Accurate and efficient expression evaluation and linear algebra, or Why it can be easier to compute accurate eigenvalues of a Vandermonde matrix than the accurate sum of 3 numbers

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  - At all? In polynomial time?
  - Depends on the model of arithmetic

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  - How does this change algorithms?

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### Motivating Example (1/2)

Def: Accurate means relative error less than 1

How do the following 3 kinds of accurate evaluation problems differ in difficulty?

- 1. Motzkin polynomial  $z^6 + x^2y^2(x^2 + y^2 3z^2)$ , or eig(V) with  $V_{ij} = x_i^j$ , where  $0 < x_1 < x_2 < \dots$
- 2. Eigenvalues of  $\nabla \cdot (\theta \nabla u) + \lambda \rho u = 0$  discretized with the FEM on a triangular mesh, or
  - x + y + z
- 3. Determinant of a Toeplitz matrix

#### Motivating Example (2/2)

Accurate alg. for Motzkin polynomial  $p = z^6 + x^2y^2(x^2 + y^2 - 3z^2)$ 

$$\begin{array}{ll} \text{if} & |x-z| \leq |x+z| \wedge |y-z| \leq |y+z| \\ p = z^4 \cdot [4((x-z)^2 + (y-z)^2 + (x-z)(y-z))] + \\ & + z^3 \cdot [2(2(x-z)^3 + 5(y-z)(x-z)^2 + 5(y-z)^2(x-z) + \\ & 2(y-z)^3)] + \\ & + z^2 \cdot [(x-z)^4 + 8(y-z)(x-z)^3 + 9(y-z)^2(x-z)^2 + \\ & 8(y-z)^3(x-z) + (y-z)^4] + \\ & + z \cdot [2(y-z)(x-z)((x-z)^3 + 2(y-z)(x-z)^2 + \\ & 2(y-z)^2(x-z) + (y-z)^3] + \\ & + (y-z)^2(x-z)^2((x-z)^2 + (y-z)^2) \\ \text{else} & \dots 7 \text{ more analogous cases} \end{array}$$

Can we automate the discovery of such algorithms? Or prove they do not exist, i.e. that extra precision is necessary? How much extra precision?

### Getting the right answer: Outline

- 1. Problem statement and (more) motivating examples
- 2. Classical Model (CM) and Black-Box Model (BBM) of arithmetic
- 3. Necessary and sufficient conditions for accurate evaluation in CM
- 4. Necessary and sufficient conditions for accurate evaluation in BBM
- 5. Consequences for finite precision arithmetic
- 6. Is it worth getting the right answer? Conditioning
- 7. Open problems

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#### **Problem statement**

Given a polynomial (or a family of polynomials) p, either produce an **accurate** algorithm to compute y = p(x), or prove that none exists.

Accurately means relative error  $\eta < 1$ , i.e.

$$\diamond ||y_{\text{computed}} - y| \leq \eta ||y|,$$

♦ 
$$\eta = 10^{-2}$$
 yields two digits of accuracy,

$$\diamond \quad y_{\text{computed}} = 0 \iff y = 0.$$

**50x50** Hilbert Matrix -  $\log_{10}$  (eigenvalues)



#### 40x40 Pascal Matrix - eigenvalues



#### 20x20 Schur Complement of 40x40 Vandermonde Matrix - eigenvalues



#### Complexity of Accurate Algorithms for General Structured Matrices

				Any			Sym
Type of matrix		$\det A$	$A^{-1}$	minor	LDU	SVD	EVD
Acyclic		n	$n^2$	n	$\leq n^2$	$n^3$	N/A
(bidiagona)	l and other)						
Total Sign	Compound	n	$n^3$	n	$n^4$	$n^4$	$n^4$
(TSC)							
Diagonally Scaled Totally		$n^3$	$n^{5}?$	$n^3$	$n^3$	$n^3$	$n^3$
Unimodular (DSTU)							
Weakly diagonally		$n^3$	$n^3$	?	$n^3$	$n^3$	$n^3$
dominant M-matrix							
	Cauchy	$n^2$	$n^2$	$n^2$	$\leq n^3$	$n^3$	$n^3$
Displace-							
ment	Vandermonde	$n^2$	?	?	?	$n^3$	$n^3$
Rank One							
	Polynomial	$n^2$	?	?	?	?	?
	Vandermonde						
Toeplitz		?	?	?	?	?	?

### Complexity of Accurate Algorithms for Totally Nonnegative (TN) Matrices

Type of			Any	Gai	lSS.	elim.	NE	Ax=b		Eig.
Matrix	$\det A$	$A^{-1}$	minor	NP	PP	CP	NP		SVD	Val.
Cauchy	$n^2$	$n^2$	$n^2$	$n^2$	$n^3$	$n^3$	$n^2$	$n^2$	$n^3$	$n^3$
Vandermonde	$n^2$	$n^3$	$n^3$	$n^2$	$n^2$	poly	$n^2$	$n^2$	$n^3$	$n^3$
Generalized	$n^2$	$n^3$	poly	$n^2$	$n^2$	poly	$n^2$	$n^2$	$n^3$	$n^3$
Vandermonde										
Any TN in	n	$n^3$	$n^3$	$n^3$	$n^3$	$n^3$	0	$n^2$	$n^3$	$n^3$
Neville form										

Def: A is Totally Positive (TP) if all minors are positive. A is Totally Nonnegative (TN) if all minors are nonnegative.

*Theorem:* The class of TN matrices for which we can do accurate linear algebra in polynomial time is closed under multiplication, taking submatrices, Schur complement, J-inverse and converse.

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- All quantities are arbitrary real numbers, or complex numbers (bits come later)
- *fl*(*a*⊗*b*) = (*a*⊗*b*)(1+δ), with arbitrary roundoff error |δ| < ε ≪ 1</li>
  Operations?

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  Operations?
  - $\diamond$  in Classical Model (CM), +, -, ×; also exact negation;

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  - ♦ in Black-Box Model (BBM), in addition to the above, polynomial expressions (e.g.  $x - y \cdot z$  (FMA), x + y + z, dot products, small determinants, ...)

• Constants?

### Availability of constants?

- Classical Model:
  - without  $\sqrt{2}$ , we cannot compute

$$x^2 - 2 = (x - \sqrt{2})(x + \sqrt{2})$$

accurately.

- no loss of generality for homogeneous, integer-coefficient polynomials.
- Black-Box Model:
  - any constants we choose can be accommodated.

- All quantities are arbitrary real numbers, or complex numbers (bits come later)
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- Constants? none in Classical Model, anything in Black-Box Model.

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• Algorithms?

 $\diamond$  exact answer in finite # of steps in absence of roundoff error

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  - $\diamond$  exact answer in finite # of steps in absence of roundoff error
  - $\diamond$  branching based on comparisons

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  - $\diamond$  branching based on comparisons
  - ◊ non-determinism (because determinism is simulable)
  - $\diamond$  domains to be  $\mathbb{C}^n$  or  $\mathbb{R}^n$  (but some domain-specific results).

#### **Problem Statement, formally:**

 $\diamond$  Notation:

- -p(x) multivariate polynomial to be evaluated,  $x = (x_1, \ldots, x_k)$ .
- $-\delta = (\delta_1, \ldots, \delta_m)$  is the vector of error (rounding) variables.
- $-p_{comp}(x, \delta)$  is the result of algorithm to compute p at x with errors  $\delta$ .
- ◇ Goal: Decide if ∃ algorithm  $p_{comp}(x, \delta)$  to accurately evaluate p(x) on  $\mathcal{D}$ :  $\forall 0 < \eta < 1$  ... for any  $\eta$  = desired relative error  $\exists 0 < \epsilon < 1$  ... there is an  $\epsilon$  = maximum rounding error  $\forall x \in \mathcal{D}$  ... so that for all x in the domain  $\forall |\delta_i| \le \epsilon$  ... and for all rounding errors bounded by  $\epsilon$  $|p_{comp}(x, \delta) - p(x)| \le \eta \cdot |p(x)|$  ... relative error is at most  $\eta$

♦ Given p(x) and  $\mathcal{D}$ , seek effective procedure ("compiler") to exhibit algorithm, or show one does not exist

Examples in classical arithmetic over  $\mathbb{R}^n$  (none work over  $\mathbb{C}^n$ ).

• 
$$M_2(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 - 2 \cdot z^2)$$

- Positive definite and homogeneous, easy to evaluate accurately

• 
$$M_3(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 - 3 \cdot z^2)$$

– Motzkin polynomial, nonnegative, zero at |x| = |y| = |z|

$$\begin{split} \text{if} & |x-z| \leq |x+z| \wedge |y-z| \leq |y+z| \\ p &= z^4 \cdot [4((x-z)^2 + (y-z)^2 + (x-z)(y-z))] + \\ &+ z^3 \cdot [2(2(x-z)^3 + 5(y-z)(x-z)^2 + 5(y-z)^2(x-z) + \\ &2(y-z)^3)] + \\ &+ z^2 \cdot [(x-z)^4 + 8(y-z)(x-z)^3 + 9(y-z)^2(x-z)^2 + \\ &8(y-z)^3(x-z) + (y-z)^4] + \\ &+ z \cdot [2(y-z)(x-z)((x-z)^3 + 2(y-z)(x-z)^2 + \\ &2(y-z)^2(x-z) + (y-z)^3] + \\ &+ (y-z)^2(x-z)^2((x-z)^2 + (y-z)^2) \\ \text{else} & \dots 2^{\# \text{vars}-1} \text{ more analogous cases} \end{split}$$

• 
$$M_4(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 - 4 \cdot z^2)$$

– Impossible to evaluate accurately

Sneak Peak.

The variety,

 $V(p) = \{ x : p(x) = 0 \} \ ,$ 

plays a necessary role.

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#### Allowable varieties in Classical Model of arithmetic.

Define *basic allowable sets*:

• 
$$Z_i = \{x : x_i = 0\},$$

• 
$$S_{ij} = \{x : x_i + x_j = 0\},\$$

• 
$$D_{ij} = \{x : x_i - x_j = 0\}.$$

A variety V(p) is *allowable* if it can be written as a finite union of intersections of basic allowable sets.

Denote by

$$\mathbf{G}(\mathbf{p}) = \mathbf{V}(\mathbf{p}) - \cup_{\mathbf{allowable}\ \mathbf{A} \ \subset \ \mathbf{V}(\mathbf{p})}\ \mathbf{A}$$

the set of points in general position.

V(p) unallowable  $\Rightarrow G(p) \neq \emptyset.$ 

### Necessary condition.

**Theorem 1:** V(p) unallowable  $\Rightarrow p$  cannot be evaluated accurately on  $\mathbb{R}^n$  or on  $\mathbb{C}^n$ .

**Theorem 2:** On a domain  $\mathcal{D}$ , if  $\operatorname{Int}(\mathcal{D}) \cap G(p) \neq \emptyset$ , p cannot be evaluated accurately.

Examples on  $\mathbb{R}^n$ , revisited.

• 
$$p(x, y, z) = x + y + z$$
  
UNALLOWABLE

• 
$$M_2(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 - 2 \cdot z^2)$$
  
ALLOWABLE,  $V(p) = \{0\}.$ 

- $M_3(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 3 \cdot z^2)$ ALLOWABLE,  $V(p) = \{|x| = |y| = |z|\}$
- $M_4(x, y, z) = z^6 + x^2 \cdot y^2 \cdot (x^2 + y^2 4 \cdot z^2)$ UNALLOWABLE
- $V(\det(\text{Toeplitz}))$ , UNALLOWABLE  $\Rightarrow$  no accurate linear algebra for Toeplitz (need arbitrary precision arithmetic, as we will see later).

### Necessary condition, real and complex.

**Theorem 1:** V(p) unallowable  $\Rightarrow p$  cannot be evaluated accurately on  $\mathbb{R}^n$  or on  $\mathbb{C}^n$ .

**Theorem 2:** On a domain  $\mathcal{D}$ , if  $\operatorname{Int}(\mathcal{D}) \cap G(p) \neq \emptyset$ , p cannot be evaluated accurately.

### Sketch of proof.

Simplest case: non-branching, no data reuse (except for inputs), non-determinism.

Algorithm can be represented as a tree with extra edges from the sources, each node corresponds to an operation  $(+, -, \times)$ , each node has a specific  $\delta$ , each node has two inputs, one output.

Let  $x \in G(p)$  and define Allow(x) as the smallest allowable set containing x.

### Necessary condition, real and complex.

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### Sketch of proof, cont'd.

*Key fact:* for a positive measure set of  $\delta$ s in  $\delta$ -space, a zero output can be "traced back" down the tree to "allowable" condition ( $x_i = 0$  or  $x_i + x_j = 0$ ), or trivial one ( $x_i - x_i = 0$ ). So for a positive measure set of  $\delta$ s, either

•  $p_{comp}(x, \delta)$  is not 0 (though p(x) = 0), or

• for all  $y \in Allow(x) \setminus V(p)$ ,  $p_{comp}(y, \delta) = 0$  (though  $p(y) \neq 0$ ).

In either case, the polynomial is not accurately evaluable arbitrarily close to x, q.e.d.

### Sufficient Condition, complex case.

**Theorem.** Let p be a polynomial over  $\mathbb{C}^n$  with integer coefficients. If V(p) is allowable, then p is accurately evaluable.

### Sketch of proof.

Can write

$$p(x) = c \prod_i p_i(x) \; ,$$

where  $p_i(x)$  is a power of some  $x_j$  or  $x_j \pm x_k$ , and c is an integer; all operations are accurate.

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**Corollary.** If p is a complex multivariate polynomial, p is accurately evaluable iff p has integer coefficients and V(p) is allowable.

#### Sufficient Condition, real case.

Trickier... Allowability *not* sufficient:

- $q = (u^4 + v^4) + (u^2 + v^2)(x^2 + y^2 + z^2), V(p) = \{u = v = 0\}$ : allowable and accurately evaluable
- $p = (u^4 + v^4) + (u^2 + v^2)(x + y + z)^2$ ,  $V(p) = \{u = v = 0\}$ : allowable but NOT accurately evaluable!
- Has to do with locally dominant behavior (in this case, near the set  $\{u = v = 0\}$ ).

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**Theorem.** If all "dominant terms" are accurately evaluable on  $\mathbb{R}^n$  then p is accurately evaluable. In non-branching case, if p is accurately evaluable on  $\mathbb{R}^n$ , then so are all "dominant terms".

#### What is dominance? Newton Polytope



#### What is dominance? Normal Fan



#### What is dominance? First orthant of -(Normal Fan)



#### What is dominance? Labeling dominant terms



# What is dominance? (x, y) regions where different terms dominant



#### Sufficient Condition, real case.

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- $q = (u^4 + v^4) + (u^2 + v^2)(x^2 + y^2 + z^2), V(p) = \{u = v = 0\}$ : allowable and accurately evaluable
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**Theorem.** If all "dominant terms" are accurately evaluable on  $\mathbb{R}^n$  then p is accurately evaluable. In non-branching case, if p is accurately evaluable on  $\mathbb{R}^n$ , then so are all "dominant terms".

Need inductive procedure of testing accurate evaluability, but so far no clear induction parameter.

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#### Allowable varieties in black-box arithmetic.

Define **black-boxes**:  $q_1, q_2, \ldots, q_k$  are polynomials  $\mathcal{V}_i = \{ V \neq \mathbb{R}^n : V \text{ can be obtained from } q_i \text{ through Process A, below} \}$ 

Process A:

**Step 1.** repeat and/or negate, or 0 out some of the inputs,

**Step 2.** of the remaining variables, keep some symbolic, and find the variety in terms of the others.

Example:  $q_1(x, y) = x - y$  has (up to symmetry)  $\mathcal{V}_1 = \{\{x = 0\}, \{x - y = 0\}, \{x + y = 0\}\},$   $q_2(x, y, z) = x - y \cdot z$  has (up to symmetry)  $\mathcal{V}_2 = \{\{x = 0\}, \{y = 0\} \cup \{z = 0\}, \{x = 0\} \cup \{x = 1\}, \{x = 0\} \cup \{x = -1\},$   $\{x = 0\} \cup \{y = 1\}, \{x = 0\} \cup \{y = -1\}, \{x - y^2 = 0\}, \{x + y^2 = 0\},$  $\{x - yz = 0\}, \{x + yz = 0\}\}.$ 

## Allowable varieties in black-box arithmetic. Define black-boxes: $q_1, q_2, \ldots, q_k$ are polynomials $\mathcal{V}_j = \{ V \neq \mathbb{R}^n : V \text{ can be obtained from } q_j \text{ through Process A} \}$

Define *basic allowable sets*:

- $\bullet Z_i = \{x : x_i = 0\},\$
- $S_{ij} = \{x : x_i + x_j = 0\},\$
- $D_{ij} = \{x : x_i x_j = 0\},\$
- any V for which there is a j such that  $V \in \mathcal{V}_j$ .

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Define **black-boxes**:  $q_1, q_2, \ldots, q_k$  are polynomials

 $\mathcal{V}_j = \{ V \neq \mathbb{R}^n : V \text{ can be obtained from } q_j \text{ through } \mathbf{Process } \mathbf{A} \}$ 

A variety V(p) is *allowable* if it is a union of irreducible parts of finite intersections of basic allowable sets.

Denote by

 $\mathbf{G}(\mathbf{p}) = \mathbf{V}(\mathbf{p}) - \cup_{\text{allowable } \mathbf{A} \subset \mathbf{V}(\mathbf{p})} \mathbf{A}$ 

the set of points in general position.

V(p) unallowable  $\Rightarrow G(p) \neq \emptyset.$ 

#### Necessary condition, real and complex.

**Theorem 1:** V(p) unallowable  $\Rightarrow p$  cannot be evaluated accurately on  $\mathbb{R}^n$  or on  $\mathbb{C}^n$ .

**Theorem 2:** On a domain  $\mathcal{D}$ , if  $\operatorname{Int}(\mathcal{D}) \cap G(p) \neq \emptyset$ , p cannot be evaluated accurately.

### Sufficient condition, complex, for all $q_i$ irreducible.

**Theorem:** If V(p) is a union of intersections of sets  $Z_i$ ,  $S_{ij}$ ,  $D_{ij}$ , and  $V(q_j)$ , then p is accurately evaluable.

**Corollary:** If all  $q_j$  are affine, then p is accurately evaluable iff V(p) is allowable.

### Consequences for Numerical Linear Algebra.

- $V(\det(\text{Toeplitz}))$  contains irreducible factors of arbitrarily large degree  $\Rightarrow$  no set of black-boxes of bounded degree will be sufficient for accurate evaluation  $\Rightarrow$  need arbitrary precision arithmetic to do NLA accurately on Toeplitz matrices.
- Same argument shows that we cannot accurately evaluate many generalized Vandermonde matrices (Schur functions as determinants).
- Conjecture: if the class of structured matrices has displacement rank  $\geq 2$ , then accurate evaluation will not always be possible.

#### Complexity of Accurate Algorithms for General Structured Matrices

				Any			Sym
Type of matrix		$\det A$	$A^{-1}$	minor	LDU	SVD	EVD
Acyclic		n	$n^2$	n	$\leq n^2$	$n^3$	N/A
(bidiagona)	l and other)						
Total Sign Compound		n	$n^3$	n	$n^4$	$n^4$	$n^4$
(TSC)							
Diagonally Scaled Totally		$n^3$	$n^{5}?$	$n^3$	$n^3$	$n^3$	$n^3$
Unimodular (DSTU)							
Weakly diagonally		$n^3$	$n^3$	No	$n^3$	$n^3$	$n^3$
dominant M-matrix							
	Cauchy	$n^2$	$n^2$	$n^2$	$\leq n^3$	$n^3$	$n^3$
Displace-							
ment	Vandermonde	$n^2$	No	No	No	$n^3$	$n^3$
Rank One							
	Polynomial	$n^2$	No	No	No	*	*
	Vandermonde						
Toeplitz		No	No	No	No	No	No

\* = "it depends"

### Getting the right answer: Outline

- 1. Problem statement and (more) motivating examples
- 2. Classical Model (CM) and Black-Box Model (BBM) of arithmetic
- 3. Necessary and sufficient conditions for accurate evaluation in CM
- 4. Necessary and sufficient conditions for accurate evaluation in BBM
- 5. Consequences for finite precision arithmetic
- 6. Is it worth getting the right answer? Conditioning
- 7. Open problems

### Choosing a finite precision arithmetic model

- In finite precision, accuracy always possible, only question is cost
- Measure bit complexity in floating point:  $(e, m) \equiv 2^e \cdot m$
- Contrasts between complexity in Floating Point and Fixed Point
  - Repeated squaring can have exponential cost in Fixed, polynomial in Float
  - $-\operatorname{Det}(A)$  polynomial in Fixed [Clarkson], unknown in Float
  - Witness for matrix singularity (null vector) in Float can have exponentially many bits
  - Computing "middle bits" of  $\prod_{i=1}^{n} (1+x_i)$  polynomial in Fixed, as hard as computing the permanent in Float
- Float seems more natural for attaining relative accuracy

### **Cost implications for Accuracy in Floating Point**

- If a problem is accurately evaluable in Classical Model, the same algorithm works in Float
  - Each operation runs in in polynomial time in size of inputs
  - -Ex: Motzkin polynomial, eig(Vandermonde)
- If a problem is accurately evaluable in Black-Box Model, then if you build an accurate library to evaluate each "black-box" operation, the same algorithm works in Float
  - If each "black-box" operation is of bounded degree and #terms, then each operation runs in polynomial time in size of inputs
  - Ex: eig(discretized scalar elliption PDE), x + y + z
  - If set of black-box operations of unbounded degree and #terms, then cost may be exponential
  - -Ex: det(Toeplitz)

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### Is it worth getting the right answer?

- If the answer is ill-conditioned, why bother?
- Many problems have enormous condition numbers, but moderate *structured* conditioned numbers
  - $-\operatorname{Ex:}$  Hilbert matrix: exponentially large condition number as n grows
  - But  $H_{ij} = 1/(x_i + x_j)$  with  $x_i = i .5$  well-conditioned wrt x().
- True for all structured matrix examples
- So high accuracy deserved!

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### **Open problems**

- **Complete** the decision procedure (analyze the dominant terms) when the domain is  $\mathbb{R}^n$  and V(p) allowable.
- **Narrow** the necessity and sufficiency conditions for the black-box case
- **Extend** to semi-algebraic domains  $\mathcal{D}$ .
- **Conjecture:** Same sufficient conditions for existence of accurate interval algorithms
- **Incorporate** division, rational functions, perturbation theory.
  - Conjecture (Demmel, '04): Accurate evaluation is possible only if condition number has only certain simple singularities (depend on reciprocal distance to set of ill-posed problems).
- **Implement** decision procedure to "compile" an accurate evaluation program given p(x),  $\mathcal{D}$ , and minimal set of "black boxes"

### Reference

For a survey with many other references, see:

"Accurate and efficient expression evaluation and linear algebra," J. Demmel, I. Dumitriu, O. Holtz, P. Koev, *Acta Numerica* (2008), v. 17, pp 87-145

- 1. Getting the right answer
  - At all? In polynomial time?
  - Depends on the model of arithmetic
- 2. Getting the same answer
  - When running same problem on two different machines?
  - When running same problem twice on same machine?
- 3. Getting a fast answer
  - Arithmetic is cheap, moving data is expensive
  - How does this change algorithms?

### Why wouldn't you get the same answer?

- Run same program twice on different machines (reproducibility)
  Different floating point semantics, compilers, ...
- Run same program twice on same machine (repeatability)
  - Floating point nonassociativity and dynamic scheduling of parallel tasks
- Who cares?
  - NA-Digest request for reproducible parallel sparse linear solver for use in a FEM package used by construction engineers with contractual obligations for repeatability
  - Subsequent informal survey of many users gave wide range of reasons for wanting repeatability
  - Debugging

#### Intel MKL is not repeatable



- Experiment:
  - Compute dot products of nearly orthogonal vectors, n = 1000.
  - Vary #thread (1-4), alignments
  - Histogram  $[\max_i v_i \min_i v_i] / \max_i |v_i|$
- Repeatability possible, question is cost [H-D. Nguyen]
  - Cost so far: 2x for n = 1000, 1.2x for  $n = 10^5$

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### Arithmetic is cheap, moving data is expensive.

- Time to do one floating point operation already hundreds of times faster than getting data from main memory, or from another processor
- Technology trends
  - Arithmetic getting faster at  $\approx 60\%/{\rm year}$
  - Communication bandwidth (moving data, either between levels of memory hierarchy or processors over a network) only improving at most 25%/year;
  - Latency worse
  - $-\operatorname{Similar}$  trends for energy

#### Impact on Linear Algebra

- Impact on Direct Linear Algebra (LU, QR, eig, SVD, ...)
  - Thm (Ballard, D., Holtz, Schwartz): Lower bound on communication for *any* of these problems
    - $\ast$  Generalizes existing lower bounds for dense matmul
    - \* Dense or sparse matrices, sequential or parallel
  - LAPACK/ScaLAPACK communicate asymptotically more than lower bounds
  - New algorithms do attain lower bounds large speedups \* Up to 13x measured (or  $\infty x$ ), 29x predicted

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- Impact on Iterative Linear Algebra (Krylov Methods) Ditto
- See bebop.cs.berkeley.edu for papers, www.cs.berkeley.edu/~demmel for short course

### Conclusion

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Don't communic....