# On hardness of some Boolean counting CSPs 

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## Counting problems

## Formally: counting problem $\mathbb{P}$

$\equiv$ a relation $R \subseteq\{0,1\}^{*} \times\{0,1\}^{*}$, such that $\exists p: \mathbb{N} \rightarrow \mathbb{N}$ polynomial: $\forall(x, y) \in R: \quad|y| \leq p(|x|)$

## Informally

a binary relation $R$ between instances $x$ and their solutions $y$; size of solutions polynomially bounded in size of instance

Aim: count solutions
For $x \in\{0,1\}^{*}: \quad S_{\mathbb{P}}(x) \equiv S_{R}(x):=\left\{y \in\{0,1\}^{*} \mid(x, y) \in R\right\}$

$$
\left|S_{\mathbb{P}}(x)\right| \equiv\left|S_{R}(x)\right|=? \in \mathbb{N}
$$

Complexity of $\mathbb{P}$ understood via Turing machines $T$ computing $\left|S_{R}\right|:\{0,1\}^{*} \rightarrow \mathbb{N}$.

## Example: \#CSP(Q)

For a finite set of relations on a finite set $A: \quad Q \subseteq \mathcal{R}_{A}$

## $\# \operatorname{csp}(Q)$

Input formula $\varphi \equiv \bigwedge_{i=1}^{\ell} \varrho_{i}\left(v_{i, 1}, \ldots, v_{i, m_{i}}\right)$

$$
\begin{aligned}
& \varrho_{i} \in Q^{\left(m_{i}\right)}, v_{i, j} \in\left\{x_{1}, \ldots, x_{n}\right\} \text { for } 1 \leq i \leq \ell, \\
& 1 \leq j \leq m_{i}
\end{aligned}
$$

Goal number of satisfying assignments (solutions)

$$
\left|\left\{s:\left\{x_{1}, \ldots, x_{n}\right\} \rightarrow A \quad \mid \quad s \models \varphi\right\}\right|
$$

## Compare:

Decision problem \# solutions: $=0$ vs. $>0$ ?
Counting problem \# solutions: $=? \in \mathbb{N}$
In this talk:
$A=\{0,1\}$ Boolean relations $\ldots \ldots \ldots \ldots \ldots \ldots \ldots$. . . . . . . . . . .

## Our Motivation

## Goal

Understanding of \#ETH
............... Exponential Time Hypothesis for Counting Problems

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What is \#ETH?
analogue of ETH for counting problems

## What is ETH? <br> (Impagliazzo, Paturi, Zane, 2001)

There is $c \in \mathbb{R}_{>0}$ such that. . .
... no deterministic algorithm solves 3-SAT in time $\mathrm{O}\left(2^{c n}\right)$
i.e., 3-SAT is not solvable in subexponential time.

## \#ЕTH

What is \#етн? (Dell, Husfeldt, Marx, Taslaman, Wahlén, 2014) There is $c \in \mathbb{R}_{>0}$ such that. . .
... no deterministic algorithm solves \#3-SAT in time $\mathrm{O}\left(2^{\text {cn }}\right)$

## ᄀ\#ETH?

$\forall \varepsilon>0 \exists$ deterministic $\mathrm{O}\left(\left(2^{\varepsilon}\right)^{n}\right)$-algorithm $A$ :
A solves \#3-SAT
Lower bound on bases of runtime
$b:=\inf \left\{c \in \mathbb{R}_{\geq 0} \mid \exists\right.$ deterministic $\mathrm{O}\left(2^{c n}\right)$-algorithm $A$ for \#3-SAT $\}$
$\neg$ \#ЕTH $\Longleftrightarrow b=0$
algos with faster and faster runtimes
\#ETH $\Longleftrightarrow b>0$ no algo better than $\mathrm{O}\left(\left(2^{b}\right)^{n}\right)$

## A characterisation of \#ETH, or 'What is $n$ ?'

## \#ETH

(Dell et al., 2014)
There is $c \in \mathbb{R}_{>0}$ such that...
... no deterministic algorithm solves \#3-SAT in time $\mathrm{O}\left(2^{\text {cn }}\right)$ where $n$ number of variables in the solution.

## $\forall k \geq 3: \quad$ ETH $\Longleftrightarrow$ <br> (Dell et al., 2014)

There is $c \in \mathbb{R}_{>0}$ such that...
... no deterministic algorithm solves \#k-SAT in time $\mathrm{O}\left(2^{c N}\right)$ where $N$ size of the formula (number of clauses/atoms).

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## Remark

- analogous to ETH and $k$-SAT by Impagliazzo, Paturi, Zane
- importance of complexity parameter already noted there


## Reductions

## Standard reductions: $\mathbb{P} \leq \mathbb{Q}$

$\exists$ deterministic polynomial-time $\mathbb{Q}$-oracle algorithm counting $\mathbb{P}$

## Reductions with linear parameter growth:

$\exists a, b \in \mathbb{N}: \exists$ deterministic $\mathbb{Q}$-oracle algorithm $A$ with oracle $B$ :
$\forall$ input $x$ of $\mathbb{P}$ of measure $n$ :

- $A$ counts $\mathrm{S}_{\mathbb{P}}(x)$
- $\forall \varepsilon>0$ : $A$ runs in at most $O\left(2^{\varepsilon n}\right)$ time-steps (subexponential)
- for each oracle call:
$A$ calls $B$ on an input of measure $N \leq a n+b$


## Example

$(\# \operatorname{CSP}(Q)$, atoms $) \leq \operatorname{lin}(\# \operatorname{CSP}(Q)$, variables $) \quad a=\max _{\rho \in Q} \operatorname{ar}(\varrho)$

## Subexponentiality

## $(\mathbb{P}, n)$ is subexponential <br> $\forall \varepsilon>0 \exists$ deterministic $\mathrm{O}\left(\left(2^{\varepsilon}\right)^{n}\right)$-algorithm $A: \quad A$ counts $(\mathbb{P}, n)$

$\neg$ \#ETH $\Longleftrightarrow$ \#3-sAT is subexponential (wrt. variables)
$\Longleftrightarrow \forall k \geq 3$ : \#k-SAT is subexponential (wrt. variables/atoms)
Lemma:
If $(\mathbb{P}, n) \leq \operatorname{lin}(\mathbb{Q}, N)$, then:
$(\mathbb{Q}, N)$ subexponential $\Longrightarrow(\mathbb{P}, n)$ subexponential

## Preserving subexponentiality

Lemma: $\quad$ For $(\mathbb{P}, n) \leq \operatorname{lin}(\mathbb{Q}, N)$ :

## $(\mathbb{Q}, N)$ subexponential $\Longrightarrow(\mathbb{P}, n)$ subexponential

- Consider any given $\delta>0$; define $\varepsilon:=\frac{\delta}{a+2}$
- Consider input $x$ for $\mathbb{P}$ of measure $n \geq b$.
- Use the $\mathbb{Q}$-oracle algorithm $A$ with an $\mathrm{O}\left(2^{\varepsilon N}\right)$-oracle $B$ on $x$
- Each oracle call to $B$ takes $O\left(2^{\varepsilon N}\right)$ time, i.e.

$$
\leq C_{1} \cdot 2^{\varepsilon N} \leq C_{1} \cdot 2^{\varepsilon(a n+b)} \leq C_{1} \cdot 2^{\varepsilon(a n+n)}=C_{1} \cdot 2^{\varepsilon(a+1) n}
$$

- Altogether $\mathrm{O}\left(2^{\varepsilon n}\right)$ steps, i.e. $\leq C_{2} \cdot 2^{\varepsilon n}$ oracle calls
- Total time for $x: \leq C_{3} \cdot 2^{\varepsilon n} \cdot 2^{\varepsilon(a+1) n}=C_{3} \cdot 2^{\varepsilon(a+2) n}=C_{3} \cdot 2^{\delta n}$


## Consequence:

$\neg$ \#ETH $\Longleftrightarrow$ \#3-sAT is subexp. (wrt. variables)
$\Longrightarrow$ \#3-sat is subexp. (wrt. atoms)

## Sparsification

\#3-sat subexp. wrt. atoms $\Longrightarrow$ \#3-sAT subexp. wrt. variables
. . . needs a different construction:

## Problem:

with $N$ variables $\quad \rightsquigarrow$ populate $n=N^{\ell} \ell$-ary atoms (constraints)

## Sparsification <br> (Impagliazzo, Paturi, Zane)

$\forall \varepsilon>0: \exists C \geq 0$ : split up any big instance of measure $N$

- into $\leq 2^{\varepsilon N}$ small subproblems
- each subproblem is sparse $n \leq C N$
- the whole algorithm of splitting and combining runs in $\mathrm{O}\left(2^{\varepsilon N}\right)$


## Counting CSPs and reductions

Importance of the complexity measure

- size of instance (\# atoms) vs. solution size (\# variables)
- $\left(\# \operatorname{CSP}(P), \operatorname{param}_{1}\right) \leq \operatorname{lin}\left(\# \operatorname{CSP}(Q), \operatorname{param}_{2}\right) \Longrightarrow$ $\left(\# \operatorname{CSP}(Q)\right.$, param $\left.{ }_{2}\right)$ subexponential
$\Longrightarrow\left(\# \operatorname{CSP}(P)\right.$, param $\left.{ }_{1}\right)$ subexponential
- for free:
(\#CSP $(Q)$, variables) subexponential
$\Longrightarrow(\# \operatorname{CSP}(Q)$, size $)$ subexponential
- needs work (e.g. sparsification):
$(\# \operatorname{CSP}(Q)$, size) subexponential
$\Longrightarrow(\# \operatorname{csp}(Q)$, variables) subexponential


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$\Longrightarrow(\# \operatorname{CSP}(Q)$, variables) subexponential


## Universal algebra

helps constructing $\leq_{\text {lin }}$-reductions

## A Galois connection

## Partial polymorphisms

$\forall W \subseteq \mathcal{R}_{A} \quad \operatorname{pPol}(W):=\left\{f \in \mathcal{P}_{A} \mid \forall \varrho \in W: f \triangleright \varrho\right\}$
Invariant relations
$\forall C \subseteq \mathcal{P}_{A} \quad \operatorname{lnv}(C):=\left\{\varrho \in \mathcal{R}_{A} \mid \forall f \in C: f \triangleright \varrho\right\}$

Theorem: for finite $A$
(Romov 1981)

- $\left\{\operatorname{pPol}(\operatorname{lnv}(W)) \mid W \subseteq \mathcal{R}_{A}\right\} \ldots . . .$. . all strong partial clones (closure under projections, composition, domain restriction)
- $\left\{\operatorname{lnv}(\operatorname{pPol}(C)) \mid C \subseteq \mathcal{P}_{A}\right\} \ldots .$. all weak systems with equality (closure under conjunctive definitions incl. =)
$[W]_{\wedge,=}=\operatorname{pPol}(\operatorname{Inv}(W))$


## Theorem on intervals of strong partial clones

Given a clone $F=\operatorname{Pol} Q$ with relational clone $Q=\operatorname{lnv} F$
Partial clones with total part $F$

$$
\begin{aligned}
\mathcal{I}(F) & =\left\{C \leq \mathcal{P}_{A} \mid C \cap \mathcal{O}_{A}=F\right\} \\
& =\left\{C \leq \mathcal{P}_{A} \mid C \cap \mathcal{O}_{A}=\operatorname{Pol} Q\right\} \\
& =\left\{\operatorname{pPol}(W) \mid W \subseteq \mathcal{R}_{A} \wedge \operatorname{Pol} W=\operatorname{pPol}(W) \cap \mathcal{O}_{A}=\operatorname{Pol} Q\right\}
\end{aligned}
$$

contains largest element $F_{\mathrm{T}}$
Weak systems with equality generating $Q$

$$
\begin{aligned}
\mathcal{I}(Q) & =\left\{\operatorname{lnvpPol}(W) \mid W \subseteq \mathcal{R}_{A} \wedge \operatorname{Pol} W=\operatorname{Pol} Q\right\} \\
& =\left\{S=[S]_{\wedge,=} \subseteq Q \mid \text { Pol } S=\operatorname{Pol} Q\right\} \\
& =\left\{S=[S]_{\wedge,=} \subseteq Q \mid[S]_{\exists, \wedge,=}=Q\right\}
\end{aligned}
$$

has a least element $S_{\perp}$

## Weak bases

Given a relational clone $Q, F:=\mathrm{Pol} Q$

## Weak basis of $Q$

(Schnoor\&Schnoor, 2008)
$S_{\perp}$ be least weak system incl. $=$ with $\left[S_{\perp}\right]_{\exists, \wedge,=}=Q$
$W$ weak base of $Q$ : any finite $W \subseteq S_{\perp}$ with $S_{\perp}=[W]_{\Lambda,=}$ (i.e. finite weak generating sets of $S_{\perp}$ )

Properties of weak bases $W, W^{\prime}$ of $Q=\left[Q_{0}\right]_{\exists, \wedge,=}$

- $[W]_{\exists, \wedge,=}=Q$
- $W \subseteq\left[Q_{0}\right]_{\wedge,=}$
$\Longrightarrow \quad(\# \operatorname{CSP}(W)$, var. $) \leq$ lin $\left(\# \operatorname{csp}\left(Q_{0}\right)\right.$, var. $)$
- $[W]_{\Lambda,=}=\left[W^{\prime}\right]_{\Lambda,=}$
V. Lagerkvist (2014) determined weak bases for Boolean rel. clones


## Result

Theorem
$\forall H \in \mathcal{H}: \quad \neg$ ETH $\Longleftrightarrow \# \operatorname{CSP}(H)$ is subexponential wrt. var.
i.e. many Boolean counting CSPs do not have subexponential algorithms under \#ETH.

What is $\mathcal{H}$ ?
$\mathcal{H}=\left\{H \subseteq\right.$ fin $\mathcal{R}_{2} \mid$ Pol $\left.H \subseteq M \vee \operatorname{Pol} H \subseteq F_{4}^{2} \vee \operatorname{Pol} H \subseteq F_{8}^{2}\right\}$


## An important problem in social relations

The T-counting problem
input a string
goal count the number of occurrences of the letter $T$

## An important problem in social relations

## The T-counting problem

input a string
goal count the number of occurrences of the letter T

## Example

input Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention. Thank you for your attention.

