# Efficiently Enumerating Hitting Sets of Hypergraphs Arising in Data Profiling

#### Martin Schirneck

Joint work with Thomas Bläsius, Tobias Friedrich, Julius Lischeid, and Kitty Meeks, to appear at ALENEX 2019.

Dagstuhl - 16 October 2018







Data profiling is the gathering of metadata from databases.

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**Task:** Enumerate all inclusion-wise minimal UCCs.



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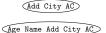


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### UCCs and the Transversal Hypergraph

There is a parsimonious polynomial reduction that preserves inclusions...

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- ...from HITTINGSET to UCC. [Bläsius, Friedrich & Sch. 2016]



#### Notation:

- n = number of vertices/attributes.
- m = number of hyperedges/minimal difference sets.
- $k^* = \text{size of the largest minimal hitting set/UCC}$ .



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- ...uses space O(mn).
- …is fast in practice!



a.k.a. the flashlight technique. [Mary & Strozecki 2016]





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Idea: decision tree pruned by an extension oracle.

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- Can X be extended to a minimal hitting set avoiding Y?



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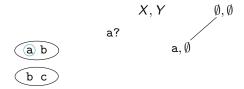
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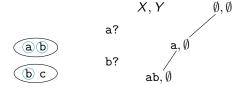
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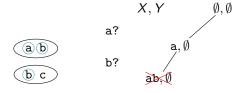
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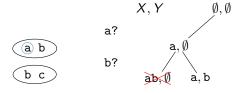
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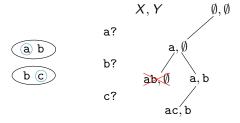
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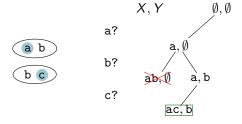
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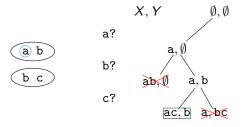
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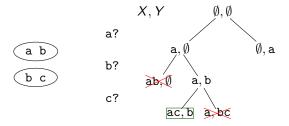
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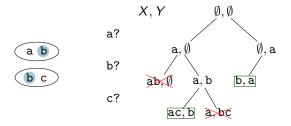
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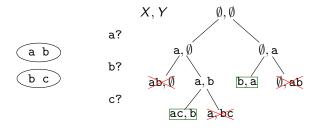
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## **Backtracking Enumeration**

a.k.a. the flashlight technique. [Mary & Strozecki 2016]

Idea: decision tree pruned by an extension oracle.



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#### Extension Problem for Minimal Hitting Sets

Let X, Y be disjoint set of vertices,  $X \cap Y = \emptyset$ .

- (i) Is there a minimal hitting set H s.t.  $X \subseteq H$  and  $H \cap Y = \emptyset$ ?
- (ii) If so, is H = X?



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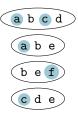
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- Under ETH: not solvable in time  $f(|X|) \cdot (m+n)^{o(|X|)}$  for any f.







(a) b (c) d

 $X = \{a,c\}$ 

a b e

(bef)

(c) d e

(a) d

b c d

(bde



- (a b c d
  - (a) h e
  - b e f
  - © d e
    - (a) d
  - $b \odot d$
  - (b d €

- $X = \{a,c\}$
- $Y = \{b\}$







e f



# HPI

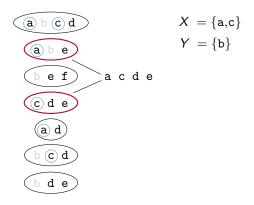
# Finding the True Witnesses

c d

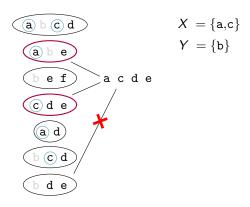
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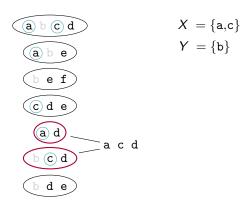




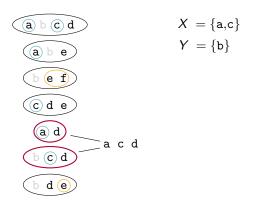




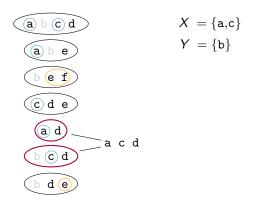






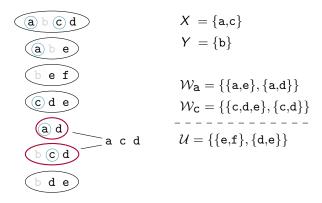






- Exactly one element of X: potential witness.
- X is extendable iff there are potential witnesses (E<sub>x</sub>)<sub>x∈X</sub> s.t. the union ⋃<sub>x∈X</sub> E<sub>x</sub> does not contain an unhit edge.





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```
1 if X = \emptyset then
      if V \setminus Y is a hitting set then return EXTEND.;
     else return NOT EXTEND.;
```



```
4 initialise set system \mathcal{U} = \emptyset;
5 foreach x \in X do initialise set system \mathcal{W}_x = \emptyset;
6 foreach edge E do
       if E \cap X = \{x\} then add E \setminus Y to W_x;
   if E \cap X = \emptyset then add E \setminus Y to \mathcal{U};
```



```
9 if \exists x \in X : \mathcal{W}_x = \emptyset then return NOT EXTEND.;
```



```
9 if \exists x \in X : \mathcal{W}_x = \emptyset then return NOT EXTEND.;
10 if \mathcal{U} = \emptyset then return MINIMAL;
```



```
11 foreach (E_{x_1}, \ldots, E_{x_{|X|}}) \in \mathcal{W}_{x_1} \times \cdots \times \mathcal{W}_{x_{|X|}} do
      W \leftarrow \bigcup_{i=1}^{|X|} E_{x_i};
      if \forall U \in \mathcal{U} : U \nsubseteq W then return EXTEND.;
14 return NOT EXTEND.:
```

# HPI

## From Run Time...

• Dominant brute-force phase:  $O(m^{|X|} \cdot mn)$ .

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# ...to Delay

**Claim:** Largest solution has constant size  $k^* \Rightarrow$  polynomial delay.



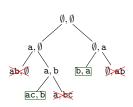
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# НРІ

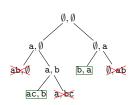
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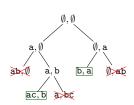
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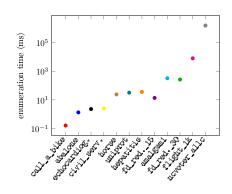
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- Maximum distance between leaves in O(n).
- If |X| ≥ k\*, oracle answer is either NOT EXTENDABLE or MINIMAL.
- Maximum delay of  $O(n) \cdot O(m^{k^*+1}n)$ .



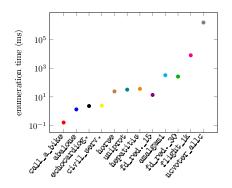


Setup: 10+2 databases on 2x 2.60GHz CPUs & 256GB RAM.





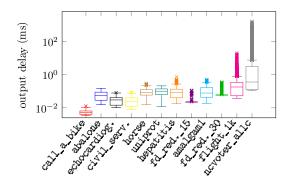
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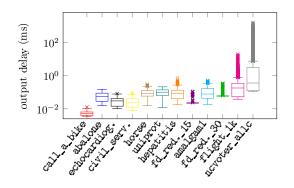
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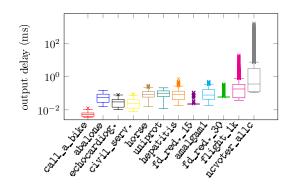


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# HPI

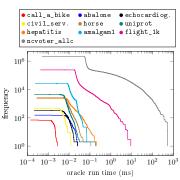
## Theory and Practice

Setup: 10+2 databases on 2x 2.60GHz CPUs & 256GB RAM.

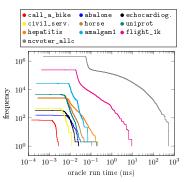


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- ncvoter\_allc (88 cols., 100k rows): n = 82, m = 448,  $k^* = 15$ .
  - Maximum delay of 1.6s, but median at 0.35ms.



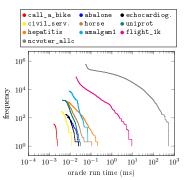






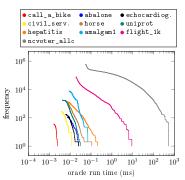
• Half the calls are trivial  $(X = \emptyset)$  or easy  $(\mathcal{W}_x = \emptyset)$  or  $\mathcal{U} = \emptyset$ .





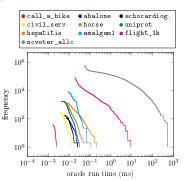
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**Practice:** The algorithm rarely hits the worst case.



- 1. Hitting set enumeration with polynomial delay is possible if the largest solution has constant size.
- 2. The extension oracle is a natural W[3]-complete problem.
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#### Thank you.