

GEMS OF TCS

FINE-GRAINED COMPLEXITY

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- **Today:** Identify reason why we're stuck

ALGORITHMIC COMPLEXITY OF SAT



2-SAT $O(m)$

1-SAT $O(m)$

ALGORITHMIC COMPLEXITY OF SAT

3-SAT 1.308^n

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ALGORITHMIC COMPLEXITY OF SAT

k -SAT $2^{n(1-O(1/k))}$

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- Strong Exponential Time Hypothesis (SETH)

SAT requires time 2^n

EDIT DISTANCE

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e | e p h a n t

r e | e v a n t

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A T A G T A C T

~~C~~ A T A C A C T



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p

A T A G T A C T
~~C~~ A T A C A C T

G

$$\tilde{O}(n^2)$$

OTHER PROBLEMS

Longest Common Subsequence

Orthogonal Vectors

Edit Distance

Hamming Closest Pair

All Pairs Max Flow

RNA-Folding

Regular Expression Matching

Graph Diameter

Subset Sum

CONJECTURED HARDNESS

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- A conjecture for each problem?
- One conjecture to rule them all?
- **Fine-grained Complexity**: Better-than-known algorithms for one problem would imply better-than-known algorithms for other problems

Orthogonal Vectors (OV)

ORTHOGONAL VECTORS PROBLEM

- S, T are sets of N vectors from $\{0, 1\}^d$. Are there $s \in S$ and $t \in T$ such that $s \cdot t = \sum_{i=1}^d s_i \cdot t_i = 0$?

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- Think of $d = \log^2 N$
- Can solve in time $d \cdot N^2$
- SETH implies that OV cannot be solved in time $N^{1.99}$

FINE-GRAINED REDUCTIONS

formula ϕ of SAT

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Algorithm for SAT

Algorithm for OV

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Algorithm for OV

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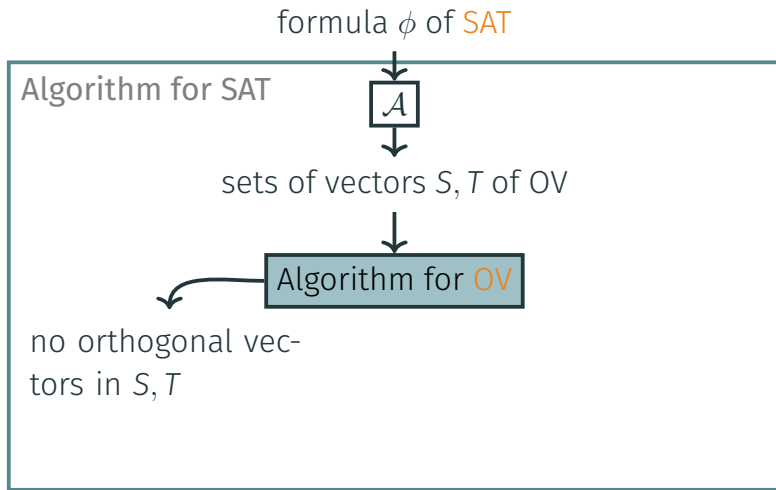
Algorithm for SAT



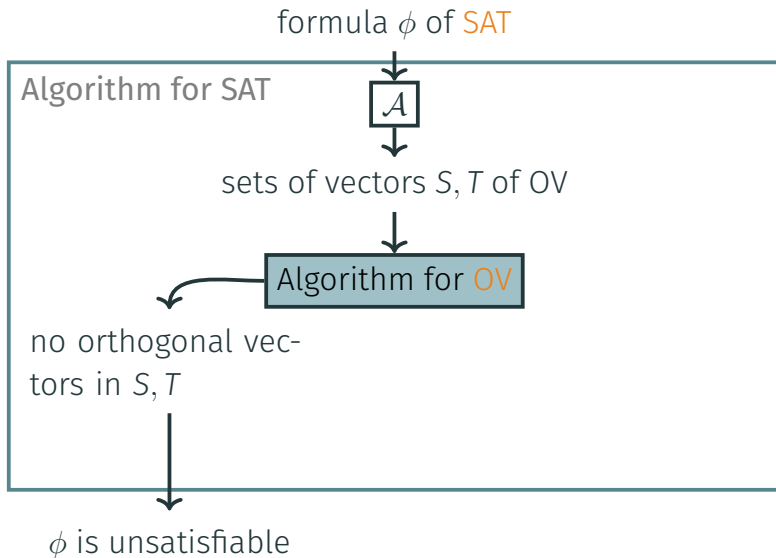
sets of vectors S, T of OV

Algorithm for OV

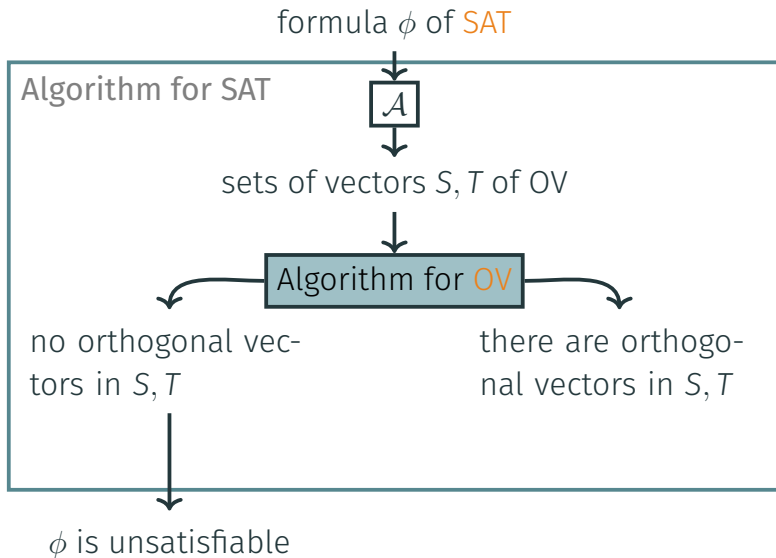
FINE-GRAINED REDUCTIONS



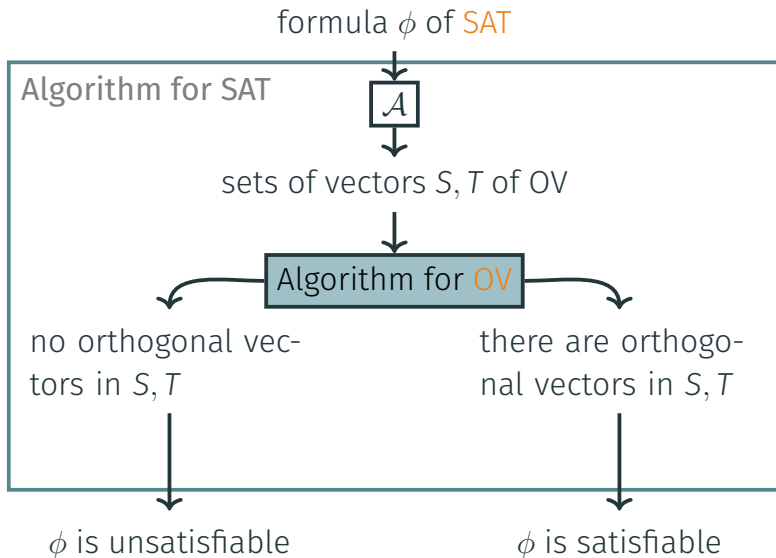
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- $N = 2^{n/2}$

SETH \implies OV

- For an assignment $x \in \{0, 1\}^{n/2}$, add $s \in \{0, 1\}^m$ to S :

$s_i = 1$ iff x **doesn't satisfy** clause C_i

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$$N^{1.99} = (2^{n/2})^{1.99} = 2^{0.995n}$$

The Dominating Set Problem

DOMINATING SET

- **k -Dominating Set:** Given $G = (V, E)$, $|V| = n$, find an $S \subseteq V$, $|S| = k$ such that

$$\forall v \in V: v \in S \text{ or } \exists u \in S: (v, u) \in E$$

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- SETH implies that k -DS cannot be solved in time $n^{k-0.01}$ for any k

SETH \implies DS

Partition vars in k groups:

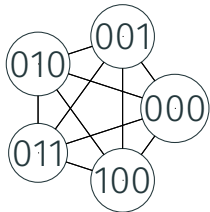
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$2^{n/k}$ vertices

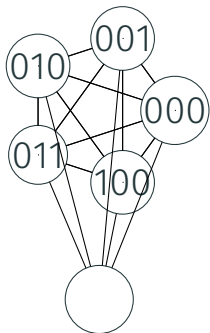


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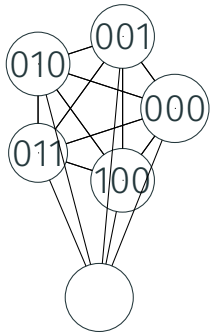


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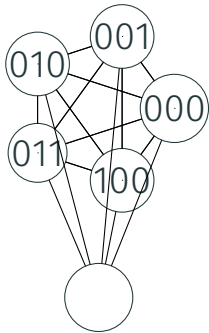
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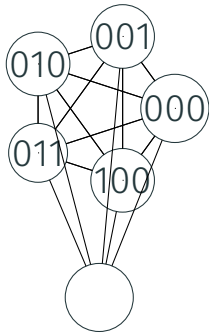
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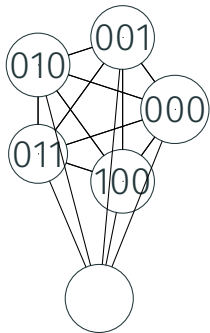
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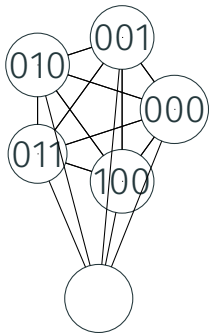
m vertices:

$2^{n/k}$ vertices



cl_1

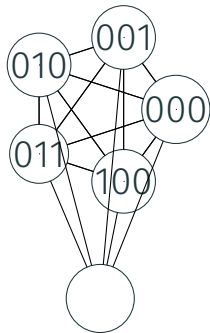
$2^{n/k}$ vertices



cl_2

cl_3

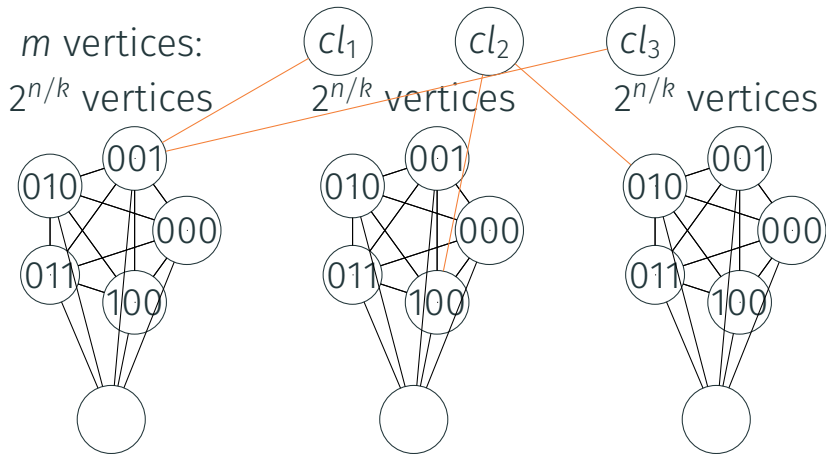
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SETH \implies DS

- For every k , we reduce SAT on n vertices k -DS with

$$\approx 2^{n/k}$$

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- If k -DS on N vertices can be solved in time $N^{k-0.1}$, then SAT can be solved in time

$$N^{k-0.1} = 2^{(n/k)(k-0.1)} = 2^{n-0.1n/k}$$