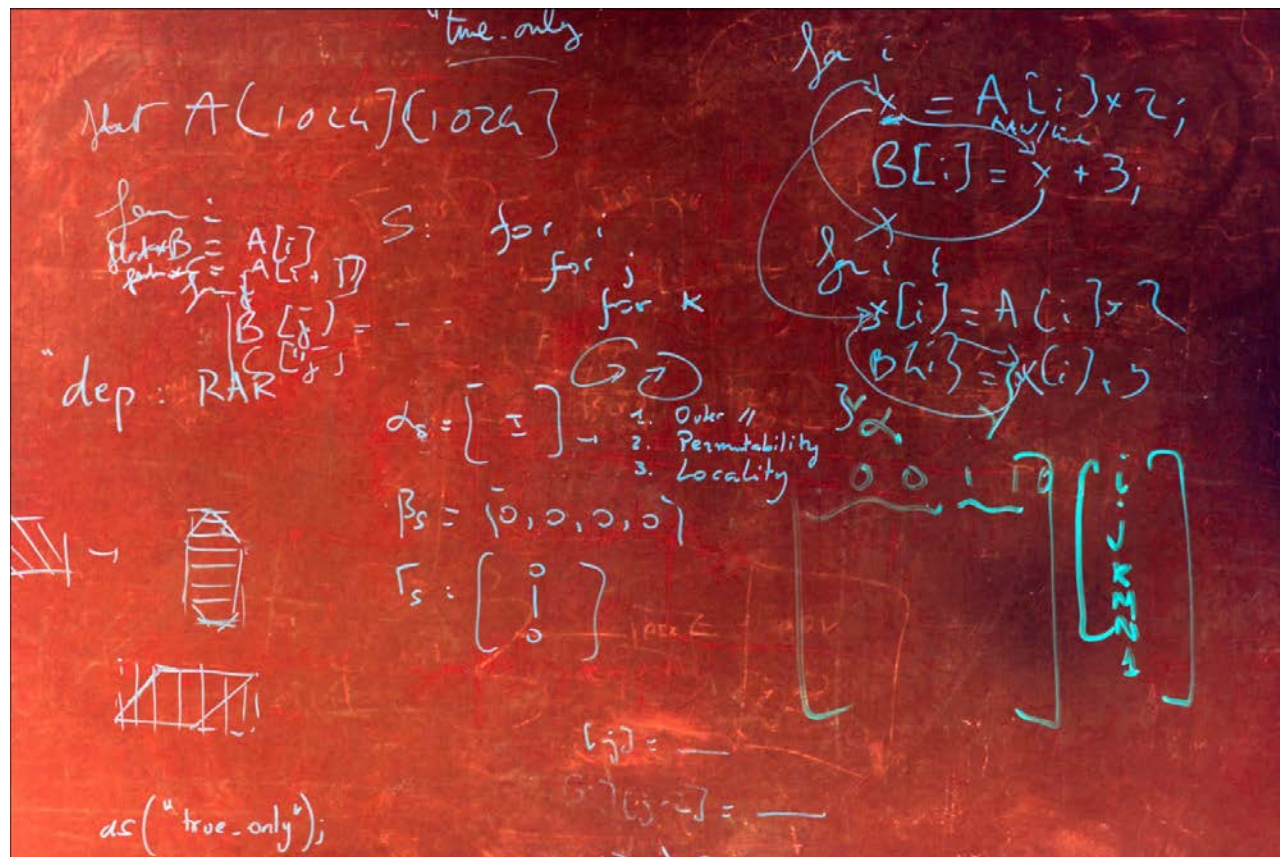


# Formal Verification of C Programs with Floating-Point Computations

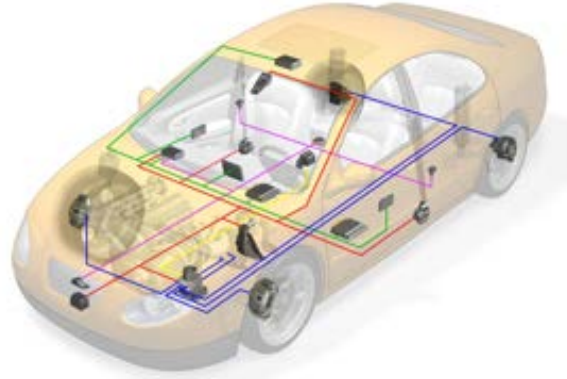
Certified Error Bounds for Signal Processing

Tahina Ramananandro

Paul Mountcastle, Benoît Meister, Richard Lethin



# Overview



- Sensor systems data processing
  - Positioning, obstacle avoidance, radar imaging, etc.
- Problems
  - Numerical optimizations for energy/time efficiency
  - Correctness and accuracy
  - Which guarantees on the actual software code?

# Overview

Can we reason about sensor systems:

- With new performance optimizations for data processing
- In a very deep way, with strong correctness guarantees down to the actual code?

Our contributions

- VCFloat: proof library and tactics to verify C programs with floating-point computations
- Example use case: a radar algorithm and its C implementation

Use cases

- Cyber-physical systems
- Lightweight UAVs (copters), cars
- Military, transportation, medicine, etc.

# The High-Level Problem

Goal: energy-efficient implementations of numerical algorithms

- Naïve implementations consume time and energy
- Main ideas for performance improvement:
  - compute in lower-precision floating-point
  - introduce approximations
- Problem: **uncertainty** introduced by errors in the result
  - How to compute some implementation **error bound**?
  - How can we **trust** this error bound?

# Our Achievements

VCFloat: a Coq library for handling floating-point computations in the verification of C programs

- Automatically compute real-number expressions with rounding error terms and their correctness proofs

Use case: SAR backprojection with linear interpolation

- Introduce approximations for square root and sine
- Tune between single- and double-precision floating-points
- Compute error bounds wrt. "ideal" mathematical real-number algorithm
- **Formal proof** of correctness using the Coq proof assistant
- Energy measurements:  $\sim 10\text{--}20\%$  saved on Intel Haswell

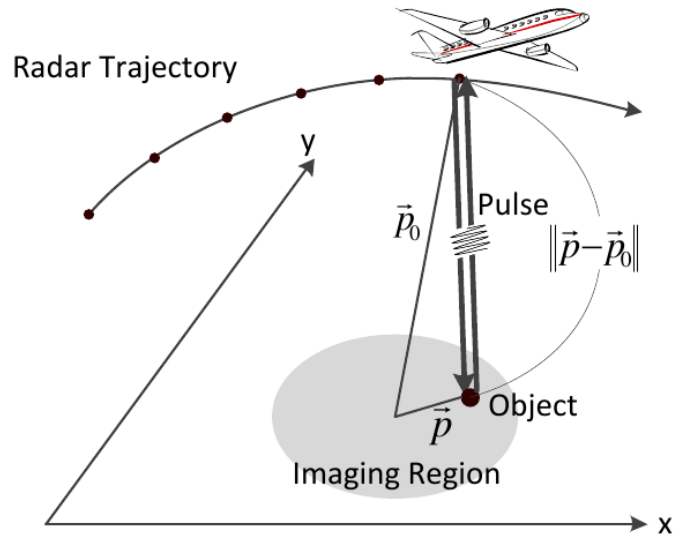
# This Presentation

- Certified error bounds for energy-efficient radar image processing
- Our Coq framework: VCFloat
- Demo
- Conclusions

Certified Error Bounds for

# **RADAR IMAGE PROCESSING**

# Synthetic Aperture Radar (SAR) Backprojection



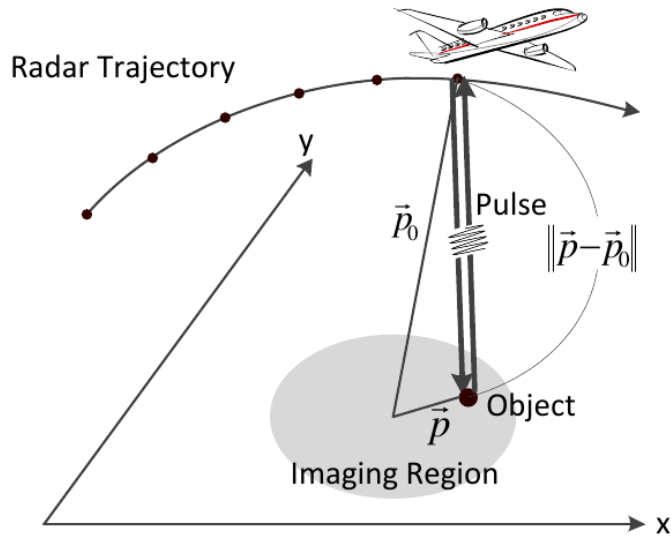
```
for all pixels  $x$  do
  for all pulses  $p$  do
     $R$  = distance between platform and pixel  $x$  for pulse  $p$ 
     $s$  = sample interpolated from pulse  $p$  in neighborhood of range  $R$ 
    Apply phase correction to sample  $s$  based on range  $R$ 
    Accumulate sample  $s$  into pixel  $x$ 
  end for
end for
```

Real-number algorithm

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012



# SAR Backprojection



```

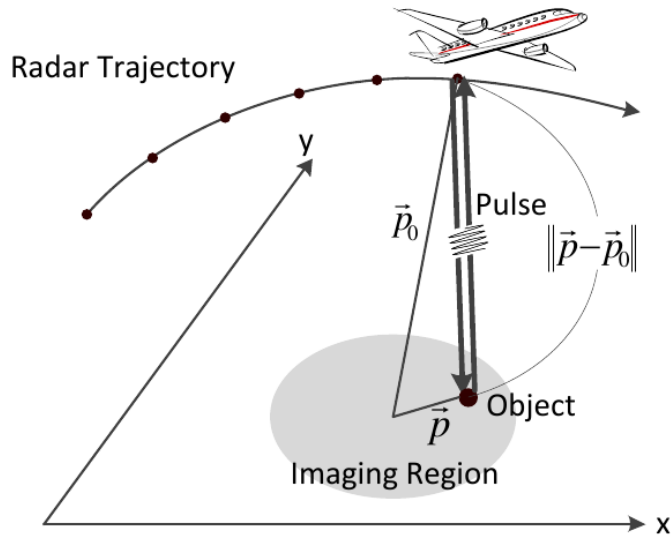
for  $y := 0$  to  $BP\_NPIX\_Y - 1$  do
   $py := (y + \frac{1-BP\_NPIX\_Y}{2}) \times dx dy$ 
  for  $x := 0$  to  $BP\_NPIX\_X - 1$  do
     $px := (x + \frac{1-BP\_NPIX\_X}{2}) \times dx dy$ 
     $\underline{image}[y][x] := 0 \in \mathbb{C}$ 
    for  $p := 0$  to  $N\_PULSES - 1$  do
       $r := \|\underline{platpos}[p] - (px, py, z[p][y][x])\|$ 
       $\underline{bin} := (r - r_0) / dr$ 
       $\underline{sample} := \underline{binSample}(N\_RANGE\_UPSAMPLED, \underline{data}[p], \underline{bin})$ 
       $\underline{matchedFilter} := \exp(2i \times ku \times r)$ 
       $\underline{image}[y][x] := \underline{image}[y][x] + \underline{sample} \times \underline{matchedFilter}$ 
    end for
  end for
end for
return  $\underline{image}$ 

```

Real-number algorithm

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# SAR Backprojection



```

for y := 0 to BP_NPIX_Y - 1 do
  py := (y +  $\frac{1-BP\_NPIX\_Y}{2}$ ) × dx dy
  for x := 0 to BP_NPIX_X - 1 do
    px := (x +  $\frac{1-BP\_NPIX\_X}{2}$ ) × dx dy
    image[y][x] := 0 ∈ ℂ
    for p := 0 to N_PULSES - 1 do
      r :=  $\| \text{platpos}[p] - (px, py, z[p][y][x]) \|$ 
      bin := (r - r0) / dr
      sample := binSample(N_RANGE_UPSAMPLED, data[p], bin)
      matchedFilter :=  $\exp(2i \times k_u \times r)$ 
      image[y][x] := image[y][x] + sample × matchedFilter
    end for
  end for
end for
return image

```

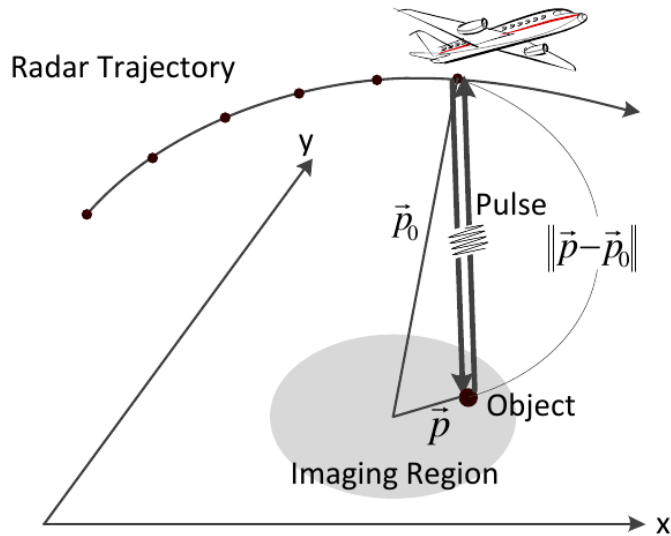
square root

sine

Real-number algorithm

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# SAR Backprojection



```

for y := 0 to BP_NPIX_Y - 1 do
  py := (y +  $\frac{1-BP\_NPIX\_Y}{2}$ ) × dx dy
  for x := 0 to BP_NPIX_X - 1 do
    px := (x +  $\frac{1-BP\_NPIX\_X}{2}$ ) × dx dy
    image[y][x] := 0 ∈ ℂ
    for p := 0 to N_PULSES - 1 do
      r := ||platpos[p] - (px, py, z[p][y][x])||
      bin := (r - r0)/dr
      sample := binSample(N_RANGE_UPSAMPLED, data[p], bin)
      matchedFilter := exp(2i × ku × r)
      image[y][x] := image[y][x] + sample × matchedFilter
    end for
  end for
end for
return image

```

square root

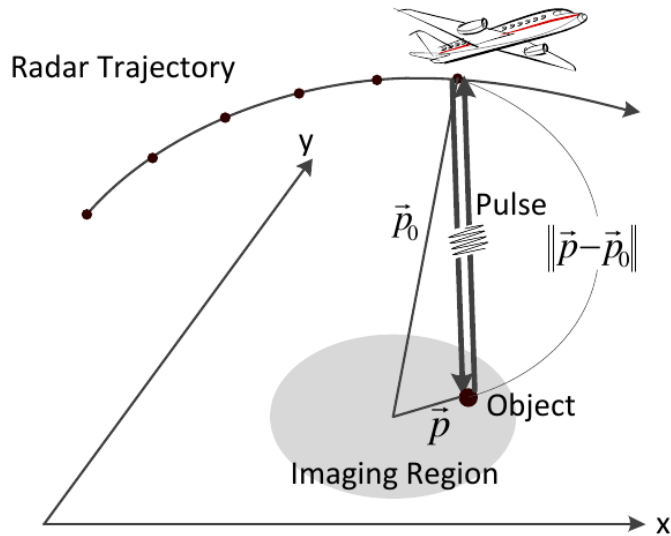
sine

floating-point computations

## Implementation

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# SAR Backprojection



```

void backprojection
(int const BP_NPIX_X, int const BP_NPIX_Y,
 int const N_PULSES, ...,
 float const **data_r, float const **data_i,
 double** image_r, double** image_i, ...) {
  for (int y = 0; y < BP_NPIX_Y, ++y) {
    ...
    for (int x = 0; x < BP_NPIX_X, ++x) {
      ...
      for (int p = 0; p < N_PULSES, ++p) {
        ...
        double ... = ... sqrt(...) ... ;
        ...
        double ... = ... sin(...) ... ;
        ...
      }
    }
  }
}

```

square root

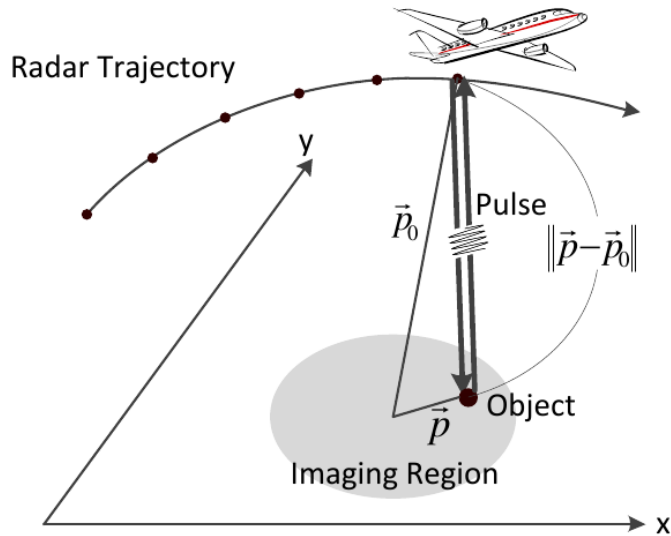
sine

floating-point computations

## C Implementation

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# SAR Backprojection



```
void backprojection
(int const BP_NPIX_X, int const BP_NPIX_Y,
 int const N_PULSES, ...,
 float const **data_r, float const **data_i,
 double** image_r, double** image_i, ...) {
  for (int y = 0; y < BP_NPIX_Y, ++y) {
```

```
    ...
    for (int x = 0; x < BP_NPIX_X, ++x) {
      ...
      for (int p = 0; p < N_PULSES, ++p) {
        ...
        double ... = ... approx_sqrt(...) ... ;
        ...
        double ... = ... approx_sin(...) ... ;
        ...
      }
    }
  }
```

approximate  
square root

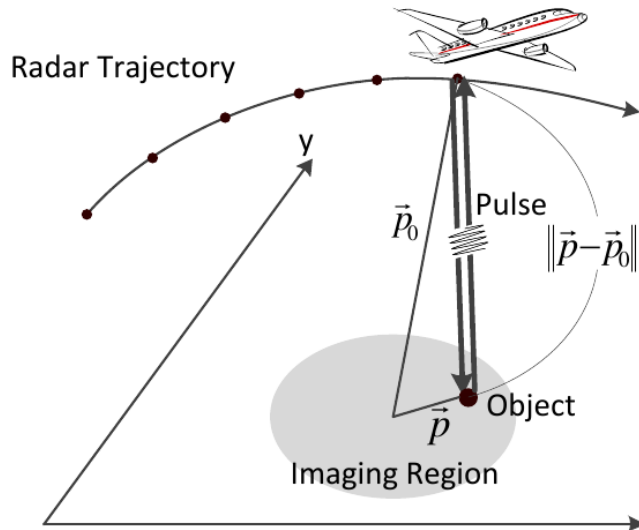
approximate  
sine

floating-point computations

## C Implementation

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# SAR Backprojection



```

void backprojection
(int const BP_NPIX_X, int const BP_NPIX_Y,
 int const N_PULSES, ...,
 float const **data_r, float const **data_i,
 float** image_r, float** image_i, ...) {
  for (int y = 0; y < BP_NPIX_Y, ++y) {
    ...
    for (int x = 0; x < BP_NPIX_X, ++x) {
      ...
      for (int p = 0; p < N_PULSES, ++p) {
        ...
        double ... = ... approx_sqrt(...) ... ;
        float ... = ... approx_sin(...) ... ;
        ...
      }
    }
  }
}
    
```

approximate square root

approximate sine

Precision tuning

floating-point computations

C Implementation  
( ~ 150 lines)

Figure from Park et al. *Efficient Backprojection-Based Synthetic Aperture Radar Computation with Many-Core Processors*, SC 2012

# Image Error Analysis

Maximize Signal-Noise Ratio:

$$SNR := \frac{\|image_0\|^2}{\|image - image_0\|^2}$$

Find an upper bound on the denominator

- Absolute error bound is enough

Error sources:

- Method errors introduced by approximation
- Rounding errors introduced by floating-point computations

# Final Correctness Statement (slightly simplified)

$\forall P \text{ ` (HYPS: SARHypotheses P) } m$

$(Hm: \text{holds } m (P \ ++$

$\text{Pperm\_int bir oir (BP\_NPIX\_X } \times \text{ BP\_NPIX\_Y) } \ ++$

$\text{Pperm\_int bii oii (BP\_NPIX\_X } \times \text{ BP\_NPIX\_Y)),$

Hypotheses

$\exists m' , \text{star Clight.step2}$

$(\text{Callstate fn\_sar\_backprojection ...}) (\text{Returnstate Vundef Kstop } m') \wedge$

$\exists \text{image\_r image\_i,}$

$\text{holds } m (P \ ++$

$\text{Parray\_int image\_r bir oir (BP\_NPIX\_X } \times \text{ BP\_NPIX\_Y) } \ ++$

$\text{Parray\_int image\_i bii oii (BP\_NPIX\_X } \times \text{ BP\_NPIX\_Y)) } \wedge$

$\forall y, (y < \text{BP\_NPIX\_Y}) \% \text{nat} \rightarrow \forall x, (x < \text{BP\_NPIX\_X}) \% \text{nat} \rightarrow$

$\text{let ir := image\_r (y } \times \text{ BP\_NPIX\_X + x)) \text{ in}$

$\text{let ii := image\_i (y } \times \text{ BP\_NPIX\_X + x)) \text{ in}$

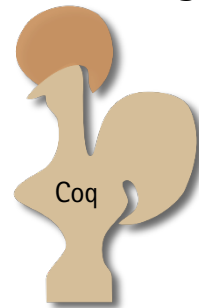
$\text{is\_finite \_ \_ ir = true } \wedge \text{ is\_finite \_ \_ ii = true } \wedge$

$\text{let (tr, ti) := SARBackProj.sar\_backprojection y x \text{ in}$

$\text{Rabs (B2R \_ \_ ir - tr) } \leq \text{pixel\_bound } \wedge$

$\text{Rabs (B2R \_ \_ ii - ti) } \leq \text{pixel\_bound.}$

Conclusions





# Final Correctness Statement (slightly simplified)

$\forall P \text{ ` (HYPS: SARHypotheses P) } m \mid \text{Hypotheses on input data}$

$(Hm: \text{holds } m (P \text{ ++}$

$P_{\text{perm\_int}} \text{ bir oir } (BP\_NPIX\_X \times BP\_NPIX\_Y) \text{ ++}$

$P_{\text{perm\_int}} \text{ bii oii } (BP\_NPIX\_X \times BP\_NPIX\_Y)),$

Memory contents  
and permissions

$\exists m', \text{star Clight.step2}$

$(\text{Callstate fn\_sar\_backprojection ...}) (\text{Returnstate Vundef Kstop } m') \wedge$

$\exists \text{image\_r image\_i,}$

$\text{holds } m (P \text{ ++}$

$\text{Parray\_int image\_r bir oir } (BP\_NPIX\_X \times BP\_NPIX\_Y) \text{ ++}$

$\text{Parray\_int image\_i bii oii } (BP\_NPIX\_X \times BP\_NPIX\_Y)) \wedge$

$\forall y, (y < BP\_NPIX\_Y)\%nat \rightarrow \forall x, (x < BP\_NPIX\_X)\%nat \rightarrow$

$\text{let } ir := \text{image\_r } (y \times BP\_NPIX\_X + x) \text{ in}$

$\text{let } ii := \text{image\_i } (y \times BP\_NPIX\_X + x) \text{ in}$

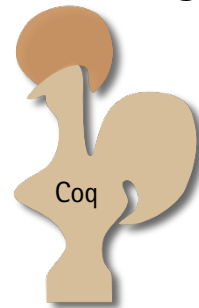
$\text{is\_finite \_\_ } ir = \text{true} \wedge \text{is\_finite \_\_ } ii = \text{true} \wedge$

$\text{let } (tr, ti) := \text{SARBackProj.sar\_backprojection } y \ x \text{ in}$

$\text{Rabs (B2R \_\_ } ir - tr) \leq \text{pixel\_bound} \wedge$

$\text{Rabs (B2R \_\_ } ii - ti) \leq \text{pixel\_bound}.$

Conclusions



# Final Correctness Statement (slightly simplified)

$\forall P$  (HYPS: SARHypotheses P) m

(Hm: holds m (P ++

Pperm\_int bir oir (BP\_NPIX\_X × BP\_NPIX\_Y) ++

Pperm\_int bii oii (BP\_NPIX\_X × BP\_NPIX\_Y)),

Hypotheses

$\exists m'$ , star Clight.step2

(Callstate fn\_sar\_backprojection ...) (Returnstate Vundef Kstop m')  $\wedge$

C code runs

$\exists$  image\_r image\_i,

holds m (P ++

Parray\_int image\_r bir oir (BP\_NPIX\_X × BP\_NPIX\_Y) ++

Parray\_int image\_i bii oii (BP\_NPIX\_X × BP\_NPIX\_Y)  $\wedge$

Memory contents

$\forall y, (y < BP\_NPIX\_Y)\%nat \rightarrow \forall x, (x < BP\_NPIX\_X)\%nat \rightarrow$

let ir := image\_r (y × BP\_NPIX\_X + x) in

let ii := image\_i (y × BP\_NPIX\_X + x) in

is\_finite \_\_ ir = true  $\wedge$  is\_finite \_\_ ii = true  $\wedge$  FP does not overflow

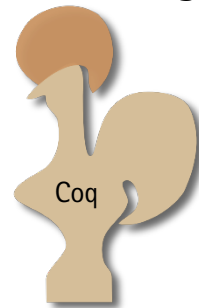
let (tr, ti) := SARBackProj.sar\_backprojection y x in

Rabs (B2R \_\_ ir - tr)  $\leq$  pixel\_bound  $\wedge$

Rabs (B2R \_\_ ii - ti)  $\leq$  pixel\_bound.

Total implementation error bound (approximation + rounding)

*computed at proof-building time*



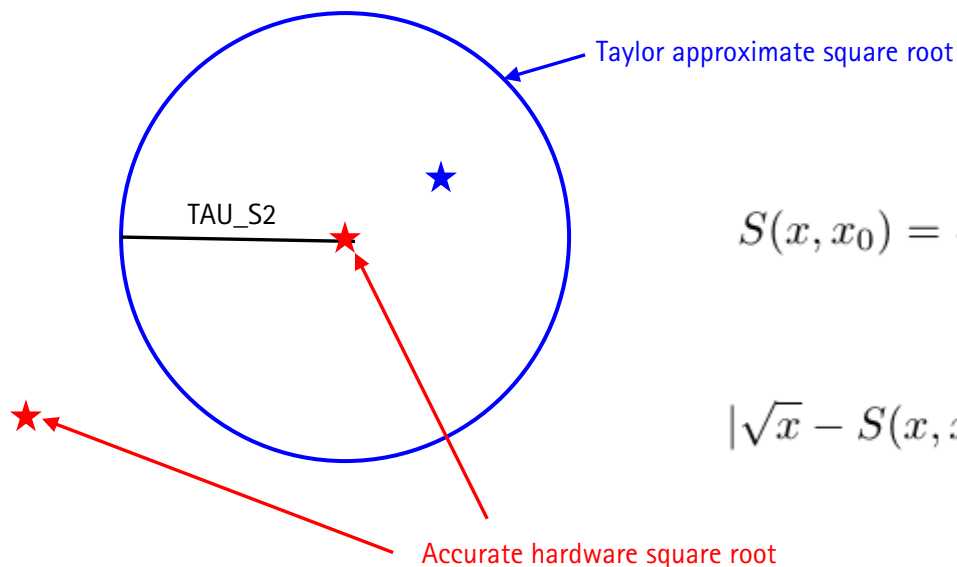
# Polynomial Approximations of Sine

Built-in hardware sine is costly in energy and time

- Replace core sine with a polynomial approximation
  - Use convex optimization
  - Compute coefficients with **unverified** numerical tools
  - Do not trust the results, use Coq to prove an error bound
- Naïve argument reduction is enough for SAR
  - Errors due to approximation of  $\pi$  and roundings
  - Lower than implementation error for core computation

# Adaptive Approximate Square Root

- Replace square root with 2-degree Taylor polynomial
  - Taylor-Lagrange inequality bounds method error
- Valid only in a convergence disc
  - Outside, use accurate hardware square root
  - **Adaptive algorithm:** Re-center the disc as needed

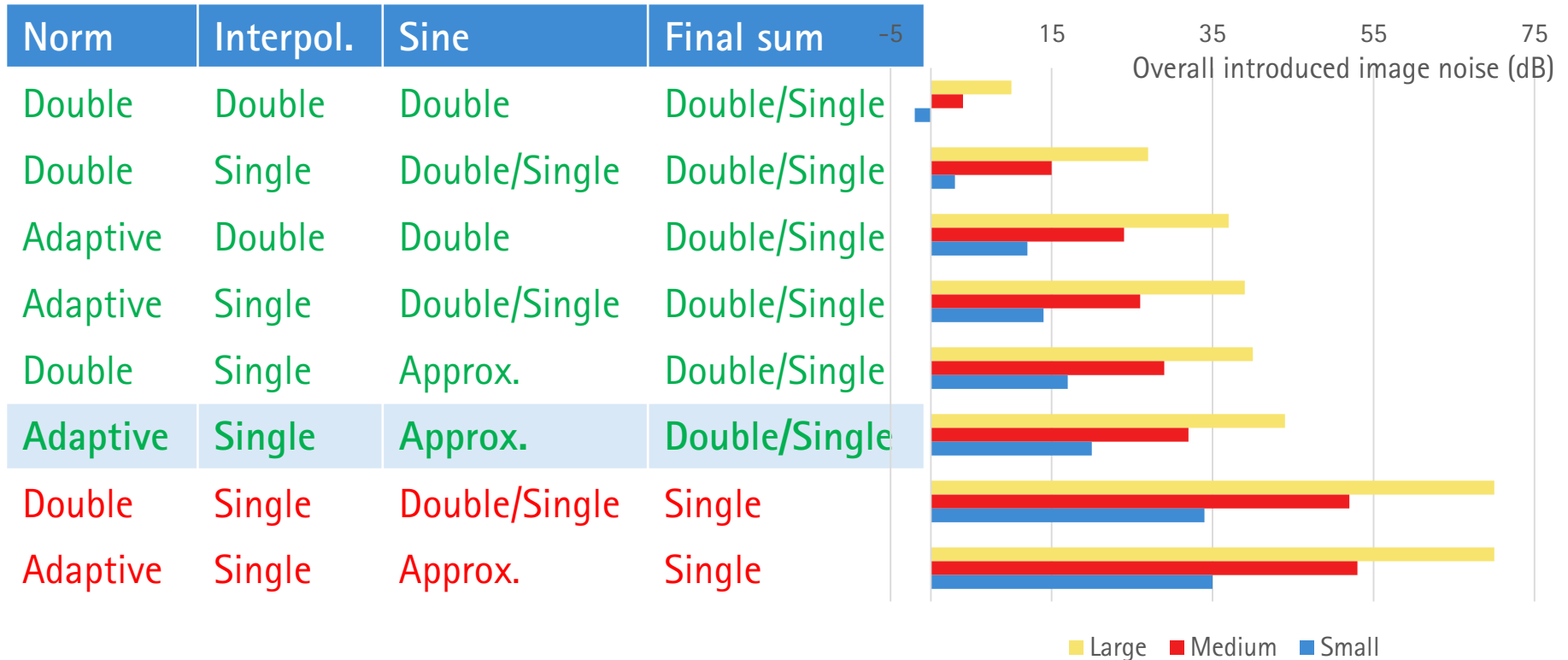


$$S(x, x_0) = \sqrt{x_0} + \frac{x - x_0}{2\sqrt{x_0}} - \frac{(x - x_0)^2}{8(\sqrt{x_0})^3}$$

$$|\sqrt{x} - S(x, x_0)| \leq \frac{(\text{TAU\_S2})^3}{16 \times (x_0 - \text{TAU\_S2})^{5/2}}$$

# Precision Results

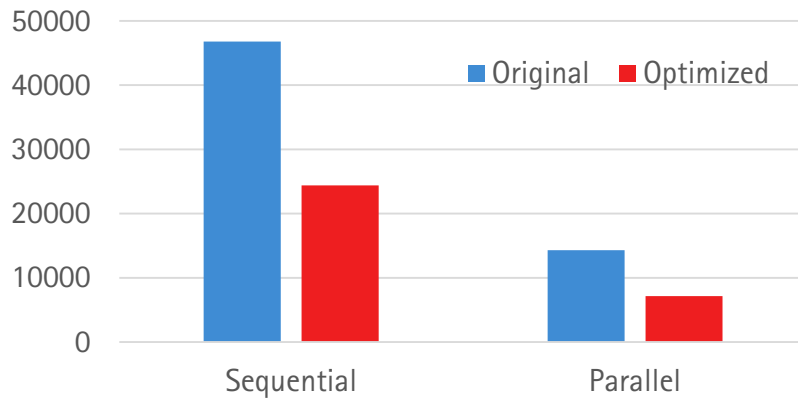
- Input data bounds from DARPA PERFECT suite
- Error grows with image size
- No statistical reasoning about data



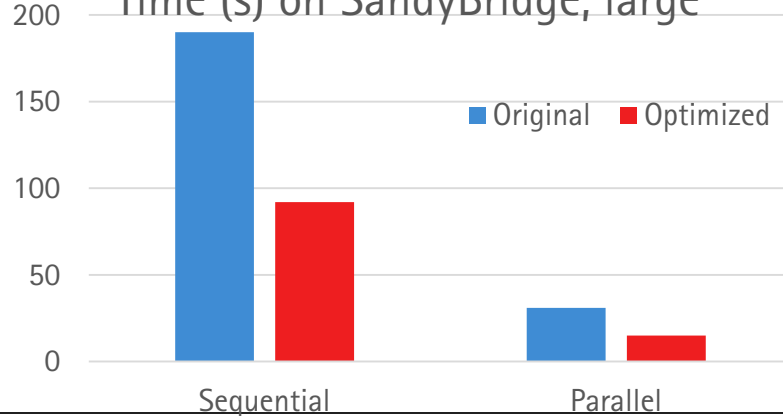
# Performance Measurements for Optimized C Code

- Intel SandyBridge: direct energy measurements
- Intel Haswell: energy model unknown, time instead

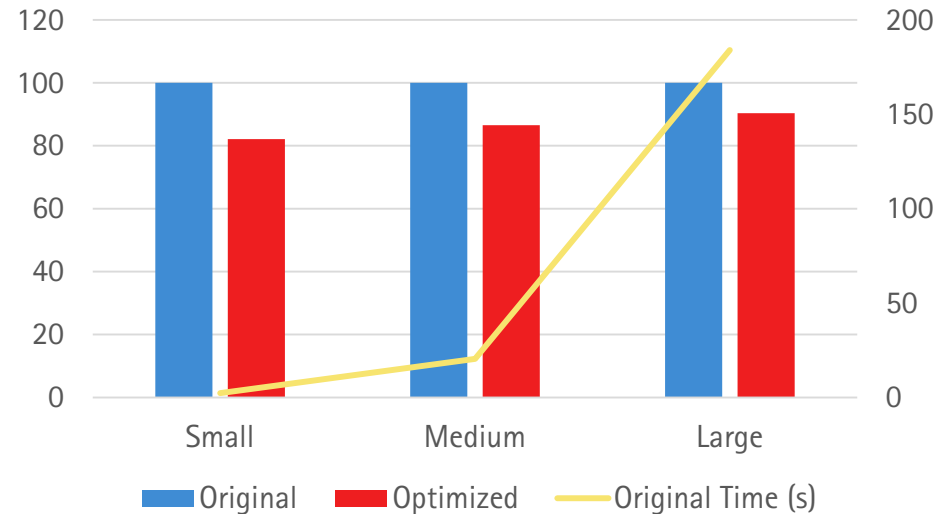
Energy (J) on SandyBridge, large



Time (s) on SandyBridge, large



Time % on Haswell, parallel



# SAR proof: facts and figures

C code size: 150 lines

Proof size:

- Previous all-manual proof: 26k lines, no connection with C
- Thanks to VCFloat: reduced to 12k lines
  - 5k lines of spec (loop invariants), 7k lines of proof
  - ~2k lines of proof for real-number reasoning
  - Remaining part due to C language constructs, could be further reduced when integrating with Verifiable C (Appel et al. 2014)

Proof building/checking time:

- 1 hour (4-core Intel Core i7, 2.10 GHz, 4 Gb RAM)
- Mostly due to interval computations

Formal Verification of Floating-Point Computations in C Programs

# **OUR COQ FRAMEWORK: VCFLOAT**



# Final Correctness Statement (slightly simplified)

$\forall P$  (HYPS: SARHypotheses P) m

(Hm: holds m (P ++

Pperm\_int bir oir (BP\_NPIX\_X × BP\_NPIX\_Y) ++

Pperm\_int bii oii (BP\_NPIX\_X × BP\_NPIX\_Y)),

Hypotheses

$\exists m'$ , star Clight.step2

(Callstate fn\_sar\_backprojection ...) (Returnstate Vundef Kstop m')  $\wedge$

C code runs

$\exists$  image\_r image\_i,

holds m (P ++

Parray\_int image\_r bir oir (BP\_NPIX\_X × BP\_NPIX\_Y) ++

Parray\_int image\_i bii oii (BP\_NPIX\_X × BP\_NPIX\_Y)  $\wedge$

$\forall y, (y < BP\_NPIX\_Y)\%nat \rightarrow \forall x, (x < BP\_NPIX\_X)\%nat \rightarrow$

let ir := image\_r (y × BP\_NPIX\_X + x) in

let ii := image\_i (y × BP\_NPIX\_X + x) in

is\_finite \_\_ ir = true  $\wedge$  is\_finite \_\_ ii = true  $\wedge$  FP does not overflow

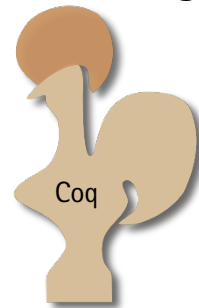
let (tr, ti) := SARBackProj.sar\_backprojection y x in

Rabs (B2R \_\_ ir - tr)  $\leq$  pixel\_bound  $\wedge$  Total implementation error bound (approximation + rounding)

Rabs (B2R \_\_ ii - ti)  $\leq$  pixel\_bound.

*computed at proof-building time*

Memory contents



# Final Correctness Statement (slightly simplified)

$\forall P \text{ ` (HYPS: SARHypotheses P) } m$

$(Hm: \text{holds } m (P \text{ ++}$

$Pperm\_int \text{ bir oir } (BP\_NPIX\_X \times BP\_NPIX\_Y) \text{ ++}$

$Pperm\_int \text{ bii oii } (BP\_NPIX\_X \times BP\_NPIX\_Y)),$

Hypotheses

$\exists m', \text{star Clight.step2}$

$(\text{Callstate fn\_sar\_backprojection ...}) (\text{Returnstate Vundef Kstop } m')$

C code runs

$\exists \text{image\_r image\_i,}$

$\text{holds } m (P \text{ ++}$

CompCert Clight

Memory contents

$\text{Parray\_int image\_r bir oir } (BP\_NPIX\_X \times BP\_NPIX\_Y) \text{ ++}$

$\text{Parray\_int image\_i bii oii } (BP\_NPIX\_X \times BP\_NPIX\_Y)) \wedge$

$\forall y, (y < BP\_NPIX\_Y)\%nat \rightarrow \forall x, (x < BP\_NPIX\_X)\%nat \rightarrow$

$\text{let ir := image\_r } (y \times BP\_NPIX\_X + x) \text{ in}$

$\text{let ii := image\_i } (y \times BP\_NPIX\_X + x) \text{ in}$

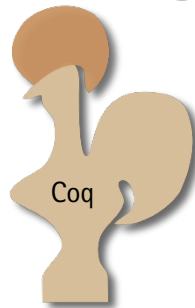
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$\text{let (tr, ti) := SARBackProj.sar\_backprojection } y \ x \text{ in}$

$\text{Rabs (B2R \_\_ ir - tr)} \leq \text{pixel\_bound} \wedge$

$\text{Rabs (B2R \_\_ ii - ti)} \leq \text{pixel\_bound}.$

Total implementation error bound (approximation + rounding) computed at proof-building time



Coq

Flocq

# Our Design Choices: Which Formal Methods?

Verification using the Coq proof assistant

Correctness fully embedded in Coq using existing libraries:

- CompCert Clight (Blazy & Leroy, J. Autom. Reason. 2009)
  - Formal semantics of a deterministic sequential subset of C
- Flocq (Boldo & Melquiond, ARITH 2009)
  - Formalization of floating-point numbers
- Coq standard library
  - Formalization of real numbers

# Our Design Choices: Which Formal Methods?

We use Coq + CompCert Clight + Flocq.

Advantages for trust:

- Unified verification framework
  - OK to combine proof libraries
- Formalization in the Gallina mathematical language of Coq
  - Can be trusted more easily than practical implementations (e.g. Fluctuat, Framac/Why3, etc.)
- Coq is the only setting where C, floating-point and real numbers are trustworthily mixed together

# Our Approach and our Trusted Computing Base

We use Coq + CompCert Clight + Flocq.

- What do we need to trust?
  - Coq's underlying logic is sound
  - Implementation of Coq is sound wrt. Coq's logic
  - Coq standard library real numbers are consistent and faithful
  - Clight is faithful wrt. the corresponding subset of ISO C99
  - Flocq is faithful wrt. IEEE 754-2008 floating-point numbers
- Formalizations in the Gallina mathematical language
- Can be assessed more easily than practical implementations of verification tools

# Verification of C Floating-Point Expressions

```
2.0f * (float) x - 3.0;
```

C floating-point  
expression



?

Real-number  
semantics

# Floating-Point Numbers

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

- A binary operation  $a \mp b$  is not computed exactly
- Rounded from its ideal value
  - Example rounding mode: rounding to nearest
- What is the shape of the **rounding error**?



# Floating-Point Numbers

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

$$\pm m \cdot 2^e \quad 0 \leq m < 2^{\text{prec}}, \quad e_{\min} \leq e \leq e_{\max}$$

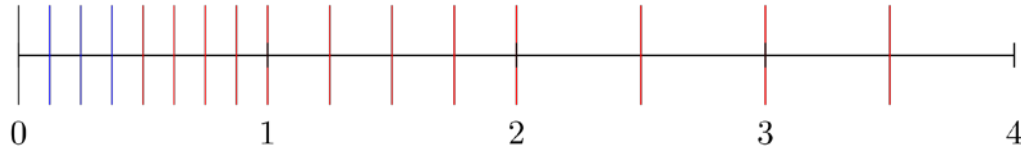


# Floating-Point Numbers

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

$$\pm m \cdot 2^e \mid 0 \leq m < 2^3 = 8; -3 \leq e \leq -1$$

Example: prec = 3, emin = -3, emax = -1

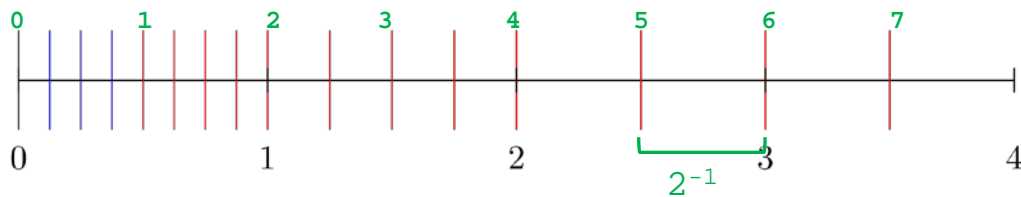
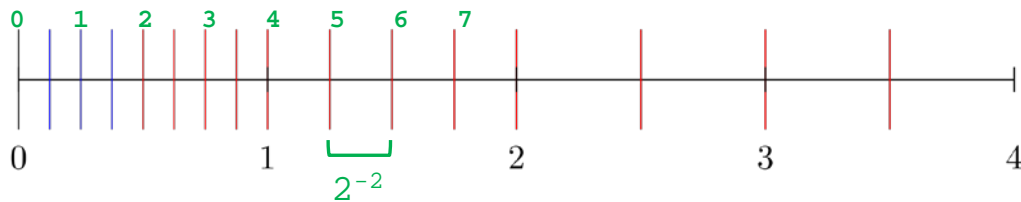


# Floating-Point Numbers

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

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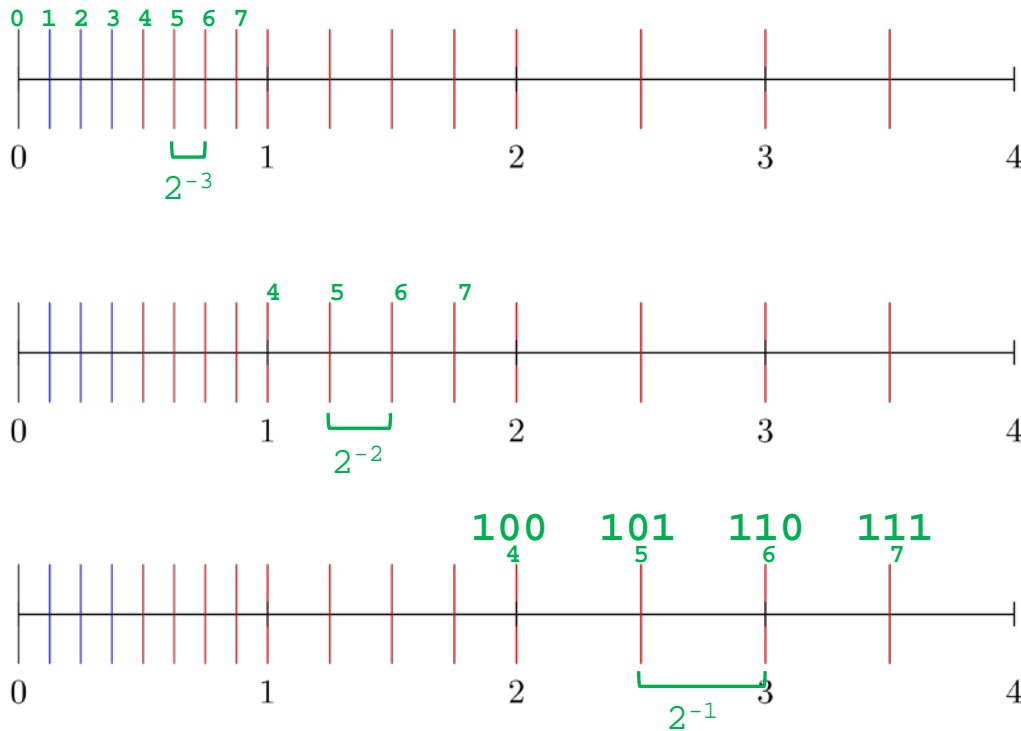


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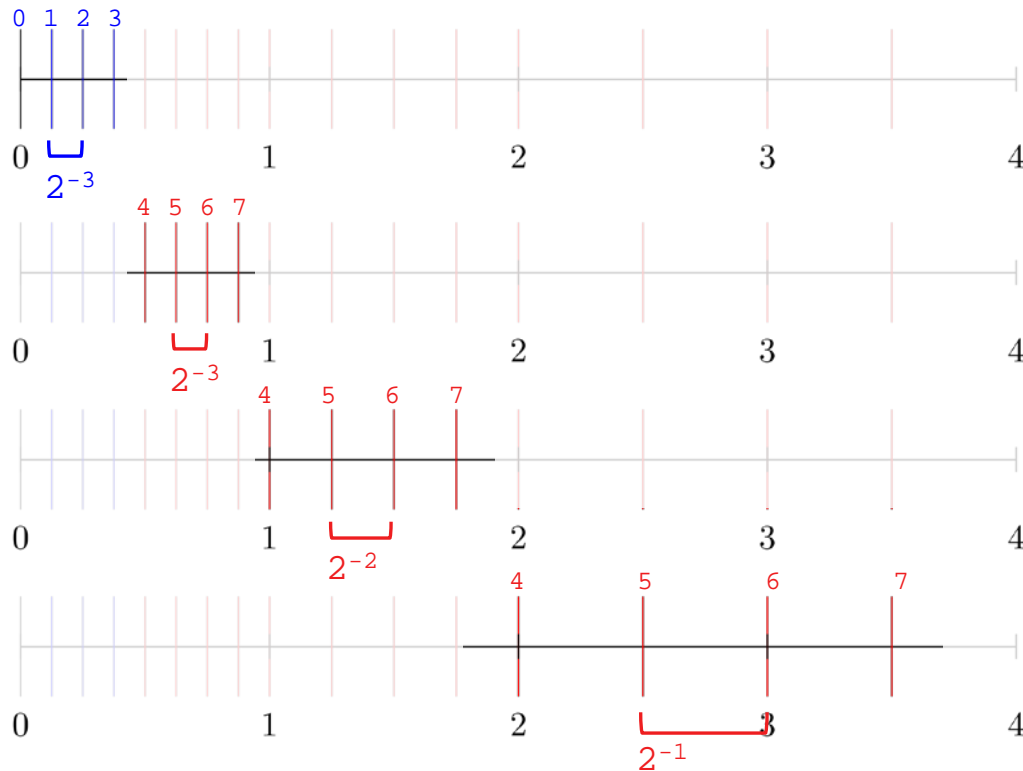


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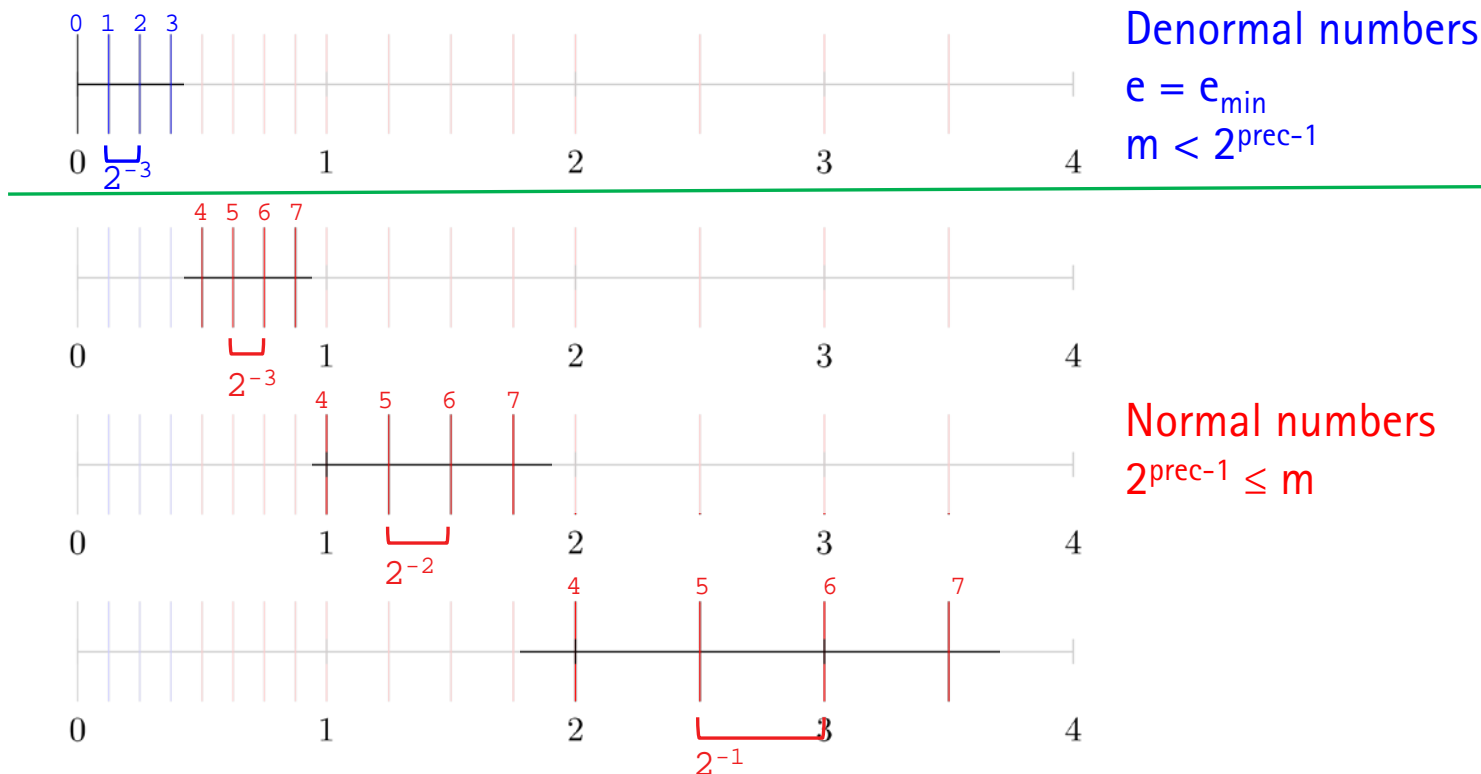


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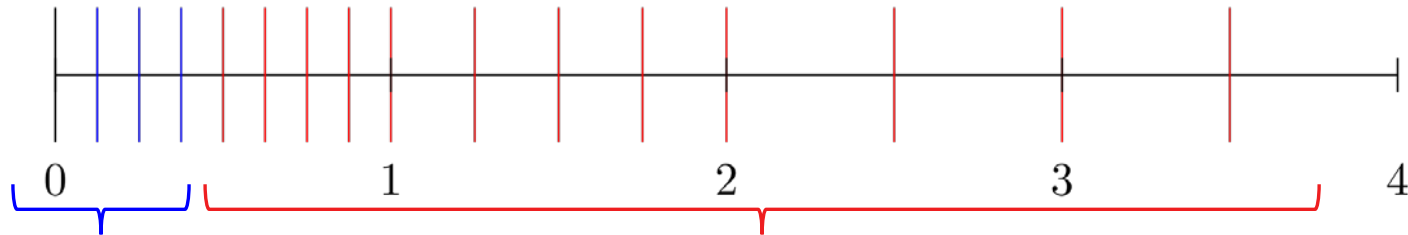
Example: prec = 3, emin = -3, emax = -1



# Floating-Point Numbers

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

$$\pm m \cdot 2^e \quad 0 \leq m < 2^{\text{prec}}, \quad e_{\min} \leq e \leq e_{\max}$$



Denormal numbers

$$e = e_{\min}$$
$$m < 2^{\text{prec}-1}$$

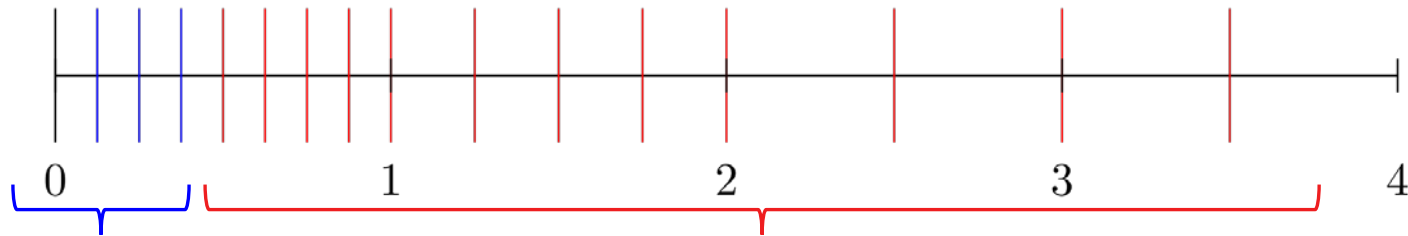
Normal numbers

$$2^{\text{prec}-1} \leq m$$

# Floating-Point Numbers and Rounding Errors

IEEE 754-2008 modelled by Flocq (Boldo et al. 2009)

- A binary operation  $a \text{ T } b$  is not computed exactly
- Rounded from its ideal value
  - Rounding mode: rounding to nearest, ties to even mantissa



Denormal:  
 $(a \text{ T } b) + c$   
 $|c| \leq 2^{\text{emin}-1}$

Normal:  
 $(a \text{ T } b) (1 + d)$   
 $|d| \leq 2^{-\text{prec}}$

General case:  $(a \text{ T } b) (1 + d) + c$   
with  $c \cdot d = 0$

# Optimized Rounding Errors

- Normal numbers:  $(a \text{ T } b) (1 + d)$ , if  $|a \text{ T } b|$  large enough and no overflow
- Denormal numbers:  $(a \text{ T } b) + e$ , if  $|a \text{ T } b|$  small enough
- Sterbenz subtraction:  $(a - b)$  if  $a/2 \leq b \leq 2a$
- Multiply by power of 2 is always exact (unless overflow)
- Divide by power of 2 is exact if no gradual underflow



# Flocq: correctness of floating-point arithmetic

Theorem Bplus\_correct :

forall plus\_nan m x y,

is\_finite x = true ->

is\_finite y = true ->

if  $\text{Rlt\_bool } (\text{Rabs } (\text{round radix2 fexp } (\text{round\_mode } m) (\text{B2R } x + \text{B2R } y))) (\text{bpow radix2 emax})$  then

$\text{B2R } (\text{Bplus plus\_nan } m \ x \ y) = \text{round radix2 fexp } (\text{round\_mode } m) (\text{B2R } x + \text{B2R } y) \wedge$

$\text{is\_finite } (\text{Bplus plus\_nan } m \ x \ y) = \text{true} \wedge$

$\text{Bsign } (\text{Bplus plus\_nan } m \ x \ y) =$

match Rcompare (B2R x + B2R y) 0 with

| Eq => match m with mode\_DN => orb (Bsign x) (Bsign y)

| \_ => andb (Bsign x) (Bsign y) end

| Lt => true

| Gt => false

end

else

$(\text{B2FF } (\text{Bplus plus\_nan } m \ x \ y) = \text{binary\_overflow } m (\text{Bsign } x) \wedge \text{Bsign } x = \text{Bsign } y).$

No overflow

Theorem relative\_error\_ex :

forall x,

$(\text{bpow } \text{emin} \leq \text{Rabs } x) \% R \leftrightarrow$

exists eps,

$(\text{Rabs } \text{eps} < \text{bpow } (-p + 1)) \% R \wedge \text{round beta fexp rnd } x = (x * (1 + \text{eps})) \% R.$

Normal numbers

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is\_fini

Bsign

matc

| Eq

| Lt

| Lt

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else

(B2FF

No overflow

Better automate  
their use

Theorem relative\_error\_ex :

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Normal numbers

# Rounding Error Terms

## Optimized cases

- Normal numbers:  $(a \text{ T } b) (1 + d)$ , if  $|a \text{ T } b|$  large enough and no overflow
- Denormal numbers:  $(a \text{ T } b) + e$ , if  $|a \text{ T } b|$  small enough
- Sterbenz subtraction:  $(a - b)$  if  $a/2 \leq b \leq 2a$
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## Our VCFloat approach:

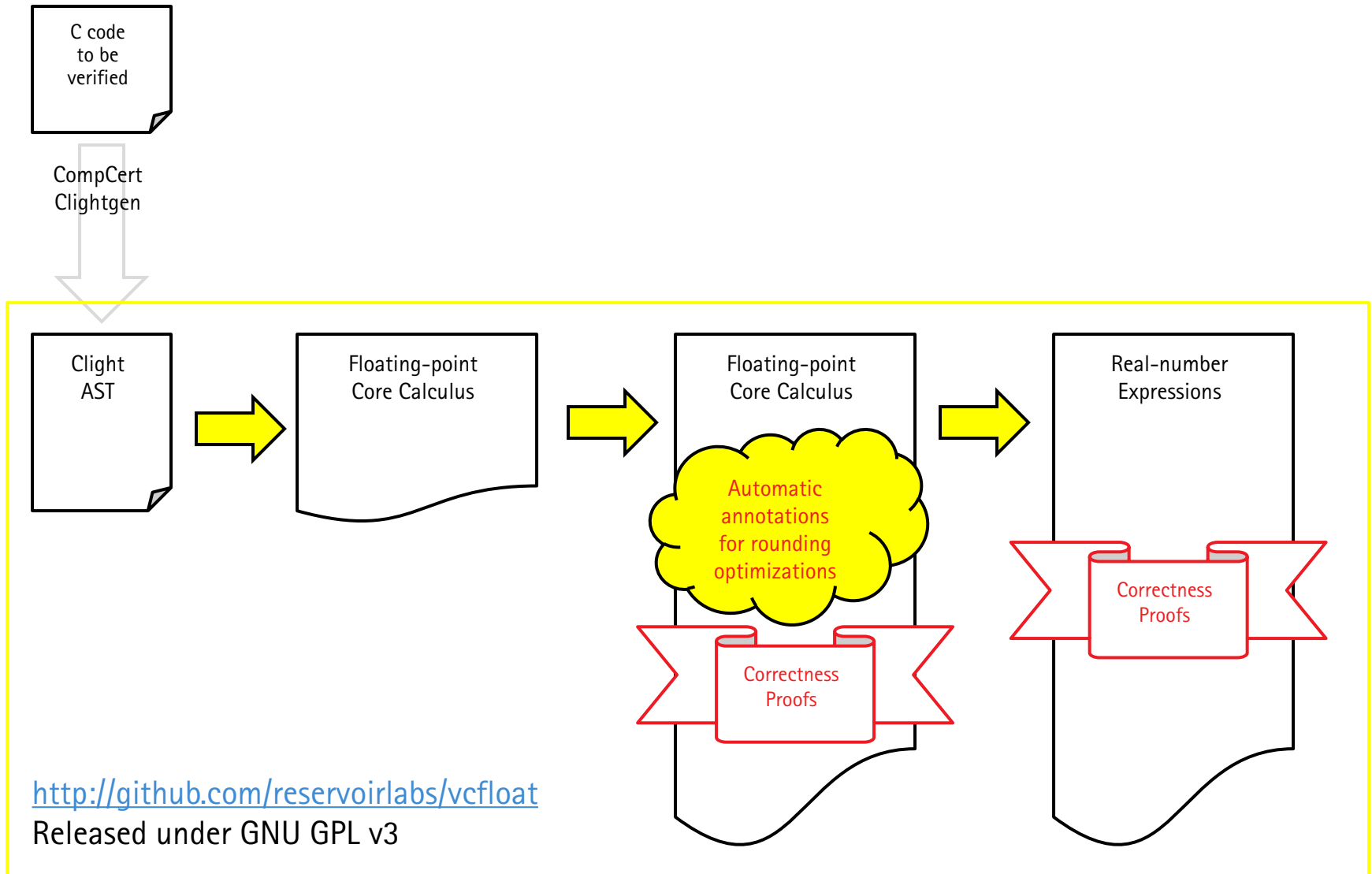
- Automatically generate validity conditions
- Automatically check them on the fly
- Add annotations for optimized rounding

# Verification of Rounding Error Terms

Use Coq-Interval (Melquiond 2015) to automatically check validity conditions

- Automatic certified interval arithmetic
- Reduce correlation issues:
  - Bisection (branch-and-bound)
  - Automatic differentiation
  - Taylor models
- Used for all rounding errors
- All computations within Coq: consumes most proof checking time and memory in overall proof
- Stress test

# Our Verification Framework: VCFloat



# Verification of Rounding Error Terms

```
2.0f * (float) x - 3.0;
```

C floating-point  
expression

# Verification of Rounding Error Terms

`2.0f * (float) x - 3.0;`

C floating-point  
expression

$(2_{(24,128)} \otimes [x]_{(24,128)}) \ominus 3_{(53,1024)}$

Core floating-point  
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$$(2_{(24,128)} \otimes [x]_{(24,128)}^{\text{Norm}}) \ominus 3_{(53,1024)}$$

Annotated floating-point  
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Because  $2^{-125} \leq |x| < 2^{128}$

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Assume  $x$  in  $[1, 2]$

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Annotated floating-point expression

Because  $2^{-125} \leq |x| < 2^{128}$

And for all  $d$  in  $[-2^{-24}, 2^{-24}]$ ,  $|2 * (x * (1 + d))| < 2^{128}$

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And for all  $d$  in  $[-2^{-24}, 2^{-24}]$ ,  $3/2 \leq 2 * (x * (1 + d)) \leq 3 * 2$

$$2 \times (x \times (1 + \delta)) - 3$$

For some  $\delta$  in  $[-2^{-24}, 2^{-24}]$

Real-number expression with error terms

Formal Verification of Error Bounds

# DEMO

Formal Verification of Error Bounds

# CONCLUSIONS

## Ongoing Work

- Floating-point steps are now automated
- Overall more practical improvements ongoing:
  - Integer handling
  - C control flow: Integrate into Verifiable C (Princeton)
  - More statistical support for hypotheses on input data to tighten error bounds

## Conclusion and Opportunities

C programs with floating-point computations can now be fully verified within Coq with a TCB smaller than ever:

- Implementation of Coq, faithfulness of Clight and Flocq

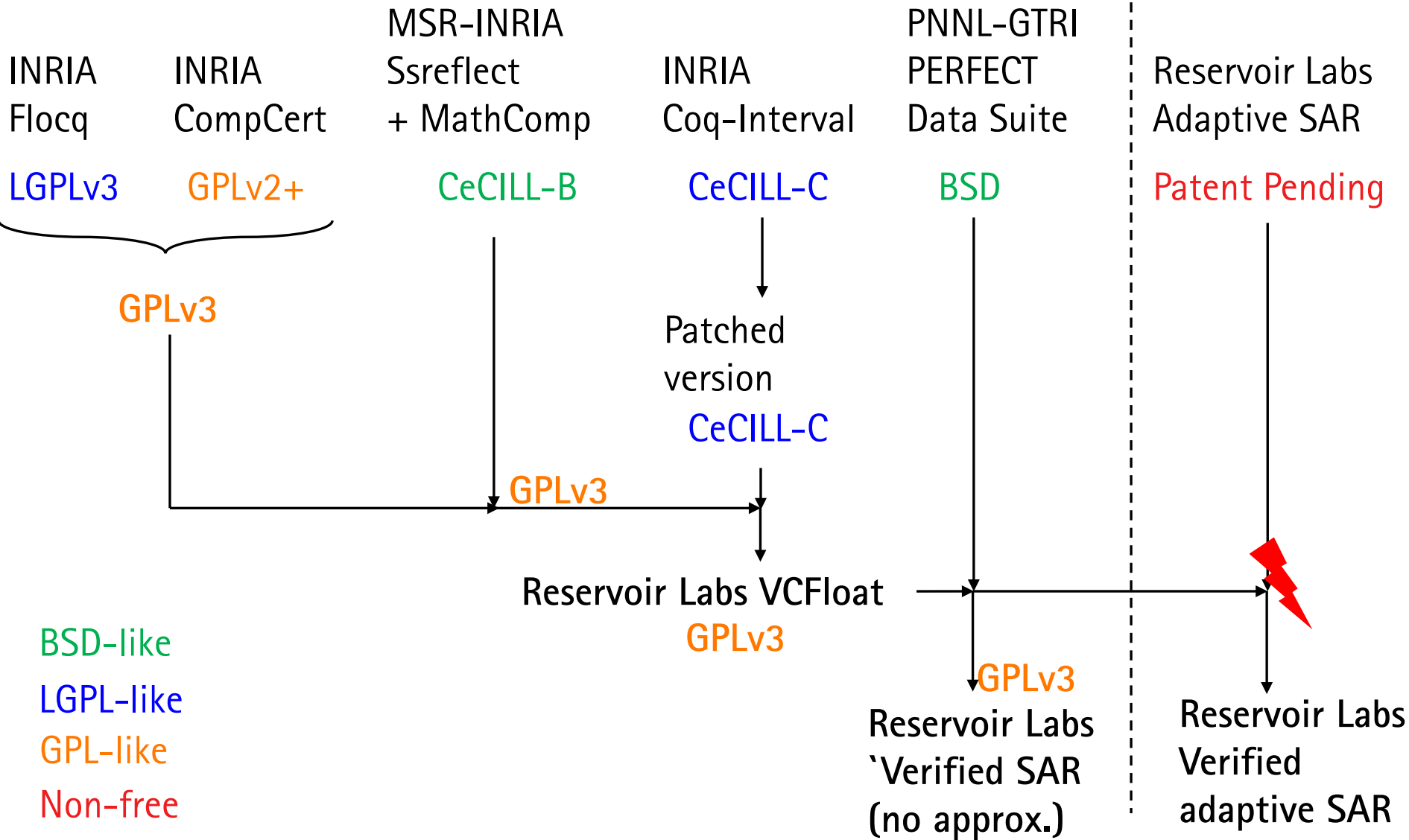
Error bounds certified with VCFloat  
allow the use of energy-efficient approximate  
implementations in critical applications

Ready for use in research and industry

- NSF DeepSpec
- End-to-end verification of cyber-physical systems



# Proofs, Code and Data Rights



# Thank you!

Coq library and proofs available online:

- <http://github.com/reservoirlabs/vcfloat>
- Stress test for Flocq, Coq-Interval, and computations within Coq
- Conference paper published at ACM/SIGPLAN CPP 2016
  - Ramananandro, Mountcastle, Meister, Lethin.  
*A Unified Coq Framework for Verifying C Programs with Floating-Point Computations*

For more information:

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- <http://www.reservoir.com/wp-publications/ramananandro2016>