

Model Counting meets Distinct Elements

Circuits meet Data Streaming

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Corresponding publications: PODS-21 and 2023 CACM Research Highlights

Model Counting

- Given
 - Boolean variables X_1, X_2, \dots, X_n
 - Formula φ over X_1, X_2, \dots, X_n
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Problem Compute (ε, δ) approximation of $|\text{Sol}(\varphi)|$

Concern Number of NP Queries

Distinct Elements

- Given a stream $\mathbf{a} = a_1, a_2, \dots, a_m$ where $a_i \in \{0, 1\}^n$
- $DE(\mathbf{a}) = |\cup_i a_i|$
 - Also known as F_0 estimation

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- **Example** $\mathbf{a} = 1, 2, 1, 1, 2, 1, 3, 5, 1, 2, 1, 3$
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- Fundamental problem in databases with a long history of work
 - Problem** Compute (ϵ, δ) approximation of F_0
 - Concern** Space Complexity

Hashing-Based Techniques

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

Distinct Elements (FM85,AMS99,GT01,BKS02,BJKST02, CM03,CLKB04,PT07, TW12,SP09)

2-wise independent Hashing

- Let H be family of 2-wise independent hash functions mapping $\{0, 1\}^n$ to $\{0, 1\}^m$

$$\forall y_1, y_2 \in \{0, 1\}^n, \alpha_1, \alpha_2 \in \{0, 1\}^m, h \stackrel{R}{\leftarrow} H$$

$$\Pr[h(y_1) = \alpha_1] = \Pr[h(y_2) = \alpha_2] = \left(\frac{1}{2^m}\right)$$

$$\Pr[h(y_1) = \alpha_1 \wedge h(y_2) = \alpha_2] = \left(\frac{1}{2^m}\right)^2$$

2-wise independent Hash Functions

- 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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 - Expected size of each XOR: $\frac{n}{2}$
- To choose $\alpha \in \{0, 1\}^m$, set every XOR equation to 0 or 1 randomly

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

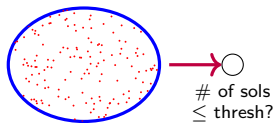
$$X_2 \oplus X_5 \oplus X_6 \cdots \oplus X_{n-1} = 1 \quad (Q_2)$$

$$\dots \quad (\dots)$$

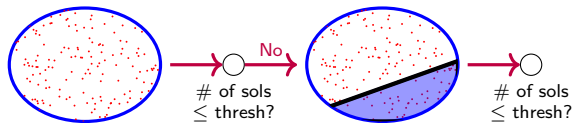
$$X_1 \oplus X_2 \oplus X_5 \cdots \oplus X_{n-2} = 1 \quad (Q_m)$$

- Therefore, $h(X) = \alpha$ can be represented as $AX = b$

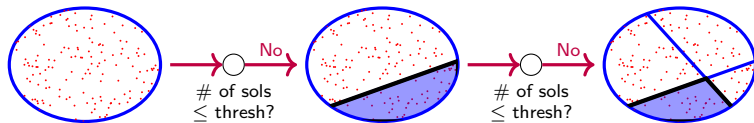
ApproxMC



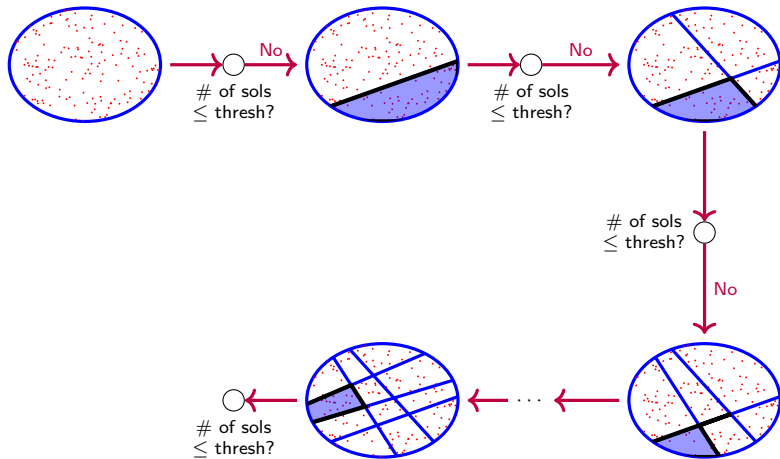
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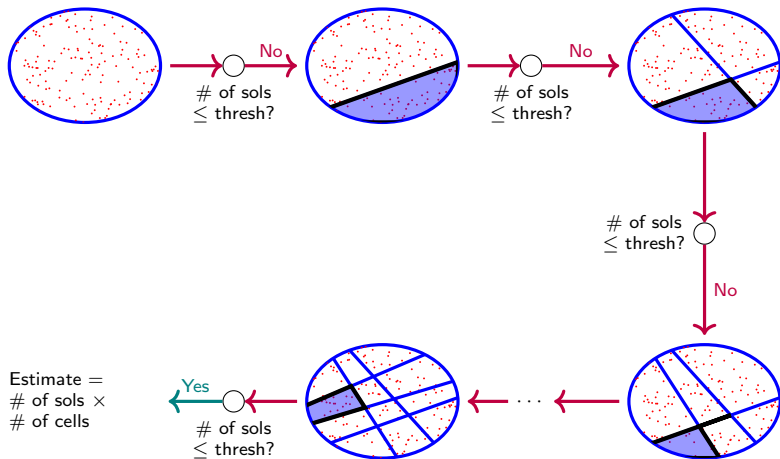
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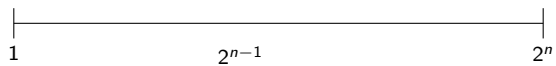
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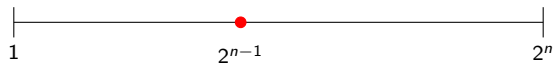
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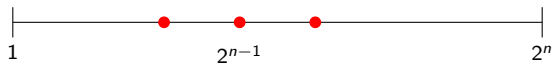
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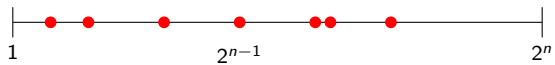
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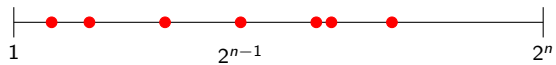
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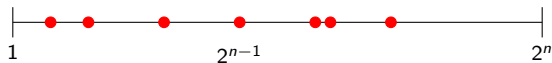
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Distinct Elements



Number of balls $\propto \frac{1}{\text{position of left most ball}}$

Algorithm DE(a)

- 1: Choose $h : \{0, 1\}^n \mapsto \{0, 1\}^n$
 - 2: minhash $\leftarrow 2^n$;
 - 3: **for** $a_i \in \mathbf{a}$ **do**
 - 4: **if** $h(a_i) < \text{minhash}$ **then**
 - 5: minhash = $h(a_i)$
 - 6: **end if**
 - 7: **end for**
 - 8: **return** $\frac{2^n}{\text{minhash}}$
-

Is there more than meets the eyes?

- From Distinct Elements to Counting
- From Counting to Distinct Elements

Hashing-based Distinct Elements

Algorithm SketchTemplate(**a**)

```
1:  $h \leftarrow$  ChooseHashFunctions
2:  $\mathcal{S} \leftarrow \{\}$ 
3: for  $a_i \in \mathbf{a}$  do
4:   ProcessUpdate( $\mathcal{S}, h, a_i$ )
5: end for
6: Est  $\leftarrow$  ComputeEst( $\mathcal{S}$ )
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Different Algorithms based on ProcessUpdate

- Minimum: Keep track of minimum $h(a_i)$
- Bucketing

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From Distinct Elements to Counting: A Two Step Recipe

\mathbf{a}_U : set of all distinct elements of the stream \mathbf{a} .

Key Idea The formula φ can be viewed as symbolic representation of some set \mathbf{a}_U such that $\text{Sol}(\varphi) = \mathbf{a}_U$.

Step 1 Capture the relationship $\mathcal{P}(\mathcal{S}, h, \mathbf{a}_U)$ between the sketch \mathcal{S} , h , and the set \mathbf{a}_U at the end of stream.

Step 2 Given a formula φ and hash function h , design an algorithm to construct sketch \mathcal{S} such that $\mathcal{P}(\mathcal{S}, h, \text{Sol}(\varphi))$ holds. And now, we can estimate $|\text{Sol}(\varphi)|$ from \mathcal{S} .

Algorithm minDE(a)

```
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2: minhash  $\leftarrow 2^n$ ;
3: for  $a_i \in \mathbf{a}$  do
4:   if minhash  $< h(a_i)$  then
5:     minhash =  $h(a_i)$ 
6:   end if
7: end for
8: return  $\frac{2^n}{\text{minhash}}$ 
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Application I: Min-based Counting Algorithm

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Bucketing-based Streaming Algorithm

Algorithm BucketDE(**a**)

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1: Choose  $h : \{0, 1\}^n \mapsto \{0, 1\}^n$ 
2:  $\ell \leftarrow 0; \mathcal{B} \leftarrow \emptyset$ 
3: for  $a_i \in \mathbf{a}$  do
4:   if  $h(a_i) \bmod 2^\ell = 0^\ell$  then
5:      $\mathcal{B}.Append(a_i)$ 
6:     if  $|\mathcal{B}| \geq \text{thresh}$  then
7:        $\ell++$ 
8:        $Filter(\mathcal{B}, h, \ell)$ 
9:     end if
10:  end if
11: end for
12: return  $|\mathcal{B}| \times 2^\ell$ 
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Elements that satisfy XOR

Add another XOR

Application II: Bucketing-based Counting Algorithm

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This is ApproxMC!

From Distinct Elements to Counting: Implications

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Theorem (FPRAS)

If construction of sketch \mathcal{S} is in PTIME for a class of formulas, then there is FPRAS for the corresponding class. E.g.: DNF, Union of XORs

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Theorem (Lower Bounds)

Lower bounds for Distributed Streaming translate to lower bounds for Distributed DNF counting

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- ApproxMC is FPRAS for DNF formulas

(CMV16,MSV17,MSV18)

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- Encompasses models such as ranges, affine spaces
- Application: Distinct Elements over Range
 - Every item $[a_i, b_i]$ can be represented using a DNF formula.
 - So just apply FPRAS for DNF

Conclusion

Summary

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Future Directions

- Practical scalability of newly devised counting techniques
- What's the relationship for other problems between circuits/formulas and streaming ?
 - Higher moments
 - Entropy