

# What is an algorithm?

Yiannis N. Moschovakis  
UCLA and University of Athens

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# 36,500,000 Google hits and no definition

- Wikipedia: *An algorithm is a step-by-step procedure for calculations*
- Knuth: *A computational method is ... (computation model)*  
*An algorithm is a computational method which terminates in finitely many steps for all [inputs]*
- Common: *An algorithm is a program* or *An algorithm or program ...*

This approach takes algorithms to be syntactic objects

*Frege: You cannot forbid the use of an arbitrarily produced process or object as a sign for something else*

- Common: *Algorithms are Turing machines*  
*or processes which can be simulated by Turing machines*

Turing machines do not express faithfully low complexity algorithms

van Emde Boas: simulation ... is hard to define as a mathematical object

# How can there be no definition?

when algorithms have been studied extensively, deeply and **rigorously** for (at least) eighty years?

- ▶ The Justice Potter Stewart argument: *I know one when I see it*  
For rigorous analysis, algorithms are specified precisely by a **computation model** or a **recursive procedure**
- ▶ In particular, **complexity theory** and (especially) the **derivation of lower bounds** with respect to various complexity measures, is always developed relative to fixed computation models
- ▶ Compare to: **probability theory**, which was developed intensively (and rigorously) for some 300 years before **random variables** were defined precisely in full generality (Kolmogorov 1933)

# Outline of the talk

- Three classical algorithms (4 slides)
- Least fixed point recursion (2 slides)
- Recursors, the set-theoretic objects which model algorithms (5 slides)
- Algorithms from specified primitives (2 slides)
- One application (if time permits) (2 slides)

Some references, all posted in [www.math.ucla.edu/~ynm](http://www.math.ucla.edu/~ynm)

- ▶ The formal language of recursion (1989)
- ▶ A mathematical modeling of pure, recursive algorithms
- ▶ On founding the theory of algorithms (1998)
- ▶ What is an algorithm? (2001)
- ▶ Is the Euclidean algorithm optimal among its peers?, with Lou van den Dries (2004)
- ▶ Elementary algorithms and their implementations, with Vassilis Paschalis (2008)

# The Euclidean algorithm

For  $x, y \in \mathbb{N}$ ,  $x \geq y \geq 1$ ,

(\*)  $\gcd(x, y) = \text{if } (\text{rem}(x, y) = 0) \text{ then } y \text{ else } \gcd(y, \text{rem}(x, y))$

where  $\text{rem}(x, y)$  is the remainder of the division of  $x$  by  $y$ ,

- (\*) expresses an algorithm  $\varepsilon$  from  $\text{rem}, =_0$  which computes  $\gcd(x, y)$

$$\begin{aligned} \text{calls}_{\varepsilon}^{\text{rem}}(x, y) &= \text{the number of calls to rem} \\ &\quad \text{required to compute } \gcd(x, y) \text{ by } \varepsilon \\ &\leq 2 \log(y) \quad (x \geq y \geq 2) \end{aligned}$$

**Conjecture** (open): For every algorithm  $\alpha$  which computes  $\gcd(x, y)$  ( $x, y \in \mathbb{N}, x \geq y > 0$ ) from  $\text{rem}$  and  $=_0$ ,

there is a sequence  $(x_n \geq y_n)_n$ , such that  $y_n \rightarrow \infty$

$$\text{and } \text{calls}_{\varepsilon}^{\text{rem}}(x_n, y_n) \leq \text{calls}_{\alpha}^{\text{rem}}(x_n, y_n)$$

- It assumes that “algorithm from  $\text{rem}, =_0$ ” and  $\text{calls}_{\alpha}^{\text{rem}}$  are defined

# The color of leaves

A (binary, colored) **forest** is a structure

$\mathbf{F} = (F, s, d, \text{Leaf}, \text{Red}, =)$  where  $\text{Leaf}, \text{Red} \subseteq F$  and  $s, d : F \rightarrow F$

A (maximal) *path from*  $x$  is any sequence  $p = (x_0, \dots)$  of length  $|p| \leq \infty$  such that

$$i < |p| \implies [\neg \text{Leaf}(x_i) \ \& \ x_{i+1} \in \{s(x_i), d(x_i)\}]$$

- $\mathbf{F}$  is **grounded** if it has no infinite paths. On grounded  $\mathbf{F}$  let

$$R(x) \iff \text{every path from } x \text{ ends in a red leaf}$$

$$(*) \quad R(x) \iff \text{if Leaf}(x) \text{ then Red}(x) \text{ else } [R(s(x)) \ \& \ R(d(x))]$$

- $(*)$  expresses a **recursive algorithm**  $\rho$  which decides  $R(x)$  on  $\mathbf{F}$ 
    - ▶ The Euclidean can be expressed by a **while program from**  $\text{rem}, =_0$
    - ▶ (Tiuryn 1989) *On some grounded forest, no algorithm expressed by a while program of  $\mathbf{F}$  decides  $R(x)$*
- ( $\mathbf{F}$  is the disjoint union of all finite, binary, colored trees)

# The sieve of Eratosthenes

Primes =  $p(u_0)$  where

$$\left\{ \begin{array}{l} u_0 = (2, 3, 4, 5, \dots), \\ p(u) = \text{Print}(\text{head}(u)) \wedge p(\text{sieve}(\text{head}(u), \text{tail}(u))), \\ \text{sieve}(x, v) = \text{if } (x \mid \text{head}(v)) \text{ then } \text{sieve}(x, \text{tail}(v)) \\ \qquad \qquad \qquad \text{else } \text{head}(v) \wedge \text{sieve}(x, \text{tail}(v)) \end{array} \right\}$$

$(S = (\mathbb{N} \rightarrow \mathbb{N}), u_0, u, v \in S, p : S \rightarrow S, x \in \mathbb{N}, \text{sieve} : \mathbb{N} \times S \rightarrow S)$

- ▶ A **system of mutual recursive equations** which expresses an algorithm  $\sigma$  from head, tail,  $\mid$ ,  $\wedge$  and (the act) Print
- ▶  $\text{sieve}(x, v)$  removes from  $v$  all numbers divisible by  $x$
- ▶  $p(u)$  prints  $\text{head}(u)$  and then calls itself on  $\text{sieve}(\text{head}(u), \text{tail}(u))$
- ▶  $\sigma$  computes successively  
 $u_0 = (2, 3, 4, \dots)$ ,  $u_1 = (3, 5, 7, \dots)$ ,  $u_2 = (5, 7, 11, \dots), \dots$   
and (as a **side effect**) “prints” the heads of these sequences
- ▶ Does the recursive system above “specify”  $\sigma$  completely?

# The basic, practical problem — too many notions!

- It seems like the basic notion should be that of **algorithm from** (given) primitives
- Too many notions associated with an algorithm: calls, recursive definitions, complexity functions, side effects (and **interaction**, which is more complex), simulation, implementability, . . .

For specific algorithms many of these are natural and simple, but a general theory might be excessively complex

- The lesson from probability theory: it is even more complex, but there is a useful and fairly simple basic notion:

*A **random variable** is a measurable function  $X : M \rightarrow \mathbb{R}$  on a sample space (a measure space of total measure 1)*

- *For algorithms, the background mathematical theory is **fixed point recursion on complete posets***



## Complete posets

- A **poset** is a pair  $(D, \leq_D)$  where  $\leq_D$  is a partial ordering of  $D$
- A poset  $D$  is (directed or chain) **complete** if every linearly ordered subset  $X \subseteq D$  (a **chain**) has a least upper bound  $\sup(X)$ . Every complete poset has a least element,  $\sup(\emptyset) = \perp$
- A set  $A$  is identified with the **flat** poset  $A_\perp = A \cup \{\perp\}$ , where
$$x \leq_{A_\perp} y \iff x = \perp \vee x = y$$
- The (naturally defined, cartesian) **product**  $D_1 \times \cdots \times D_n$  of complete posets is complete
- A function  $f : D \rightarrow E$  is **monotone** if

$$x \leq_D y \implies f(x) \leq_E f(y),$$

and **strict** if in addition  $f(x) \neq \perp \implies x$  is **total** (maximal) in  $D$

- $\text{Mon}(D, E)$ ,  $\text{Strict}(D, E)$  are the posets of monotone and strict functions ordered pointwise. They are complete, if  $D, E$  are complete

## Least fixed point recursion

- A function  $f : D \rightarrow E$  on complete posets is (Scott) **continuous** if
$$\sup_E \{f(x) \mid x \in D\} = f(\sup_D X) \quad (\text{for every chain } X \subseteq D)$$
- The poset  $\text{Cont}(D, E)$  of all continuous functions is complete and
$$\text{Strict}(D, E) \subseteq \text{Cont}(D, E) \subseteq \text{Mon}(D, E)$$

### Theorem (classical)

Every monotone function  $f : D \rightarrow D$  on a complete poset has a **least fixed point**  $\bar{d} = \min(d \in D)[f(d) = d]$ , characterized by

$$f(\bar{d}) = \bar{d}, \quad (\forall d)[f(d) \leq d \implies \bar{d} \leq d]$$

Moreover: if  $f : X \times D \rightarrow D$  is monotone, then the function

$$g(x) = \min(d \in D)[f(x, d) = d] \quad (x \in X)$$

is also monotone, and if  $f$  is continuous, then so is  $g$

## ★ (Monotone) recursors

- A **recursor**  $\alpha : X \rightsquigarrow W$  on one complete poset to another is a tuple

$$\alpha = (\alpha_0, \alpha_1, \dots, \alpha_k),$$

such that for suitable, complete posets  $D_1, \dots, D_k$ :

- (1) Each **part**  $\alpha_i : X \times D_1 \times \dots \times D_k \rightarrow D_i$ , ( $i = 1, \dots, k$ ) is monotone
  - (2) The **output part**  $\alpha_0 : X \times D_1 \times \dots \times D_k \rightarrow W$  is also monotone
- $\alpha_0$  is the **head** of  $\alpha$ ;  $(\alpha_1, \dots, \alpha_k)$  its **body**;  $D_\alpha = D_1 \times \dots \times D_k$  is its **solution poset**, and its **transition mapping**  $\tau_\alpha : X \times D_\alpha \rightarrow D_\alpha$  is

$$\tau_\alpha(x, d) = (\alpha_1(x, d), \dots, \alpha_n(x, d)) \quad (x \in X, d \in D_\alpha)$$

- The function  $\bar{\alpha} : X \rightarrow W$  **computed by**  $\alpha$  is

$$\bar{\alpha}(x) = \alpha_0(x, \bar{d}_x), \text{ where } \bar{d}_x = \min\{d \in D_\alpha \mid \tau_\alpha(x, d) = d\}$$

- We express all this succinctly by writing

$$\alpha(x) = \alpha_0(x, d) \text{ where } \{d = \tau_\alpha(x, d)\}, \quad (\text{recursor})$$

$$(\text{function}) \quad \bar{\alpha}(x) = \alpha_0(x, d) \text{ where } \overline{\{d = \tau_\alpha(x, d)\}}$$

# The importance of the solution poset

$\alpha(x) = \alpha_0(x, d)$  where  $\{d = \tau_\alpha(x, d)\}$ ,  $(x \in X, d \in D_\alpha = D_1 \times \dots \times D_k)$

- The Morris example (Manna 1975)

$$\boxed{p(s, t) = \text{if } (s = 0) \text{ then } 0 \text{ else } p(s - 1, p(s, t))} \quad (s, t \in \mathbb{N})$$

- The “official” associated recursor is

$$\alpha(s, t) = p(s, t)$$

where  $\{p = \lambda(s, t)(\text{if } (s = 0) \text{ then } 0 \text{ else } p(s - 1, p(s, t)))\}$

Solutions of the Morris recursive equation:

- If  $p$  varies over  $\text{Strict}(\mathbb{N}^2, \mathbb{N})$  (**call by value**),

$$\bar{p}(s, t) = \text{if } (s = 0) \text{ then } 0 \text{ else } \perp$$

- If  $p$  varies over  $\text{Cont}(\mathbb{N}^2, \mathbb{N})$  or  $\text{Mon}(\mathbb{N}^2, \mathbb{N})$  (**call by name**),  $\bar{p}(s, t) = 0$
- In the recursor representing the sieve of Eratosthenes we should use **streams** and **continuous function spaces** to insure “implementability”

## ★ Recursor isomorphism (identity)

Suppose  $\alpha, \beta : X \rightsquigarrow W$  are recursors

$$\alpha(x) = \alpha_0(x, d) \text{ where } \{d = \tau_\alpha(x, d)\}, \quad D = D_1 \times \cdots, D_k$$

$$\beta(x) = \beta_0(x, e) \text{ where } \{e = \tau_\beta(x, e)\}, \quad E = E_1 \times \cdots \times E_l$$

- *A recursor does not change if we replace its posets by isomorphic copies and permute the order of the parts in its body*

We say that  $\alpha$  is **naturally isomorphic** (equal) with  $\beta$ ,  $\alpha \cong \beta$ , if

- $k = l$ , i.e.,  $\alpha$  and  $\beta$  have the same number of parts
- There is a permutation  $\pi : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$  and for each  $i = 1, \dots, k$ , a poset isomorphism  $\rho_i : D_i \rightarrow E_{\pi(i)}$ , such that the induced isomorphism  $\rho_\pi : D \rightarrow E$  preserves the parts, i.e.,

$$\begin{aligned} \alpha_0(x, d) &= \beta_0(x, \rho_\pi(d)), \\ \rho_i(\alpha_i(x, d)) &= \beta_i(x, \rho_\pi(d)) \quad (i = 1, \dots, k) \end{aligned}$$

- Natural recursor isomorphism is a very fine notion—perhaps too fine

# Operations on recursors, I

- **Degenerate recursors.** Each function  $f : X \rightarrow W$  can be viewed as a degenerate recursor  $(f)$  with empty body,

$$\delta_f(x) = f(x) \text{ where } \{ \}$$

- **Composition of a recursor with a function.** For  $\beta : Y \rightsquigarrow W$  and  $g : X \rightarrow Y$  a monotone function, put

$$\alpha(x) = \beta(g(x)) = \beta_0(g(x), d) \text{ where } \{d = \tau_\beta(g(x), d)\};$$

then  $\boxed{\bar{\alpha}(x) = \bar{\beta}(g(x))}$

- **Recursor composition.** For  $\gamma : X \rightsquigarrow V$  and  $\beta : V \times Z \rightsquigarrow W$ , put

$$\begin{aligned} \alpha(x, z) &= \beta(\gamma(x), z) \\ &= \beta_0(v, z, d) \text{ where } \{v = \gamma_0(x, e), e = \tau_\gamma(x, e), d = \tau_\beta(v, z, d)\}; \end{aligned}$$

then  $\boxed{\bar{\alpha}(x, z) = \bar{\beta}(\bar{\gamma}(x), z)}$

- $\delta_f(g(x)) = f(g(x))$  where  $\{ \} \neq \delta_f(\delta_g(x)) = f(v)$  where  $\{v = g(x)\}$

## Operations on recursors, II

- **$\lambda$ -substitution**. For given  $\beta, \gamma$ , put

$$\alpha(x) = \beta(\lambda u \gamma(x, u)) = \beta_0(r, e) \text{ where } \{e = \tau_\beta(r, e), \\ r = \lambda u \gamma_0(x, u, d(u)), d = \lambda u \tau_\gamma(x, u, d(u))\};$$

then  $\boxed{\bar{\alpha}(x) = \bar{\beta}(\lambda u \bar{\gamma}(x, u))}$

- **Recursor recursion**. For given recursors  $\beta^0, \dots, \beta^k$ , put

$$\begin{aligned} \alpha(x) &= \beta^0(x, d) \text{ where } \{d_1 = \beta^1(x, d), \dots, d_k = \beta^k(x, d)\} \\ &= \beta_0^0(x, d, e^0) \text{ where } \{d_1 = \beta_0^1(x, d, e^1), \dots, d_k = \beta_0^k(x, d, e^k), \\ &\quad e^0 = \tau_{\beta^0}(x, d, e^0), \\ &\quad e^1 = \tau_{\beta^1}(x, d, e^1), \\ &\quad \vdots \\ &\quad e^k = \tau_{\beta^k}(x, d, e^k)\}; \end{aligned}$$

then  $\boxed{\bar{\alpha}(x) = \bar{\beta}^0(x, d) \text{ where } \{d_1 = \bar{\beta}^1(x, d), \dots, d_k = \bar{\beta}^k(x, d)\}}$

# Monotone and strict structures

- A **monotone structure** is a pair  $\mathbf{M} = (\mathcal{D}, \mathcal{F})$ , where
  - (1)  $\mathcal{D}$  is a set of complete posets which contains the (flat) **Boolean poset**  $\{\mathbb{t}, \text{ff}\}_\perp$
  - (2) Each  $f \in \mathcal{F}$  is a monotone function

$$f : D_1 \times \cdots \times D_n \rightarrow D \quad (D_1, \dots, D_k, D \in \mathcal{D})$$

- With each (first order) structure

$$\mathbf{A} = (A, \{\phi^{\mathbf{A}}\}_{\phi \in \Phi}) \quad (\Phi \text{ a set of function and relation symbols})$$

we associate the **strict monotone** (in fact **continuous**) **structure**

$$\mathbf{A}_s = (\{A_\perp, \{\mathbb{t}, \text{ff}\}_\perp\}, \{\phi_s^{\mathbf{A}}\}_{\phi \in \Phi}), \text{ where}$$

if  $\phi$  is a  $k$ -ary function symbol, then  $\phi_s^{\mathbf{A}} : A_\perp^k \rightarrow A_\perp$  is the strict extension of  $\phi^{\mathbf{A}}$ ,

and if  $\phi$  is a  $k$ -ary relation symbol, then  $\phi_s : A_\perp^k \rightarrow \{\mathbb{t}, \text{ff}\}_\perp$  is the strict extension of the *characteristic function* of  $\phi^{\mathbf{A}}$

- We do not assume that  $=_A$  is one of the primitives of  $\mathbf{A}$



## ★ Algorithms from specified primitives

*Slogan: an algorithm of  $\mathbf{M}$  is a recursor which is explicitly definable from the primitives of  $\mathbf{M}$*

- An **algorithm of** a monotone structure  $\mathbf{M} = (\mathcal{D}, \mathcal{F})$  (or **from**  $\mathcal{F}$ ) is a recursor which belongs to every collection of recursors  $\mathcal{R}$  with the following properties:

(A1)  $\mathcal{R}$  contains all the (degenerate) recursors  $\delta_f$  with  $f \in \mathcal{F}$

(A2)  $\mathcal{R}$  contains  $\delta_f$  for every “relevant” **call** or **conditional** function

$\text{ev}^{D,W}(p, x) = p(x) \quad (p : D_1 \times \cdots \times D_n \rightarrow E \text{ with } D_i, W \in \mathcal{D})$

$\text{cond}^W(r, x, y) = \text{if } r \text{ then } x \text{ else } y \quad (r \in \{\mathbf{tt}, \mathbf{ff}\}_\perp, x, y \in W \in \mathcal{D})$

(A3)  $\mathcal{R}$  is closed under **composition with the functions**  $\text{ev}^{D,W}$  and every **projection**  $\pi_i(x_1, \dots, x_n) = x_i \quad (1 \leq i \leq n)$

(A4)  $\mathcal{R}$  is closed under **recursor composition**,  **$\lambda$ -substitution** and **recursor recursion**

- For a first order structure  $\mathbf{A}$ , **the algorithms of  $\mathbf{A}$**  (i.e.,  $\mathbf{A}_s$ ) compute the (call-by-value, McCarthy)  **$\mathbf{A}$ -recursive partial functions** on  $\mathbf{A}$

The most important thing missing is an account of **implementations** and the connection between an implementable algorithm and its implementations. A very little of this has been worked out.

## Intrinsic complexities in a structure $\mathbf{A} = (A, \{\phi^{\mathbf{A}}\}_{\phi \in \Phi})$

- **Calls complexity.** For each algorithm  $\alpha : A^n \rightsquigarrow \{\text{tt}, \text{ff}\}_\perp$  of  $\mathbf{A}$  and each  $\Phi_0 \subseteq \Phi$ , we can define

$\text{calls}_\alpha^{\Phi_0}(x)$  = the number of calls to primitives  $\phi^{\mathbf{A}}$  with  $\phi \in \Phi_0$   
made by  $\alpha$  in the computation of  $\bar{\alpha}(x)$  ( $\bar{\alpha}(x) \downarrow$ )

This agrees with the usual calls-complexity for “concrete algorithms”

### Theorem

*With each relation  $R \subseteq A^n$  which is decidable by some algorithm of  $\mathbf{A}$  and each  $\Phi_0 \subseteq \Phi$ , there is a function  $\text{calls}_R^{\Phi_0} : A^n \rightarrow \mathbb{N}$  such that for every algorithm  $\alpha$  of  $\mathbf{A}$  which decides  $R$ ,*

$$\text{calls}_R^{\Phi_0}(\vec{x}) \leq \text{calls}_\alpha^{\Phi_0}(\vec{x}) \quad (\vec{x} \in A^N)$$

- The **intrinsic calls-complexity**  $\text{calls}_R^{\Phi_0}$  is usually not trivial (next page)
- There are similar results for a large variety of complexity measures which can be defined for  $\mathbf{A}$ -algorithms

## An example

- **Coprimeness**,  $x \perp\!\!\!\perp y \iff \gcd(x, y) = 1$

### Theorem (van den Dries, ynm)

Suppose  $\mathbf{A} = (\mathbb{N}, \{\phi^{\mathbf{A}}\}_{\phi \in \Phi})$  is a structure on  $\mathbb{N}$  whose primitives are **Presburger** (piecewise linear) functions and relations. There is a rational  $r > 0$ , such that

$$\text{calls}_{\perp\!\!\!\perp}^{\Phi} (a, a^2 - 1) \geq r \log(a) \quad (a > 2)$$

In particular, the **binary** (Stein) algorithm is “suboptimal” up to a multiplicative constant for deciding coprimeness from Presburger primitives.