# Analyzing Average-Case Complexity by Meta-Complexity 

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## Outline

## 1. Toward Excluding Heuristica

2. Limits of Black-Box Reductions
3. Our Results, Meta-Complexity, and Proof Techniques

## The $\mathbf{P} \neq \mathbf{N P}$ Conjecture and Cryptography

$$
\mathbf{P}=\mathbf{N} \mathbf{P} \quad \text { or } \quad \mathbf{P} \neq \mathbf{N} \mathbf{P}
$$

(-) Any problem in NP can be solved efficiently.Automated theorem proving can be done efficiently.
;- Any public-key cryptosystem can be broken.


Bitcoin loses its value.
(:) There is a problem in NP that can't be solved efficiently.
-) There might be a secure cryptosystem (?)

Using a public-key cryptosystem, Bitcoin prevents those who do not own a secret key from spending a coin.

## Impagliazzo's Five Possible Worlds


[Impagliazzo '95] classified five possible worlds consistent with our current knowledge.

Algorithmica

$$
P=N P
$$

## Impagliazzo's Five Possible Worlds


[Impagliazzo '95] classified five possible worlds consistent with our current knowledge.

## $P \neq N P$

() Any problem in NP can be solved efficiently.

Automated theorem proving is possible.
Impossible to construct a secure cryptosystem.

Algorithmica

$$
P=N P
$$

## Impagliazzo's Five Possible Worlds

Cryptomania
[Impagliazzo '95] classified five possible worlds consistent with our current knowledge.


Algorithmica

$$
P=N P
$$

## Impagliazzo's Five Possible Worlds

## The Ultimate Goal of Complexity Theory

is to decide which world corresponds to our world.
(In particular, we would like to resolve the conjecture that our world is Cryptomania.)

$$
P \neq N P \quad \& \quad \text { DistNP } \subseteq \operatorname{Avg} P
$$

## Algorithmica

$$
P=N P
$$

## Known Facts and Open Questions

Cryptomania
$\exists$ public-key crypto.
Minicrypt $\sqrt{\text { lil? }}$
$\exists$ secret-key crypto.

Pessiland


DistNP $\ddagger$ AvgP
("P $=$ NP on average")

Heuristica


Algorithmica


## Toward Public-key Crypto.

Cryptomania
$\exists$ public-key crypto.


Important Open Question
Can we exclude Pessiland?
DistNP $\ddagger$ AvgP
("P $\neq \mathrm{NP}$ on average")
Heuristica


$$
P \neq N P
$$

Algorithmica


## Important Open Question

$\mathrm{P} \neq \mathrm{NP}$ (Can we exclude Algorithmica?)

Proving the four implications
Our world is Cryptomania!

Proving one implication $\Leftrightarrow$
Excluding one world

## Limits of Current Proof Techniques

Cryptomania
$\exists$ public-key crypto.

$\mathbb{B}$ : Barrier results
Several types of proof techniques are insufficient to resolve the open question.

DistNP $\ddagger$ AvgP
("P $=\mathrm{NP}$ on average")


## Limits of Current Proof Techniques

Cryptomania
$\exists$ public-key crypto.

$\mathbb{B}$ : Barrier results
Several types of proof techniques are insufficient to resolve the open question.
$\exists$ secret-key crypto.


Algorithmica

# A New Paradigm: Meta-Complexity 

Cryptomania
$\exists$ public-key crypto

The complexity of problems asking for complexity

$\exists$ secret-key crypto.

Pessiland


DistNP $\ddagger$ AvgP
("P = NP on average")

Heuristica


Algorithmica

MINKT (Minimum Time-Bounded Kolmogorov Complexity Problem) The problem of computing the minimum program to compute $x$ efficiently

## MINKT

## Overcoming Limits of Black-box Reductions

## Cryptomania

$\exists$ public-key crypto.

$\exists$ secret-key crypto

Pessiland


Theorem [H. (FOCS 2018)]
Worst- and average-case complexities of MINKT are equivalent.

DistNP $\ddagger$ AvgP
(" $\mathrm{P} \neq \mathrm{NP}$ on average")

## Heuristica



Algorithmica
$>$ Limits: NP/poly $\cap$ coNP/poly
$>$ Conjecture [Rudich'97]: GapMINKT $\notin$ coNP/poly
$>$ This is the first result that goes beyond the limits!

## A Long-Standing Open Question

## Cryptomania

$\exists$ public-key crypto.
Minicrypt


NP is hard on average


A long-standing open question on worst- versus average-cas

UP is exponentially hard in the worst case

DistNP $\ddagger$ AvgP
(" $\mathrm{P} \neq \mathrm{NP}$ on average")


# Overcoming two barriers simultaneously 

## Cryptomania



## Overcoming two barriers simultaneously

## Cryptomania



## A New Relativization Barrier

## Cryptomania

$\exists$ public-key crypto.


## "Fine-Grained" Five Worlds [Chen-H.-Vafa (ITCS'22)]

Cryptomania $\quad$| Poly.-time |
| :--- |
| adversary |




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## Complexity Classes



PSPACE : polynomial space

PH : polynomial(-time) hierarchy

NP : non-deterministic polynomial-time
UP : unambiguous polynomial-time
(solvable by a non-deterministic polynomial-time machine with at most one accepting path for each input.)

P: polynomial time
[Ko'85, Grollmann \& Selman'88]
$\mathrm{UP} \neq \mathrm{P} \Leftrightarrow$ There is a one-to-one one-way function that is hard to invert in the worst case.

## (Black-Box) Reductions



These are proved by black-box reductions:

$\forall L \in S Z K$, there is a reduction $R^{(\cdot)}$ such that for any oracle $A$ that solves some $\left(L^{\prime}, \mathcal{D}\right) \in \operatorname{DistNP}$, $R^{A}(x)$ outputs the correct answer $L(x)$ for every input $x$.

A "non-black-box" reduction $\Leftrightarrow$ The reduction might fail if the oracle is inefficient.

## Limits of Black-Box Reductions

## Theorem [Feigenbaum \& Fortnow'93, Bogdanov \& Trevisan'06]

There is no nonadaptive black-box reduction from $L$ to DistNP, for any $L \notin \mathrm{NP} /$ poly $\cap$ coNP/poly.
> Nonadaptive black-box reductions are too strong to be useful for worst-case-to-average-case connections outside coNP/poly.
$>$ We need to use either non-black-box or adaptive reductions! We exploit the efficiency of an oracle using "meta-complexity".


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## Our Results

## Main Theorems [H. STOC'21]

(1) UP $\ddagger \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \quad \Rightarrow \quad$ DistNP $\nsubseteq \operatorname{Avg} \mathrm{P}$
(2) $\mathrm{PH} \nsubseteq \operatorname{DTIME}\left(2^{o(n / \log n)}\right) \quad \Rightarrow \quad$ DistPH $\nsubseteq \operatorname{Avg} \mathrm{P}$
(3) NP $\nsubseteq \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \Rightarrow \quad$ DistNP $\nsubseteq \operatorname{Avg}_{\mathrm{P}} \mathrm{P}$
$>n$ denotes the length of inputs (encoded as binary strings).
$>\operatorname{Avg}_{\mathrm{P}} \mathrm{P}(\subseteq \operatorname{AvgP})$ : the class of $(L, \mathcal{D})$ solvable by average-case polynomial-time algorithms whose running time can be "estimated."

## Our Results

Inverting a size-verifiable oneway function in the worst-case

The hard distribution is the uniform distribution $U$ or the tally distribution $\mathcal{T}$.
(1) $\operatorname{NTIME}_{\text {sv }}\left(2^{n^{1-\delta}}\right) \nsubseteq \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \Rightarrow \operatorname{coNP} \times\{\mathcal{U}, \mathcal{T}\} \nsubseteq \operatorname{Avg}_{1-n^{-c}}^{1} \mathrm{P}$
(2) $\operatorname{PHTIME}\left(2^{n^{1-\delta}}\right) \nsubseteq \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \Rightarrow \operatorname{PH} \times\{\mathcal{U}, \mathcal{T}\} \nsubseteq \operatorname{Avg}_{1-n^{-c}}^{1} \mathrm{P}$
(3) $\operatorname{NTIME}\left(2^{n^{1-\delta}}\right) \nsubseteq \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \Rightarrow \mathrm{NP} \times\{\mathcal{U}, \mathcal{T}\} \nsubseteq \operatorname{Avg}_{\mathrm{P}} \mathrm{P}$
$2^{n^{1-\delta}}$-time version of NP with success probability $n^{-c}$. (Refutation)

[^0]
## Time-Bounded Kolmogorov Complexity

$>t$-time-bounded Kolmogorov complexity of $x$ $\mathrm{K}^{t}(x):=$ (the length of a shortest program that prints $x$ in $t$ steps)

## Examples


$\mathrm{K}^{t}(x) \leq n+O(1) \quad$ for $t \gg n$ and for every $x \in\{0,1\}^{n} . \quad \leftarrow$ print "x" $\mathrm{K}^{\infty}(x) \geq n-2 \quad$ with probability $\geq \frac{3}{4}$ over a random $x \sim\{0,1\}^{n}$.

## Meta-Complexity - Complexity of Complexity

> Examples of meta-computational problems: MCSP, MKTP, MINKT, ...

## MINKT [Ko'91] = "Compute the time-bounded Kolmogorov complexity"

- $t$-time-bounded Kolmogorov complexity of $x$
$\mathrm{K}^{t}(x):=$ (the length of a shortest program that prints $x$ in $t$ steps)
- $\operatorname{MINKT}=\left\{\left(x, 1^{t}, 1^{s}\right) \mid \mathrm{K}^{t}(x) \leq s\right\}$.
- GapMINKT $=\left(\Pi_{\mathrm{Yes}}, \Pi_{\mathrm{No}}\right) \quad$ An " $O(\log n)$-additive approximation" version

$$
\begin{aligned}
& \Pi_{\mathrm{Yes}}=\left\{\left(x, 1^{t}, 1^{s}\right) \mid \mathrm{K}^{t}(x) \leq s\right\} . \quad \text { p: some polynomial } \\
& \Pi_{\mathrm{No}}=\left\{\left(x, 1^{t}, 1^{s}\right) \mid \mathrm{K}^{p(|x|+t)}(x)>s+\log p(|x|+t)\right\} .
\end{aligned}
$$

## Meta-Complexity - Complexity of Complexity

> Examples of meta-computational problems: MCSP, MKTP, MINKT, ...

$$
\text { MINKT }^{A}\left[\mathrm{Ko'}^{\prime} 91\right]=\text { "Compute the } A \text {-oracle time-bounded Kolmogorov complexity" }
$$

- $A$-oracle $t$-time-bounded Kolmogorov complexity of $x$ $\mathrm{K}^{t, A}(x):=\left(\right.$ the length of a shortest program $M^{A}$ that prints $x$ in $t$ steps)
- $\operatorname{MINKT}^{A}=\left\{\left(x, 1^{t}, 1^{s}\right) \mid \mathrm{K}^{t, A}(x) \leq s\right\}$.

Remark: In general, we may have $A \Phi_{m}^{p}$ MINKT $^{A}$.
It is easy to see MINKT ${ }^{A} \in \mathrm{NP}^{A}$.
Open: $\mathrm{NP} \leq$ MINKT? $\mathrm{NP} \leq \mathrm{MINKT}^{\mathrm{PH}}$ ?

## Average-Case Complexity $=$ Meta-Complexity

## Theorem [H. (FOCS'20)]

## DistPH $\subseteq \operatorname{AvgP} \quad \Leftrightarrow \quad$ GapMINKT ${ }^{\text {PH }} \in P$

$>$ GapMINKT $^{A}$ : an $O(\log n)$-additive approximation version of MINKT ${ }^{A}$.
$>$ Corollary: A new technique of analyzing average-case complexity by meta-complexity.


## Theorem [H. STOC'21]

$$
\text { (2') NP } \nsubseteq \operatorname{DTIME}\left(2^{o(n / \log n)}\right) \Rightarrow \text { DistPH } \nsubseteq \operatorname{AvgP}
$$



## Universal Heuristic Scheme - A key notion in this work

$>$ A universal heuristic scheme is "universal" in the following sense.

Proposition (universality of universal heuristic schemes)

## Assume DistNP $\subseteq$ AvgP.

For every $L:\{0,1\}^{*} \rightarrow\{0,1\}$, the following are equivalent.

1. There is a universal heuristic scheme for $L$.
2. $\{L\} \times \mathrm{PSamp} \subseteq \operatorname{Avg}_{\mathrm{P}} \mathrm{P}$.
```
P-computable
average-case
    poly-time
```


## The Definition of Universal Heuristic Scheme

$>$ Computational Depth [Antunes, Fortnow, van Melkebeek, Vinodchandran'06]

$$
\operatorname{cd}^{t}(x):=\mathrm{K}^{t}(x)-\mathrm{K}^{\infty}(x)
$$

$>(t, s)$-Time-Bounded Computational Depth

$$
\operatorname{cd}^{t, s}(x):=\mathrm{K}^{t}(x)-\mathrm{K}^{s}(x)
$$

$>$ An algorithm $A$ is called a universal heuristic scheme for $L$ if for some polynomial $p$, for every $x \in\{0,1\}^{*}$ and every $t \geq p(|x|)$,

1. $A(x, t)=L(x)$ and
2. $A(x, t)$ halts in time $2^{O\left(\mathrm{~cd}^{t, p(t)}(x)+\log t\right)}$.

## Theorem [H. STOC'21]

## (2') NP $\nsubseteq \operatorname{DTIME}\left(2^{O(n / \log n)}\right) \Rightarrow$ DistPH $\nsubseteq \operatorname{AvgP}$



## Fast Algorithms from Universal Heuristic Schemes

## Lemma

If there is some universal heuristic scheme $A$ for $L$, then

$$
L \in \operatorname{DTIME}\left(2^{O(n / \log n)}\right)
$$

Proof Idea: Find a parameter $t$ so that the input $x$ is "computationally shallow" (i.e., $\mathrm{cd}^{t, p(t)}(x)=O(n / \log n)$ ). Proof: Consider the following telescoping sum for a parameter $I=\epsilon \log n(\epsilon>0$, constant):

$$
\begin{aligned}
& \operatorname{cd}^{t, p(t)}(x)+\operatorname{cd}^{p(t), p \circ p(t)}(x)+\cdots+\operatorname{cd}^{p^{I-1}(t), p^{I}(t)}(x)=\mathrm{K}^{t}(x)-\mathrm{K}^{p^{I}(t)}(x) \leq n+O(1) \\
& \Rightarrow \text { for some } i \in\{1,2, \ldots, I\}, \text { we have } \mathrm{cd}^{p^{i-1}(t), p^{i}(t)}(x) \leq \frac{n+O(1)}{I}=O\left(\frac{n}{\log n}\right) .
\end{aligned}
$$

Algorithm $B$ :
Run $A(x, t), A(x, p(t)), A\left(x, p^{2}(t)\right), \ldots, A\left(x, p^{I-1}(t)\right)$ in parallel. Take the first one that halts, and output what it outputs.

Correctness: $B(x)=L(x)$ for every input $x$.

A universal heuristic scheme $A$ for $L: \exists p(t)=t^{O(1)}$,

1. $A(x, t)=L(x)$
2. $A(x, t)$ runs in time $2^{o\left(\mathrm{~cd}^{t, p(t)}(x)+\log t\right)}$.
(The running time of $B) \lesssim \min _{i}\left\{2^{o\left(\operatorname{cd}^{p^{i-1}(t), p^{i}(t)}(x)+\log p^{i}(t)\right)}\right\} \leq 2^{O(n / \log n)}$

$$
\left(p^{I}(t) \lesssim n^{c^{I}} \leq 2^{O(n / \log n)} \text { for } I=\epsilon \log n\right)
$$

## How we overcame limits of black-box reductions

Let $p(n)$ be the
runtime of AvgP.

DistPH $\subseteq$ AvgP

$\xrightarrow{\left[\mathrm{H} . \text { FOCS' }^{\prime} 18, \text { CCC' }^{20]}\right.}$ GapMINKT ${ }^{\text {NP }} \in \mathrm{P}$



```
                                    [H. STOC'21]
```

                                    based on [H. ITCS'20, STOC'20]
    
$>$ The reduction is non-black-box because we exploit the efficiency of AvgP.
I.e., the proof is not subject to the barrier of [Bogdanov \& Trevisan'06].

## Theorem [H. STOC'21]

## (2') $\operatorname{NP} \nsubseteq \operatorname{DTIME}\left(2^{o(n / \log n)}\right) \Rightarrow$ DistPH $\nsubseteq \operatorname{AvgP}$



## $k$-Wise Direct Product Generator [H. Stoc'20]

$$
\begin{aligned}
& \mathrm{DP}_{k}:\{0,1\}^{n} \times\left(\{0,1\}^{n}\right)^{k} \rightarrow\{0,1\}^{n k+k} \\
& \qquad \mathrm{DP}_{k}\left(x ; z_{1}, \ldots, z_{k}\right)=\left(z_{1}, \ldots, z_{k},\left\langle z_{1}, x\right\rangle, \ldots,\left\langle z_{k}, x\right\rangle\right)
\end{aligned}
$$

A pseudorandom generator construction based on a "hard" truth table $x$ that extends seed $z$ by $k$ bits.
$\left\langle z_{i}, x\right\rangle$ : the inner product between $z_{i}$ and $x$ modulo 2 .
A Reconstruction Property of $\mathrm{DP}_{k}$ : (under DistNP $\subseteq$ AvgP or a derandomization assumption)
For every oracle $D:\{0,1\}^{n k+k} \rightarrow\{0,1\}$ and every $x \in\{0,1\}^{n}$, if $\mathrm{K}^{t, D}(x) \geq k+O(\log n)$, then $\mathrm{DP}_{k}(x ;-)$ is pseudorandom against $D$; that is,

$$
\operatorname{Pr}_{z \sim\{0,1\}^{n k}}\left[D\left(\mathrm{DP}_{k}(\boldsymbol{x} ; \mathbf{z})\right)=1\right] \approx \operatorname{Pr}_{w \sim\{0,1\}^{n k+k}}[D(w)=1] .
$$

The Key Point: (The advice complexity of $\left.\mathrm{DP}_{k}\right)=k+O(\log n)$

## Claim: DistNP $\subseteq$ AvgP $\Rightarrow$ GapMINKT $\in P$

$>$ For simplicity, $t:=n^{2}$.
$>$ Consider the following distributional problem (MINKT, $\mathcal{U}^{\prime}$ ) $\in$ DistNP:
Given $x \sim\{0,1\}^{n}$ as input, decide whether $\mathrm{K}^{t}(x)<n-2$ or not.
$>$ Let $A$ be an errorless heuristic algorithm that solves (MINKT, $U^{\prime}$ ) with probability $\geq 1-o(1)$.
$A(x)$ outputs the correct answer or $\perp$ ("time out").

$$
\mathrm{K}^{t}(x)<n-2 \Rightarrow A(x) \in\{1, \perp\}
$$

$$
\operatorname{Pr}_{x \sim\{0,1\}^{n}}[A(x)=\perp] \leq o(1)
$$

> A randomized algorithm $B$ for solving GapMINKT:
(YES case)

$$
B\left(x, 1^{s}\right):=1 \Leftrightarrow A\left(\operatorname{DP}_{k}(x ; z)\right) \in\{1, \perp\} \text { for a random } z \sim\{0,1\}^{n k+k} \text { and } k:=s+O(\log n) .
$$

(No case) $\quad \mathrm{K}^{t^{\prime}}(x) \gg s+O(\log n)=k \Rightarrow \operatorname{Pr}_{z}\left[A\left(\mathrm{DP}_{k}(x ; z)\right) \in\{1, \perp\}\right] \approx \underset{w}{\operatorname{Pr}}[A(w) \in\{1, \perp\}] \leq \frac{1}{4}+o(1)$

$$
\begin{aligned}
\mathrm{K}^{t}(x) \leq s & \Rightarrow \mathrm{~K}^{2 t}\left(\mathrm{DP}_{k}(x ; z)\right) \leq \mathrm{K}^{t}(x)+|z|+O(1) \leq s+n k+O(1) \ll k+n k-2 . \\
& \Rightarrow A\left(\operatorname{DP}_{k}(x ; z)\right) \in\{1, \perp\} \text { with probability } 1
\end{aligned}
$$

This is a non-black-box reduction: $t^{\prime} \approx($ the running time of $A)=\operatorname{poly}(t, n)$.
GapMINKT is a meta-computational problem!

## Summary and Open Questions

> Meta-complexity is a powerful tool to analyze average-case complexity.

- Especially because it allows us to overcome the limits of black-box reductions
$>$ A lot of interesting questions remain open:
- Non-relativizing proof techniques in this context?
- NP-hardness of GapMINKT
- Can we prove NP $\nsubseteq \operatorname{DTIME}\left(2^{o(n)}\right) \Longrightarrow$ DistNP $\nsubseteq$ AvgP?
- Does the exponential-time hypothesis (ETH) imply DistPH $\nsubseteq$ AvgP?
- Can we prove PH $\ddagger$ io-DTIME $\left(2^{o(n)}\right) \Rightarrow$ DistPH $\ddagger$ io-AvgP?

Viola's barrier comes into play in this setting!


[^0]:    $\mathcal{T}:=\left\{\mathcal{J}_{n}\right\}_{n \in \mathbb{N}} ; \mathcal{T}_{n}$ is the singleton distribution on $1^{n}$.

