
Heuristic time hierarchies via hierarchies for sampling distributions

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First steps

HARTMANIS AND STEARNS, 1965

For any $k > 0$ we have that

$$\mathbf{P} \not\subseteq \mathbf{DTime}(n^k).$$

COOK, 1973; ZAK, 1983

For any $k > 0$ holds

$$\mathbf{NP} \not\subseteq \mathbf{NTime}(n^k).$$

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Probabilistic algorithms

BOUNDED PROBABILISTIC ALGORITHMS

Language $L \in \mathbf{BPTIME}(n^k)$ iff there is randomized $O(n^k)$ -time algorithm A such that

$$\forall x \in \{0, 1\}^* \Pr[A(x) = L(x)] > \frac{3}{4}.$$

We also denote $\mathbf{BPP} = \bigcup_k \mathbf{BPTIME}(n^k)$.

OPEN QUESTION

Is it true that for any $k > 0$ holds that

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Derandomization

FOLKLORE

If there is pseudorandom generator that maps $\log(n)$ bits to $\text{poly}(n)$ then $\mathbf{BPP} \not\subseteq \mathbf{BPTIME}(n^k)$.

ITSYKSON, KNOP, SOKOLOV, 2015

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Deterministic algorithms

HEURISTIC DETERMINISTIC ALGORITHM

Language $L \in \text{Heur}_\delta \mathbf{DTime}(n^k)$ iff there is $O(n^k)$ -time algorithm A such that

$$\forall n \in \mathbb{N} \quad \Pr_{x \in \{0,1\}^n} [A(x) = L(x)] > 1 - \delta.$$

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For every $k > 0$ and $\epsilon > 0$ holds that

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For each $k > 0$ and $\epsilon > 0$ holds that

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Nondeterministic algorithms

HEURISTIC NONDETERMINISTIC ALGORITHMS

Language $L \in \text{Heur}_\delta \mathbf{NTime}(n^k)$ iff there is nondeterministic $O(n^k)$ -time algorithm A such that

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State of art for heuristic hierarchies

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For any $k > 0$ and $\epsilon > 0$ holds

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$$\mathbf{NP} \not\subseteq \text{Heur}_{\frac{1}{2}-\epsilon} \mathbf{NTime}(n^k).$$

Generalized hierarchy

ITSYKSON, KNOP, SOKOLOV, 2015

For any $k > 0$, $\epsilon > 0$ and $a > 1$ holds that

$$\text{Heur}_\epsilon \mathbf{FBPP} \not\subseteq \text{Heur}_{1-\frac{1}{a}-\epsilon} \mathbf{FBPTIME}(n^k).$$

Moreover there is $F: \{0, 1\}^n \rightarrow \{0, \dots, b-1\}$ such that $F \in \text{Heur}_\epsilon \mathbf{FBPP}$ and $F \notin \text{Heur}_{1-\frac{1}{a}-\epsilon} \mathbf{FBPTIME}(n^k)$.

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Samplable random variables

SAMPLABLE RANDOM VARIABLES

Ensemble of random variables $\gamma \in \mathbf{DSamp}(n^k)$ iff there is a randomized $O(n^k)$ -time algorithm A such that γ_n and $A(1^n)$ are equally distributed. We also denote $\mathbf{PSamp} = \bigcup_k \mathbf{DSamp}(n^k)$.

WATSON, 2014

For any $k > 0$, $\epsilon > 0$ and $a > 1$ there is an ensemble of random variables $\gamma \in \mathbf{PSamp}$ such that for every $\beta \in \mathbf{DSamp}(n^k)$ holds $\Delta(\gamma, \beta) > 1 - \frac{1}{a} - \epsilon$ and $\text{supp}(\gamma) = \{0, \dots, a - 1\}$.

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Proof for $a = 2$

① Consider the language $L = \{r \mid 0.r > \Pr[\gamma_n = 1]\}$.

② Note that $L \in \text{Heur}_c \mathbf{BPP}$.

Consider the following algorithm:

- ▶ Sample r_1, \dots, r_m from γ_n ;
- ▶ Return 1 if $0.r \geq \frac{1}{m} \sum_{i=0}^m r_i$;
- ▶ Return 0 in other case.

③ Note that $L \notin \text{Heur}_{1-\frac{1}{2}-\epsilon} \mathbf{BPTIME}(n^k)$.

Let us assume the opposite that L are decidable by algorithm D and consider the following algorithm:

- ▶ sample random $r \in \{0, 1\}^n$;
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Open questions

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- ▶ Hierarchy theorem for heuristic version of **RTime**(n^k);
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