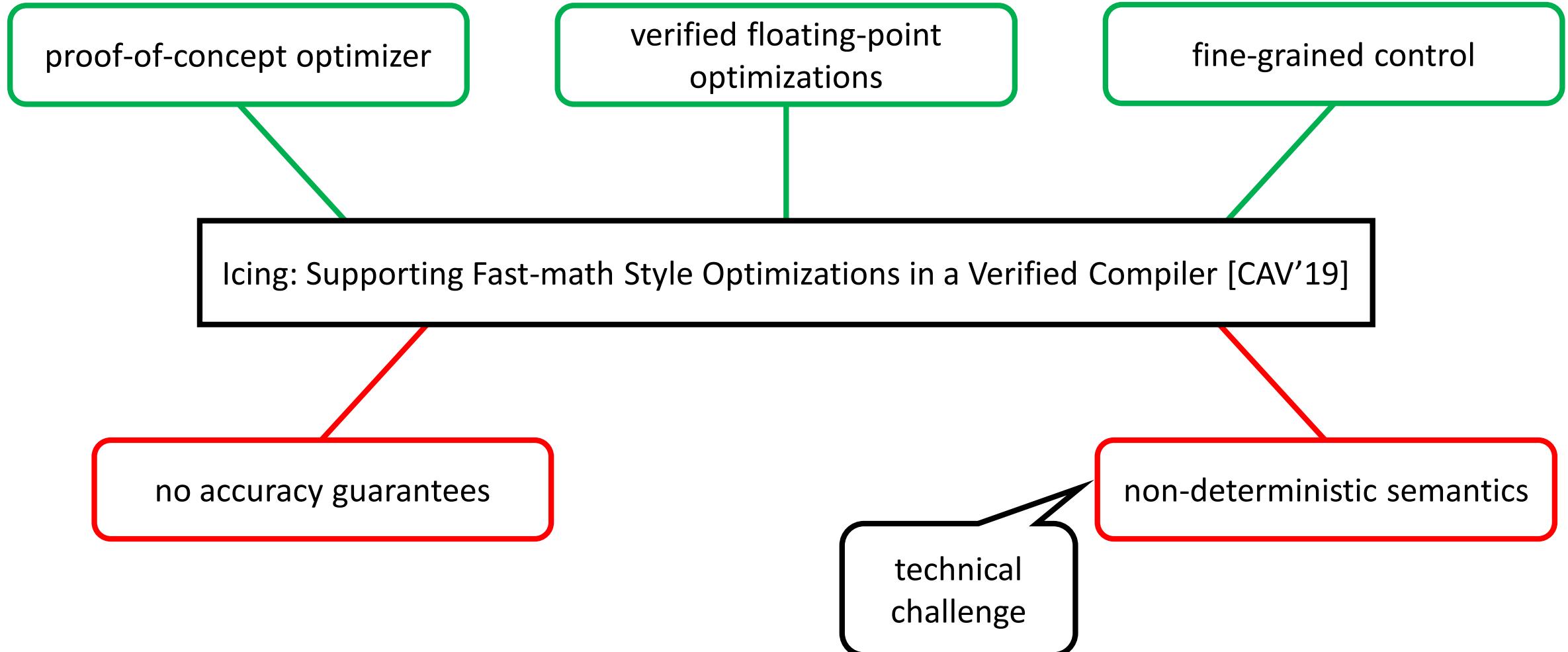
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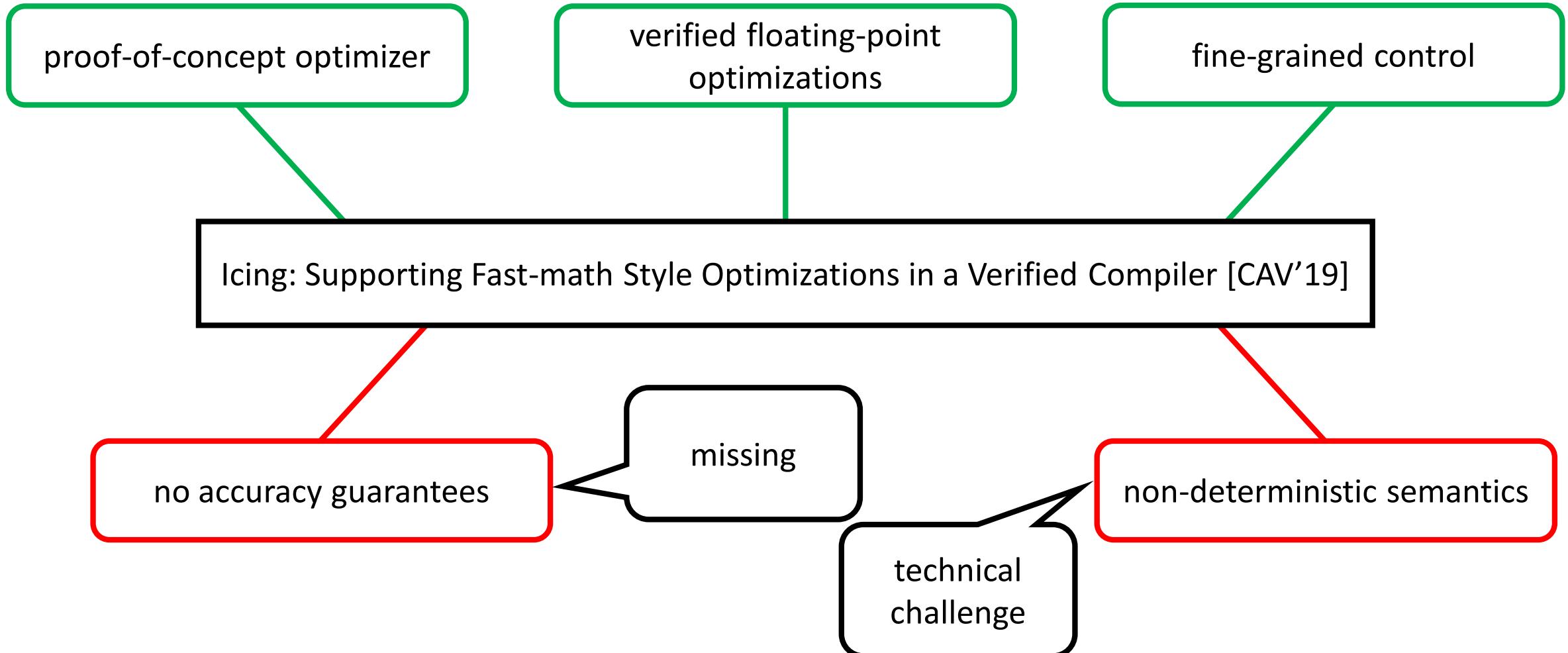
fine-grained control

Icing: Supporting Fast-math Style Optimizations in a Verified Compiler [CAV'19]

technical
challenge

non-deterministic semantics



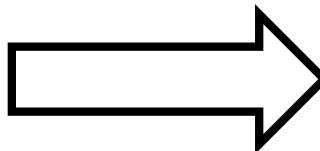


Why Accuracy Matters

```
(* require(1.0 ≤ x ≤ 100.0 ∧  
    1.0 ≤ y ≤ 100.0)  
    *)  
fun cartToPol_x (x:double, y:double):double =  
    sqrt((x * x) + (y * y))
```

Why Accuracy Matters

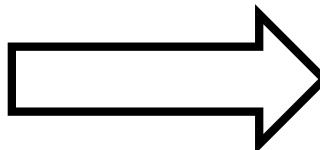
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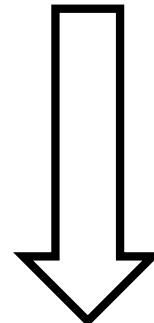
A Verified Implementation of ML

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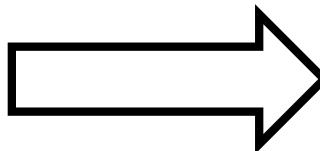
A Verified Implementation of ML



```
machine code
```

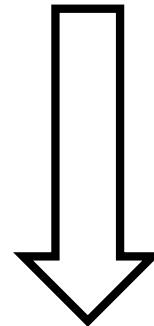
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A Verified Implementation of ML

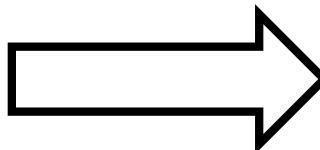
IEEE-754



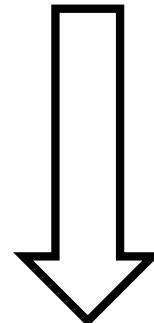
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A Verified Implementation of ML

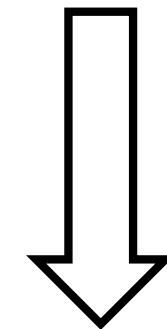
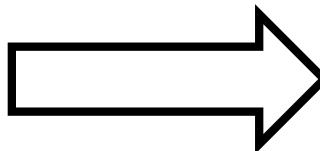


machine code

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program designed for unavoidable
input and output errors

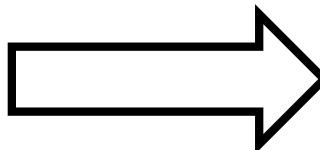


machine code

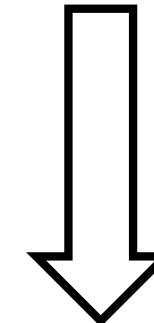
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roundoff error \leq
output error

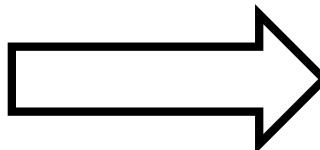


machine code

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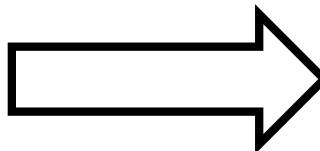
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program designed for unavoidable
input and output errors



roundoff error \leq
output error

machine code

machine code

machine code

error refinement
any optimized implementation
below output error is fine

Contributions

RealCake:

- extends CakeML with **relaxed non-deterministic floating-point semantics**
- optimizes with a **fast-math optimizer**
- **soundly proves roundoff errors** of floating-point kernels with automated tools
- proves **error refinement**

Contributions

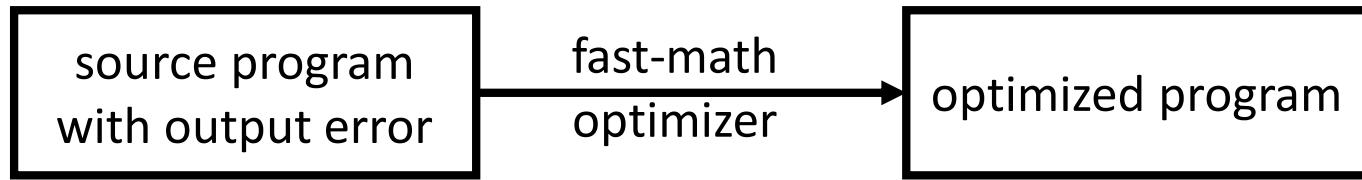
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- extends CakeML with **relaxed non-deterministic floating-point semantics**
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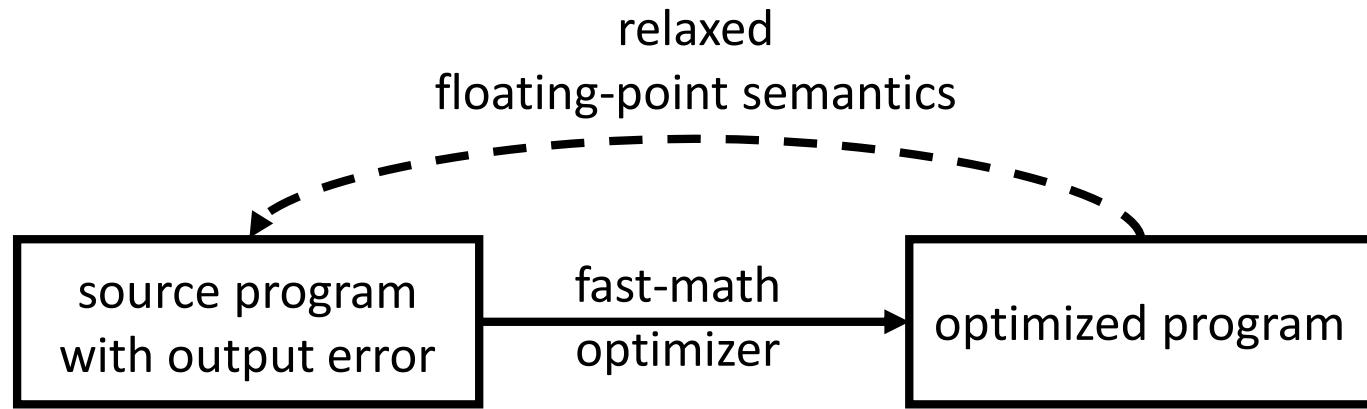
The RealCake Compiler Zoomed In

source program
with output error

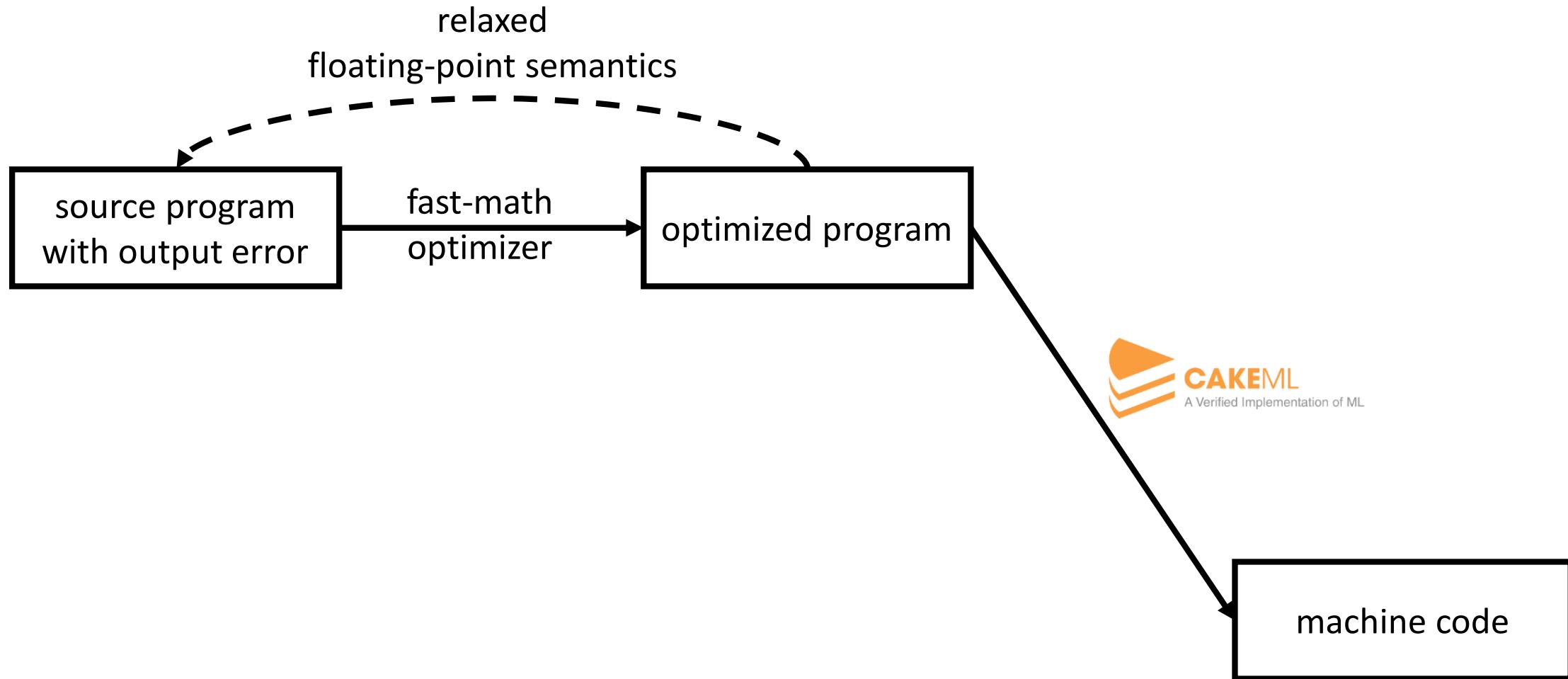
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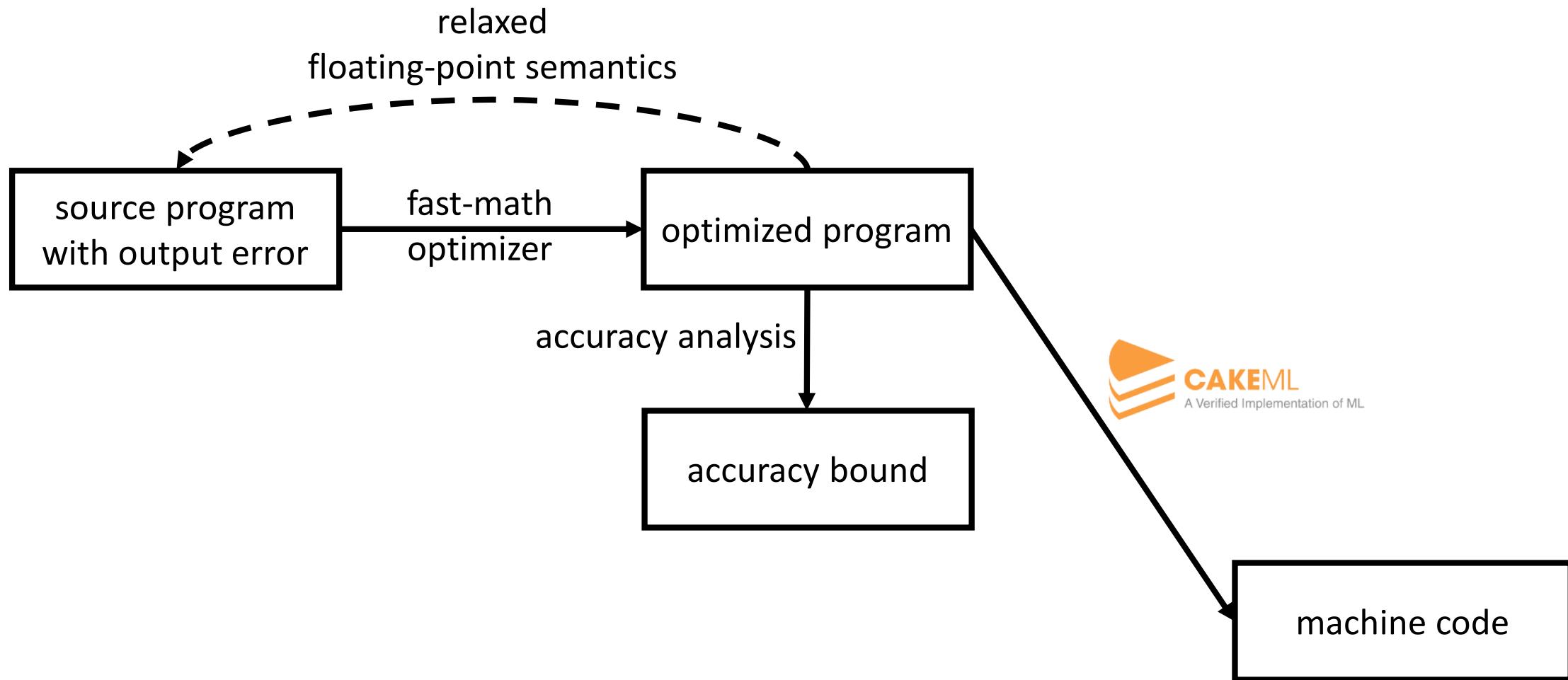
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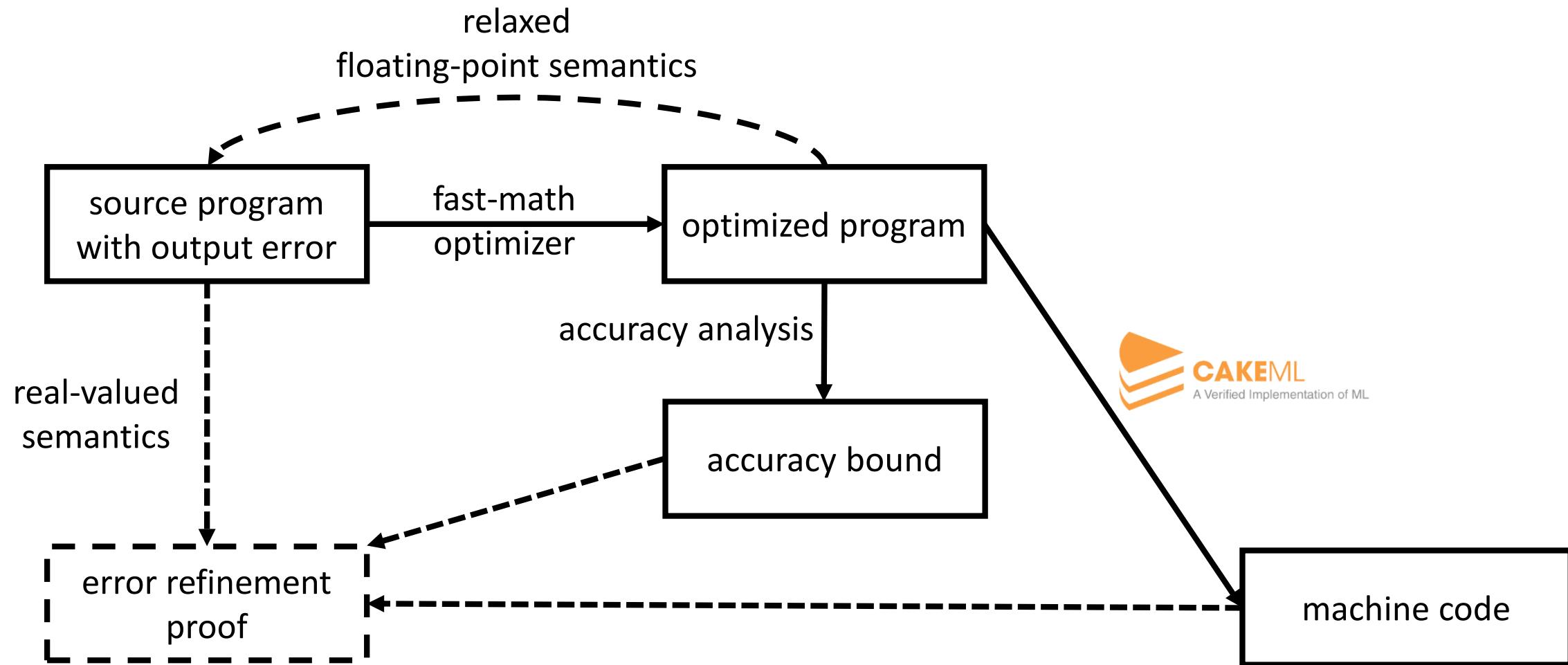
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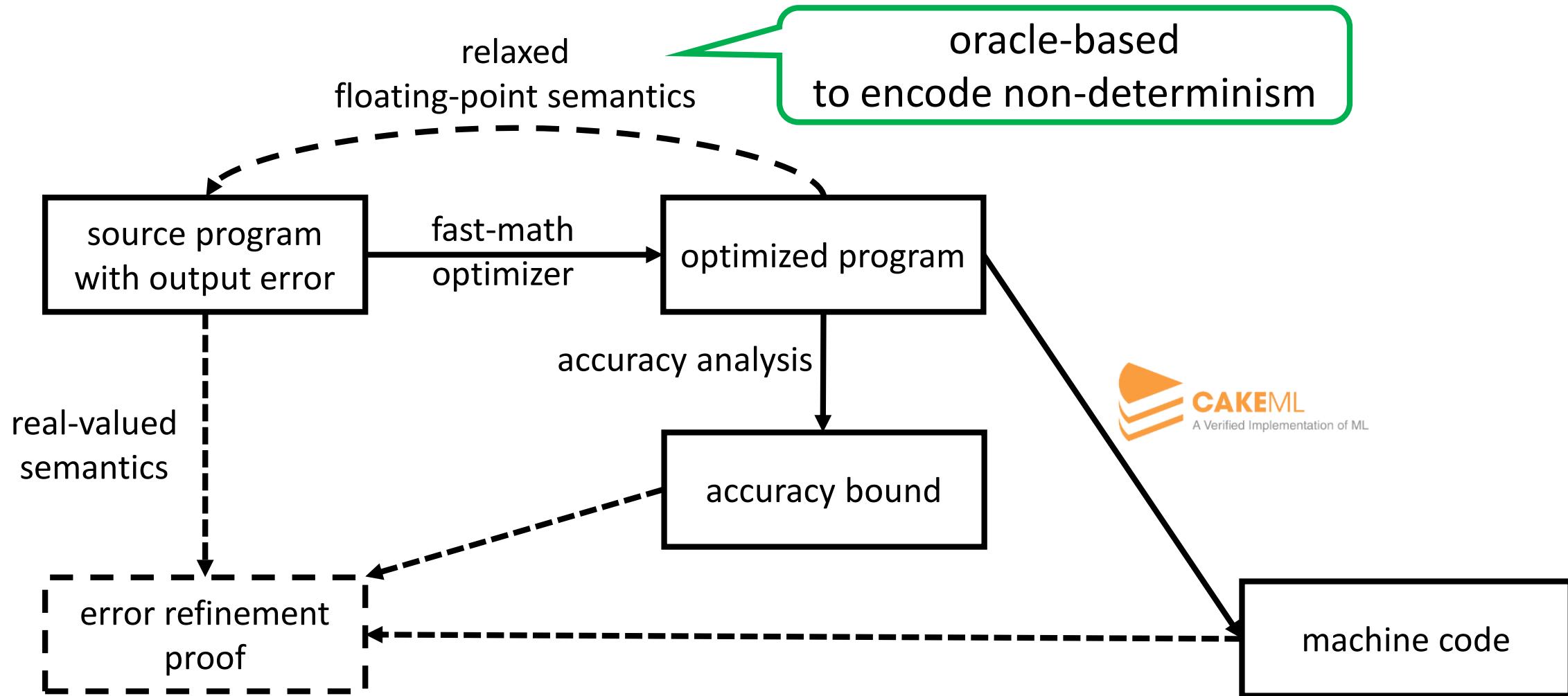
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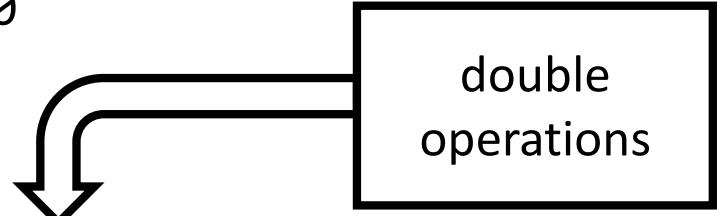


Floating-Point Programs in CakeML

```
fun jetEngine(x1:double, x2:double):double =
  opt: (let val t = (((3.0 * x1) * x1) + (2.0 * x2)) - x1
        val t2 = (((3.0 * x1) * x1) - (2.0 * x2)) - x1
        val d = (x1 * x1) + 1.0
        val s = t / d
        val s2 = t2 / d
      in
        x1 + (((((((2.0 * x1) * s) * (s - 3.0)) +
          ((x1 * x1) * ((4.0 * s) - 6.0))) * d) +
          (((3.0 * x1) * x1) * s)) +
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    end)
```

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      ((x1 * x1) * ((4.0 * s) - 6.0))) * d) +
      (((3.0 * x1) * x1) * s)) +
      ((x1 * x1) * x1)) + x1) + (3.0 * s2))
  end)
```



A diagram illustrating the flow of data in the floating-point program. A black rectangular box labeled "double operations" contains the mathematical expressions. An arrow points from the output of this box back to the input of the first multiplication term in the innermost expression ($(2.0 * x1) * s$), indicating that the result of the previous operations is used as input for the current step.

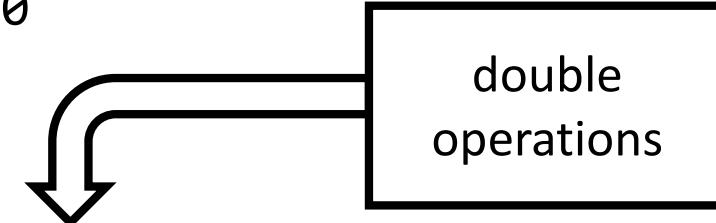
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              ((x1 * x1) * x1)) + x1) + (3.0 * s2))  
    end)
```

optimization
annotation



double
operations



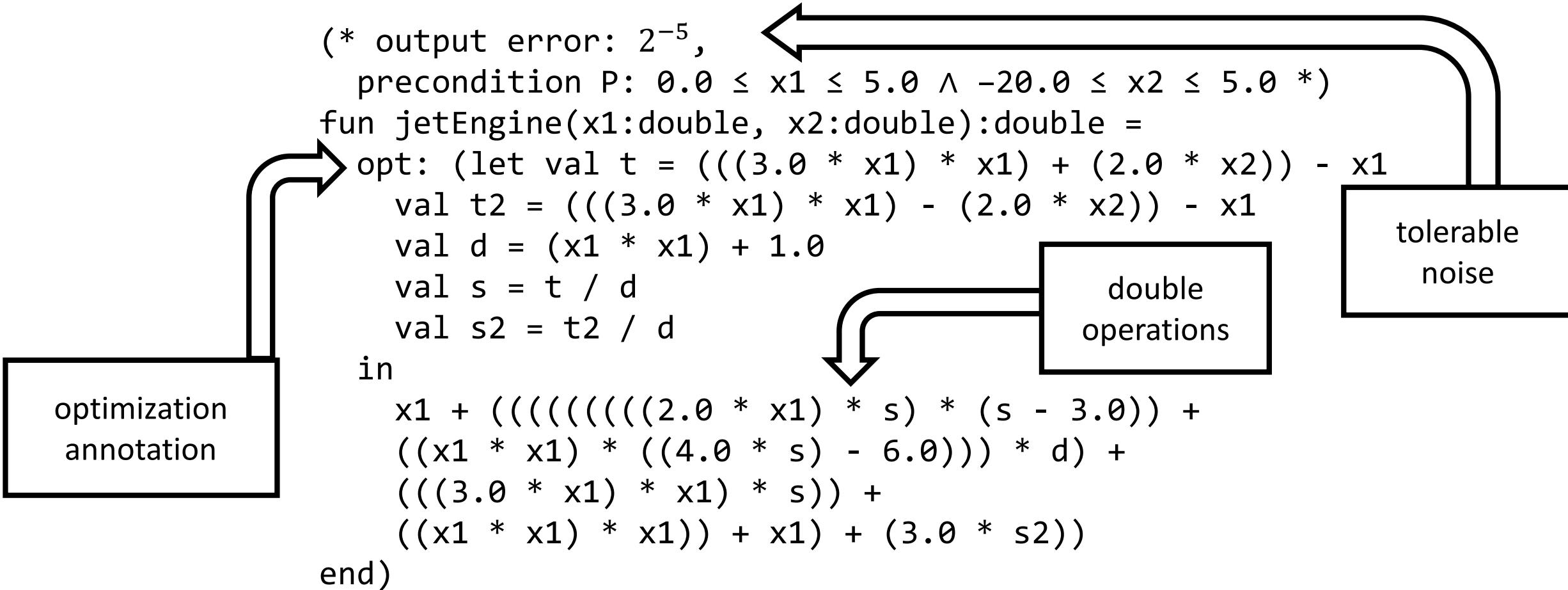
Floating-Point Programs in CakeML

```
(* output error:  $2^{-5}$ ,  
 precondition P:  $0.0 \leq x_1 \leq 5.0 \wedge -20.0 \leq x_2 \leq 5.0$  *)  
fun jetEngine(x1:double, x2:double):double =  
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    end)
```

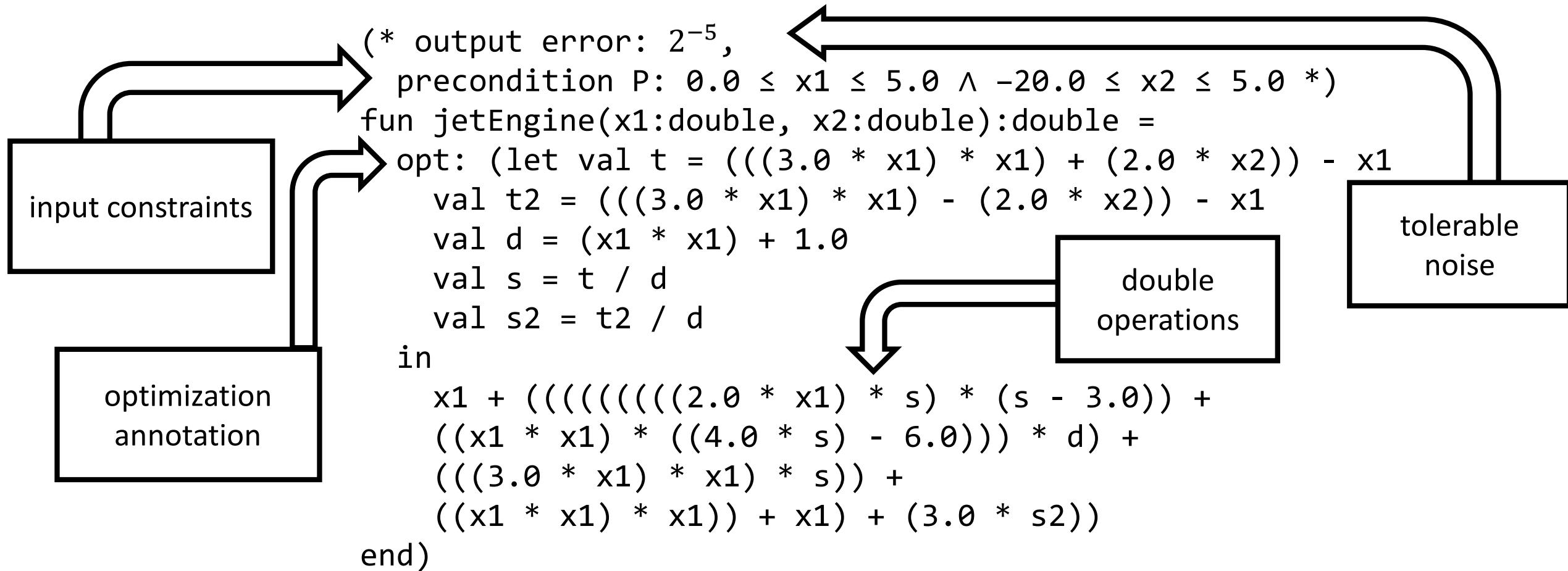
optimization
annotation

double
operations

Floating-Point Programs in CakeML



Floating-Point Programs in CakeML



Optimized Floating-Point Programs

```
(* guaranteed error bound: 2-5,  
precondition P: 0.0 ≤ x1 ≤ 5.0 ∧ -20.0 ≤ x2 ≤ 5.0 *)  
fun jetEngine(x1:double, x2:double):double =  
  noopt: (let  
    val t = fma((x1+x1)+x1, x1, (x2 + x2) - x1)  
    val t2 = fma((x1+x1)+x1, x1, fma(-2.0, x2, -x1))  
    val d = fma(x1, x1, 1.0)  
    val s = t / d  
    val s2 = t2 / d  
  in  
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      x1 * fma(4.0, s, -6.0)),  
      fma(x1 * x1, ((s + s) + s) + x1,  
        x1 + ((s2 + s2) + s2)))  
  end)
```

Optimized Floating-Point Programs

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  end)
```

verified by
accuracy
analysis

Optimized Floating-Point Programs

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  end)
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verified by
accuracy
analysis

faster, locally more accurate

Optimized Floating-Point Programs

Disallow further optimization

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  end)
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verified by
accuracy
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faster, locally more accurate

Optimized Floating-Point Programs

machine code from
IEEE-754 preserving
compilation in CakeML

Disallow further
optimization

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faster, locally more accurate

The Final Specification Theorem

$$\text{jetEngineInputsInPrecond } (s_1, s_2) (w_1, w_2) \wedge \text{environmentOk } ([\text{jetEngine}; s_1; s_2], fs) \Rightarrow \exists w r. \text{CakeMLevaluatesAndPrints } (\text{jetEngineCode}, s_1, s_2, fs) (\text{toString } w) \wedge \text{initialFPcodeReturns } \text{jetEngineUnopt } (w_1, w_2) w \wedge \text{realSemanticsReturns } \text{jetEngineUnopt } (w_1, w_2) r \wedge \text{abs } (\text{fpToReal } w - r) \leq 2^{-5}$$

The Final Specification Theorem

inputs in
specified
constraints

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output error sound

The Final Specification Theorem

inputs in specified constraints

program returns double word w

the program is run with the correct inputs

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CakeMLEvaluatesAndP
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RealCake is the first verified compiler that proves end-to-end accuracy bounds for **compiled** fast-math optimized programs

$tEngine; s_1; s_2], fs) \Rightarrow$

$w) \wedge$

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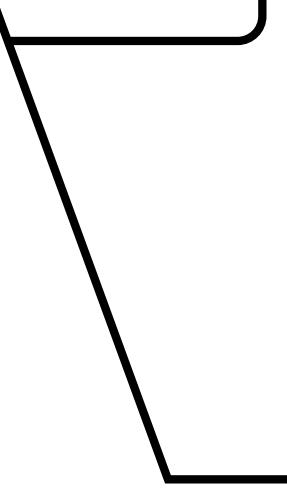
output error sound

RealCake's Four-Phase-Optimizer

canonical form

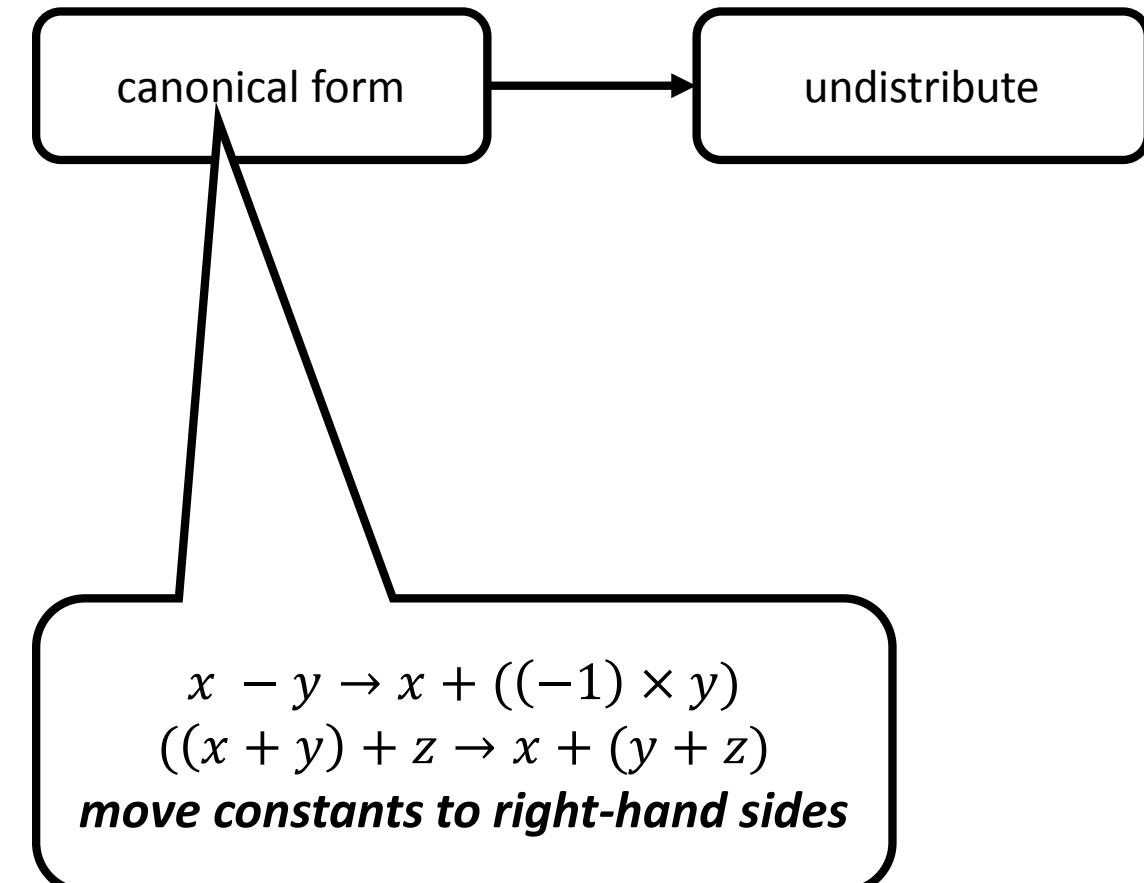
RealCake's Four-Phase-Optimizer

canonical form

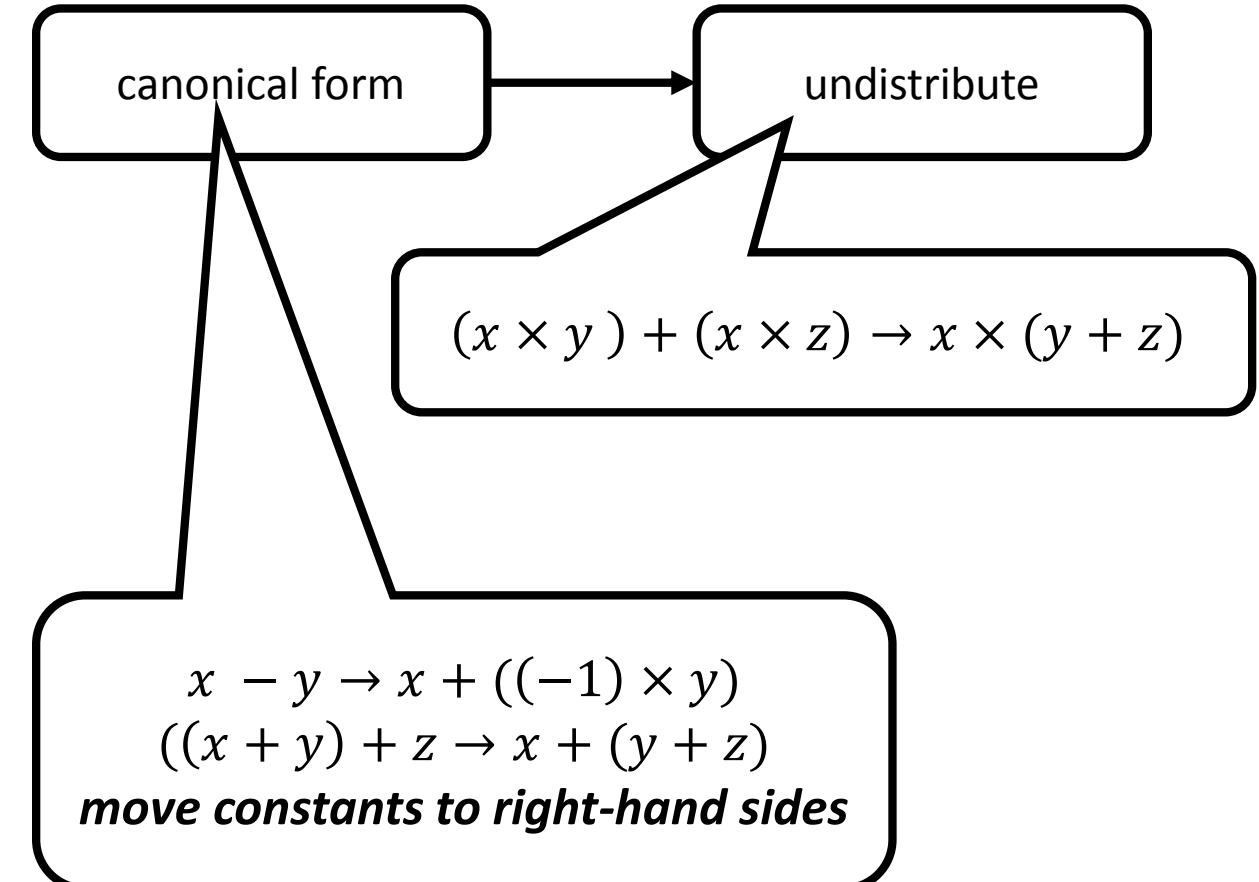

$$x - y \rightarrow x + ((-1) \times y)$$
$$((x + y) + z \rightarrow x + (y + z))$$

move constants to right-hand sides

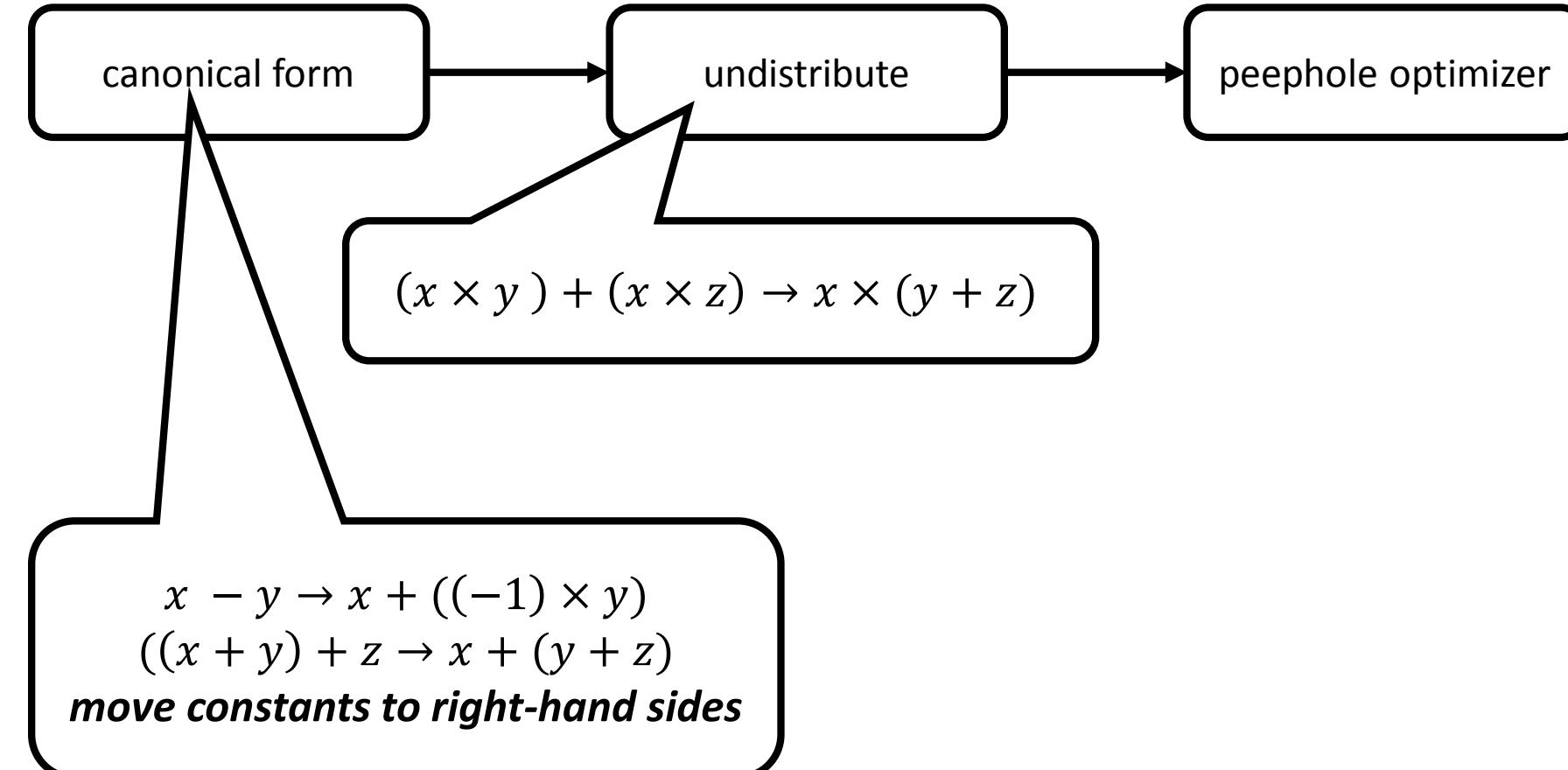
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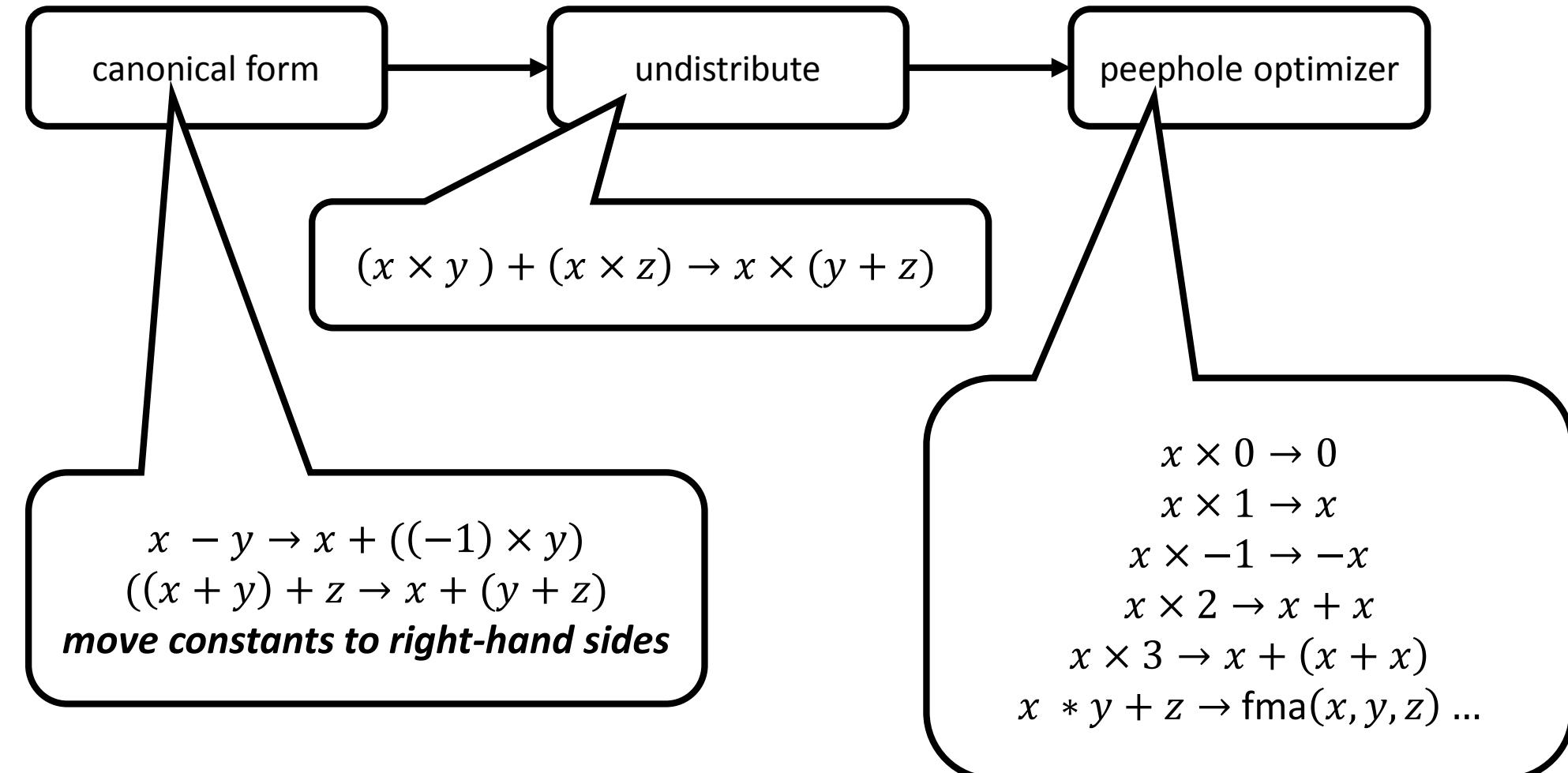
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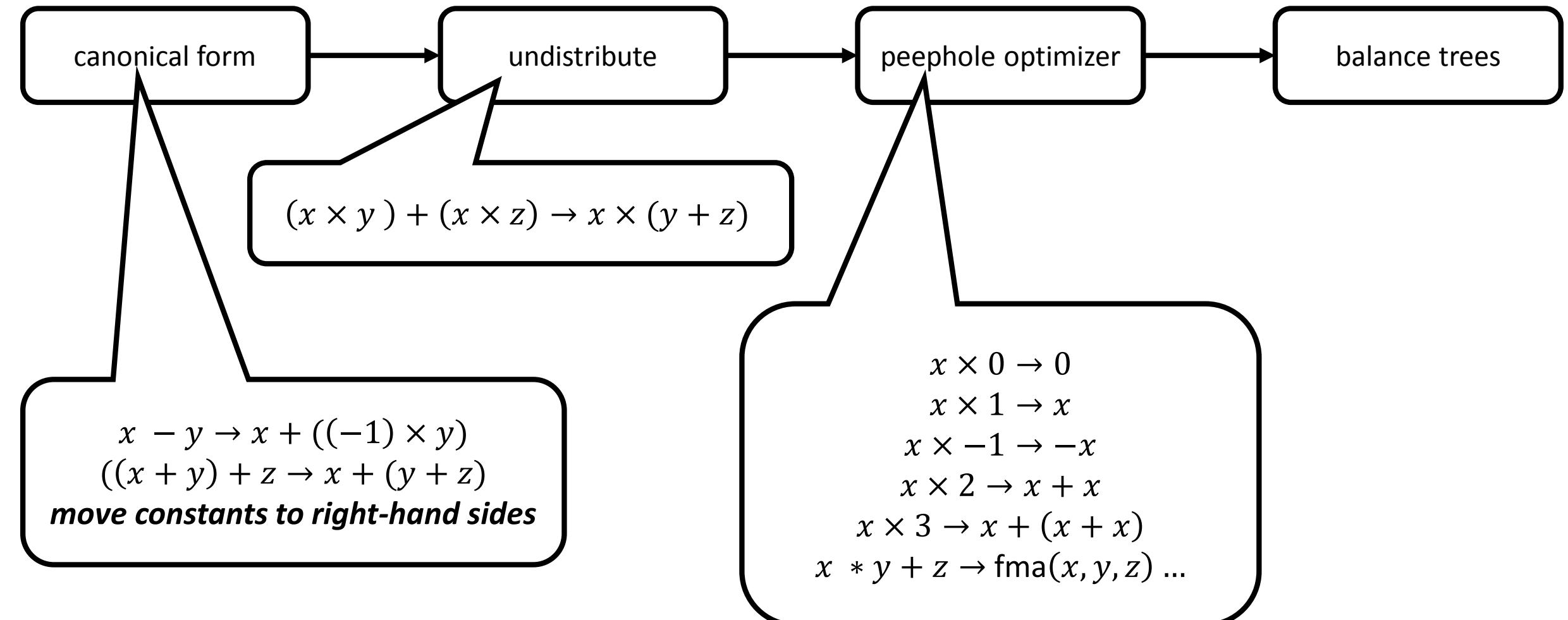
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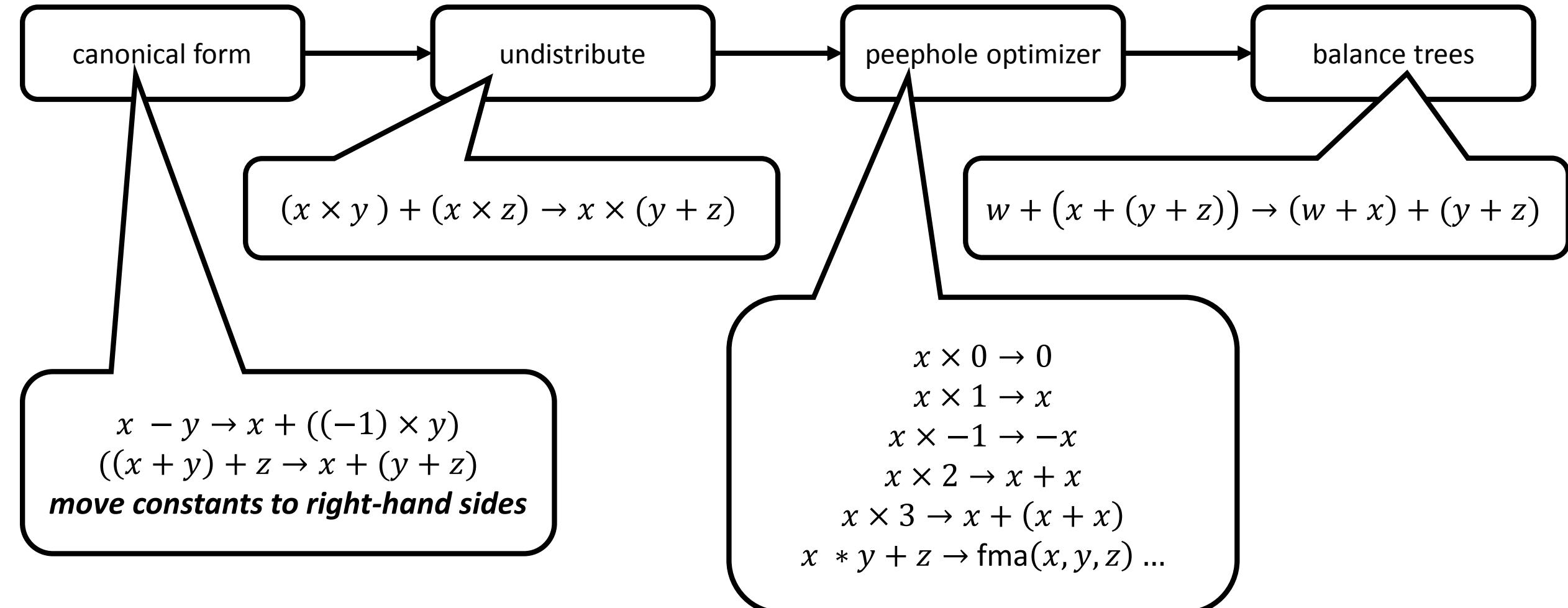
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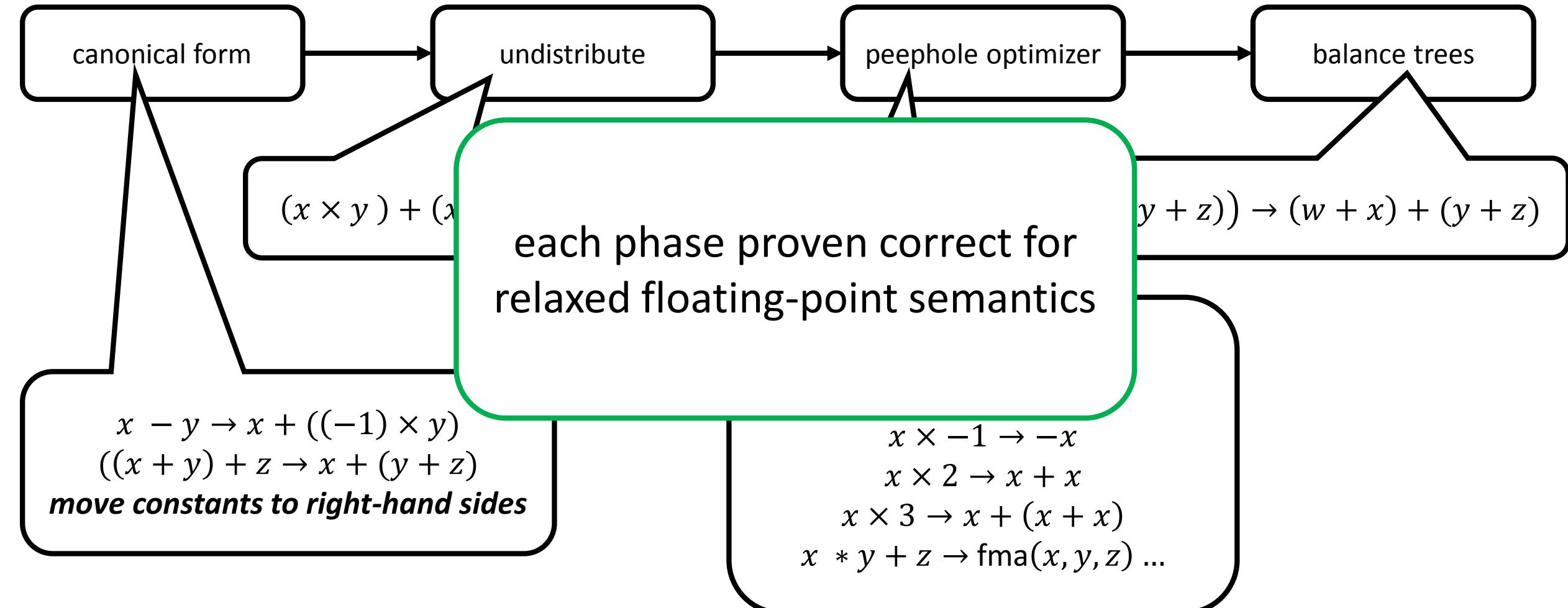
RealCake's Four-Phase-Optimizer



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RealCake's Four-Phase-Optimizer



Experimental Results – Roundoff Errors

Name

cartToPol

delta

doppler2

pid

sine_newton

sqroot

turbine1

Experimental Results – Roundoff Errors

Name	Original
cartToPol	2.815e-09
delta	1.970e-13
doppler2	6.534e-13
pid	7.621e-15
sine_newton	7.495e-15
sqroot	1.115e-15
turbine1	1.588e-13

Experimental Results – Roundoff Errors

Name	Original	fast-math
cartToPol	2.815e-09	2.463e-09
delta	1.970e-13	2.940e-12
doppler2	6.534e-13	1.639e-12
pid	7.621e-15	7.727e-15
sine_newton	7.495e-15	6.27e-15
sqroot	1.115e-15	1.059e-15
turbine1	1.588e-13	1.541e-13

Experimental Results – Roundoff Errors

Name	Original	fast-math	Improvement
cartToPol	2.815e-09	2.463e-09	13%
delta	1.970e-13	2.940e-12	-198%
doppler2	6.534e-13	1.639e-12	50%
pid	7.621e-15	7.727e-15	-1%
sine_newton	7.495e-15	6.27e-15	16%
sqroot	1.115e-15	1.059e-15	5%
turbine1	1.588e-13	1.541e-13	3%

Experimental Results – Roundoff Errors

Name	Original	fast-math	Improvement
cartToPol	2.815e-09	2.463e-09	13%
delta	1.970e-13	2.940e-12	-198%
doppler2	6.534e-13	1.639e-12	50%
pid	7.621e-15	7.727e-15	-1%
sine_newton	7.495e-15	6.27e-15	16%
sqroot	1.115e-15	1.059e-15	5%
turbine1	1.588e-13	1.541e-13	3%

trade accuracy for performance

Experimental Results – Performance

Name

cartToPol

delta

doppler2

pid

sine_newton

sqroot

turbine1

Experimental Results – Performance

Name	Original
cartToPol	2.05
delta	13.49
doppler2	36.00
pid	104.11
sine_newton	126.34
sqroot	87.06
turbine1	121.02

Experimental Results – Performance

Name	Original	fast-math
cartToPol	2.05	9%
delta	13.49	16%
doppler2	36.00	6%
pid	104.11	0%
sine_newton	126.34	0%
sqroot	87.06	5%
turbine1	121.02	0%

Experimental Results – Performance

Name	Original	Csts	fast-math
cartToPol	2.05	1%	9%
delta	13.49	1%	16%
doppler2	36.00	91%	6%
pid	104.11	96%	0%
sine_newton	126.34	92%	0%
sqroot	87.06	95%	5%
turbine1	121.02	96%	0%

Conclusion

RealCake:

- proves **error refinement** for CakeML programs
- extends CakeML with **oracle-based floating-point semantics**
- optimizes with **fast-math-style optimizations**

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in the paper:

- verified constant lifting optimization
- heuristic to avoid slow-downs
- integration into CakeML toolchain
- implementation of real-numbered and IEEE-754 semantics