

Fine-Grained Complexity Classification of Counting Problems

A golden wheat field at sunset with birds flying in the sky. The sun is low on the horizon, casting a warm, golden glow over the scene. The wheat stalks are in sharp focus in the foreground, with some blurred in the background. Three birds are visible in the sky, flying towards the right.

Holger Dell

Saarland University and Cluster of Excellence (MMCI)

Simons Institute for the Theory of Computing

In P or not in P ?

Exact counting

P | #P-hard

Approximate counting

FPRAS | #BIS-complete | #SAT-hard

FPRAS | no FPRAS unless RP=NP

Motivation for fine-grained complexity

#P-hard problems could have algorithms running in time:

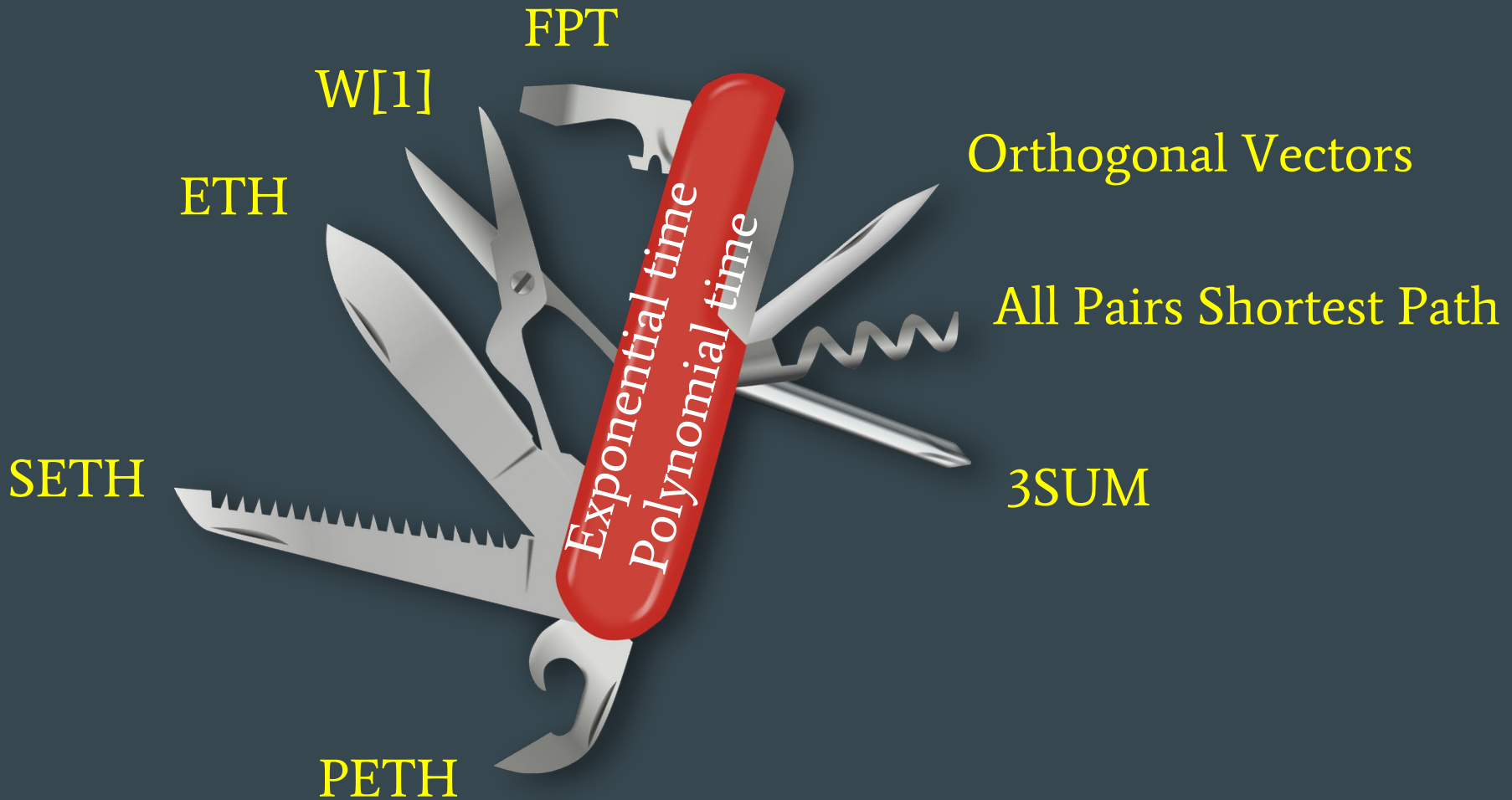
$$O(2^n) \quad O(1.01^n) \quad O(2^{\sqrt{n}}) \quad O(n^k) \quad O(2^k n)$$

Problems in P could have algorithms running in time:

$$O(n) \quad O(n^2) \quad O(n^{100})$$

Fine-grained complexity is a **toolbox**
to pin down best running time more precisely

Available conjectures, problems, and classes



3-CNF-SAT faster than exhaustive search

Schöning's algorithm

Expected running time: $(4/3)^n$

- Sample random assignment $x \in \{0,1\}^n$
- While there is clause $(a \vee b \vee c)$ not satisfied by x :
 - Choose random literal in $\{a,b,c\}$
 - **Flip** its value in x
- Restart process if too long

Satisfying
assignment

Random
assignment

Hamming
distance

x^*

x

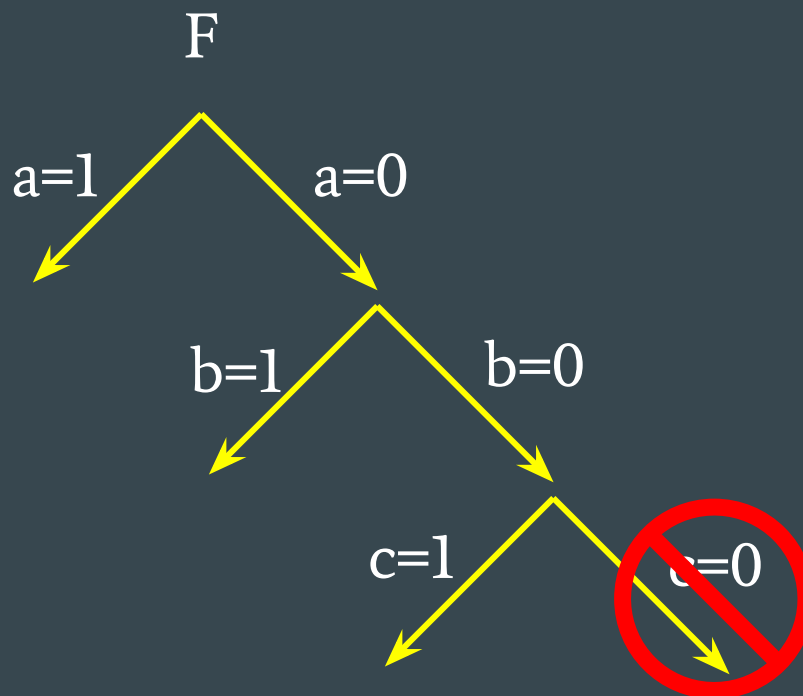


$\Pr \geq 1/3$

$\Pr \leq 2/3$

Branching algorithms

Formula F
contains clause
 $(a \vee b \vee c)$



For every clause $(a \vee b \vee c)$, only
need to look at $\frac{7}{8}$ of all assignments

→ Time $7^{n/3} \sim 1.913^n$

Counting Satisfying Assignments of 3-CNFs

Kutzkov 07: exact counting in time $O(1.6423^n)$

Thurley 12: ϵ -approximation in time $O(\epsilon^{-2} 1.5366^n)$

Exponential time hypothesis (#ETH)

$\exists \delta > 0$. #SAT for 3-CNFs is not in time $(1+\delta)^n$

Sparsification Lemma

(Impagliazzo Paturi Zane 01; Calabro Impagliazzo Paturi 06; D, Husfeldt, Wahlén 10)

Can assume #clauses $\sim (1/\delta)^3 n$

Sparsification Lemma

(Impagliazzo Paturi Zane 01; Calabro Impagliazzo Paturi 06; D Husfeldt Wahlén 10)

Input:

- k -CNF formula F with n variables and m clauses
- $\epsilon > 0$

Output:

$F_1 \dots F_t$ such that

- $\text{sat}(F) = \text{sat}(F_1) \sqcup \dots \sqcup \text{sat}(F_t)$
- each F_i has $(k/\epsilon)^k n$ clauses
- $t = 2^{\epsilon n}$

General CNFs

Chan and Williams 15:

Compute **#SAT** for a CNF formula **F** in time $2^{n(1 - \text{savings})}$

- **F** is a **k**-CNF \rightarrow savings $\sim 1 / k$
- **F** has **cn** clauses \rightarrow savings $\sim 1 / \log c$

Strong exponential time hypothesis (**#SETH**)

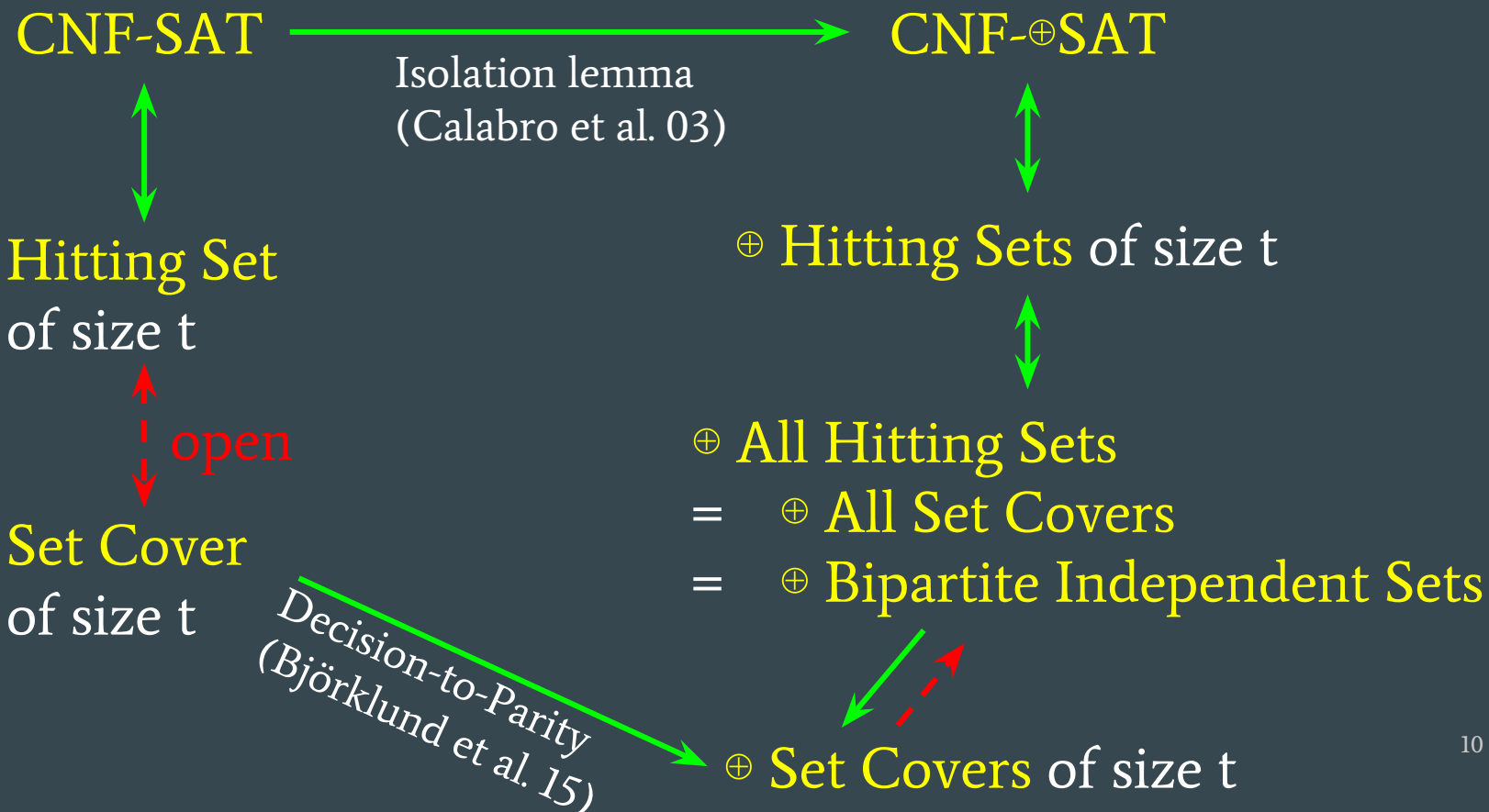
$\forall \delta > 0 \exists k.$ **#SAT** for **k**-CNFs cannot have savings $\geq \delta$

Problems equivalent under SETH

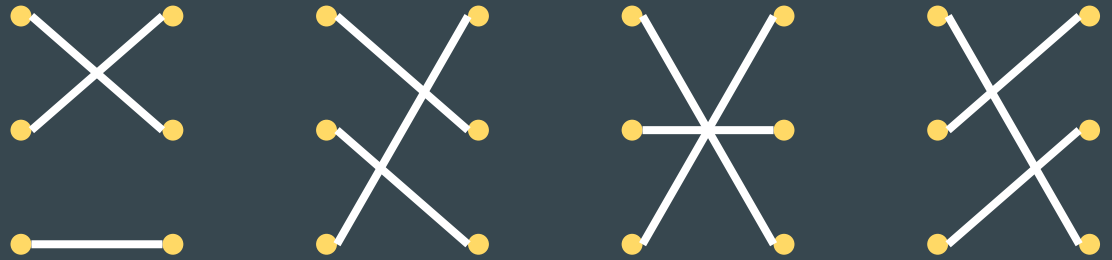
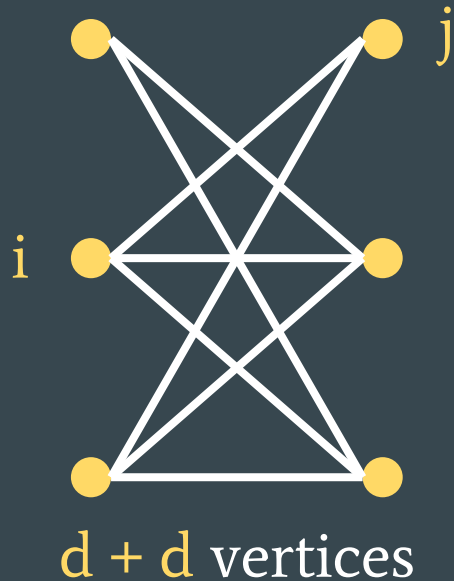
Cygan et al. 2012

Decision problems

Counting modulo two



Perfect Matchings in Bipartite Graphs



Perfect Matchings

$$= \text{per}(A)$$

$$= \sum_{\text{permutation } \pi} \prod_{i \in \{1..d\}} A_{i \pi(i)}$$

$d \times d$ matrix A

$A_{ij} = 1$ iff $\{i, j\}$ is an edge

Computing the permanent

Evaluation time
 $\sim d! \sim 2^d \log d$

$$\text{per}(d \times d \text{ matrix } A) = \sum_{\text{permutation } \pi} A_{1 \pi(1)} \cdots A_{d \pi(d)}$$

$$= \sum_{\text{all functions } f: \{1..d\} \rightarrow \{1..d\}} A_{1 f(1)} \cdots A_{d f(d)}$$

$$- \sum_j \sum_{f: \{1..d\} \rightarrow \{1..d\} \setminus \{j\}} A_{1 f(1)} \cdots A_{d f(d)}$$

$$+ \sum_{j,k} \sum_{f: \{1..d\} \rightarrow \{1..d\} \setminus \{j,k\}} A_{1 f(1)} \cdots A_{d f(d)}$$

...

$$\prod_{i \in \{1..d\}} \sum_{j \in \{1..d\}} A_{ij}$$

Evaluation time
 $O(d2^d)$

Ryser's Inclusion-Exclusion Formula (1963)

$$= \sum_{S \subseteq \{1..d\}} (-1)^{|S|} \prod_{i \in \{1..d\}} \sum_{j \in \{1..d\} \setminus S} A_{ij}$$

Fine-Grained Complexity of the Permanent

Curticapean 15; D Husfeldt Marx Taslaman Wahlén 10

If $\text{per}(d \times d \text{ matrix } A)$ can be computed in time $2^{o(d)}$,
then #ETH is false

Servedio and Wan 05

If A has $\leq cd$ nonzero entries,
 $\text{per}(A)$ can be computed in time $(2-\delta)^d$ where $\delta(c) < 1$

Permanent Strong Exponential Time Hypothesis (PETH)

$\forall \delta > 0 \exists c.$ $\text{per}(c\text{-sparse } A)$ not in time $(2-\delta)^d$

Count Perfect Matchings in General Graphs

$\text{per}((n/2) \times (n/2) \text{ matrix})$

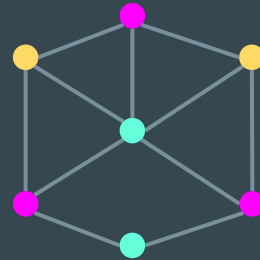
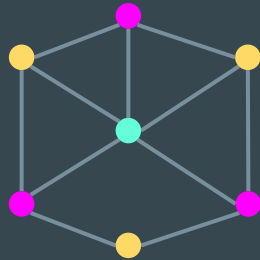
= # Perfect Matchings of bipartite graph with $n/2 + n/2$ vertices

→ $2^{n/2}$ algorithm

Björklund 11

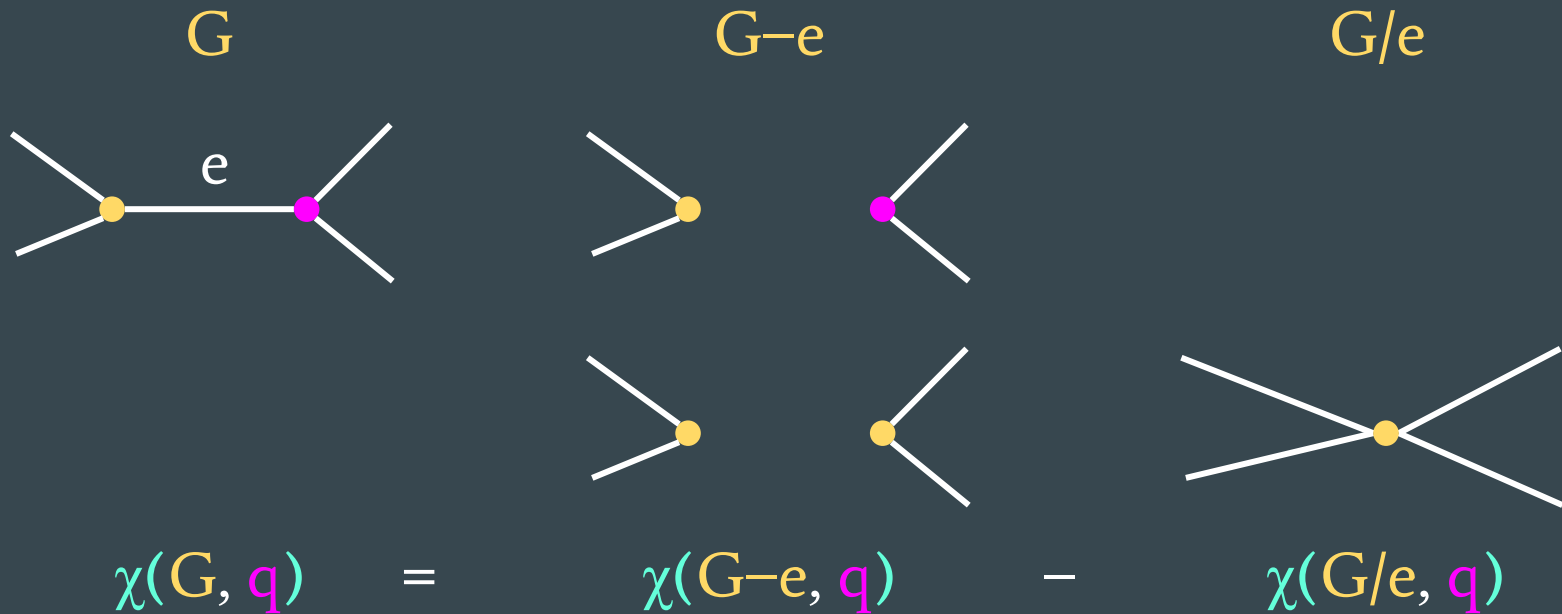
Count perfect matchings in general graphs in time $2^{n/2}$.

Proper q-colorings



Chromatic polynomial & Deletion-Contraction

$\chi(G, q) = \#$ proper q -colorings of G



$$\chi(\text{k-independent set}, q) = q^k$$

$\rightarrow \chi(G, q)$ is a degree- n polynomial in q .

Compute # q-Colorings

The deletion-contraction algorithm takes time 2^m .

Björklund Husfeldt Koivisto 09

Compute the number of q-colorings in time 2^n .

Impagliazzo Paturi Zane 01

ETH \rightarrow no $2^{o(n)}$ algorithm for q-coloring.

The Tutte Polynomial

$$T(\mathbf{G}, \mathbf{x}, \mathbf{y}) = \sum_{A \subseteq E} (\mathbf{x}-1)^{k(A)-k(\mathbf{G})} (\mathbf{y}-1)^{k(A)+|A|-|V|}$$

Generalizes

- chromatic polynomial $\chi(\mathbf{G}, \mathbf{q}) = (-1)^{n-k(\mathbf{G})} \mathbf{q}^{k(\mathbf{G})} T(\mathbf{G}, \mathbf{1}-\mathbf{q}, \mathbf{0})$
- Ising model, \mathbf{q} -state Potts model
- many combinatorial problems

Computing the Tutte polynomial

The trivial algorithm runs in time 2^m

Björklund Husfeldt Kaski Koivisto 08

Time 2^n algorithm

Curticapean 15; D Husfeldt Marx Taslaman Wahlén 10; Jaeger Vertigan Welsh 90

#ETH \rightarrow no $2^{o(n)}$ algorithm

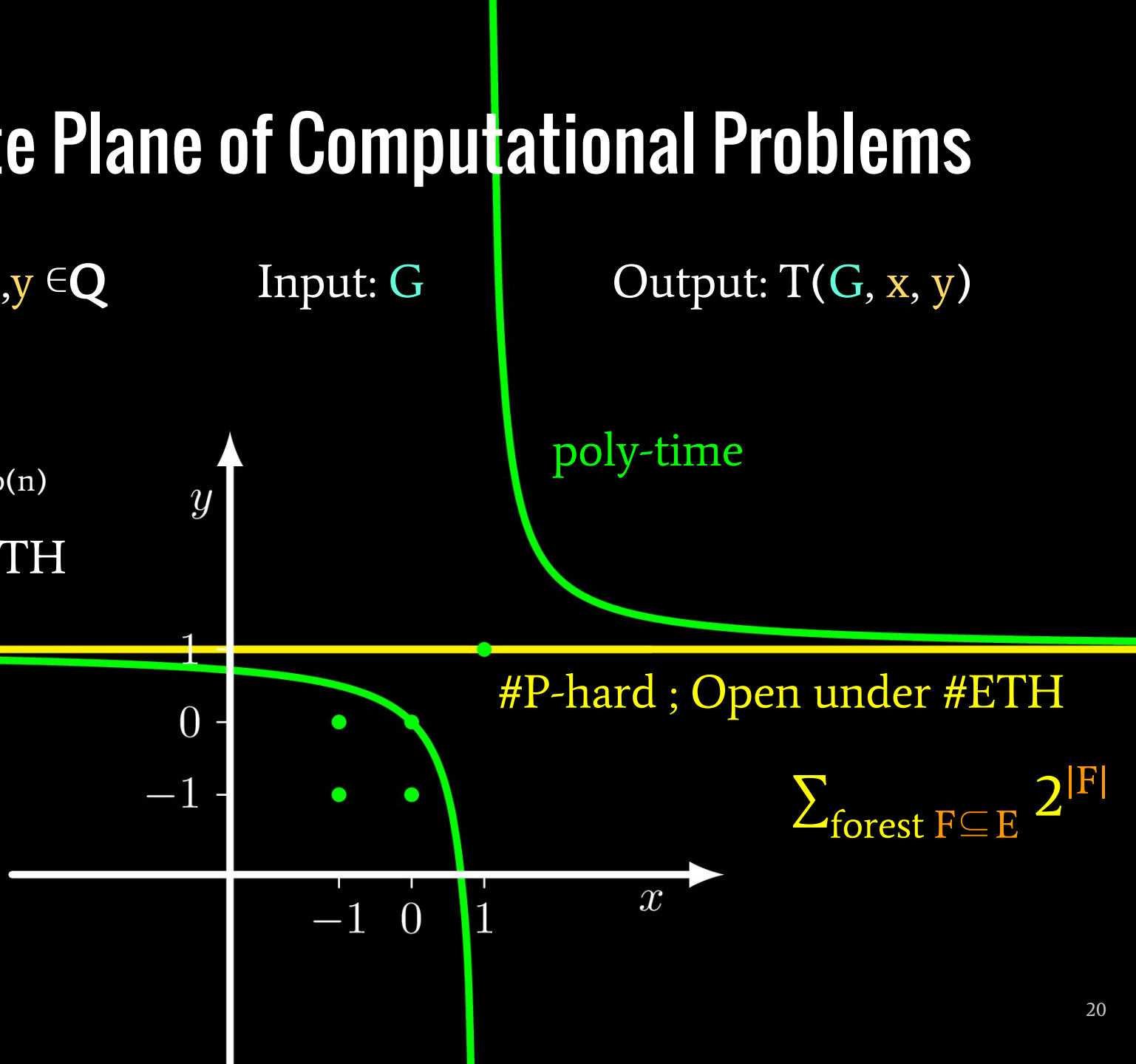
The Tutte Plane of Computational Problems

Fix $x, y \in \mathbb{Q}$

Input: G

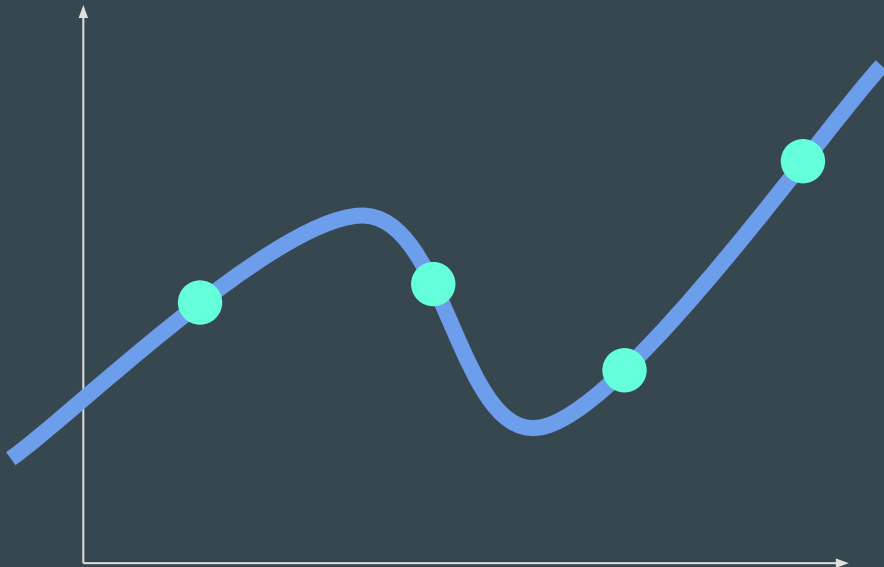
Output: $T(G, x, y)$

Black:
not in $2^{o(n)}$
under #ETH



Polynomial Interpolation

→ compute p in poly-time from samples



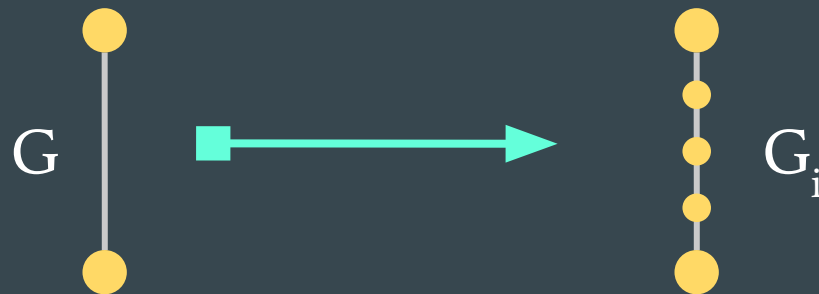
univariate polynomial p
degree m

$m + 1$ samples
($a, p(a)$)

Interpolation in Counting Complexity

Only rules out $2^{o(m/\log m)}$ time algorithms under #ETH

$$p(G, a_i) = p(G_i, a)$$

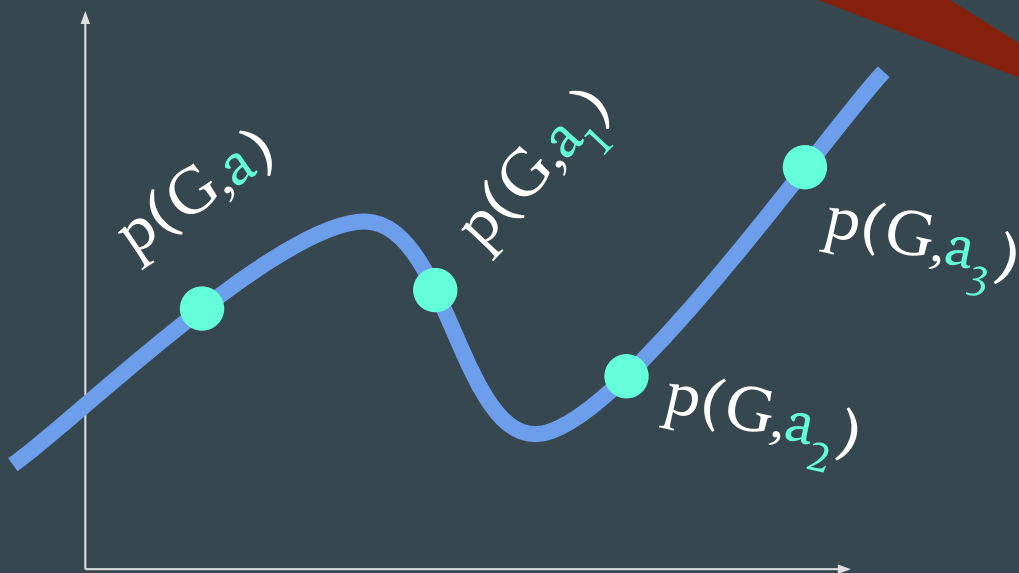


Need **$m+1$ samples**

→ $m+1$ different gadgets

→ $m(G_i) \sim m \log m$

Graph polynomial $p(G, z)$
degree m



Block interpolation

Curticapean 15

Univariate $p(G, z) \rightarrow$ Multivariate $q(G, z_1, \dots, z_{m/r})$

such that

- $p(G, z) = q(G, z_1, \dots, z_{m/r})$
- Each z_i has degree $r=O(1)$

\rightarrow Multivariate interpolation

$\sim r^{m/r} = \exp(\varepsilon m)$ samples

$r+1=O(1)$ distinct gadgets per variable

Can rule out $2^{o(m)}$
time algorithms
under #ETH

Approximate Counting

Valiant Vazirani 86

FPRAS for # Sat when given access to an NP-oracle

Traxler 14

If CNF-Sat is in $c^n \cdot \text{poly}(m)$ time,

we can $(1+1.1^{-n})$ -approximate # CNF-Sat in time $(c + .00001)^n$

Is Counting really harder than Decision?

If **SETH** is true,

- **CNF-SAT** takes time 2^n
- **# CNF-SAT** takes time 2^n (even to approximate)
- **QBF-SAT** takes time 2^n

In applications: **CNF-SAT** **much** easier than **QBF-SAT**.

Is there a tight reduction from **QBF-SAT** to **# CNF-SAT** ?

Open problems

- is computing $\sum_{\text{forest } F \subseteq E} 2^{|F|}$ hard under ETH or #ETH ?
- is the permanent hard under SETH ?
- which problems are hard under PETH ?
- fine-grained inapproximability for # CNF-SAT ?
- is counting really harder than decision?
can we tightly reduce QBF-SAT to # CNF-SAT ?