

Implementation of affine arithmetic using floating-point arithmetic: how to handle roundoff errors

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affine arithmetic and applications

computer arithmetic
floating-point arithmetic
roundoff errors

Question: Is the inclusion property preserved?

Question: Are roundoff errors accounted for?

implementation of affine arithmetic
using floating-point arithmetic:
fast and accurate

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implementation of affine arithmetic
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Contributions

Roundoff errors:

Use of directed roundings:

Influence of the architecture?

Agenda

Affine arithmetic: state of the art

Affine arithmetic: definition

Affine arithmetic: handling of roundoff errors

Affine arithmetic: new approach

Bounding the roundoff errors

Computing the roundoff errors

Accumulating the roundoff errors

Directed rounding modes

Experiments

Evaluation of Shekel5

Remaining questions

Conclusion

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Affine arithmetic

Comba, de Figueiredo, Stolfi – Vu, Sam-Haroud, Faltings

Variant of interval arithmetic, thus guaranteed enclosures of the sought result.

Why? To counteract the dependency problem.

With affine arithmetic, $\mathbf{x} - \mathbf{x} = [0]$.

How? Each quantity is a linear combination of noise symbols:

$$\hat{\mathbf{x}} = x_0 + \sum_i x_i \epsilon_i \text{ where } x_i \in \mathbb{R} \text{ and } \epsilon_i \text{ lives in } [-1, 1].$$

Affine arithmetic: operations

Comba, de Figueiredo, Stolfi

Notations

$$\alpha \in \mathbb{R}, \beta \in \mathbb{R},$$

$$\hat{\mathbf{x}} = x_0 + \sum_i x_i \epsilon_i \text{ where } x_i \in \mathbb{R} \text{ and } \epsilon_i \text{ lives in } [-1, 1],$$

$$\hat{\mathbf{y}} = y_0 + \sum_i y_i \epsilon_i \text{ where } y_i \in \mathbb{R} \text{ and } \epsilon_i \text{ lives in } [-1, 1].$$

$$\alpha + \beta \hat{\mathbf{x}} + \hat{\mathbf{y}} = (\alpha + \beta x_0 + y_0) + \sum_i (\beta x_i + y_i) \epsilon_i.$$

$$\hat{\mathbf{x}} \times \hat{\mathbf{y}} = \dots$$

Affine arithmetic: multiplication... and limiting the number of noise symbols

$$\hat{\mathbf{x}} \times \hat{\mathbf{y}} = x_0 \times y_0 + \dots + \sum_i (x_0 y_i + x_i y_0) \epsilon_i + \epsilon_{n+1}$$

ϵ_{n+1} is a new symbol created to account for nonlinear terms.

Explosion of the number of symbols: handled by fixing the maximal number of such symbols.

Handling the non-created symbols? through a dedicated symbol ϵ_{\pm} .

Variations exist...

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Variations exist...

Affine arithmetic: handling of roundoff errors

Implementation using floating-point arithmetic.

Question: **Is the inclusion property preserved?**

Question: **Are roundoff errors accounted for?**

Affine arithmetic: handling of roundoff errors

Comba, de Figueiredo, Stolfi

$$\alpha + \beta \hat{\mathbf{x}} + \hat{\mathbf{y}} = (\alpha + \beta x_0 + y_0) + \sum_i (\beta x_i + y_i) \epsilon_i.$$

For each operation, either + or \times : roundoff error. Ex.: $\beta \times x_i$.

Roundoff error computed as:

$$e = \max(\text{RU}(\beta \times x_i) - \text{RN}(\beta \times x_i), \text{RN}(\beta \times x_i) - \text{RD}(\beta \times x_i)).$$

Roundoff errors accumulated in ϵ_{\pm} :

$$\epsilon_{\pm} = \text{RU}(\epsilon_{\pm} + e).$$

Affine arithmetic: handling of roundoff errors

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Affine arithmetic: handling of roundoff errors

Hansen, Messine

$$\alpha + \beta \hat{\mathbf{x}} + \hat{\mathbf{y}} = (\alpha + \beta x_0 + y_0) + \sum_i (\beta x_i + y_i) \epsilon_i.$$

For each operation, either + or \times : roundoff error. Ex.: $\alpha + y_0$.

Actual value belongs to the interval $[RD(\alpha + y_0), RU(\alpha + y_0)]$.

Coefficients are intervals:

$$\hat{\mathbf{x}} = \mathbf{x}_0 + \sum_i \mathbf{x}_i \epsilon_i \text{ where } \mathbf{x}_i \in \mathbb{IR} \text{ and } \epsilon_i \text{ lives in } [-1, 1],$$

with operations performed using interval arithmetic.

Affine arithmetic: handling of roundoff errors

Rump in IntLab v.8, Kashiwagi

$$\alpha + \beta \hat{\mathbf{x}} + \hat{\mathbf{y}} = (\alpha + \beta x_0 + y_0) + \sum_i (\beta x_i + y_i) \epsilon_i.$$

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Roundoff error computed as ???

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Bounding roundoff errors

$$\alpha + \beta \hat{\mathbf{x}} + \hat{\mathbf{y}} = (\alpha + \beta x_0 + y_0) + \sum_i (\beta x_i + y_i) \epsilon_i.$$

For each operation, either $+$ or \times : roundoff error.

Each roundoff error is bounded: approach à la COSY (u is the machine roundoff unit):

- ▶ error on $a + b$ is less than $2u \max(|a|, |b|)$;
- ▶ error on $a \times b$ is less than $2u \text{RN}(|a \times b|)$.

Roundoff errors accumulated in \mathbf{I} :

\mathbf{I} is an interval coefficient corresponding to ϵ_{\pm} .

$$\mathbf{I} = \mathbf{I} + \text{ bounds on roundoff errors}$$

with operations performed using interval arithmetic

Computing the roundoff errors

Two useful properties of floating-point arithmetic:

- ▶ in rounding-to-nearest, the roundoff error for $+$, $-$, \times is a floating-point number;
- ▶ this error can be computed using floating-point arithmetic. Codes that transform $a \diamond b$ into $r + e$ with $r = \text{RN}(a \diamond b)$ and e the roundoff error are called **EFT: Error Free Transforms**.

Let's make use of EFT!

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EFT for +

TwoSum: $s + e = a + b$

$$s = \text{RN}(a + b)$$

$$a' = \text{RN}(s - b)$$

$$b' = \text{RN}(s - a')$$

$$\delta_a = \text{RN}(a - a')$$

$$\delta_b = \text{RN}(b - b')$$

$$e = \text{RN}(\delta_a + \delta_b)$$

The equality $s + e = a + b$ holds in **exact arithmetic**.

EFT for \times

FMA: Fused Multiply and Add is a floating-point operator that performs $a \times b + c$ with only one roundoff error of the exact result.

TwoProd: $p + e = a \times b$

$$p = \text{RN}(a \times b)$$

$$e = \text{RN}(\text{FMA}(a, b, -p))$$

Without FMA: it is also possible to compute e , the code is a bit longer (17 operations).

EFT for other operations?

What about other operations? $/$, $\sqrt{}$

No EFT.

The “remainder” is a floating-point number that can be computed... but not very useful for our purpose.

Roundoff errors are bounded.

Accumulating the roundoff errors

Roundoff errors are computed exactly. What to do with them?

Accumulate.

How?

- ▶ No need to be very accurate.
- ▶ Need to get an upper bound.

As only non-negative values are accumulated, this gives a bound:

$$\text{accu} = \text{RU}(\text{accu} + e).$$

Directed rounding modes or not?

Directed rounding modes can incur time penalty.

From 10 to 100 when rounding modes are set using global registers and pipelines must be flushed when the rounding mode changes.

Nothing when rounding modes are set in the instruction code (cf. CUDA for GPU, assembler for Itanium).

Directed rounding modes cannot be set. In OpenMP, OpenCL, BLAS...: the only rounding mode is rounding-to-nearest.

Get free of the rounding modes!

As only non-negative terms are accumulated, this gives a bound:

$$\text{accu} = \text{RN}((1 + 4u) \times (\text{accu} + e)).$$

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Framework

Variants of affine arithmetic are implemented in **IBEX**:

- ▶ double = no guarantee, arithmetic provided by the processor;
- ▶ AF_no: affine arithmetic without control of roundoff errors;
- ▶ sAF: de Figueiredo and Stolfi's version;
- ▶ iAF: Hansen's and Messine's variant: affine arithmetic with interval coefficients;
- ▶ fAF: roundoff errors are bounded and accumulated in an interval;
- ▶ fAF_v2: roundoff errors are computed and accumulated – in rounding to nearest – in a floating-point value.

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Evaluation of Shekel 5: time and accuracy

Points	x_1	x_2	$f(x_2)$
double	0.12s	-	-
Interval	148.4s	154.0s	$[-10.1663975282993153, -7.28436947861868234]$
AF_no	1072.5s	1377.1s	$[-10.1611323182907558, -7.04912944184162704]$
sAF	3096.5s	3593.6s	$[-10.1611323182908002, -7.04912944184159329]$
iAF	1898.4s	3763.7s	$[-10.1611323182907629, -7.04912944184162082]$
fAF	1900.1s	1946.8s	$[-10.1611323182912301, -7.04912944184125667]$
fAF_v2	1306.3s	1579.4s	$[-10.1611323182907913, -7.04912944184159862]$

Table: CPU-time of 10^8 evaluations of the Shekel-5 function.

Caution

Still experimental code:

- ▶ compiler's options probably need a finer tuning;
- ▶ with FMA: prototype machine, buggy code for the time being.

Evaluation of Shekel 5: timings

Points	x_1	x_2
double	0.01s	-
Interval	1.48s	1.54s
Interval	8.45s	9.86s
AF_no	10.72s	13.77s
AF_no	42.5s	57.6s
sAF	30.96s	35.94s
sAF	562.9s	371.6s
iAF	18.98s	37.64s
iAF	73.6s	141.7s
fAF	19.00s	19.47s
fAF	75.0s	97.2s
fAF_v2	13.06s	15.79s
fAF_v2	71.8s	95.6s

≈ CPU-time of 10^6 evaluations of the Shekel-5 function on a Xeon
CPU-time of 10^6 evaluations of the Shekel-5 function on an AMD
(with EMA)

On different architectures

Caution (reminder): experimental and probably buggy code.

Use of fAF (bounds on the roundoff errors, only one interval: the remainder):

- ▶ on Xeon: 1 to 2 times faster than other variants of affine arithmetic,
50% to 100% slower than affine arithmetic without roundoff errors,
10,000 times slower than double arithmetic;
- ▶ on AMD: up to 7 times faster than other variants of affine arithmetic,
twice slower than affine arithmetic without roundoff errors,
100 times slower than double arithmetic;
- ▶ in both cases, 10 times slower than interval arithmetic.

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On different architectures

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Use of fAF_v2 (computations of the roundoff errors, no changes of the rounding modes):

- ▶ on Xeon: from 50% to 2.5 times faster than other variants of affine arithmetic,
slightly slower than affine arithmetic without roundoff errors,
1,500 times slower than double arithmetic;
- ▶ on AMD: from 2 to 10 times faster than other variants of affine arithmetic,
similar to affine arithmetic without roundoff errors, 50 times
slower than double arithmetic;
- ▶ in both cases, 6 times slower than interval arithmetic.

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Conclusion and future work

For an implementation of interval/affine/... arithmetic using floating-point arithmetic:

- ▶ to preserve the inclusion property: handle roundoff errors;
- ▶ many ways to handle roundoff errors: bound them, compute them:
think in terms of significant bits
- ▶ influence of the architecture and of the instruction set: to be further explored.

Conclusion and future work

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Global optimization

IBEX has code to solve global optimization problems: technique chosen here = Affine Relaxation Technique.

COCONUT has benchmark problems: 113 problems are tested here.

Global optimization

Version of Affine arithmetic used in IBEX	non-reliable		reliable							
	AF_no		sAF		iAF		fAF		fAF_v2	
	nb	t (s)	nb	t (s)	nb	t (s)	nb	t (s)	nb	t (s)
Average for solved problems	111	72	113	73	113	72	113	67	113	65
Average for problems solved by all versions	111	72	111	75	111	73	111	69	111	66