

A Theory of Complexity, Condition and Roundoff

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Background

L. Blum, M. Shub, and S. Smale [1989]

*On a theory of computation and complexity over the real numbers:
NP-completeness, recursive functions and universal machines.*

$P_{\mathbb{R}}$ $NP_{\mathbb{R}}$

“Finally, to bring machines over \mathbb{R} closer to the subject of numerical analysis, it would be useful to incorporate round-off error, condition numbers and approximate solutions into our development.”

Finite-precision computations

Floating-point number system: $\mathbb{F} \subset \mathbb{R}$.

$$y = \pm 0.b_1 b_2 \dots b_t \times 2^e \quad |e| \leq e_{\max}$$

$$\text{Range}(\mathbb{F}) := [-2^{e_{\max}}(1 - 2^{-t}), -2^{-e_{\max}-1}] \cup \{0\} \cup [2^{-e_{\max}-1}, 2^{e_{\max}}(1 - 2^{-t})].$$

Unit roundoff: $u_{\text{mach}} := 2^{-t}$.

Rounding function: $\text{fl} : \text{Range}(\mathbb{F}) \rightarrow \mathbb{F}$

for all $x \in \text{Range}(\mathbb{F})$, $\text{fl}(x) = x(1 + \delta)$ for some δ with $|\delta| < u_{\text{mach}}$.

Floating-point arithmetic: to $\circ \in \{+, -, \times, /\}$ we associate

$$\tilde{\circ} : \mathbb{F} \times \mathbb{F} \rightarrow \mathbb{F}$$

$$x \tilde{\circ} y = (x \circ y)(1 + \delta) \text{ for some } \delta \text{ with } |\delta| < u_{\text{mach}}.$$

Unrestricted exponents: $e_{\max} = \infty$ ($\text{Range}(\mathbb{F}) = \mathbb{R}$).

Most analyses in the literature assume unrestricted exponents.

Stability and condition

Two factors in the accumulation of errors in a computation:

(1) How sensitive is the result of the computed function φ to perturbations of the data d ?

Condition number $\text{cond}^\varphi(d)$

it depends only on φ and d

(2) How badly does the algorithm at hand accumulate errors?

Stability analysis

it depends on the algorithm and the dimension of d

Example Linear equation solving: $(A, b) \mapsto x = A^{-1}b$. Under the assumption of unrestricted exponents, we have

$$\text{cond}^\varphi(A) = \kappa(A) := \|A\| \|A^{-1}\|.$$

The computed (using Householder QR decomposition) solution \tilde{x} satisfies, for some constant C ,

$$\frac{\|\tilde{x} - x\|}{\|x\|} \leq Cn^3 \mathbf{u}_{\text{mach}} \kappa(A) + o(\mathbf{u}_{\text{mach}}). \quad (1)$$

Remark Hestenes and Stiefel showed that $\kappa(A)$ also plays a role in complexity analyses.

Important remark: A wide variety of computational problems:

- decisional
- functional
- set-valued

...

results in a variety of condition numbers.

Condition numbers are defined “ad hoc”.

The Theory

Decision problems

Data has discrete and continuous components:

$$\mathcal{I} := \{0, 1\}^\infty \times \mathbb{R}^\infty.$$

Here

$$\mathbb{R}^\infty := \bigsqcup_{i=0}^{\infty} \mathbb{R}^i \qquad \{0, 1\}^\infty := \bigsqcup_{i=0}^{\infty} \{0, 1\}^i.$$

Definition A *decision problem* is a pair (A, μ) where $A \subset \mathcal{I}$ and $\mu : \mathcal{I} \rightarrow [1, \infty]$. Here μ is the *condition number*.

We denote by Σ the set $\{(u, x) \in \mathcal{I} \mid \mu(u, x) = \infty\}$ and we say that elements in Σ are *ill-posed*.

Remark Different condition numbers for the same subset $A \subset \mathcal{I}$ define different decision problems. This is akin to the situation in classical (i.e., both discrete and infinite-precision BSS) complexity theory where different encodings of the intended input data define (sometimes radically) different problems.

Finite-precision machines, input size, and cost

Definition A *finite-precision BSS machine* is a BSS machine performing finite-precision computations. To define the latter, we fix a number $\mathbf{u}_{\text{mach}} \in (0, 1)$ (the *unit roundoff*) and let

$$\mathbf{k}_{\text{mach}} := \left\lceil \log_2 \frac{1}{\mathbf{u}_{\text{mach}}} \right\rceil.$$

In a \mathbf{u}_{mach} -*computation*, built-in constants, input values, and the result of arithmetic operations, call any such number z , are systematically replaced by $\text{fl}(z)$ satisfying

$$\text{fl}(z) = z(1 + \delta) \quad \text{for some } |\delta| < \mathbf{u}_{\text{mach}}. \quad (2)$$

We will refer to $\mathbf{k}_{\text{mach}} \in \mathbb{N}$ as the *precision* of M .

Complexity = dependence of cost on size.

For $(u, x) \in \{0, 1\}^s \times \mathbb{R}^n \subset \mathcal{I}$, we let $\text{length}(u, x)$ to be $s + n$ and

$$\text{size}(u, x) := \text{length}(u, x) + \lceil \log_2 \mu(u, x) \rceil.$$

Note that if (u, x) is ill-posed then $\text{size}(u, x) = \infty$ and that otherwise $\text{size}(u, x) \in \mathbb{N}$.

Arithmetic cost: number of steps performed before halting. We denote it by $\text{ar_cost}_M(u, x)$.

Accuracy cost: smallest value of k_{mach} guaranteeing a correct answer.

Close to the cost in practice of operating with floating-point numbers since, assuming the exponents of such numbers are moderately bounded, this cost is at most quadratic on k_{mach} .

Clocked computations

Definition Let $\text{Arith} : \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$ and $\text{Prec} : \mathbb{N} \rightarrow \mathbb{N}$. A decision problem (S, μ) is *solved with cost* $(\text{Arith}, \text{Prec})$ when there exists a machine M satisfying the following. For every $(u, x) \in \mathcal{I}$ with $\mu(u, x) < \infty$ the computation of M with input (u, x) satisfies

$$\text{ar_cost}_M(u, x) \leq \text{Arith}(\text{length}(u, x), k_{\text{mach}}).$$

Furthermore, if

$$k_{\text{mach}} \geq \text{Prec}(\text{size}(u, x))$$

then all computations of M correctly decide whether $(u, x) \in S$.

- (i) Computations are **clocked**, i.e., their arithmetic cost is bounded by a function on two parameters immediately available: length of the input data and machine precision.
- (ii) Computations are **unreliable**: there is no guarantee that the precision used is enough to ensure a correct output. Even for exact computations correctness is not guaranteed.

A hierarchy theorem

Proposition (Precision Hierarchy Theorem) *Let $T : \mathbb{N} \rightarrow \mathbb{N}$ be time constructible and $P_1, P_2 : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that P_2 is continuous and increasing and $P_1 < \frac{P_2}{2}$. There exists a decision problem (B, μ) which can be decided with $\text{ar_cost}(u, x) \leq \mathcal{O}(T(\text{length}(u, x)))$ and $k_{\text{mach}} = P_2(\text{size}(u, x)) + 3$, but cannot be decided with $k_{\text{mach}} = P_1(\text{size}(u, x))$ (no matter the arithmetic cost).*

General polynomial time: the class P_{ro}

Definition A decision problem (S, μ) belongs to P_{ro} (*roundoff polynomial cost*) when there exists a finite-precision BSS machine M solving S with cost $(\text{Arith}, \text{Prec})$ and such that

- (i) Prec is bounded by a polynomial function, and
- (ii) the function $\text{Arith}(\text{length}(u, x), \text{Prec}(\text{size}(u, x)))$ is bounded by a polynomial in $\text{size}(u, x)$, for all $(u, x) \in \mathcal{I}$.

Direct algorithms: the class P_{dir}

Definition A decision problem (S, μ) belongs to P_{dir} (*direct polynomial cost*) when there exists a machine M satisfying the following. For every $(u, x) \in \mathcal{I}$ the computation of M with input (u, x) satisfies

$$\text{ar_cost}_M(u, x) \leq (\text{length}(u, x))^{\mathcal{O}(1)}.$$

Furthermore, if

$$k_{\text{mach}} \geq (\text{size}(u, x))^{\mathcal{O}(1)}$$

then all computations of M correctly decide whether $(u, x) \in S$. If correctness is ensured as soon as $k_{\text{mach}} \geq (\log \text{size}(u, x))^{\mathcal{O}(1)}$ we say that (S, μ) can be solved with *logarithmic precision*.

Examples. Deciding whether $\det(A) > 0$, whether S is p.s.d., ...

Proposition We have $P_{\text{dir}} \subsetneq P_{\text{ro}}$.

Notation:

\mathcal{C} an algebraic circuit (n input gates, 1 output gate)

$f_{\mathcal{C}} : \mathbb{R}^n \rightarrow \mathbb{R}$ function computed by the circuit

$$S_{\mathcal{C}} := \{x \in \mathbb{R}^n \mid f_{\mathcal{C}}(x) \geq 0\}.$$

Example Instances for CircEval are algebraic circuits \mathcal{C} together with a point $x \in \mathbb{R}^n$. The problem is to decide whether $x \in S_{\mathcal{C}}$. To specify a condition number we first define

$$\rho_{\text{eval}}(\mathcal{C}, x) := \begin{cases} \sup\{\varepsilon < 1 \mid \text{all } \varepsilon\text{-evaluations of } \mathcal{C} \text{ at } x \text{ yield } x \in S_{\mathcal{C}}\} & \text{if } x \in S_{\mathcal{C}} \\ \sup\{\varepsilon < 1 \mid \text{all } \varepsilon\text{-evaluations of } \mathcal{C} \text{ at } x \text{ yield } x \notin S_{\mathcal{C}}\} & \text{otherwise.} \end{cases}$$

We then take as condition number

$$\mu_{\text{eval}}(\mathcal{C}, x) := \frac{1}{\rho_{\text{eval}}(\mathcal{C}, x)}.$$

Nondeterministic Polynomial Cost

Problems in (all versions of) NP are sets S for which membership of an element x to S can be established through a “short” proof y .

$NP, NP_{\mathbb{R}}$ short = small length

NP_{ro}^U short = small length + small condition

NP_{ro}^B short = small length + small condition + small magnitude

The class $\text{NP}_{\text{ro}}^{\text{U}}$

Definition A decision problem (W, μ_W) belongs to $\text{NP}_{\text{ro}}^{\text{U}}$ (*non-deterministic roundoff polynomial cost*) when there exist a decision problem (B, μ_B) , a machine M deciding (B, μ_B) in P_{ro} , and polynomials p, Q , such that for $(u, x) \in \mathcal{I}$,

- (i) if $(u, x) \in W$ then there exists $y^* \in \mathbb{R}^m$, such that $(u, x, y^*) \in B$, and $\log \mu_B(u, x, y^*) \leq Q(\log \mu_W(u, x))$, and
- (ii) if $(u, x) \notin W$ then, for all $y \in \mathbb{R}^m$ we have $(u, x, y) \notin B$ and $\log \mu_B(u, x, y) \leq Q(\log \mu_W(u, x))$.

Here $m = p(\text{length}(u, x))$.

Example Instances for **CircFeas** are algebraic circuits \mathcal{C} (with input variables Y_1, \dots, Y_m). The problem is to decide whether there exists $y \in \mathbb{R}^m$ such that $y \in S_{\mathcal{C}}$ (in which case, we say that \mathcal{C} is *feasible*). We take as condition number

$$\mu_{\text{feas}}(\mathcal{C}) := \frac{1}{\varrho_{\text{feas}}(\mathcal{C})}$$

where

$$\varrho_{\text{feas}}(\mathcal{C}) := \begin{cases} \sup_{y \in S_{\mathcal{C}}} \varrho_{\text{eval}}(\mathcal{C}, y) & \text{if } \mathcal{C} \text{ is feasible,} \\ \inf_{y \in \mathbb{R}^m} \varrho_{\text{eval}}(\mathcal{C}, y) & \text{otherwise.} \end{cases}$$

Note that in the feasible case, $\mu_{\text{feas}}(\mathcal{C})$ is the condition of its best conditioned solution, and in the infeasible case, it is the condition of the worst conditioned point in \mathbb{R}^m .

Proposition $\text{CircFeas} \in \text{NP}_{\text{ro}}^{\text{U}}$.

Proposition $\text{P}_{\text{ro}} \subset \text{NP}_{\text{ro}}^{\text{U}}$.

Definition A P_{ro} -reduction from (W, μ_W) to (S, μ_S) is a machine \overline{M} which, given a point $(u, x) \in \mathcal{I}$ and a number $k \in \mathbb{N}$, performs a discrete computation and returns a pair $(v, z) \in \mathcal{I}$ with $\text{ar_cost}_{\overline{M}}(u, x)$ polynomially bounded on $\text{length}(u, x)$ and k .

In addition, we require the existence of some $D, p > 0$ such that for all $k \geq D \text{size}(u, x)^p$ one has

- (i) $(u, x) \in W \iff (v, z) \in S$, and
- (ii) $\log \mu_S(v, z)$ is polynomially bounded in $\text{size}_W(u, x)$.

If all of the above holds, we write $(W, \mu_W) \preceq_{ro} (S, \mu_S)$.

Proposition If $(W, \mu_W) \preceq_{\text{ro}} (S, \mu_S)$ and $(S, \mu_S) \in P_{\text{ro}}$ then $(W, \mu_W) \in P_{\text{ro}}$.

Definition A problem (S, μ_S) is $\text{NP}_{\text{ro}}^{\text{U}}$ -hard when for any problem $(W, \mu_W) \in \text{NP}_{\text{ro}}^{\text{U}}$ we have $(W, \mu_W) \preceq_{\text{ro}} (S, \mu_S)$.

It is $\text{NP}_{\text{ro}}^{\text{U}}$ -complete when it is $\text{NP}_{\text{ro}}^{\text{U}}$ -hard and belongs to $\text{NP}_{\text{ro}}^{\text{U}}$.

Theorem CircFeas is $\text{NP}_{\text{ro}}^{\text{U}}$ -complete.

Corollary $P_{\text{ro}} = \text{NP}_{\text{ro}}^{\text{U}} \iff \text{CircFeas} \in P_{\text{ro}}$.

Open Question Does one have $P_{\text{ro}} = \text{NP}_{\text{ro}}^{\text{U}}$? As usual, we believe this is not the case.

The class $\text{NP}_{\text{ro}}^{\text{B}}$

Given $k \in \mathbb{N}$ we consider the floating-point system F_k with

$$t = k, \quad \text{and} \quad e_{\max} = 2^k - 1.$$

For $x \in \mathbb{R}$ we define the *magnitude* of x to be

$$\text{mgt}(x) := \min\{k \geq 1 \mid x \in \text{Range}(F_k)\},$$

and for $x \in \mathbb{R}^n$, $\text{mgt}(x) := \max_{i \leq n} \text{mgt}(x_i)$.

Definition A decision problem (W, μ_W) belongs to $\text{NP}_{\text{ro}}^{\text{B}}$ (*bounded non-deterministic roundoff polynomial cost*) when there exist a decision problem (B, μ_B) , a machine M deciding (B, μ_B) in P_{ro} , and polynomials p, q, Q , such that for $(u, x) \in \mathcal{I}$,

(i) if $(u, x) \in W$ then there exists $y^* \in \mathbb{R}^m$, such that $(u, x, y^*) \in B$, $\log \mu_B(u, x, y^*) \leq Q(\log \mu_W(u, x))$, and $\text{mgt}(y^*) \leq q(\text{size}_W(u, x))$, and

(ii) if $(u, x) \notin W$ then, for all $y \in \mathbb{R}^m$ we have $(u, x, y) \notin B$ and $\log \mu_B(u, x, y) \leq Q(\log \mu_W(u, x))$.

Here $m = p(\text{length}(u, x))$.

Example Instances for **CircBFeas** are algebraic circuits \mathcal{C} (with input variables Y_1, \dots, Y_m). The problem is to decide whether there exists $y \in \mathbb{R}^m$ such that $y \in S_{\mathcal{C}}$. What makes this problem different from **CircFeas** is its condition number. Here we take

$$\mu_{\text{Bfeas}}(\mathcal{C}) := \frac{1}{\varrho_{\text{Bfeas}}(\mathcal{C})}$$

where

$$\varrho_{\text{Bfeas}}(\mathcal{C}) := \begin{cases} \sup_{y \in S_{\mathcal{C}}} \varrho_{\text{eval}}(\mathcal{C}, y) 2^{-\text{mgt}(y)} & \text{if } \mathcal{C} \text{ is feasible,} \\ \inf_{y \in \mathbb{R}^m} \varrho_{\text{eval}}(\mathcal{C}, y) & \text{otherwise.} \end{cases}$$

Theorem **CircBFeas** is $\text{NP}_{\text{ro}}^{\text{B}}$ -complete.

Corollary $\text{P}_{\text{ro}} = \text{NP}_{\text{ro}}^{\text{B}} \iff \text{CircBFeas} \in \text{P}_{\text{ro}}$.

Exponential cost

Definition A decision problem (S, μ) belongs to EXP_{ro} (*roundoff exponential cost*) when there exists a finite-precision BSS machine M deciding S with cost $(\text{Arith}, \text{Prec})$ and such that

- (i) Prec is bounded by an exponential function, and
- (ii) the function $\text{Arith}(\text{length}(u, x), \text{Prec}(\text{size}(u, x)))$ is bounded by an exponential in $\text{size}(u, x)$, for all $(u, x) \in \mathcal{I}$.

In both (i) and (ii) by exponential we understand a function of the kind $n \mapsto a^{n^d}$ for some $a > 1$ and $d > 0$.

Theorem $\text{CircBFeas} \in \text{EXP}_{ro}$.

Corollary $\text{NP}_{ro}^B \subset \text{EXP}_{ro}$ *and the inclusion is strict.*

Open Question A major open question in this is whether NP_{ro}^U is included in EXP_{ro} or, equivalently, whether CircFeas belongs to EXP_{ro} . We conjecture that this question has a positive answer.

