#### CPSC 768: Scalable and Private Graph Algorithms

Lecture 20: Dynamic Graph Algorithms

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**CPSC 768** 

#### Announcements

- Final project report and presentation: April 24<sup>th</sup> (last day of class)
  - Final project presentation is a 30 min presentation

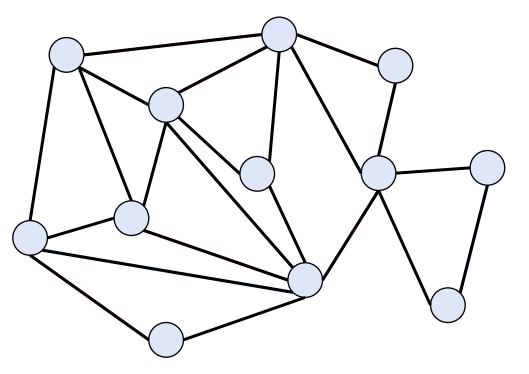
#### **Dynamic Graph Algorithms**

 Updates to the graph occur where edges are added and deleted from the graph

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Edge insertions/deletions arrive sequentially

Maintain graph property after each update



#### Minimize Update Time

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Sublinear Runtime: strive for poly(log n)

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#### Minimize Update Time

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- Sometimes need to do preprocessing
  - Small polynomial in the input graph
- Sometimes have queries (e.g. connectivity queries)

Sublinear Runtime: strive for poly(log n)

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- Dynamic maximum matching (find a matching of maximum size):
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- Approximate Densest Subgraph:
  - **Best known:** poly(log *n*) update time [SW STOC '20, CCHHQRS SODA '24]

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- Dynamic meets distributed:
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- Learning-augmented Dynamic Algorithms [LS '23, BFNP '23, HSSY '23]

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  - Polynomial or exponential gaps in runtimes

**Best Fully Dynamic** 

**Best Partially Dynamic** 

Planar Digraph APSP	$\widetilde{O}\left(n^{2/3} ight)$	[FR06, Kle05]	$\widetilde{O}(\sqrt{n})$	[DGWN22]
Triconnectivity	$O(n^{2/3})$	[GIS99]	$\widetilde{O}\left(1 ight)$	[HR20, PSS17]
k-Edge Connectivity	$n^{o(1)}$	[JS22]	$\widetilde{O}(1)$	$[CDK^+21]$
Dynamic DFS Tree	$\widetilde{O}\left(\sqrt{mn} ight)$	[BCCK19]	$\widetilde{O}\left(n ight)$	[BCCK19, CDW <sup>+</sup> 18]
Minimum Spanning Forest	$\widetilde{O}(1)$	[HDLT01]	$\widetilde{O}(1)$	[Epp94]
APSP	$igg(rac{256}{k^2}igg)^{4/k} ext{-Approx} \ \widetilde{O}\left(n^k ight)  ext{ update} \ \widetilde{O}(n^{k/8})  ext{ query}$	[FGNS23]	$(2r-1)^k ext{-} ext{Approx}\ \widetilde{O}\left(m^{1/(k+1)}n^{k/r} ight)$	$[CGH^+20]$
AP Maxflow/Mincut	$O(\log(n)\log\log n) ext{-}\operatorname{Approx} \ \widetilde{O}\left(n^{2/3+o(1)} ight)$	$[CGH^+20]$	$O\left(\log^{8k}(n) ight)$ -Approx. $\widetilde{O}\left(n^{2/(k+1)} ight)$	[Gor19, GHS19]
MCF	$egin{array}{llllllllllllllllllllllllllllllllllll$	[CGH <sup>+</sup> 20]	$egin{aligned} O(\log^{8k}(n)) ext{-}\operatorname{Approx.} \ \widetilde{O}\left(n^{2/(k+1)} ight)  ext{ update} \ \widetilde{O}(P^2)  ext{ query} \end{aligned}$	[Gor19, GHS19]
Strongly Connected Components	$\Omega(m^{1-\varepsilon})$ query or update	[AW14]	$\widetilde{O}(m)$	[Rod13]
	$2^{O(\log^{5/6}(n))}$ -Approx $2^{O(\log^{5/6}(n))}$ update		$O\left(\log^{8k}(n) ight) ext{-}\mathrm{Approx}\ \widetilde{O}\left(n^{2/(k+1)} ight)$	
Uniform Sparsest Cut	$O(\log^{1/6}(n))$ query	[GRST21]	O(1) query	[Gor19, GHS19]
Submodular Max	$\widetilde{O}(k^2)$	$[\mathrm{DFL}^+23]$	$egin{aligned} 0.3178 ext{-} ext{Approx}\ \widetilde{O}\left( ext{poly}(k) ight) \end{aligned}$	$[FLN^+22]$

[**L**S '23]

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- Amortized runtimes:
  - Lazy updates strategy where updates are delayed and processed all at once

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    - Rerun maximal matching static algorithm after  $\varepsilon \cdot n$  updates
    - Amortized update time:  $O\left(\frac{m}{\epsilon n}\right) = o(n)$  when m = o(n)

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  - Oblivious:
    - Sequence of updates determined before algorithm starts
    - Updates come one at a time online

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  - Adaptive:
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- Large gap between oblivious and adaptive adversaries:
  - Example: dynamic connectivity, polynomial deterministic worst-case, polylog oblivious worst-case

# **Dynamic Connectivity**

- Offline Dynamic [Eppstein '92]
  - Including offline dynamic minimum spanning tree
- Oblivious [Kapron-King-Mountjoy SODA '13]
- Deterministic [Frederickson's Algorithm '85]

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- Oblivious [Kapron-King-Mountjoy SODA '13]
- Deterministic [Frederickson's Algorithm '85] (classic, won't discuss today—similar theme to newer algorithms: <u>http://www.cs.cmu.edu/afs/cs.cmu.edu/academic/class/15850-f20/www/notes/lec3.pdf</u>)

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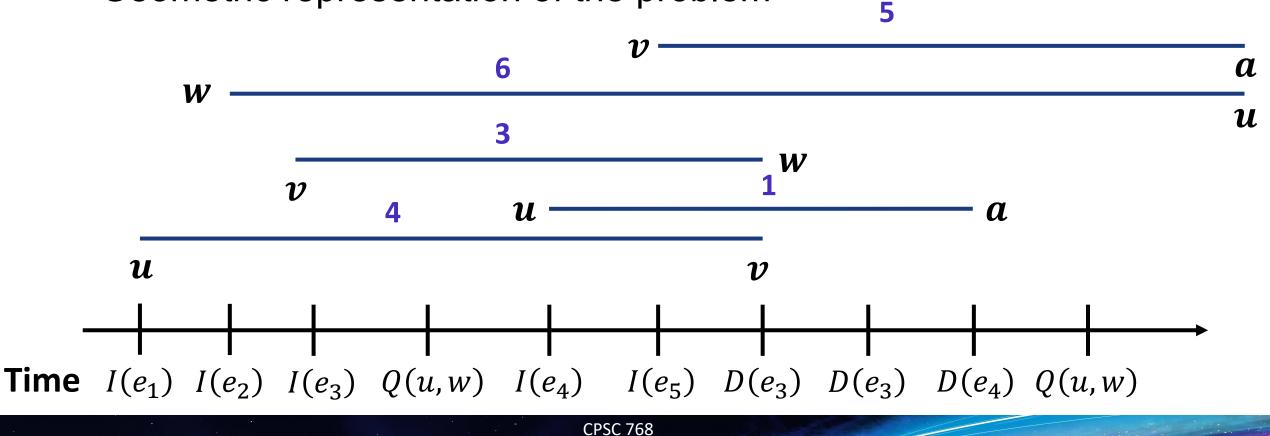
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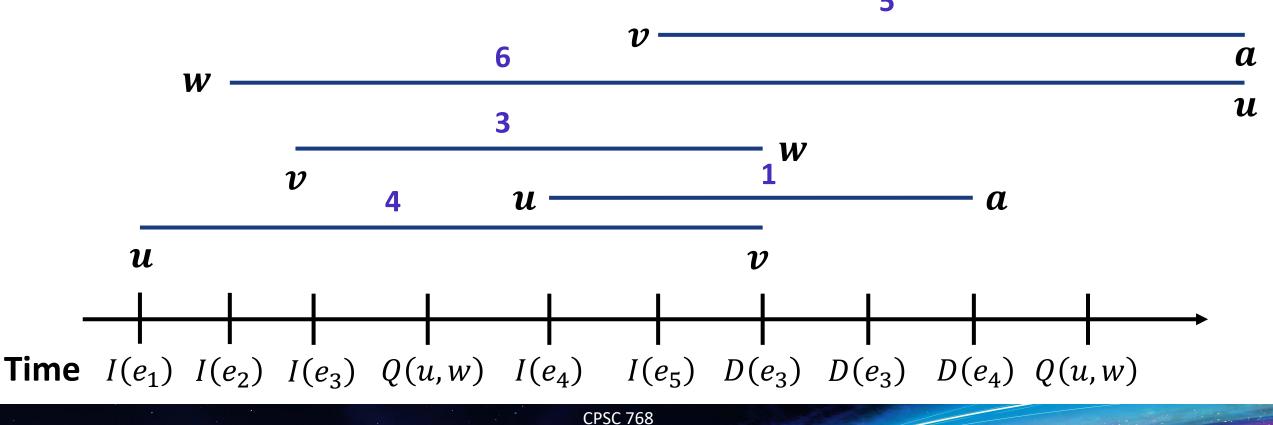
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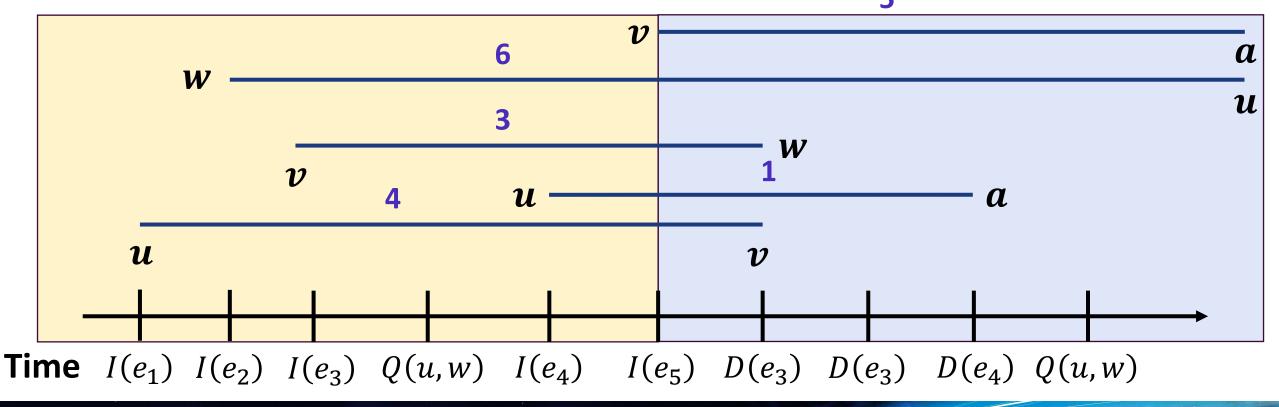
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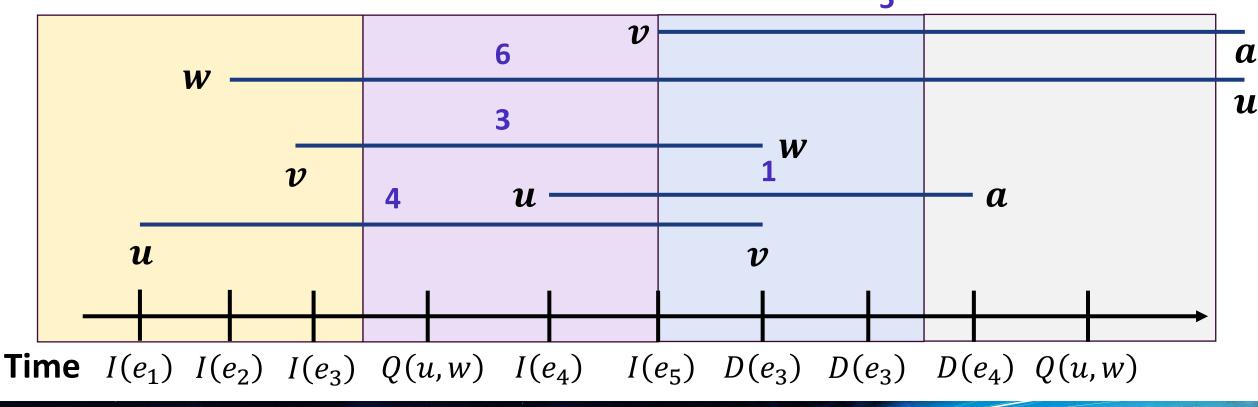
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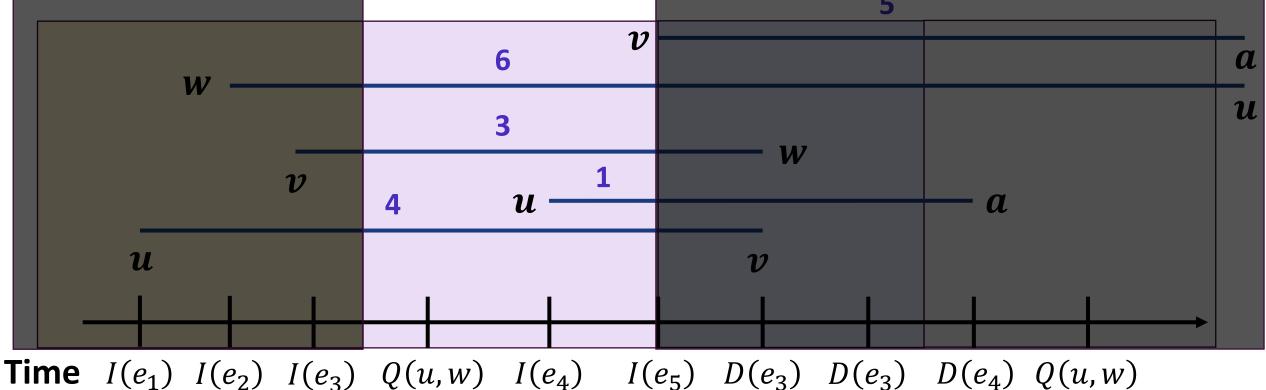
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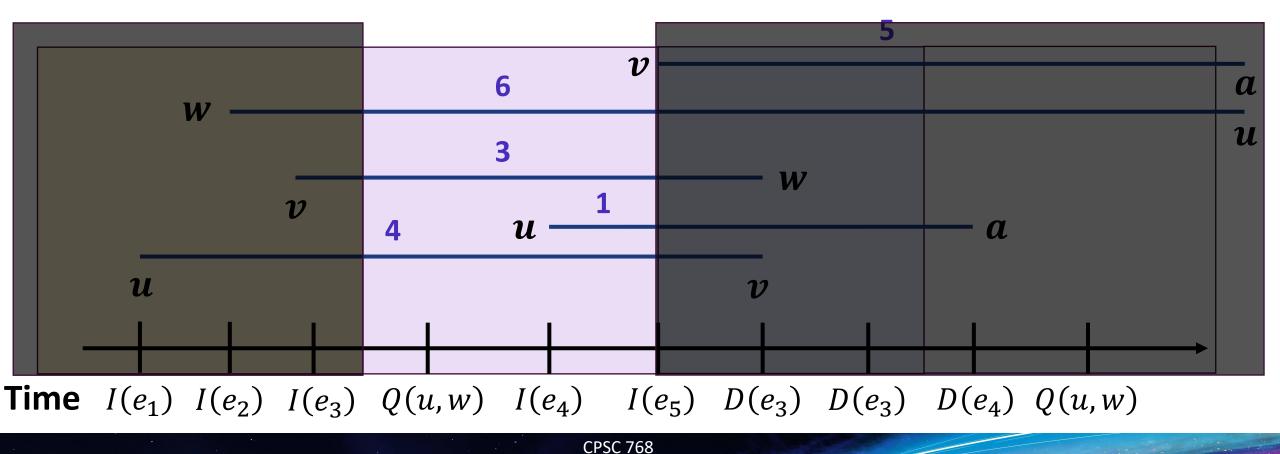
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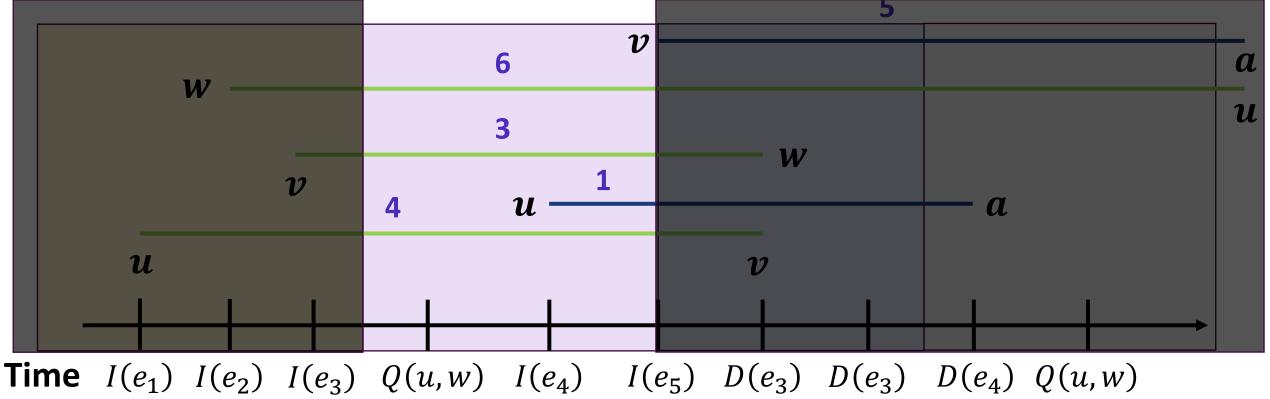
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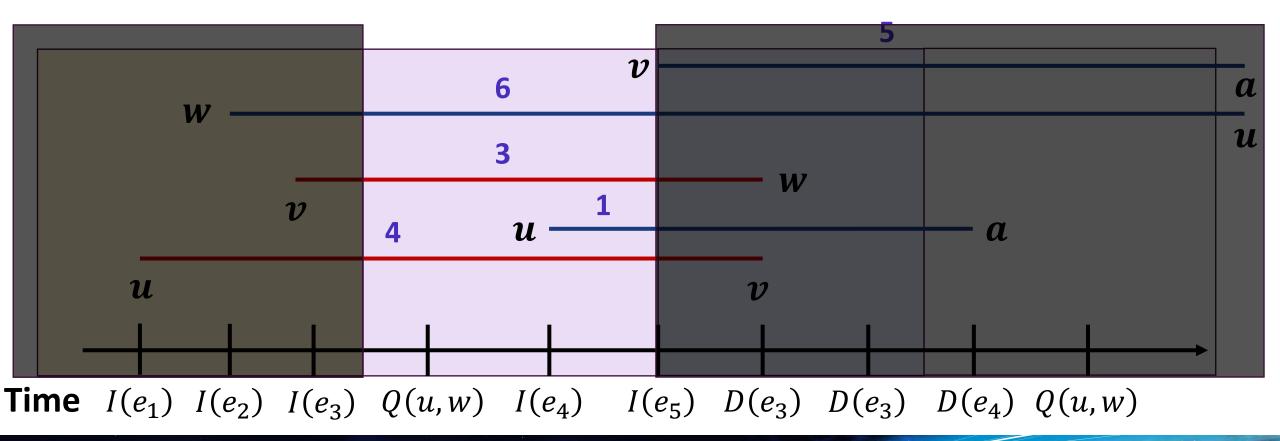
First, consider all permanent edges (edges that go across the subproblem)



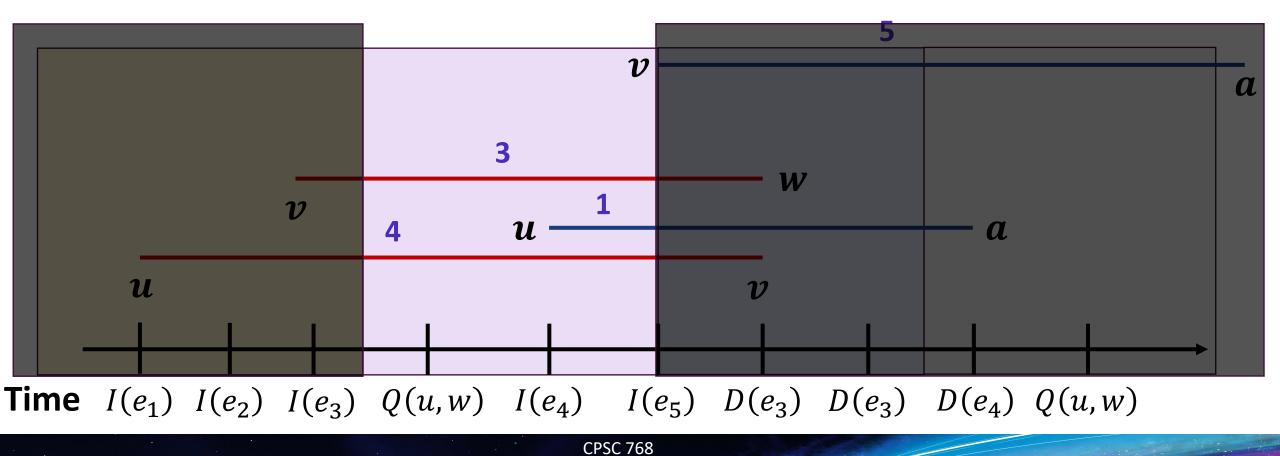
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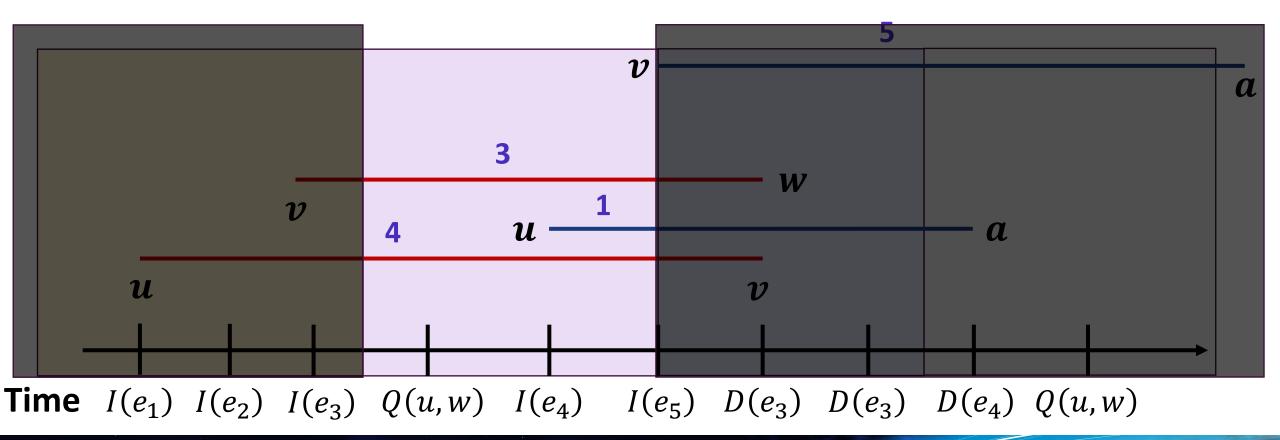
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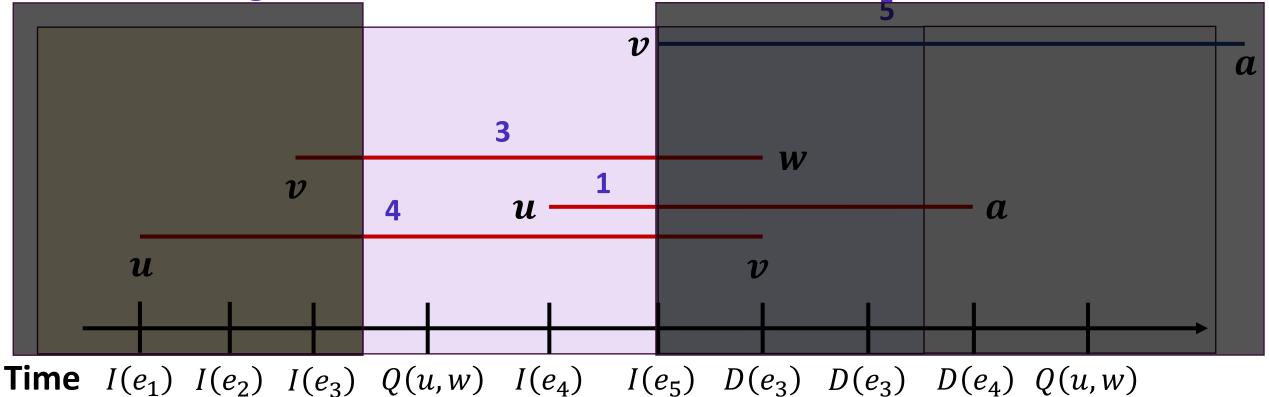
- Run any linear time MST algorithm on all considered edges
- Red edges are in the MST; Delete permanent edges not red



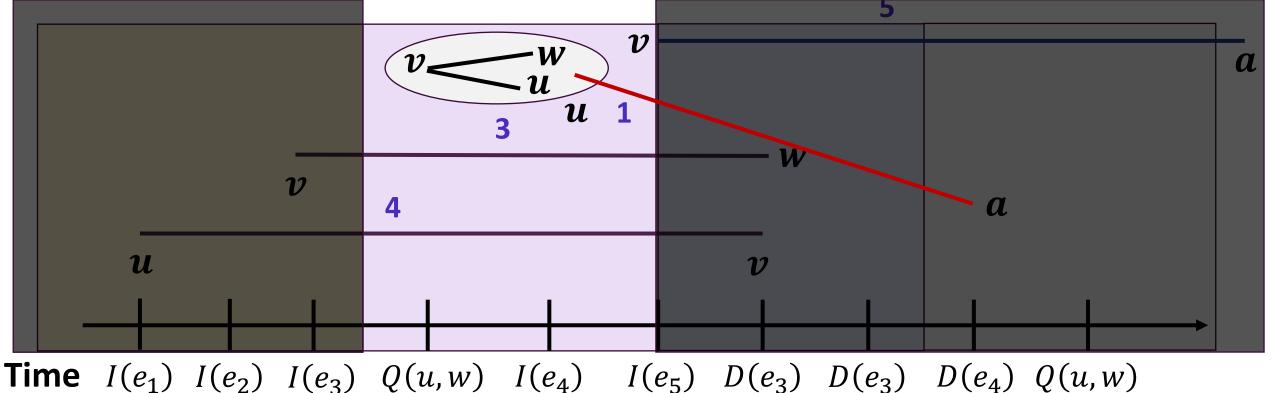
- Red edges are in the MST; Delete permanent edges not red
- Now consider all edges in subproblem



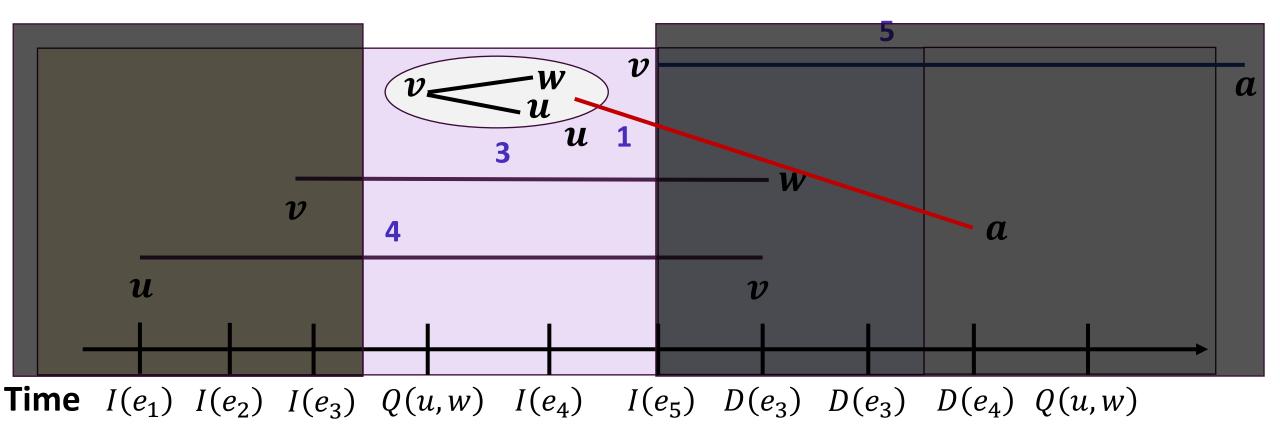
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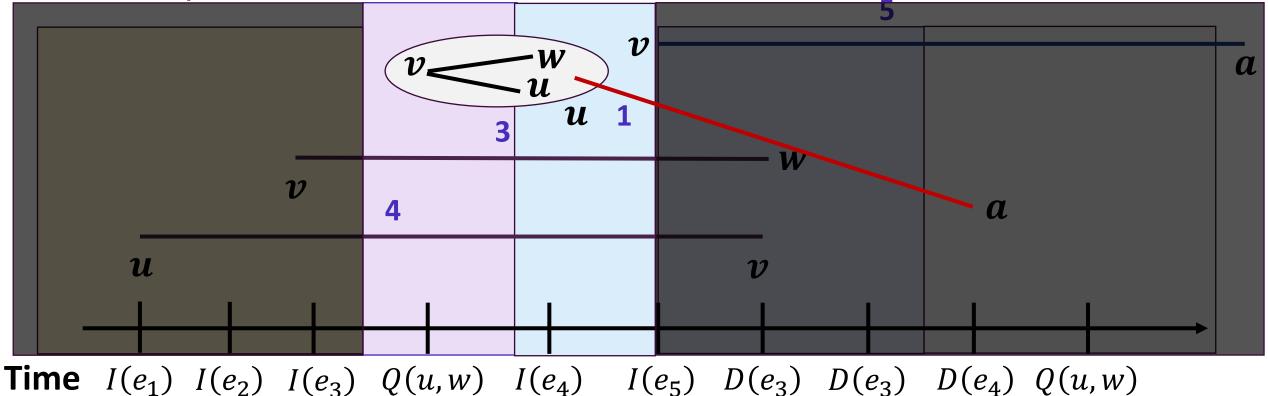
- Now consider all edges in subproblem; Run any linear time MST algorithm
- Contract any permanent edges in the MST; link-cut tree



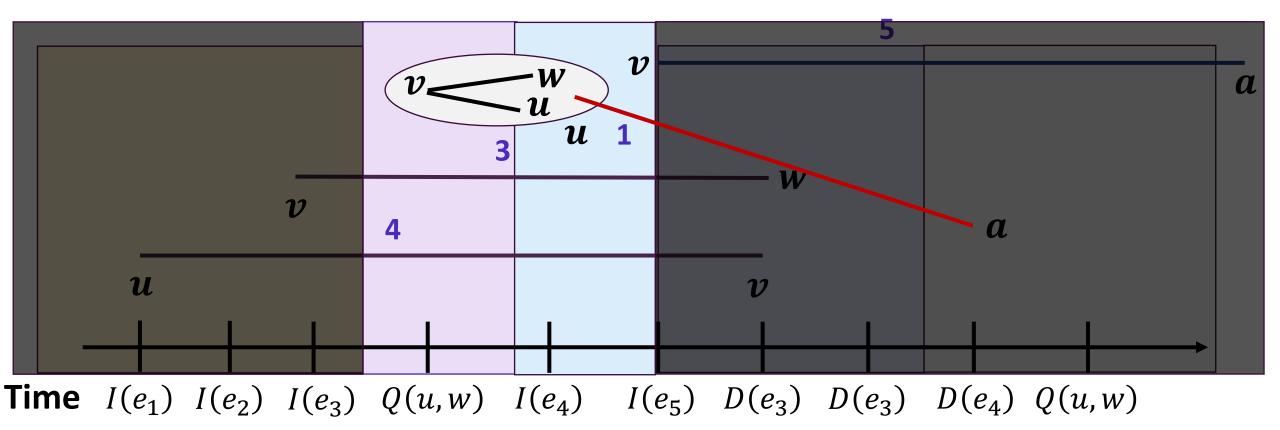
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- Pass data structure to next smaller subproblem (persistence)
- Consider non-contracted and not deleted edges in subproblem



 For queries, look at the data structure and edges of smallest subproblem containing the query



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Total Runtime:  $O(T \log(T))$  by

**Master Theorem** 

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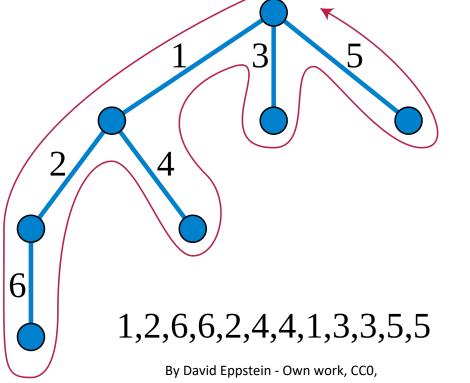
- Level Data Structure Algorithm of [Kapron-King-Mountjoy SODA '13]
- High-Level Idea:
  - Data structure for quickly determining: given a cut if there's an edge (whp) going in between the cut
  - Data structure for maintaining connected vertices
  - Easy access to determine if vertices are in the same connected component

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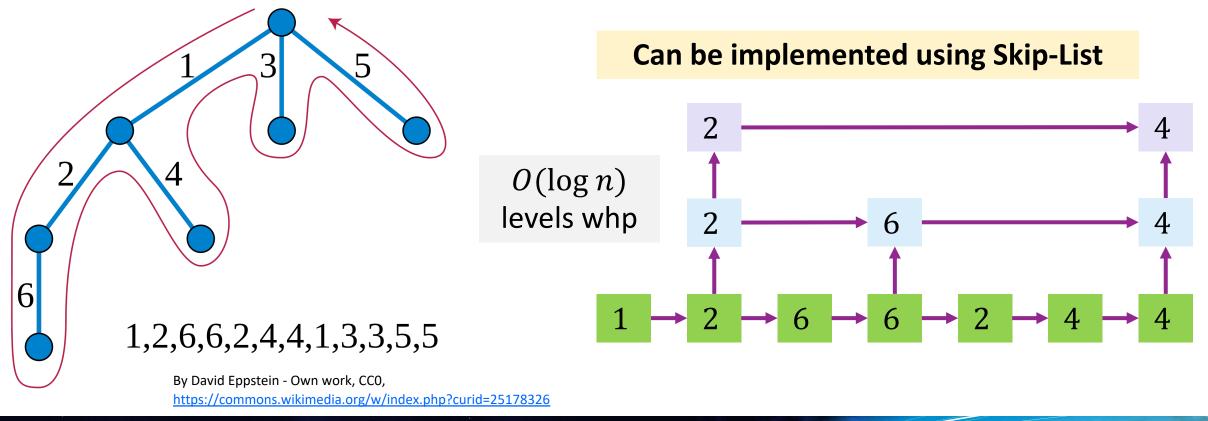
- Data structure:
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    - Break a cycle in  $O(\log n)$  time
    - Find SUM or XOR (any commutative, associative operation) of subtree in  $O(\log n)$  time

• Euler tour tree (Tarjan-Vishkin '94) high level description:

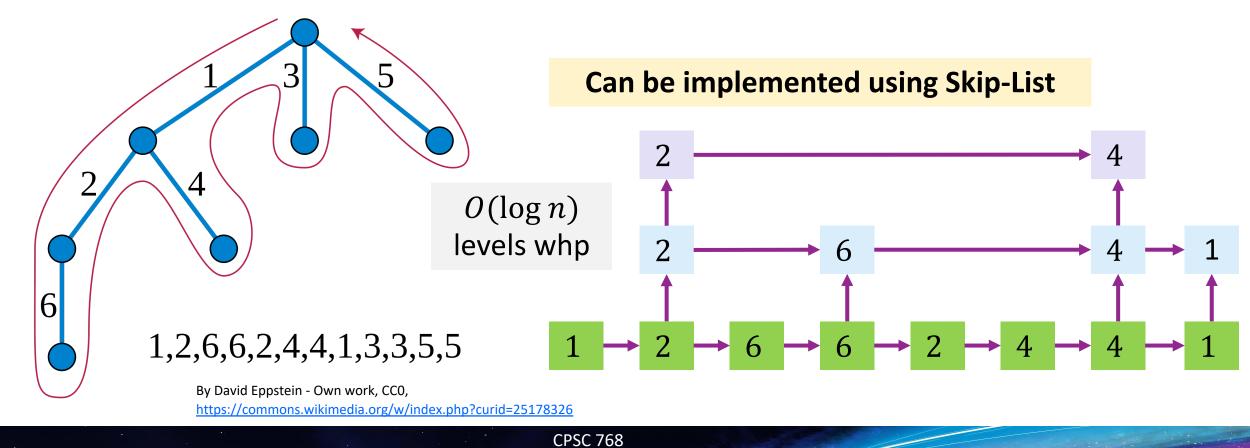


https://commons.wikimedia.org/w/index.php?curid=25178326

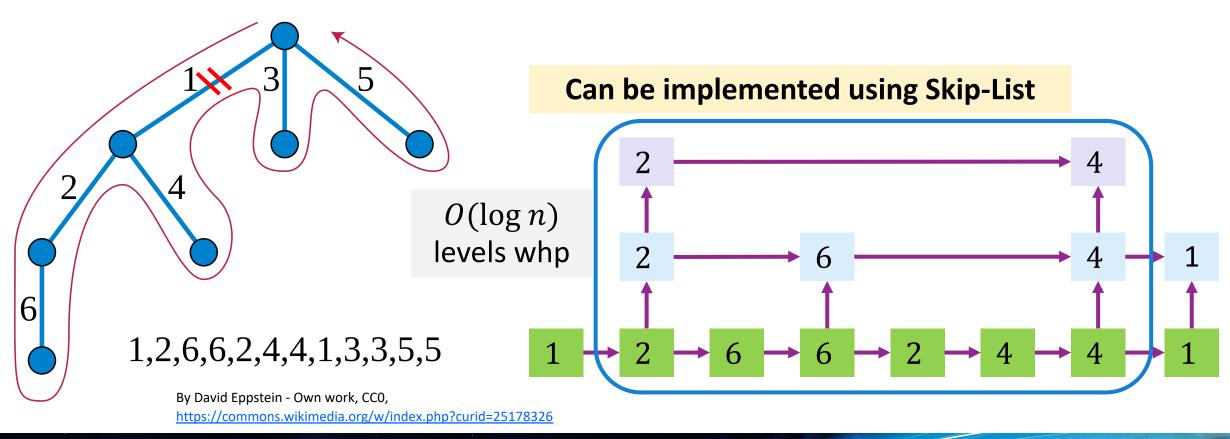
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Euler tour tree allows you to remove a subtree very easily



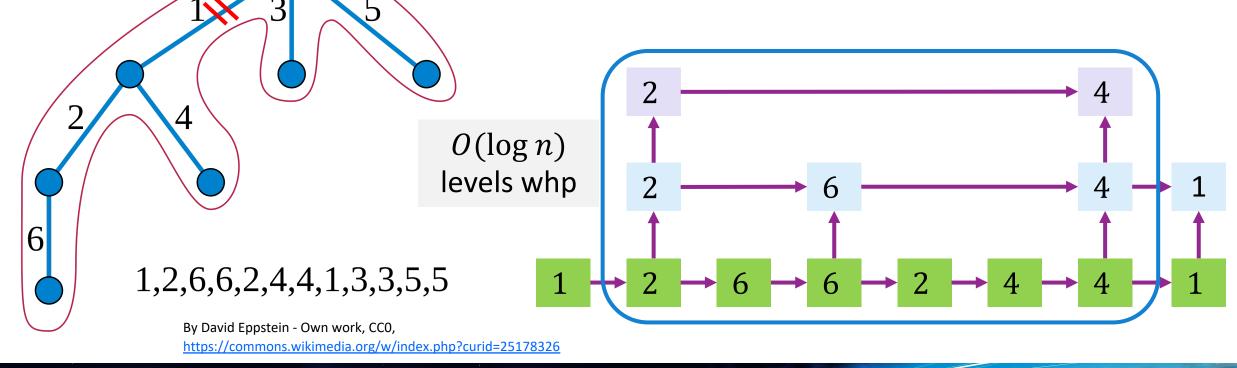
Delete edge 1



# Monte Carlo Oblivious Adversary Dynamic Connectivity

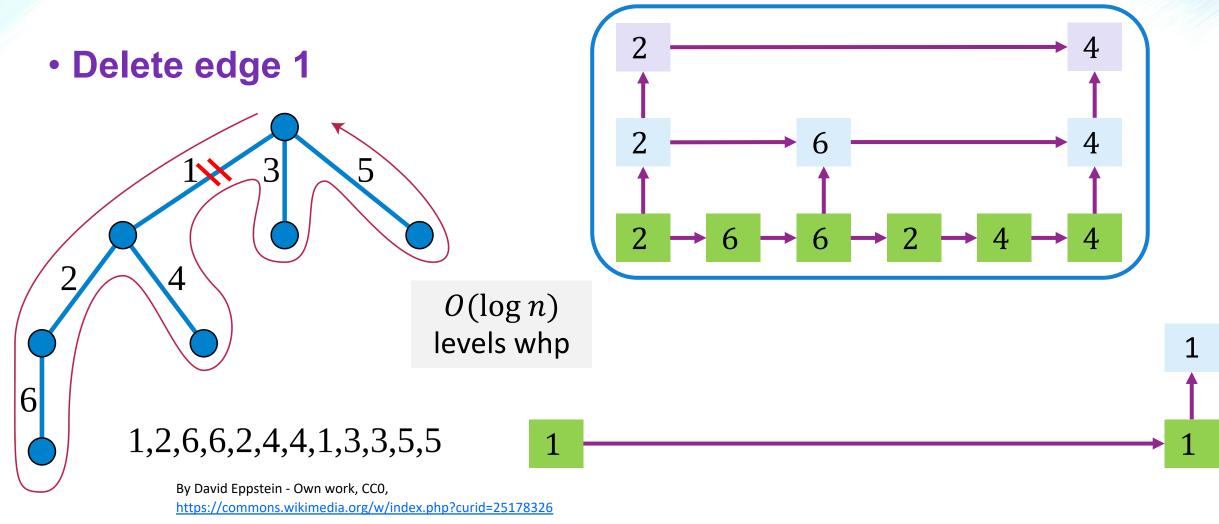
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Remove relevant contiguous section of skip-list and link together the ends



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# Monte Carlo Oblivious Adversary Dynamic Connectivity



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Start from initially empty graph

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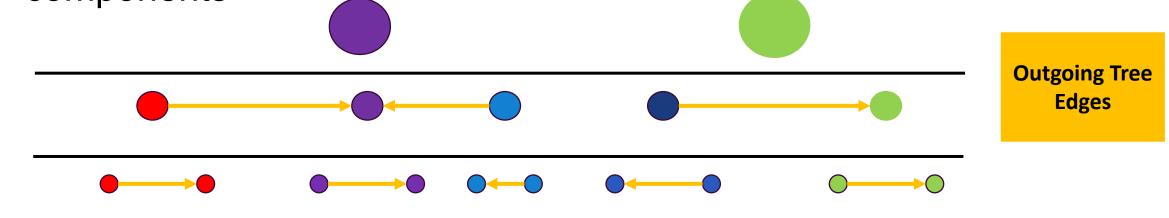
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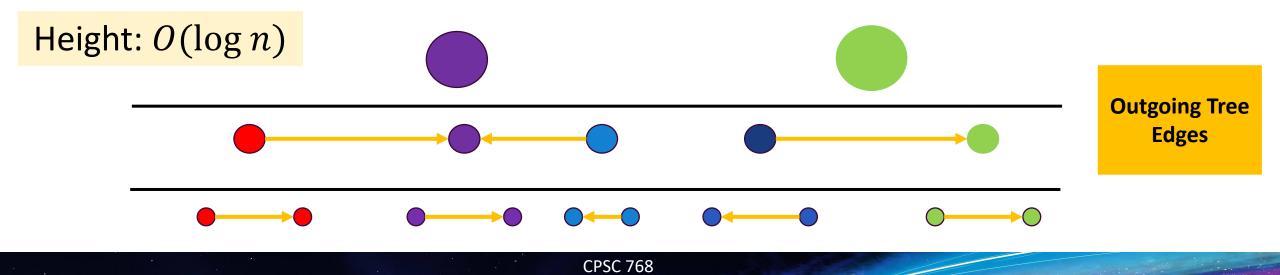
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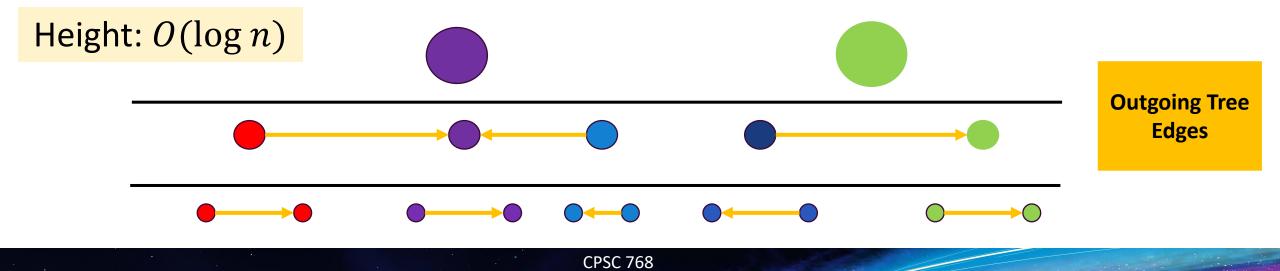
**CPSC 768** 



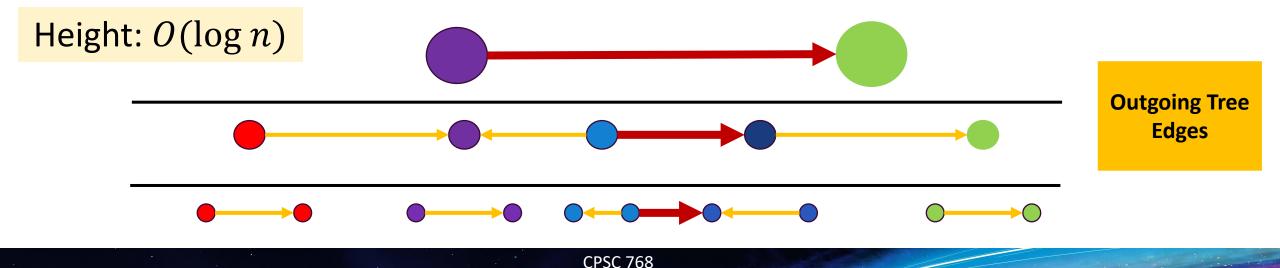
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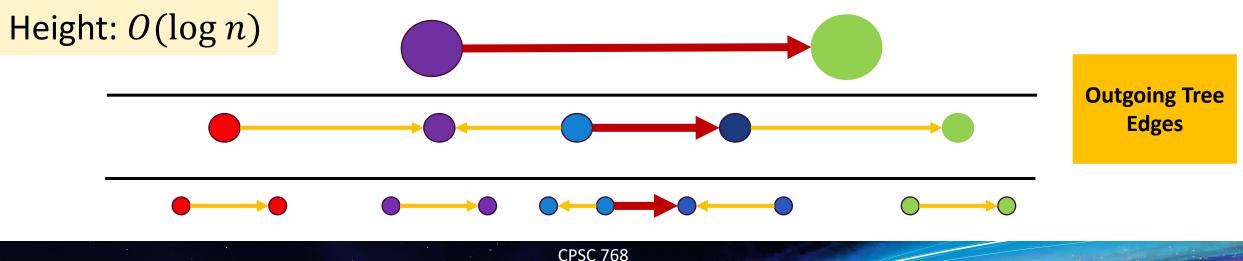
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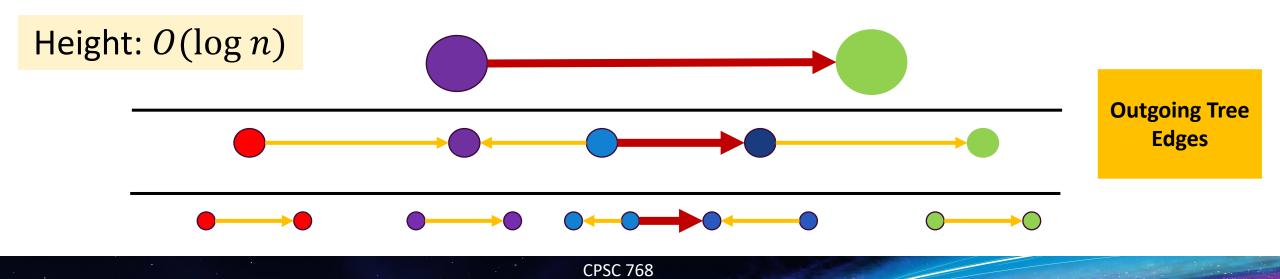
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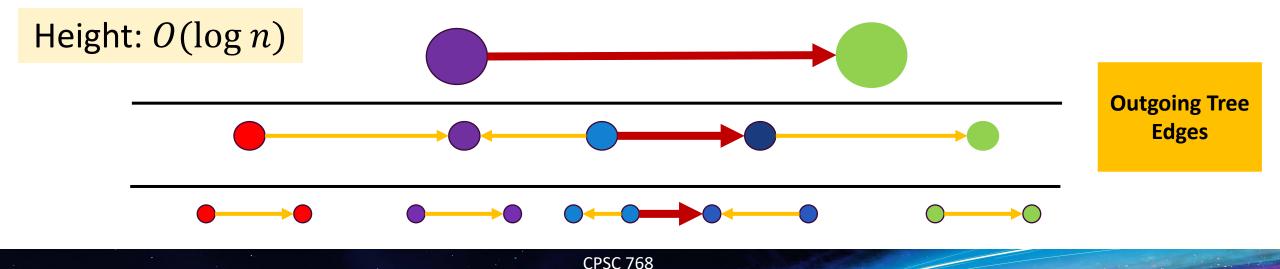
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- Update XOR data structure



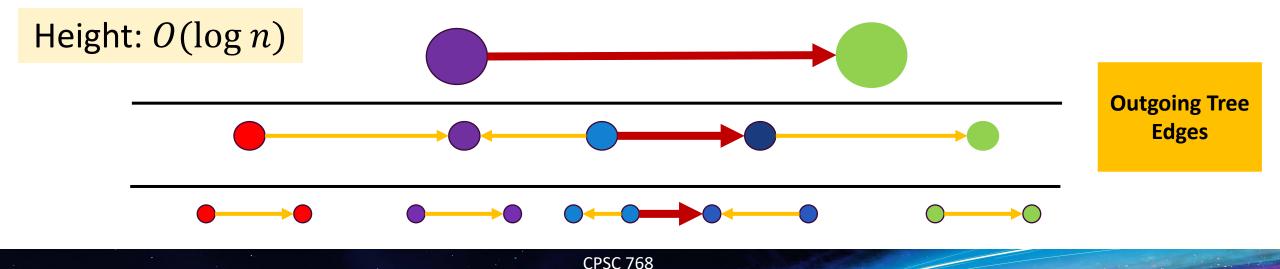
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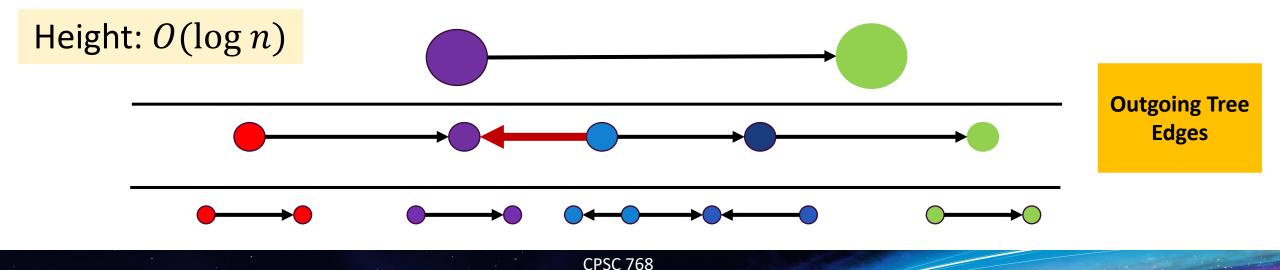
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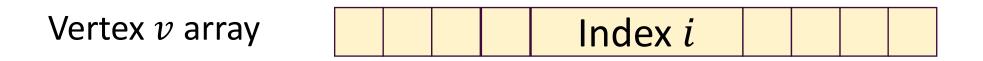
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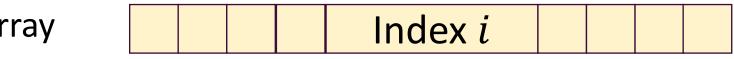
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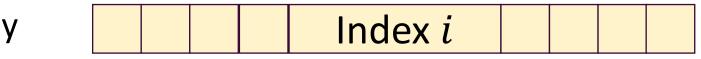
Index *i* 

Store edge  $(ID_v, ID_u)$  in index *i* with probability  $\frac{1}{2^i}$ 

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Vertex v array



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#### XOR data structure

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If XOR data structure of tree only contains tree edges, returns 0

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 $ID_e$ 

Otherwise, if contains one

outgoing edge, returns ID,

- Why have different probabilities of sampling:
  - Due to XOR data structure, return exactly one edge in cut whp

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 IDe

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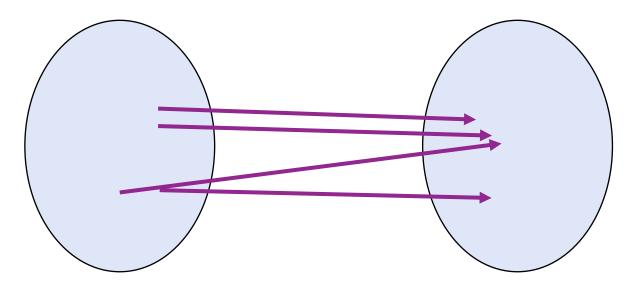
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Cutset data structure

ID<sub>e</sub> Otherwise, if contains one outgoing edge, returns ID<sub>e</sub>

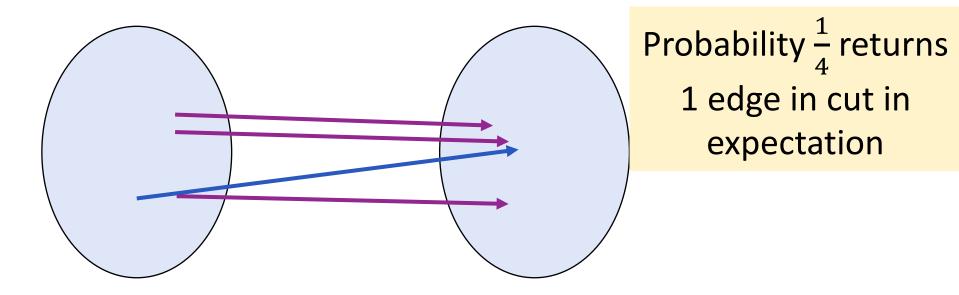
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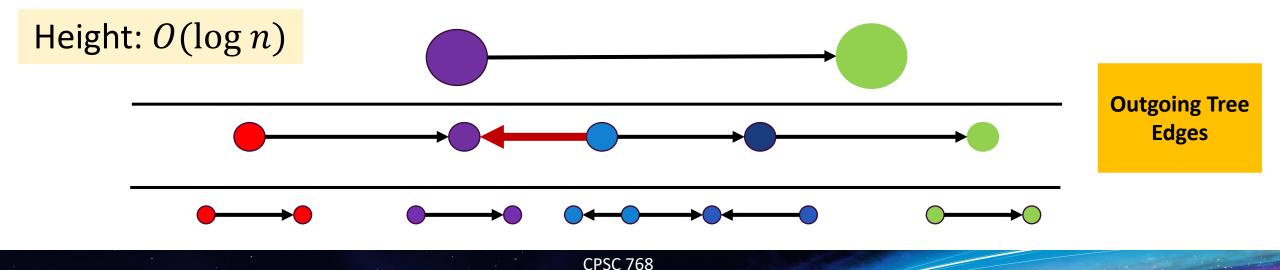
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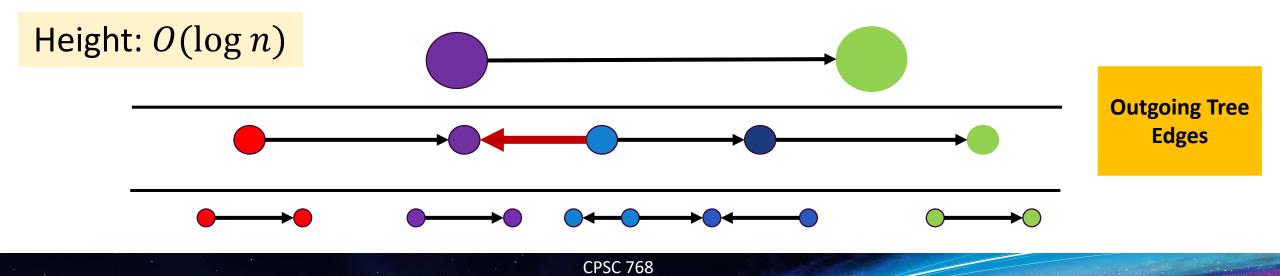
Cutset data structure



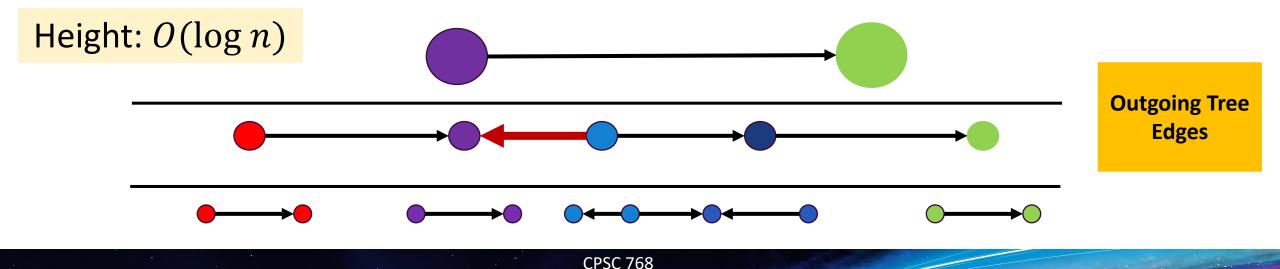
- On edge deletion: if deletion of an **outgoing tree edge**:
  - Delete from every level
  - Search for replacement edge from XOR data structure
- Update XOR data structure



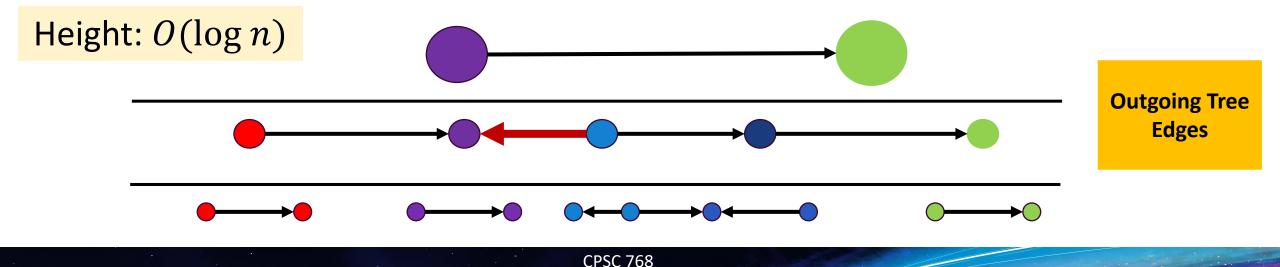
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  - Total: poly(log *n*) time per operation



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    - Thus, by Chernoff,  $O(\log n)$  levels suffice