

Sparse Hashing for Scalable Approximate Model Counting: When Theory and Practice Finally Meet

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Joint work with S. Akshay

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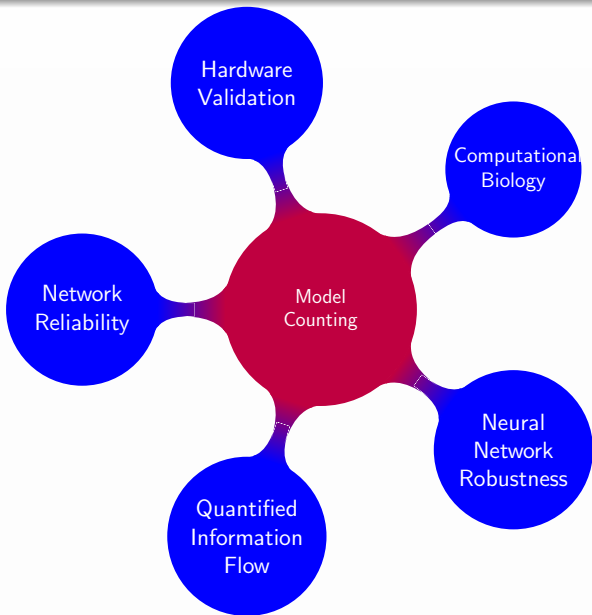
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- $|\text{Sol}(F)| = 3$



Different Shades of Approximation

- Probabilistic $(1 + \varepsilon)$ -Approximation

$$\Pr \left[\frac{|\text{Sol}(F)|}{1 + \varepsilon} \leq \text{ApproxCount}(F, \varepsilon, \delta) \leq |\text{Sol}(F)|(1 + \varepsilon) \right] \geq 1 - \delta$$

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$$\frac{|\text{Sol}(G)|}{4} \leq C \leq 4 \cdot |\text{Sol}(G)| \implies \frac{|\text{Sol}(F)|}{2} \leq \sqrt{C} \leq 2 \cdot |\text{Sol}(F)|$$

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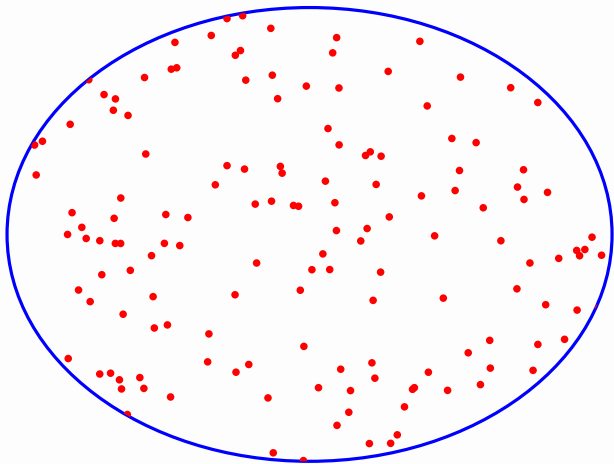
- From 4 to $(1 + \varepsilon)$ -factor

Construct $G = F_1 \wedge F_2 \dots F_{\frac{1}{\varepsilon}}$ And then we can take $\frac{1}{\varepsilon}$ -root

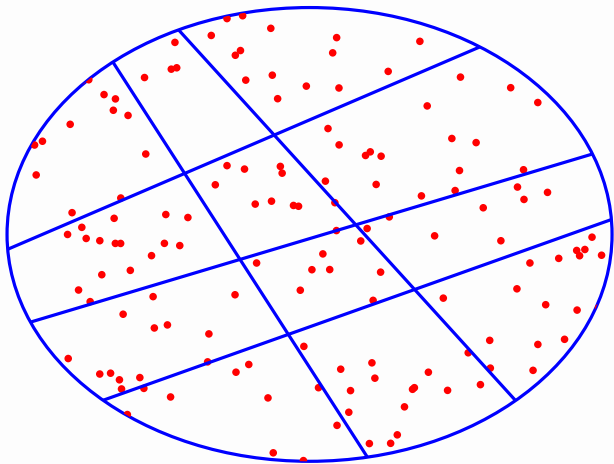
The Rise of Hashing-based Approach: Promise of Scalability and Guarantees

(S83,GSS06,GHSS07,CMV13b,EGSS13b,CMV14,CDR15,CMV16,ZCSE16,AD16
KM18,ATD18,SM19,ABM20,SGM20)

As Simple as Counting Dots

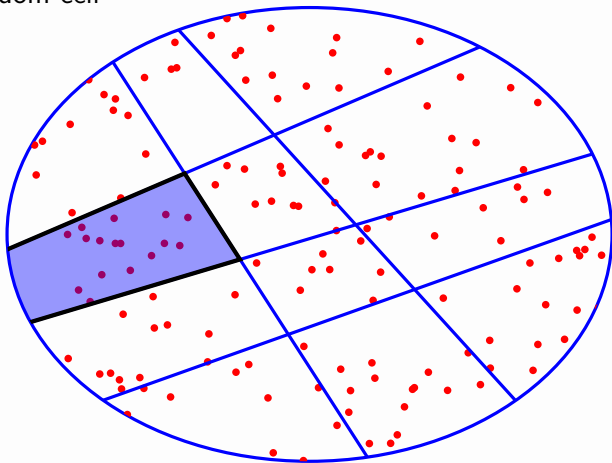


As Simple as Counting Dots



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Pick a random cell



Estimate = Number of solutions in a cell \times Number of cells

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Challenge 3 How many cells?

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- Two choices for thresh.
 - thresh = constant \rightarrow 4-factor approximation
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dispersion index: $\frac{\sigma^2[Z_m]}{(E[Z_m])^2} \leq$ some constant
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Techniques based on thresh = $\mathcal{O}(\frac{1}{\varepsilon^2})$ such as ApproxMC scale significantly better than those based on thresh = constant.

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- Designing function h : assignments \rightarrow cells (hashing)
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- Choose h randomly from a specially constructed large family H of hash functions

Carter and Wegman 1977

Pairwise Independent Hashing

- Variables: X_1, X_2, \dots, X_n
- To construct $h : \{0, 1\}^n \rightarrow \{0, 1\}^m$, choose m random XORs
- Pick every X_i with prob. $\frac{1}{2}$ and XOR them
 - $X_1 \oplus X_3 \oplus X_6 \cdots \oplus X_{n-2}$
 - Expected size of each XOR: $\frac{n}{2}$

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- To choose $\alpha \in \{0, 1\}^m$, set every XOR equation to 0 or 1 randomly

$$X_1 \oplus X_3 \oplus X_6 \cdots \oplus X_{n-2} = 0 \quad (Q_1)$$

$$X_2 \oplus X_5 \oplus X_6 \cdots \oplus X_{n-1} = 1 \quad (Q_2)$$

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- Solutions in a cell: $F \wedge Q_1 \cdots \wedge Q_m$

The Performance Bottleneck: SAT Calls

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- Solutions in a cell: $F \wedge Q_1 \cdots \wedge Q_m$
- Performance of state of the art SAT solvers degrade with increase in the size of XORs (SAT Solvers != SAT oracles)

The Hope of Short XORs

- View the set of XORs as Matrices: $AX = b$ where $\cdot = \wedge$ and $+$ $= \oplus$
 - A is 0-1 matrix of size $m \times n$
 - b is 0-1 matrix of size $m \times 1$
- If we pick every variable X_i with probability p .
 - Expected Size of each XOR: np
- $\Pr[\sigma_1 \text{ is in Cell}] = \Pr[A\sigma_1 = b] = \frac{1}{2^m}$
 - $E[Z_m] = \sum_{\sigma \in \text{Sol}(F)} \Pr[\sigma_1 \text{ is in Cell}] = \frac{|\text{Sol}(F)|}{2^m}$
- Now,

$$\begin{aligned}\Pr[\sigma_1 \text{ and } \sigma_2 \text{ are in Cell}] &= \Pr[A\sigma_1 = b = A\sigma_2] \\ &= \Pr[A\sigma_1 = b] \Pr[A(\sigma_2 - \sigma_1) = 0] \\ &= \frac{1}{2^m} \left(\frac{1}{2} + \frac{(1-2p)^w}{2} \right)^m\end{aligned}$$

- $\sigma^2[Z_m] \leq E[Z_m] + \sum_{\sigma_1 \in \text{Sol}(F)} \sum_{\substack{\sigma_2 \in \text{Sol}(F) \\ w=d(\sigma_1, \sigma_2)}} r(w, m)$
 - where, $r(w, m) = \frac{1}{2^m} \left(\left(\frac{1}{2} + \frac{(1-2p)^w}{2} \right)^m - \frac{1}{2^m} \right)$
- For $p = \frac{1}{2}$, we have $\frac{\sigma^2[Z_m]}{E[Z_m]} \leq 1$

The First Decade

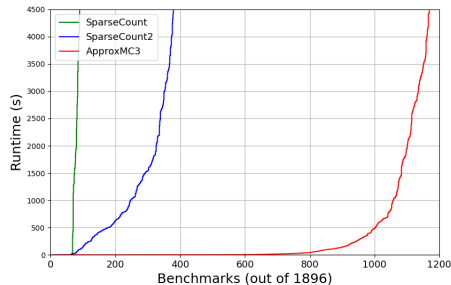
- The first decade (GSS07,EGSS14,ZCSE16,AD17,ATD18)
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 - The weak bounds lead to significant slowdown: typically $100\times$ to $1000\times$ factor of slowdown! (ADM20)



The Power of Isoperimetric Inequalities

- $$\sum_{\sigma_1 \in \text{Sol}(F)} \sum_{\substack{\sigma_2 \in \text{Sol}(F) \\ w=d(\sigma_1, \sigma_2)}} r(w, m) = \sum_{w=1}^n C_F(w) r(w, m)$$
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- Well, $C_F(1) \leq |\text{Sol}(F)| \binom{n}{1}$
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- Possibilities: $\{(0, 0, 0), (1, 0, 0), (0, 1, 0), (0, 0, 1)\}$

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Theorem (Harper's Theorem (1962))

$$C_F(1) \leq |\text{Sol}(F)| \binom{\ell}{1} \text{ where } \ell = \log |\text{Sol}(F)|$$

Lemma (Rashtchian and Raynaud 2019)

$$\sum_{w=1}^n C_F(w) \leq \sum_{w=1}^n \binom{8e\sqrt{n \cdot \ell}}{w} \text{ where } \ell = \log |\text{Sol}(F)|$$

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- Improvement from $\binom{n}{w}$ to $\binom{8e\sqrt{n\cdot\ell}}{w}$
- $\frac{\binom{n}{w}}{\binom{8e\sqrt{n\cdot\ell}}{w}} \approx \left(\frac{n}{\ell}\right)^{\frac{w}{2}}$

Theorem (Informal)

For all q, k , $|\text{Sol}(F)| \leq k \cdot 2^m$, $p = \mathcal{O}(\frac{\log m}{m})$ we have

$$\frac{\sigma^2[Z_m]}{\mathbb{E}[Z_m]} \leq q(\text{a constant})$$

Recall, average size of XORs: $n \cdot p$

Improvement of p from $\frac{m/2}{m}$ to $\frac{\log m}{m}$

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Challenge: No meaningful bounds on $|\text{Sol}(F)|$

How many cells?

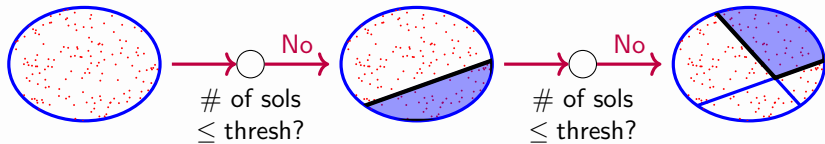
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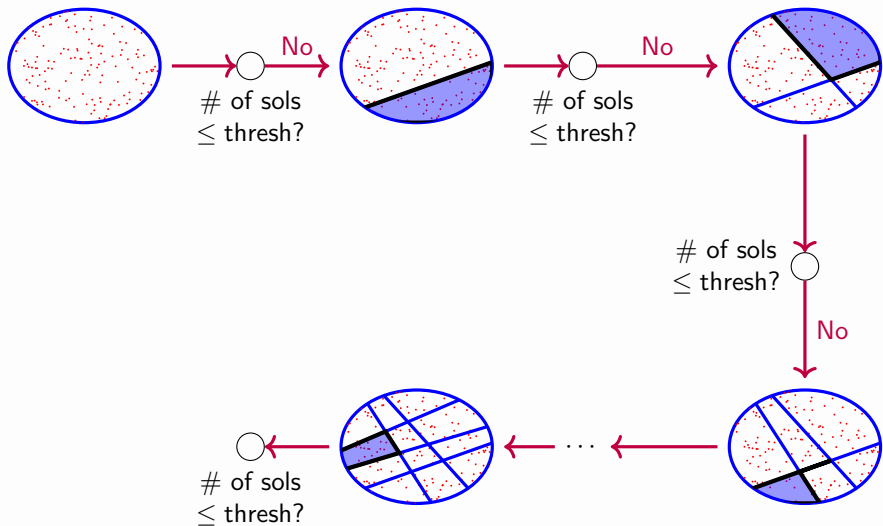
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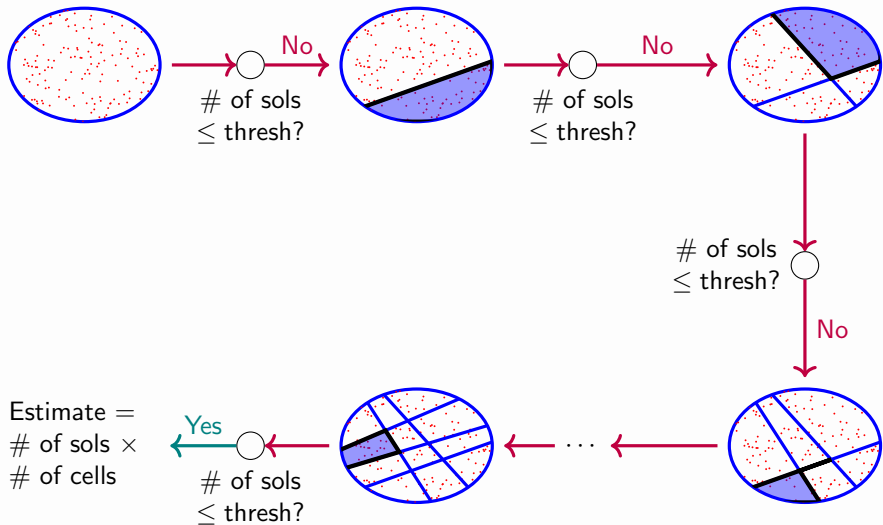
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The Secrets of Hashing-based Techniques

Challenge How do we obtain meaningful bounds on $|\text{Sol}(F)|$?

Solution : We do not need to!

Key Insight : When adding m -th XOR, theoretical analysis only requires $\frac{\sigma^2[Z_m]}{\mathbb{E}[Z_m]} \leq q$ whenever $|\text{Sol}(F)| \leq \text{thresh} \cdot 2^m$

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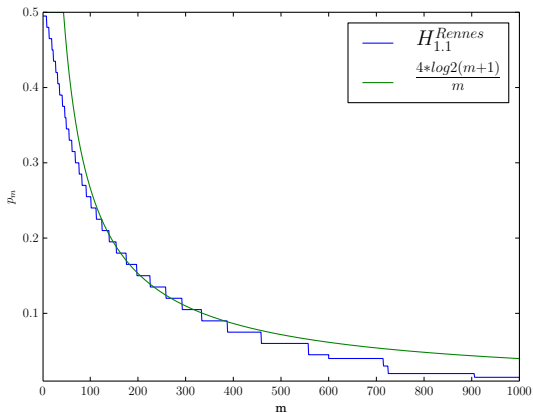
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- Add m -th XOR with $p_m = \mathcal{O}\left(\frac{\log m}{m}\right)$

Sparse Hash Functions



$H_{1.1}^{Rennes}$: Sparse hash functions that guarantee $q = 1.1$

Experimental Evaluation

Benchmark	Vars	$\log_2(\text{Count})$	ApproxMC4	ApproxMC5	Speedup
03B-4	27966	28.55	983.72	1548.96	0.64
squaring23	710	23.11	0.66	1.21	0.55
case144	765	82.07	102.65	202.06	0.51
modexp8-4-6	83953	32.13	788.23	920.34	0.86
min-28s	3933	459.23	48.63	35.83	1.36
s9234a_7_4	6313	246.0	4.77	2.45	1.95
min-8	1545	284.78	8.86	4.59	1.93
s13207a_7_4	9386	699.0	34.94	17.05	2.05
min-16	3065	539.88	33.67	16.61	2.03
90-15-4-q	1065	839.25	273.1	135.75	2.01
s35932_15_7	17918	1761.0	-	72.32	-
s38417_3_2	25528	1663.02	-	71.04	-
75-10-8-q	460	360.13	-	4850.28	-
90-15-8-q	1065	840.0	-	3717.05	-

Remember; thresh = $\mathcal{O}\left(\frac{\sigma^2[Z_m]}{\mathbb{E}[Z_m]} \cdot \frac{1}{\epsilon^2}\right)$

$\frac{\sigma^2[Z_m]}{\mathbb{E}[Z_m]} \leq 1$ for 2-wise independent; $\frac{\sigma^2[Z_m]}{\mathbb{E}[Z_m]} \leq q = 1.1$ for $H_{1.1}^{\text{Rennes}}$.

The first sparse XOR-based scheme to achieve speedup without loss of theoretical guarantees

- Hashing-based techniques employ random XORs, and promise theoretical guarantees and scalability
- The runtime of SAT solvers depend on the size of XORs
- Meaningful bounds on $\frac{\sigma^2[Z_m]}{E[Z_m]}$ via Isoperimetric inequalities.
- The first sparse XOR scheme to attain speedup improvement without loss of theoretical guarantees
- Future Directions:
 - **Theoretical** Lower bounds on the sparsity of XORs
 - **Algorithmic** Achieving speedup without slow down for any instance
 - **System** Design of Sparse XOR-based XOR solving modules
- Open-source Tool: <https://github.com/meelgroup/approxmc>