The complexity of approximate counting Part 1

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The Complexity of Approximate Counting

- Relative Complexity and #BIS: This talk
- Markov Chain Monte Carlo: Ivona Bezáková, Nayantara Bhatnagar (next talk!)
- Approximate Counting and Constraint Satisfaction Problems David Richerby (Part 2)
- Correlation Decay and Phase Transitions Yitong Yin

Cai, Chebolu, Dyer, Galanis, Greenhill, Guo, Gysel, Jerrum, Kelk, Lapinskas, Martin, Paterson, Štefankovič, Vigoda

Interaction strength $\gamma \ge -1$. Set of spins [q]. Graph G = (V, E). 1 if spins are the same and 0 otherwise. partition function $Z_{\text{Potts}}(G;q,\gamma) = \sum \left[(1 + \gamma \,\delta(\{\sigma(u),\sigma(v)\})) \right]$ $\sigma: V \rightarrow [q] e = \{u, v\} \in E$ Configuration σ : assigns spins to vertices

Interaction strength
$$\gamma \ge -1$$
.
Set of spins $[q]$. Using $q = 2$
Graph $G = (V, E)$.
partition function
 $Z_{Potts}(G; q, \gamma) = \sum_{\sigma: V \to [q]} \prod_{e=\{u,v\} \in E} (1 + \gamma \, \delta(\{\sigma(u), \sigma(v)\}))$

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• "computational counting": computing sums of products.

• In FP^{#P}. We'll focus on approximate counting within #P (or within FP^{#P}).

In #P up to "easily-computable factor"

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The goal: an FPRAS

A randomised approximation scheme (RAS) is an algorithm for approximately computing the value of a function $f : \Sigma^* \to \mathbb{R}$.

Input:

- finite alphabet Σ
- instance $x \in \Sigma^*$

• rational error tolerance $\epsilon \in (0, 1)$

e.g., if $f = Z_{\text{Potts}}$ then x encodes G

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Output: Rational number z such that, for all x,

$$\Pr\left(e^{-\varepsilon}f(x)\leqslant z\leqslant e^{\varepsilon}f(x)\right)\geqslant\frac{3}{4}.$$

z is a random variable, depending on the "coin tosses" made by the algorithm"

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FPRAS: Running time bounded by a polynomial in |x| and ε^{-1} .

e.g., if $f = Z_{\text{Potts}}$ then x encodes G and |x| = n.

"No FPRAS": typically, we can't even get close!

• $k \cdot G$: k disjoint copies of G.

• $Z_{\text{Potts}}(k \cdot G; q, \gamma) = Z_{\text{Potts}}(G; q, \gamma)^k$.

• Set
$$k = O\left(\frac{1}{\varepsilon}\right)$$

• given a constant factor approximation to $Z_{\text{Potts}}(k \cdot G; q, \gamma)$

contrast with optimisation!

- take k'th root
- get FPRAS for $Z_{\text{Potts}}(G; q, \gamma)$.

An approximation within a polynomial factor would also suffice.

How difficult is it to FPRAS a problem in #P?

under (randomised) polynomial-time Turing reductions

• It can be NP-hard

Obviously, an FPRAS for counting satisfying assignments will tell you, with high probability, whether there is one. How difficult is it to FPRAS a problem in #P?

- It can be NP-hard
- But it can't be much harder
 - Valiant, Vazirani 1986 bisection technique

• **#SAT** can be approximated by a probabilistic polynomial-time Turing machine using an oracle for **SAT**.



How difficult is it to FPRAS a problem in #P?

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• **#SAT** can be approximated by a probabilistic polynomial-time Turing machine using an oracle for **SAT**.

• Given an FPRAS for #SAT, obtain an FPRAS for any

problem in #P

Cook's theorem is parsimonious

number of accepting computations of Turing machine/input pair = number of satisfying assignments of the constructed formula

f, *g*: functions from Σ^* to \mathbb{N} .

AP-reduction from *f* to *g*:

apologies to Pilu Crescenzi (1997)

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AP-reduction from f to g: randomised algorithm \mathcal{A} for

computing *f* using an oracle for *g*. Input: $(x, \varepsilon) \in \Sigma^* \times (0, 1)$. *x* is an instance of *f*

f, g: functions from Σ^* to N.

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1 \mathcal{A} makes oracle calls (w, δ)

w is an instance of g. $0 < \delta < 1$ is an error bound satisfying $\delta^{-1} \leq \text{poly}(|x|, \varepsilon^{-1})$

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- **1** \mathcal{A} makes oracle calls (w, δ)
- A meets the specification for being a RAS for f whenever the oracle meets the specification for being a RAS for g

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- **(**) the run-time of A is polynomial in |x| and ε^{-1} .

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- **1** \mathcal{A} makes oracle calls (w, δ)
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- **1** the run-time of \mathcal{A} is polynomial in |x| and ε^{-1} .

The class of functions with an FPRAS is closed under AP-reducibility.

An impossible goal (if NP \neq RP)

A dichotomy within #P:

- FPRASable problems.
- The rest.

All AP-interreducible

All AP-interreducible.
No FPRAS unless NP = RP

An impossible goal (if NP \neq RP)

A dichotomy within #P:

- FPRASable problems.
- The rest.

Bordewich 2010

Like Ladner 1975 for P versus NP.

Let π be a problem in #P such that there is no FPRAS for π . Then there is a problem $\pi' \in$ #P such that

- there is no FPRAS for π' , and
- π ≰_{AP} π'.

Three classes of interreducible classes within #P

- FPRASable problems
- Problems AP-interreducible with #BIS
- Problems AP-interreducible with #SAT

Name **#BIS** Instance A bipartite graph *B*. Output The number of independent sets in *B*.

All problems in #P are AP-reducible to #SAT (since a parsimonious reduction is an AP-reduction) Another impossible goal

A trichotomy within #P:

- FPRASable problems
- Problems ≡_{AP} #BIS
- Problems ≡_{AP} #SAT

Bordewich 2010

If there is no FPRAS for #BIS then there is a problem π in #P that does not have an FPRAS such that #BIS $\leq_{AP} \pi$.

(Infinite hierarchy below #BIS) We have candidates above #BIS but nothing natural below

some settings where trichotomies arise ...

Graph Homomorphisms

Homomorphism from G to H

$$\sigma\colon V(G)\to V(H)$$

for every edge $(u, v) \in E(G)$, $(\sigma(u), \sigma(v)) \in E(H)$





Name #HOMSTO(*H*). Instance Graph *G*. Output The number of homomorphisms from *G* to *H*.

connected 3-vertex H



Weighting function $w: V(H) \to \mathbb{Q}_{\geq 0}$ assigns non-negative rational weight to each vertex of *H*.

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weighting function for each $v \in V(G)$

 $W(G,H) = \{w_v \mid v \in V(G)\}.$

Weighting function $w: V(H) \to \mathbb{Q}_{\geq 0}$ assigns non-negative rational weight to each vertex of *H*.

 $W(G,H) = \{w_v \mid v \in V(G)\}.$

$$Z_{H}(G, W(G, H)) = \sum_{\sigma \in \operatorname{Hom}(G, H)} \prod_{\nu \in V(G)} w_{\nu}(\sigma(\nu)).$$







Connection to the Potts model

For any q > 2 and any (efficiently approximable) γ , counting homomorphisms to J_q is AP-equivalent to computing the partition of the ferromagnetic *q*-state Potts model $Z_{\text{Potts}}(\cdot; q, \gamma)$.

 J_q is like J_3 but with q branches

We'll come back to the Potts model

Approximate counting problems which are $\equiv_{AP} \#BIS$

- Counting downsets in a partial order
- Graph homomorphism counting problems
- Counting Constraint Satisfaction (#CSP) problems
- Ferromagnetic Ising with mixed fields
- Ferromagnetic Ising in a hypergraph (even without fields)
- Counting stable matchings (in general, or for geometric preference models)

We've seen some.

follows also from

#CSP results

See also [Kelk 2003]

part 2

The ferro Ising partition function of a hypergraph

Interaction strength $\gamma > 0$.

$$Z_{\text{Ising}}(H; \gamma) = \sum_{\sigma: V \to \{0,1\}} \prod_{f \in E} \left(1 + \gamma \, \delta(\{\sigma(v) \mid v \in f\}) \right)$$

 $\delta(S) = 1$ if its argument is a singleton and 0 otherwise.

Back to Binary Matroid Ising

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By (Fortuin, Kasteleyn 1972) this is the same as the Tutte polynomial version

$$Z_{\text{Tutte}}(H; \gamma) = \sum_{F \subseteq E} 2^{\kappa(F)} \gamma^{|F|},$$

 $\kappa(F)$ is the number of connected components in (V, F): think of the connected components of the underlying graph if you replace hyperedges by cliques

Back to Binary Matroid Ising

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By (Fortuin, Kasteleyn 1972) this is the same as the Tutte polynomial version

$$Z_{\text{Tutte}}(H; \gamma) = \sum_{F \subseteq E} 2^{\kappa(F)} \gamma^{|F|},$$

Proof: "Integrate out" one of the sums in $\sum_{\sigma: V \to \{0,1\}} \sum_{F \subseteq E} \prod_{f \in F} \gamma \, \delta(\{\sigma(v) \mid v \in f\})$

Back to Binary Matroid Ising



IS with F on RHS: $2^{|U|-|\Gamma(F)|}$

Back to FerroPotts

$$Z_{\text{Tutte}}(H; \mathbf{1}) = \sum_{F \subseteq E} 2^{\kappa(F)}$$

- Vertices $U \cup \{v\}$
- Hyperedges $R = \{a, b, c, v\}$ $B = \{c, d, v\}$ $G = \{d, e, v\}$

Contribution of *F*: $2^{\kappa(F)} = 2^{|U| - \Gamma(F) + 1}$.

Ferromagnetic Hypergraph Ising \leq_{AP} Downsets

Downsets in a partial order: Represent partial order as directed graph (drawn on the slides with edges pointing down). Spin 1 forces all vertices below to have spin 1

Ferromagnetic Hypergraph Ising \leq_{AP} Downsets

$$Z_{\text{Ising}}(H; 1) = \sum_{\sigma: V \to \{0,1\}} \prod_{f \in E} \left(1 + \delta(\{\sigma(v) \mid v \in f\} \right)$$



Then stretch and thicken to get other γ

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Now place #BIS in logically defined class