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# Topic 3: Efficient query evaluation Unit 2: Cyclic queries Lecture 19

Wolfgang Gatterbauer

CS7240 Principles of scalable data management (sp22)

https://northeastern-datalab.github.io/cs7240/sp22/

3/29/2022

## Outline: T3-2: Cyclic conjunctive queries

- T3-1: Acyclic conjunctive queries
- T3-2: Cyclic conjunctive queries
  - 2SAT (a detour)
  - Tree decompositions
  - Decompositions of hypertrees
  - Duality in Linear programming (a quick primer)
  - AGM bound (maximal result size for full CQs)
  - Worst-case optimal joins & the triangle query
  - Worst-case optimal joins & the 4-cycle
  - Optimal joins & the 4-cycle

cycles make everything more complicated ⊗

#### Why cyclic queries (other than social networks)

Likes(person, drink) Frequents(person, bar) Serves(bar, drink, cost)

#### 2. Specify or choose a Query

Supported grammar

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104 Bars: Persons who frequent some bar that serves some drink they like.

#### Why cyclic queries (other than social networks)

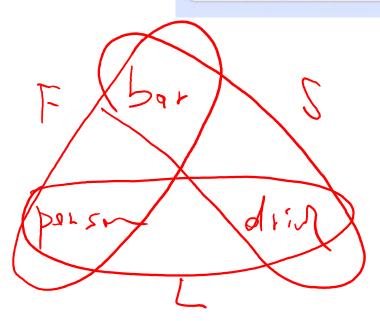
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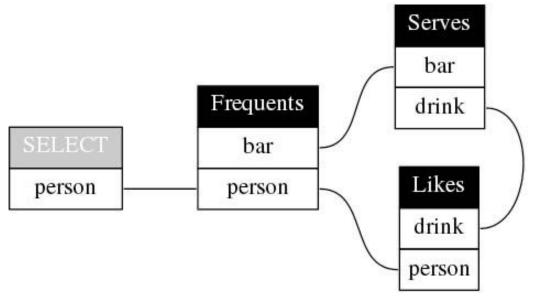
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#### Source: http://demo.queryvis.com

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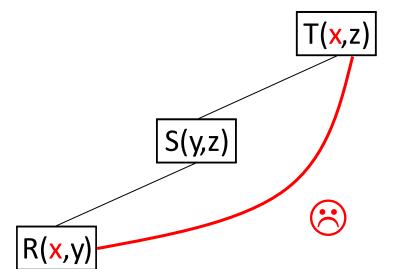
SELECT F1.person FROM Frequents F1 exists WHERE (SELECT \* FROM Serves S2 WHERE S2.bar = F1.barAND exists (SELECT \* Likes L3 FROM WHERE L3.person = F1.personS2.drink = L3.drink)) AND

#### Joins in databases: one-at-a-time

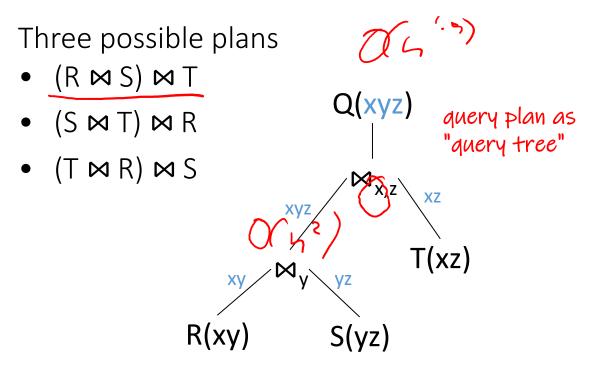
How can we efficiently process multi-way joins with cycles?

Q(x,y,z) := R(x,y), S(y,z), T(x,z).

Recall:



There is no join tree! You can't fulfill the running intersection property...



 $\overline{\mathbf{S}}$ 

- there is no full semijoin reducer
- intermediate result size bigger than output

#### Can we do better for cyclic queries? ③

#### Outline: T3-2: Cyclic conjunctive queries

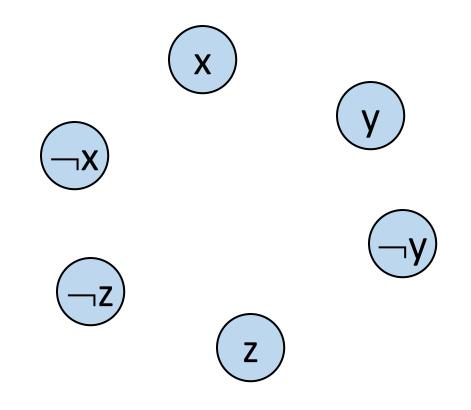
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#### 2SAT

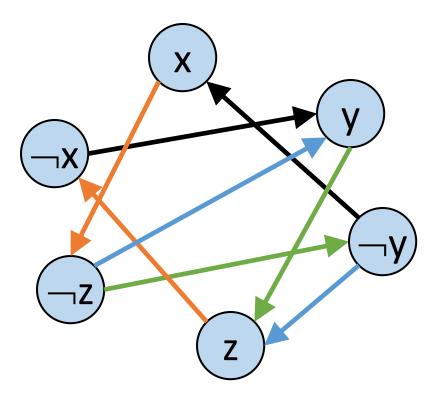
$$\varphi = (x \lor y) \land (\neg y \lor z) \land (\neg x \lor \neg z) \land (z \lor y)$$

- Instance: A 2-CNF formula  $\boldsymbol{\phi}$
- Problem: To decide if  $\boldsymbol{\phi}$  is satisfiable
- Theorem: 2SAT is polynomial-time decidable.
  - Proof: We'll show how to solve this problem efficiently using path searches in graphs...
- Background: Given a graph G=(V,E) and two vertices s,t∈V, finding if there is a path from s to t in G is polynomial-time decidable. Use some search algorithm (DFS/BFS).

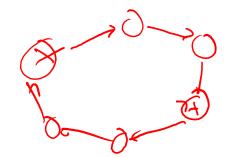
• Vertex for each variable and a negation of a variable

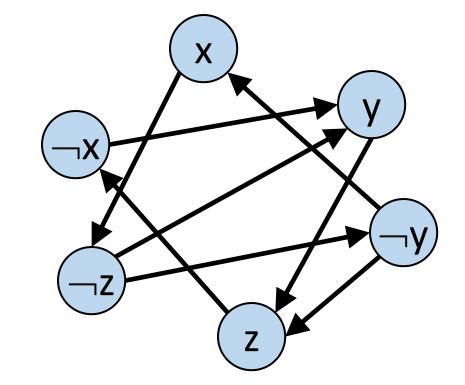


- Vertex for each variable and a negation of a variable
- Edge  $(\neg x \rightarrow y)$  iff there exists a clause equivalent to  $(x \lor y)$ 
  - Recall  $(x \lor y)$  same as  $(\neg x \Rightarrow y)$  and  $(\neg y \Rightarrow x)$ , thus also  $(\neg y \rightarrow x)$

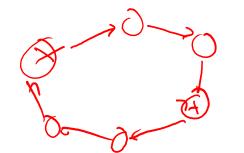


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- Claim: a 2-CNF formula  $\phi$  is unsatisfiable iff there exists a variable x, such that:
  - there is a path from x to  $\neg x$  in the graph, and
  - there is a path from  $\neg x$  to x in the graph





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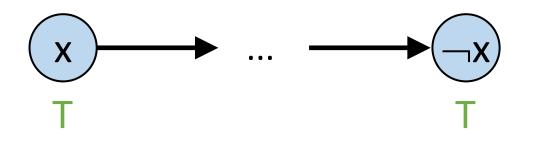
not enough, needs both directions! X



#### Correctness (1)

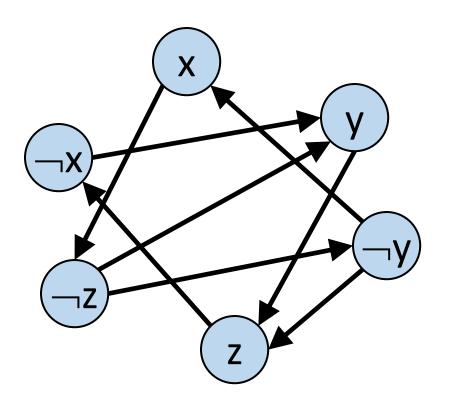
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- Suppose there are paths x..¬x and ¬x..x for some variable x, but there's also a satisfying assignment ρ.
  - If  $\rho(x)=T$ :



– Similarly for  $\rho(x)=F...$ 

recall, needs to hold in both directions!



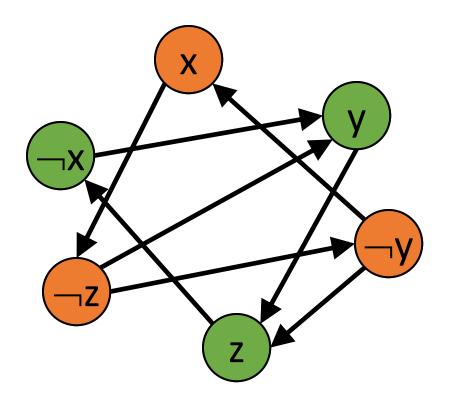
#### Correctness (2)

$$\varphi = (x \lor y) \land (\neg y \lor z) \land (\neg x \lor \neg z) \land (z \lor y)$$

- Suppose there are no variables with such paths.
- Construct an assignment as follows:

1. pick an unassigned literal  $\alpha$ , with no path from  $\alpha$  to  $\neg \alpha$ , and assign it T

- 2. assign T to all reachable vertices
- 3. assign F to their negations
- 4. Repeat until all vertices are assigned



#### 2SAT is in P

We get the following PTIME algorithm for 2SAT:

- For each variable x find if there is a path from x to  $\neg x$  and vice-versa.
- Reject if any of these tests succeeded.
- Accept otherwise.

#### $\Rightarrow$ 2SAT $\in$ P.

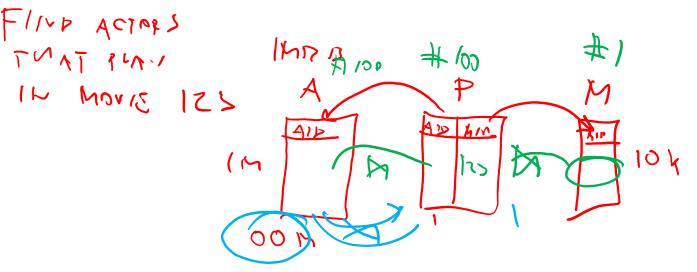
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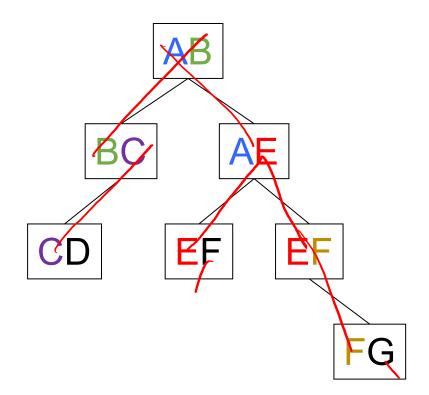
#### Join Processing: two approaches

#### 1. Cardinality-based

- binary joins, consider the sizes of input relations as to reduce the intermediate sizes
- commercial DBMSs: series of pairwise joins, system R (Selinger), join size estimation
- 2. Structural approaches (next)
  - acylicity: Yannakakis, GYO algorithm, join tree
  - bounded "width": query width, hypertree width (hw), generalized hw (ghw). All go back to notion of treewidth (work by Robertson & Seymour on graph minors)
- AGM: fractional hw (fhw):
  - consider both statistics on relations and query structure



#### Definition of an <u>attribute-connected</u> tree



DEFINITION: A tree is attributeconnected if the subtree induced by each attribute is connected

Same as the running intersection property from join trees

Also called "coherence"

#### Tree decomposition

A tree decomposition of graph G(N, E) is a tree T(V, F) and a subset

 $N_v \subseteq N$  assigned to each vertex (or "supernode")  $v \in V$  s.t.:

(1) Node coverage: Every vertex of G is assigned at least one vertex in T

(2) Edge coverage: For every edge e of G, there is a vertex in T that contains both ends of e

(3) Coherence: The tree is "attribute-connected"

The width of a tree decomposition is the size of its largest set minus one

## Tree decomposition example 1: a tree



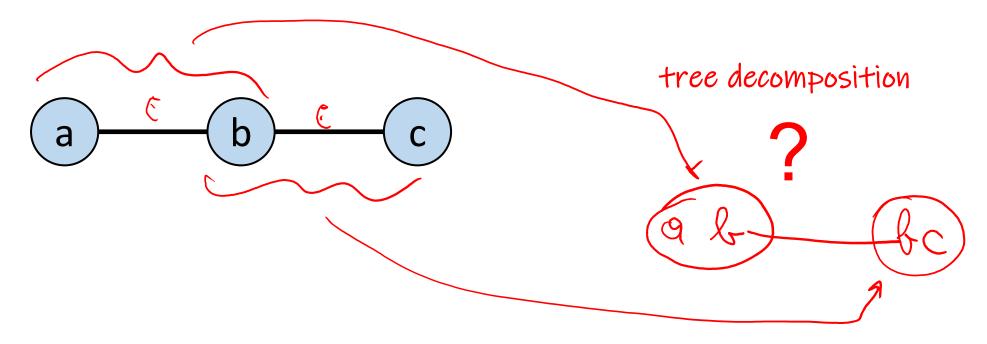
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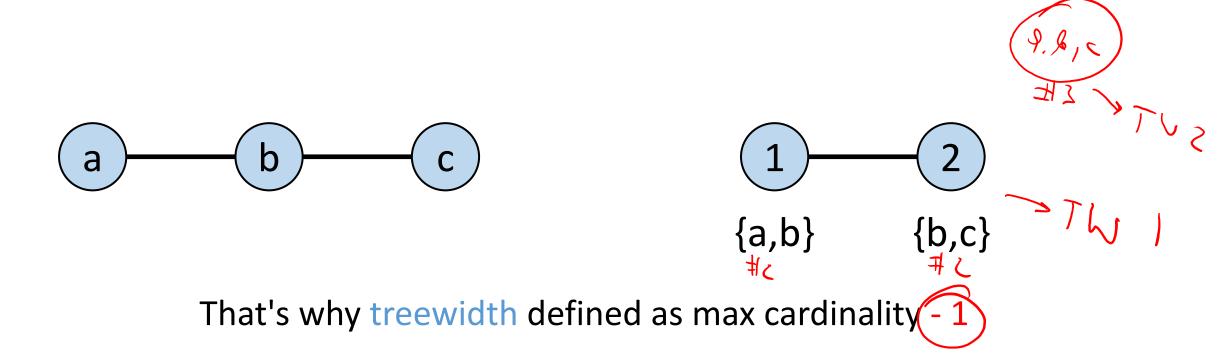
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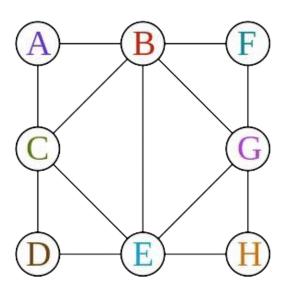
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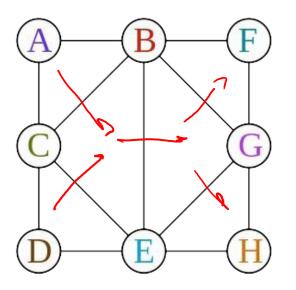
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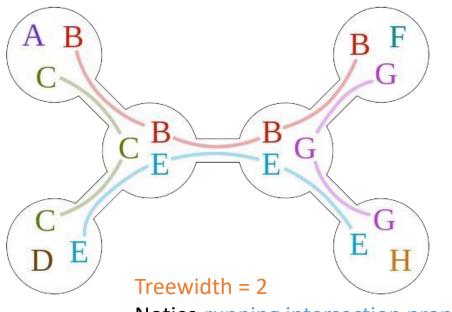
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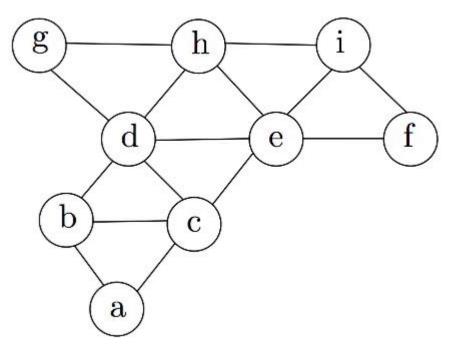




Notice running intersection property



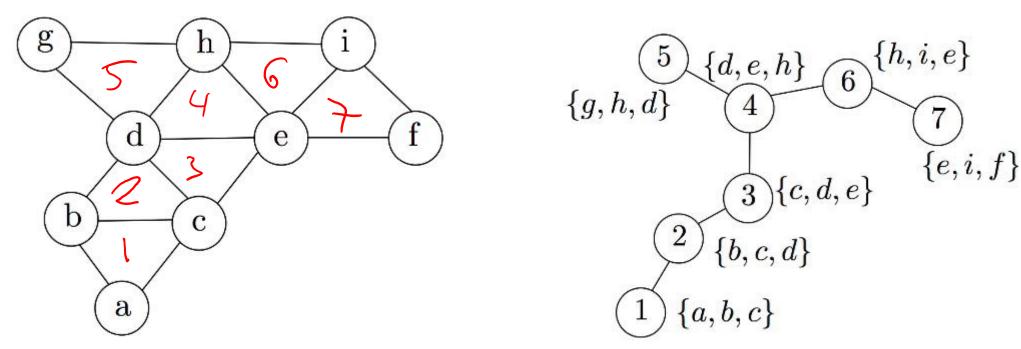
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Example from: https://www.mi.fu-berlin.de/en/inf/groups/abi/teaching/lectures/lectures past/WS0910/V Discrete Mathematics for Bioinformatics P1/material/scripts/treedecomposition1.pd Wolfgang Gatterbauer. Principles of scalable data management: https://northeastern-datalab.github.io/cs7240/

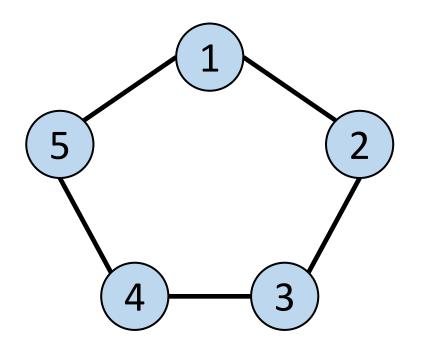
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Wolfgang Gatterbauer. Principles of scalable data management: <a href="https://northeastern-datalab.github.io/cs7240/">https://northeastern-datalab.github.io/cs7240/</a>



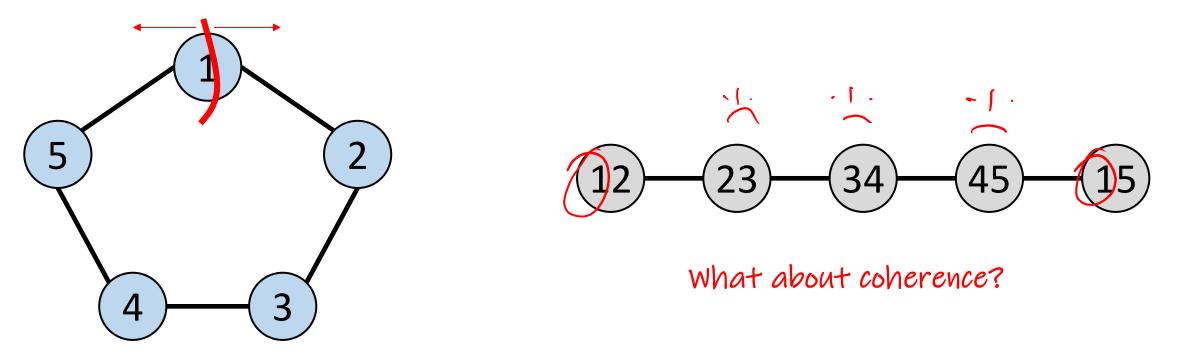
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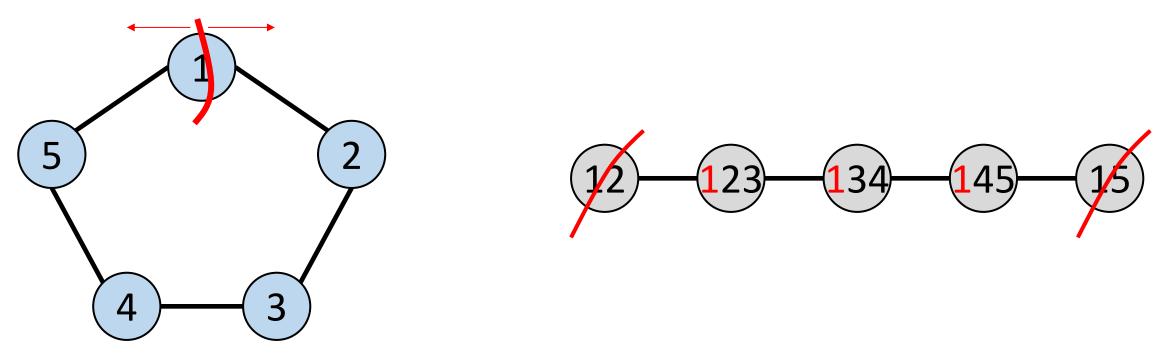


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