

Distributional proving problems

Edward A. Hirsch

(based on results of
E.A.H., D. Itsykson, I. Monakhov, V. Nikolaenko, A. Smal, D. Sokolov)

Steklov Institute of Mathematics at St.Petersburg, RAS
St.Petersburg Academic University, RAS

Motivation: Optimal algorithms

- ▶ Given a problem, even if there is no fast algorithm, what is the best one?
- ▶ Levin's optimal algorithm for **NP search** problems is known since 1973:
 - ▶ Run all algorithms A_1, A_2, \dots "in parallel".
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- ▶ ...and for any decision problem in $\mathbf{co-NP} \setminus \mathbf{P}$.
Optimal algorithms for **TAUT** are tightly related to optimal proof systems
- ▶ The best we can do is $\mathbf{E} \setminus \mathbf{P}$, for immune sets [Messner; Chen, Flum, Müller].

Distributional proving problems

Distributional proving problem (D, L) consists of

- ▶ a language L of “theorems”,
- ▶ a polynomial-time samplable distribution $D = \{D_n\}_{n \in \mathbb{N}}$ on \bar{L} .

Motivation:

- ▶ a small (wrt D) amount of wrong theorems is acceptable;
- ▶ not interested in what happens on statements that are not claimed;
- ▶ polynomial-time samplable distributions are concentrated on **NP** languages, thus the definition is natural for $L \in \mathbf{co-NP}$.

Distributional problems

- ▶ A distribution on all inputs.
- ▶ Gives no information about the problem.

Distributional proving problems

- ▶ A distribution on negative instances.
- ▶ Allows to verify an algorithm on counterexamples.
- ▶ There are natural polynomial-time samplable distributions on all negative instances (e.g., planted **SAT**).

PAC learning

- ▶ A distribution providing correct answers.
- ▶ Allows to verify an algorithm on all samples.
- ▶ Polynomial-time samplable distributions on all inputs are unlikely to exist for **NP**-complete problems.

Heuristic acceptors

Definition

(Classical) acceptor A for L :

(completeness) A accepts every $x \in L$.

(correctness) A does not stop on any $x \notin L$.

Complexity parameter: running time on L .

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Heuristic acceptor $A(x, d)$ for (D, L) : (d is the desired “confidence”)

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- ▶ Time $\tau_A(x, d)$ is a **random variable**.
- ▶ $t_A(x, d)$ is the **median** (w.r.t. random bits) running time of $A(x, d)$.
- ▶ Polynomial time \sim polynomial in $|x|$ and d .

Heuristic acceptors

Are there hard problems?

Theorem

\exists polynomial-time samplable $D \exists L \in \mathbf{co-NP} \nexists$ polynomial-time heuristic acceptor for $(D, L) \iff \exists$ infinitely-often one-way function.

Proof.

- ▶ i.o. o.w.f. \Rightarrow i.o. PRG
(similar to [Håstad, Impagliazzo, Levin, Luby])
- ▶ i.o. PRG \Rightarrow hard problem for heuristic acceptors
(hint: PRG is a distribution)
- ▶ hard problem for heuristic acceptors \Rightarrow average-case o.w.f.
(hint: the sampler is difficult to invert)
- ▶ average-case o.w.f. \Rightarrow i.o. o.w.f.
(padding)

Optimal heuristic acceptor

Definition

(Classical) acceptor S **simulates** W if it runs almost as fast for each x , i.e., there is a polynomial p such that $\forall x \in L, \quad t_S(x) \leq p(t_W(x) \cdot |x|)$.

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Idea: Certify A_i by testing it on samples $x \leftarrow D_n$.

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 2. If it accepts (in T_i steps), test its correctness:
let $E_i = 0$ and execute k times:
 - ▶ $r \leftarrow D_{|x|}$,
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Let $d' = 4d|x|$, $k = 2d^3|x|^3$, $\delta = \frac{1}{2d|x|}$.

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(completeness) There is a proof accepted whp:

$$\forall x \in L \forall d \in \mathbb{N} \exists w \quad \Pr\{\Pi(x, w, d) = 1\} > \frac{1}{2}.$$

(Such w is a Π -proof with confidence d .)

(correctness) Most non-theorems don't have such proofs:

$$\Pr_{r \leftarrow D_n} \{\exists w \{\Pr\{\Pi(r, w, d) = 1\} > \frac{1}{8}\}\} < \frac{1}{d}.$$

Turning **AM** protocols into heuristic proof systems

- ▶ Assume $L \in \mathbf{AM}$.
(E.g., $L = \mathbf{GNI}$, D samples random isomorphic graphs.)
- ▶ Consider a protocol (A, M) for L
(w.l.o.g., with perfect completeness and exponentially small error):

$$x \in L \implies \forall r \exists w A(x, w, r) = 1,$$

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- ▶ Consider $L' = \{(x, r) \mid x \in L\}$, where the length of r is enough to make the public random choices.
- ▶ Consider $D' = D \times U$, where D is any “original” distribution on \bar{L} and U is the uniform distribution.

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Theorem (Itsykson, Sokolov)

(1) (L', D') has a polynomially bounded heuristic p.s.

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- ▶ Proof: Simulate A (first round) using r .

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Implies randomized heuristic algorithm for (L, D) .

Theorem (Itsykson, Sokolov)

(1) (L', D') has a polynomially bounded heuristic p.s.

(2) if $L' \in \mathbf{NP}$, then $L \in \mathbf{NP}$.

(3) if (L', D') has polynomial-time heuristic acceptor, then (L, D) does. 10 / 15

Turning AM protocols into heuristic proof systems

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- (2) if $\forall L L' \in \mathbf{NP}$, then $\mathbf{NP} = \mathbf{co-NP}$.
- (3) if $\forall L, D (L', D')$ has poly-time h.acc., then $(\mathbf{NP}, \mathbf{PSamp})$ does.

Optimal proof systems

Classical case

- ▶ A proof system Σ **simulates** a proof system Ω iff Σ -proofs are at most as long as Ω -proofs (up to a polynomial p):
$$\forall F \in L \quad |\text{shortest } \Sigma\text{-proof of } F| \leq p(|\text{shortest } \Omega\text{-proof of } F|, |F|).$$
- ▶ **p -simulation** is a constructive version: For any w -size Ω -proof, one can compute a $p(w)$ -size Σ -proof in polynomial time.
- ▶ **(p) -optimal** proof system (p) -simulates any other proof system.
- ▶ **Does it exist?..**

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Theorem

\exists **p -optimal proof system** $\iff \exists$ **optimal acceptor**.

For **TAUT**: [Krajíček, Pudlák].

For paddable languages: [Messner].

For **co-NP**-complete languages: [Chen, Flüm, Müller].

Definition

L is **paddable** if there is an injective non-length-decreasing polynomial-time padding function $\text{pad}_L: \{0, 1\}^* \times \{0, 1\}^* \rightarrow \{0, 1\}^*$ that is polynomial-time invertible on its image and such that $\forall x, w (x \in L \iff \text{pad}_L(x, w) \in L)$.

Optimal proof [Messner, 99]:

- ▶ A proof π of x in some system Π ;
- ▶ padding.

Verification:

- ▶ run optimal acceptor on $\text{pad}_L(x, \pi)$;
- ▶ for a correct proof π , it accepts in a polynomial time because for a correct system Π , the set $\{\text{pad}_L(x, \pi) \mid x \in L, \Pi(x, \pi) = 1\} \subseteq L$ can be accepted in a polynomial time.

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- ▶ Does not go for heuristic, average-case algorithms.

Simulations

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\exists polynomial $p \forall x$

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$\forall \epsilon > 0 \exists c > 0$

$$\mathbf{E}_{x \leftarrow D_n} [t_{\mathcal{A}}^c(x)] = O(n \mathbf{E}_{y \leftarrow D_n} [t_{\mathcal{B}}^\epsilon(y)])$$

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- ▶ (weaker) simulation **scheme**:

simulate everywhere except for the set of D -prob. $1/2d$.

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pointwise-optimal algorithm for bi-immune sets [Chen, Flum, Müller].

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- ▶ **Distributional proving** problem (D, L) : $\text{supp } D \subseteq \bar{L}$.
Solved by heuristic acceptors, may allow false positives only.
 - ▶ pointwise optimal randomized heuristic acceptor for p.-t.s. D , r.e. L

Open questions

- ▶ \exists optimal proof system \iff \exists optimal heuristic acceptor;
- ▶ \exists optimal heuristic proof system $\stackrel{?}{\iff}$ \exists optimal heuristic acceptor;
- ▶ \exists optimal proof system with advice $\stackrel{?}{\iff}$ \exists optimal acceptor with advice;
- ▶ \exists average-case optimal acceptor?
- ▶ \exists optimal acceptor for GNI or any other $\text{co-NP} \setminus \mathbf{P}$ problem?
- ▶ \exists optimal proof system for any problem outside \mathbf{P} ?
- ▶ $\exists (D, L) \in (\text{co-NP}, \text{PSamplable})$ with no polynomially-bounded heuristic proof system \iff ?
- ▶ **AM** protocols make deterministic (heuristic) proof systems with very small error; suggest another example: randomized and with larger error.