### The axiomatic derivation of absolute lower bounds

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Algorithms from primitives — the Euclidean algorithm For  $a, b \in \mathbb{N} = \{0, 1, ...\}, a \ge b \ge 1$ ,

(
$$\varepsilon$$
)  $gcd(a, b) = if (rem(a, b) = 0)$  then b else  $gcd(b, rem(a, b))$ 

where a = iq(a, b)b + rem(a, b)  $(0 \le rem(a, b) < b)$ 

 $\operatorname{calls}(\varepsilon, a, b) = ext{the number of divisions } \varepsilon ext{ needs to compute } \operatorname{gcd}(a, b) \\ \leq 2 \log(b) \qquad (x \geq y \geq 2)$ 

Is the Euclidean optimal for computing gcd(a, b) from rem?

Is the Euclidean optimal for deciding coprimeness from rem?

$$a \bot b \iff \gcd(a, b) = 1$$

- And is this true among all algorithms from rem,  $=_0, =_1$ ?
- Aim: derive provably robust (with respect to the choice of computation model) and plausibly absolute lower bounds for algorithms which compute a function from specified primitives

Outline

Slogan: Absolute lower bound results are the undecidability facts about decidable problems

- (1) Preliminaries
- (2) Uniform processes
- (3) Comprimeness in ℕ
- (4) Polynomial 0-testing

*Is the Euclidean algorithm optimal among its peers?* (with vDD, 2004) *Arithmetic complexity* (with vDD, 2009)

Y. Mansour, B. Schieber, and P. Tiwari (1991)
A lower bound for integer greatest common divisor computations, Lower bounds for computations with the floor operation
J. Meidânis (1991): Lower bounds for arithmetic problems
P. Bürgisser and T. Lickteig (1992) Verification complexity of linear prime ideals
P. Bürgisser, T. Lickteig, and M. Shub (1992), Test complexity of

generic polynomials

## (Partial) structures

• A (partial) structure is a tuple  $\mathbf{M} = (M, \Phi^{\mathbf{M}})$ 

where  $\Phi$  is a set of function (and relation) symbols and  $\Phi^{M} = \{\phi^{M}\}_{\phi \in \Phi}$ , where

$$\phi^{\mathsf{M}}: \mathcal{M}^{n_{\phi}} \rightharpoonup \mathcal{M}_{s}$$
 i.e.,  $\phi^{\mathsf{M}}: \mathcal{M}^{n_{\phi}} \rightharpoonup \mathcal{M}$  or  $\phi^{\mathsf{M}}: \mathcal{M}^{n_{\phi}} \rightharpoonup \{\mathfrak{t}, \mathsf{ff}\}$ 

▶ 
$$\mathbf{N}_{\varepsilon} = (\mathbb{N}, \text{rem}, =_0, =_1)$$
, the Euclidean structure  
▶  $\mathbf{N}_{\varepsilon} \upharpoonright U = (U, \text{rem} \upharpoonright U, =_0 \upharpoonright U, =_1 \upharpoonright U)$  where  $U \subseteq \mathbb{N}$  and  
 $(f \upharpoonright U)(x, y) = w \iff \vec{x} \in U^n, w \in U_s \& f(\vec{x}) = w$ 

 The (equational) diagram of a Φ-structure is the set of its basic equations,

$$\mathsf{eqdiag}(\mathsf{M}) = \{(\phi, \vec{x}, w) : \vec{x}, w \in M \text{ and } \phi^{\mathsf{M}}(\vec{x}) = w\}$$

▶ We may assume that **M** is completely determined by eqdiag(**M**)

#### Homomorphisms and substructures

▶ A homomorphism  $\pi : \mathbf{U} \rightarrow \mathbf{V}$  is any  $\pi : U \rightarrow V$  such that for all  $\phi \in \Phi, x_1, \dots, x_n \in U, w \in U_s$ , (with  $\pi(\mathtt{t}) = \mathtt{t}, \pi(\mathtt{ff}) = \mathtt{ff}$ )

$$\phi^{\mathsf{U}}(x_1,\ldots,x_n)=w\implies\phi^{\mathsf{M}}(\pi x_1,\ldots,\pi x_n)=\pi w$$

It is an embedding if it is injective

Substructures:

 $\mathbf{U} \subseteq_{p} \mathbf{M} \iff U \subseteq M$ & the identity  $I : \mathbf{U} \rightarrow \mathbf{M}$  is an embedding  $\iff U \subseteq M \& \operatorname{eqdiag}(\mathbf{U}) \subseteq \operatorname{eqdiag}(\mathbf{M})$ 

We use finite substructures U ⊆<sub>p</sub> M to represent calls to the primitives executed during a computation in M

## Algorithms from primitives - the basic intuition

An *n*-ary algorithm  $\alpha$  of  $\mathbf{M} = (M, \Phi)$  (or from  $\Phi$ ) "computes" some *n*-ary partial function

$$\overline{\alpha} = \overline{\alpha}^{\mathsf{M}} : M^n \rightharpoonup M_s$$

using the primitives in  $\Phi$  as oracles

We understand this to mean that in the course of a "computation" of  $\overline{\alpha}(\vec{x})$ , the algorithm may request from the oracle for any  $\phi^{M}$  any particular value  $\phi^{M}(\vec{u})$ , for arguments  $\vec{u}$  which it has already computed, and that if the oracles cooperate, then "the computation" of  $\overline{\alpha}(\vec{x})$  is completed in a finite number of "steps"

- The notion of a uniform process attempts to capture minimally (in the style of abstract model theory) these aspects of algorithms from primitives
- It does not capture their effectiveness, but their uniformity —that an algorithm applies "the same procedure" to all arguments in its domain

### I The Locality Axiom

A uniform process  $\alpha$  of arity n (and sort s) of a structure  $\mathbf{M} = (M, \Phi^{\mathbf{M}})$  assigns to each substructure  $\mathbf{U} \subseteq_{p} \mathbf{M}$  an n-ary partial function

 $\overline{\alpha}^{\mathbf{U}}: U^n \rightharpoonup U_s$ 

It computes the partial function  $\overline{\alpha}^{\mathsf{M}}: M^n \rightharpoonup M_s$ 

For an algorithm α, intuitively, α<sup>U</sup> is the restriction to U of the partial function computed by α when the oracles respond only to questions with answers in eqdiag(U)

We write

$$\mathbf{U}\vdash \alpha(\vec{x})=w\iff \vec{x}\in U^n, w\in U_s \text{ and } \overline{\alpha}^{\mathbf{U}}(\vec{x})=w$$

### II The Homomorphism Axiom

If  $\alpha$  is an n-ary uniform process of  $\mathbf{M}$ ,  $\mathbf{U}, \mathbf{V} \subseteq_{p} \mathbf{M}$ , and  $\pi : \mathbf{U} \to \mathbf{V}$  is a homomorphism, then

 $\mathbf{U}\vdash\alpha(\vec{x})=w\implies\mathbf{V}\vdash\alpha(\pi\vec{x})=\pi w\quad(x_1,\ldots,x_n\in U,w\in U_s)$ 

In particular, if  $\mathbf{U} \subseteq_{p} \mathbf{M}$ , then  $\overline{\alpha}^{\mathbf{U}} \sqsubseteq \overline{\alpha}^{\mathbf{M}}$ 

- For algorithms: when asked for φ<sup>U</sup>(x̄), the oracle for φ may consistently provide φ<sup>V</sup>(πx̄), if π is a homomorphism
- This is obvious for the identity embedding *I* : U → M, but it is a strong restriction for algorithms from rich primitives (stacks, higher type constructs, etc.)
   It can be verified for the standard (deterministic and non-deterministic) computation models

#### III The Finiteness Axiom

If  $\alpha$  is an n-ary uniform process of **M**, then

 $\mathbf{M}\vdash\alpha(\vec{x})=w$ 

 $\implies$  there is a finite  $\mathbf{U} \subseteq_p \mathbf{M}$  generated by  $\vec{x}$  such that  $\mathbf{U} \vdash \alpha(\vec{x}) = w$ 

For every call \(\vec{u}\) to the primitives, the algorithm must construct the arguments \(\vec{u}\), and so the entire computation takes place within a finite substructure generated by the input \(\vec{x}\)
 We write

$$|\mathbf{U} \vdash_{c} \alpha(\vec{x}) = w \iff \mathbf{U}$$
 is finite, generated by  $\vec{x}$  and  $\mathbf{U} \vdash \alpha(\vec{x}) = w$ ,

and we think of  $(\mathbf{U}, \vec{x}, w)$  as a computation of  $\alpha$  on the input  $\vec{x}$ 

## Complexity measures for uniform processes

Suppose  $\alpha$  is an *n*-ary u.p. of **M**,  $\Phi_0 \subseteq \Phi$ , **M**  $\vdash \alpha(\vec{x}) = w$ , and  $\mu$  is a substructure norm in **M**. Set:

- calls<sub>Φ0</sub>(α, x) = min{|eqdiag(U↾Φ0)| : U ⊢<sub>c</sub> α(x) = w} (the least number of calls to φ ∈ Φ0 α must do to compute α<sup>M</sup>(x))
- size<sub>α</sub>(x) = min{|U| : U ⊢<sub>c</sub> α(x) = w} (the least number of elements of M that α must see)
- depth<sub>α</sub>(x) = min{depth<sub>x</sub>(U) : U ⊢<sub>c</sub> α(x) = w}
   (the least number of calls α must execute in sequence)

$$\mathsf{Thm} \, \left| \, \mathsf{depth}(\alpha, \vec{x}) \leq \mathsf{size}(\alpha, \vec{x}) \leq \mathsf{calls}(\alpha, \vec{x}) \right| \, (= \mathsf{calls}_{\Phi}(\alpha, \vec{x}))$$

These notions agree with standard definitions for concrete algorithms

### $\star$ The forcing and certification relations

Suppose  $f: M^n \to M_s$ ,  $f(\vec{x}) \downarrow$ ,  $\mathbf{U} \subseteq_p \mathbf{M}$ .

• A homomorphism  $\pi: \mathbf{U} \to \mathbf{M}$  respects f at  $\vec{x}$  if

$$\vec{x} \in U \& f(\vec{x}) \in U_s \& \pi(f(\vec{x})) = f(\pi(\vec{x}))$$

 $\mathbf{U} \Vdash^{\mathbf{M}} f(\vec{x}) = w \iff \text{every homomorphism } \pi : \mathbf{U} \to \mathbf{M} \text{ respects } f \text{ at } \vec{x}$  $\mathbf{U} \Vdash^{\mathbf{M}}_{c} f(\vec{x}) = w \iff \mathbf{U} \text{ is finite, generated by } \vec{x} \text{ and } \mathbf{U} \Vdash^{\mathbf{M}} f(\vec{x}) = w$ 

The intrinsic complexities of f in  $\mathbf{M}$ 

- $C_{\mu}(\mathbf{M}, f, \vec{x}) = \min\{\mu(\mathbf{U}, \vec{x}) : \mathbf{U} \Vdash_{c} f(\vec{x}) = w\}$
- ► calls<sub> $\Phi_0$ </sub>(**M**, *f*, *x*) = min{|eqdiag(**U** |  $\Phi_0$ )| : **U** || $_c^{\mathbf{M}} \alpha(\vec{x}) = w$ }
- ► size( $\mathbf{M}, f, \vec{x}$ ) = min{ $|U| : \mathbf{U} \Vdash_{c}^{\mathbf{M}} \alpha(\vec{x}) = w$ }
- depth( $\mathbf{M}, f, \vec{x}$ ) = min{depth<sub> $\vec{x}$ </sub>( $\mathbf{U}$ ) :  $\mathbf{U} \Vdash_{c}^{\mathbf{M}} \alpha(\vec{x}) = w$ }

# The best uniform process for $f: M^n \rightarrow M_s$ in **M** Define $\beta_{f,\mathbf{M}}$ by

$$\overline{eta}_{f,\mathsf{M}}^{\mathsf{U}}(ec{x}) = w \iff \mathsf{U} \Vdash^{\mathsf{M}} f(ec{x}) = w \quad (\mathsf{U} \subseteq_{
ho} \mathsf{M})$$

#### Theorem

The following are equivalent for a  $\Phi$ -structure **M** and  $f: M^n \rightarrow M_s$ :

(i) Some uniform process  $\alpha$  of **M** computes f. (ii)  $(\forall \vec{x}, w) (f(\vec{x}) = w \implies (\exists \mathbf{U} \subseteq_p \mathbf{M}) [\mathbf{U} \Vdash_c^{\mathbf{M}} f(\vec{x}) = w])$ . (iii)  $\beta_{f,\mathbf{M}}$  is a uniform process of **M** which computes f. Moreover, if these conditions hold, then for every uniform process  $\alpha$  which computes f in **M** and all complexity measures  $C_{\mu}$  as above,

$$C_{\mu}(\mathbf{M}, f, \vec{x}) = C_{\mu}(\beta_{f,\mathbf{M}}, \vec{x}) \le C_{\mu}(\alpha, \vec{x}) \qquad (f(\vec{x})\downarrow).$$

## The Homomorphism Test

#### Lemma

Suppose  $\mu$  is a substructure norm (calls $_{\Phi_0}$ , size, depth) on a  $\Phi$ -structure **M**,  $f : M^n \rightarrow M_s$ ,  $f(\vec{x}) \downarrow$ , and

for every finite 
$$\mathbf{U} \subseteq_{p} \mathbf{M}$$
 which is generated by  $\vec{x}$ ,  
 $(f(\vec{x}) \in U_{s} \& \mu(\mathbf{U}, \vec{x}) < m) \implies (\exists \pi : \mathbf{U} \to \mathbf{M})[f(\pi(\vec{x})) \neq \pi(f(\vec{x}))];$ 

then  $C_{\mu}(\mathbf{M}, f, \vec{x}) \geq m$ .

### A lower bound for coprimeness on $\ensuremath{\mathbb{N}}$

Let  $\mathbf{M} = (\mathbf{N}_{\varepsilon}, \Psi)$  with  $\Psi$  any finite set of *Presburger functions* Theorem (van den Dries, ynm, 2004, 2009) If  $\xi > 1$  is quadratic irrational, then for some r > 0 and all sufficiently large coprime (a, b),

$$\left|\xi - \frac{a}{b}\right| < \frac{1}{b^2} \implies \operatorname{depth}(\mathbf{M}, \mathbb{L}, a, b) \ge r \log \log a.$$
 (1)

In fact, the conclusion of (1) holds with some r

- for all positive solutions (a, b) of Pell's equation  $a^2 = 2b^2 + 1$ , and
- for all successive Fibonacci pairs  $(F_{k+1}, F_k)$ . with  $k \ge 3$ .

#### Theorem (Pratt, unpublished)

There is a non-deterministic algorithm  $\varepsilon_{nd}$  of  $\mathbf{N}_{\varepsilon}$  which decides coprimeness, is at least as effective as the Euclidean everywhere and

$$calls(\varepsilon_{nd}, F_{k+1}, F_k) \leq K \log \log F_{k+1}$$

▶ The theorem is best possible from its hypotheses

#### Non-uniform complexity

Given N, how good can a coprimeness algorithm be if we only insist that it works for n-bit numbers?

 $\mathbf{M} = (\mathbf{N}_{\varepsilon}, \mathbf{\Psi})$  as before. For any *N*, and any one of the intrinsic complexities as above, let

$$C_{\mu}(\mathbf{M}, f, N) = \max\{C_{\mu}(\mathbf{M} \upharpoonright [0, 2^N), f, a, b) : a, b < 2^N\}$$

#### Theorem (van den Dries, ynm 2009)

For some rational number r > 0 and all sufficiently large N,

$$\mathsf{calls}(\mathbf{M}, \bot, 2^N) \ge \mathsf{size}(\mathbf{M}, \bot, 2^N) \ge r \log N.$$

▶ Non-uniform lower bound for depth( $\mathbf{M}, \perp, 2^N$ )?

#### Horner's rule

For any field F and  $n \ge 1$ , the value of an n'th degree polynomial can be computed using no more than n multiplications and n additions in F:

$$a_0 + a_1x + a_2x^2 + \dots + a_nx^n = a_0 + x(a_1 + a_2x + \dots + a_nx^{n-1})$$

Divisions might help:

$$1 + x + x^{2} + \dots + x^{n} = \frac{x^{n+1} - 1}{x - 1}$$

#### Theorem (Pan 1966, (Winograd 1967, 1970))

Every straight line algorithm from the real field operations requires at least n multiplications/divisions and at least n additions/subtractions to compute  $a_0 + a_1x + a_2x^2 + \cdots + a_nx^n$ , when  $\vec{a}, x$  are algebraically independent real numbers The optimality of Horner's rule for polynomial 0-testing

The nullity relation on a field F:

$$N_F(a_0,\ldots,a_n,x) \iff a_0+a_1x+a_2x^2+\cdots+a_nx^n=0$$

#### Theorem

Let  $\mathbf{R} = (\mathbb{R}, 0, 1, +, -, \cdot, \div, =)$ . If  $n \ge 1$  and  $a_0, \ldots, a_n, x$  are algebraically independent, then:

(1) 
$$\operatorname{calls}_{\{\cdot, \div\}}(\mathbf{R}, N_{\mathbb{R}}, \vec{a}, x) = n$$
  
(2)  $\operatorname{calls}_{\{+, -\}}(\mathbf{R}, N_{\mathbb{R}}, \vec{a}, x) = n - 1$   
(3)  $\operatorname{calls}_{\{+, -, =\}}(\mathbf{R}, N_{\mathbb{R}}, \vec{a}, x) = n + 1$ 

For algebraic decision trees, (1) is due to Bürgisser and Lickteig (1992), and a result equivalent to (3) is due to Bürgisser, Lickteig and Shub (1992)

The lemma for calls $_{\{+,-\}}(\mathbf{R}, N_{\mathbb{R}}, \vec{a}, x) = n-1$ 

 $Roots(a_1, \dots, a_n) = \{a_i^{\frac{1}{m}} : m > 0, i = 1, \dots, n\} \qquad (a_1, \dots, a_n > 0)$ An operation  $u \circ v$  is trivial if  $u, v \in \mathbb{K}(x, z)$ 

#### Lemma

Suppose  $n \ge 2$ ,  $\overline{g} \in \mathbb{K}$  (= real algebraic numbers),  $\overline{g} \ne 0$ ,  $z, a_1, \ldots, a_n, x$  are positive, algebraically independent real numbers, and **U** is a finite substructure of **R** generated by

$$(U \cap \mathbb{K}) \cup \{x, z\} \cup (U \cap \operatorname{Roots}(a_1, \ldots, a_n))$$

which has < (n-1) non-trivial additions and subtractions. Then there is a field homomorphism  $\pi : \mathbb{K}(x, z, \text{Roots}(\vec{a})) \to \mathbb{K}(x, \text{Roots}(\vec{a}))$ such that