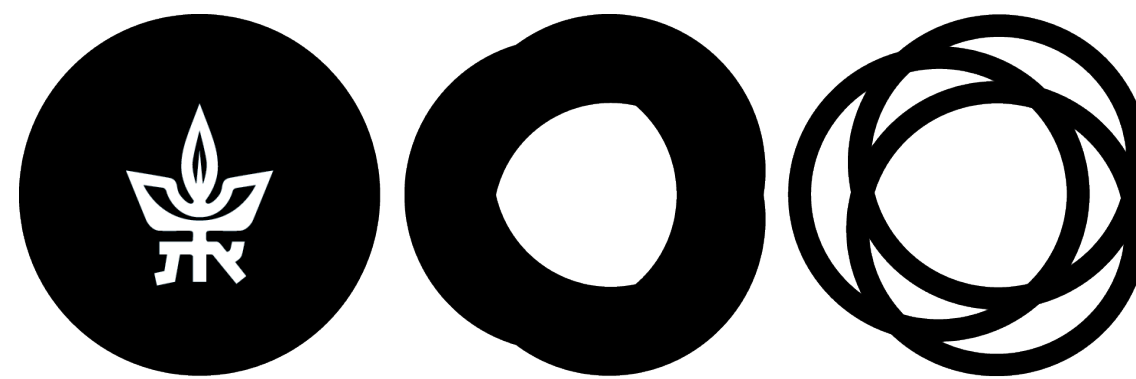


# Faster **Deterministic** **Worst-Case** Dynamic **All-Pairs** Shortest Paths

Shiri Chechik

Tianyi Zhang



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UNIVERSITY תל אביב

# Dynamic All-Pairs Shortest Paths

- Given a weighted digraph  $G = (V, E, \omega)$
- A sequence of **vertex updates**, maintain **pairwise exact distances**

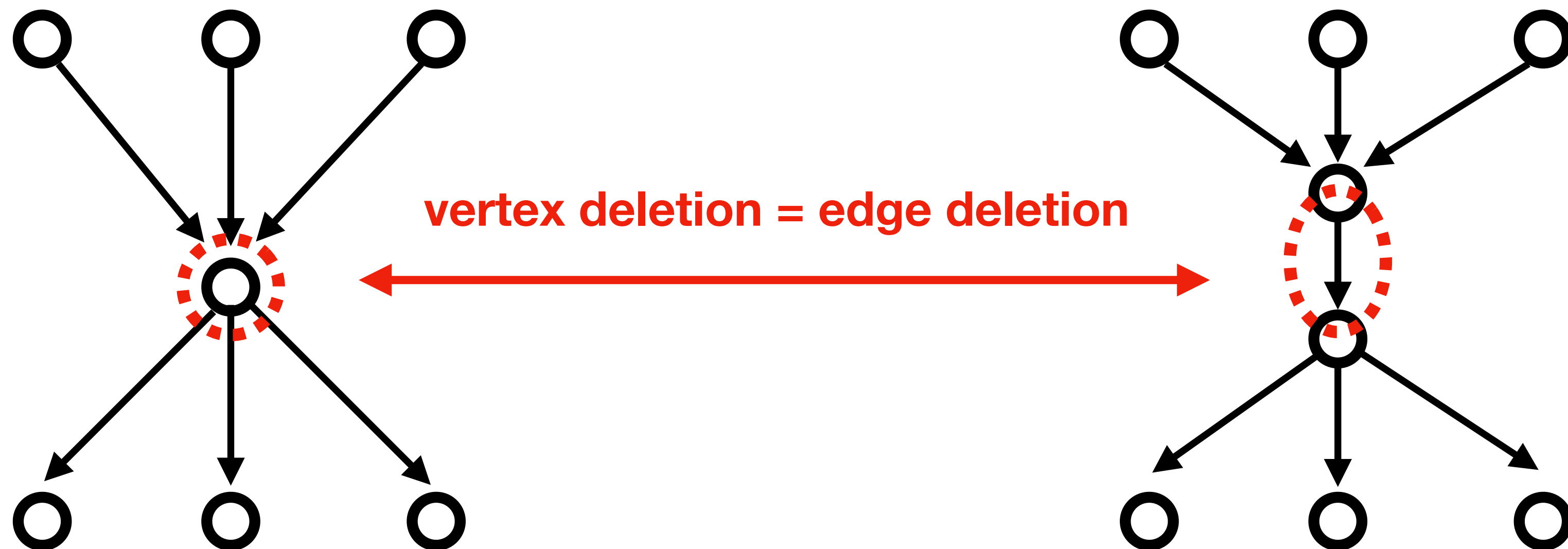
More specifically, want a data structure:

- **Ins**( $v, \text{adj}(v)$ ) / **Del**( $v$ )  
Insert/delete vertex  $v$  in  $G$  with adjacency list  $\text{adj}(v)$ ; **want  $n^2$  runtime**
- **Query**( $u, v$ )  
Return the shortest distance from  $u$  to  $v$  in  $G$ ; **want  $O(1)$  runtime**

# Dynamic All-Pairs Shortest Paths

- Given a weighted digraph  $G = (V, E, \omega)$
- A sequence of **vertex updates**, maintain **pairwise exact distances**

Why vertex updates, not edge updates?



# History

reference	vertex update time	deterministic / randomized	worst-case / amortized
King, 1999	$\tilde{O}(n^{2.5}\sqrt{W})$	deterministic	amortized
Demetrescu, Italiano, 2004	$\tilde{O}(n^2)$	deterministic	amortized
Thorup, 2005	$\tilde{O}(n^{3-1/4})$	deterministic	worst-case
Abraham, Chechik, Krinninger, 2017	$\tilde{O}(n^{3-1/3})$	randomized	worst-case
Probst, Wulff-Nilsen, 2020	$\tilde{O}(n^{3-2/7})$	deterministic	worst-case
<b>New</b>	$\tilde{O}(n^{3-20/61})$	deterministic	worst-case

$n$  is the number of vertices in the graph,  $W$  refers to the maximum edge weight

Previous approaches

# Reduction to batch deletion

Batch deletion data structure:

- **Prep**(G)  
Preprocess the graph G and be ready for **one batch deletion** and queries
- **Batch**(B)  
Remove a subset  $B \subseteq V$  of vertices from graph G
- **Query**(u, v)  
Return the shortest distance from u to v in  $G \setminus B$  in  $O(1)$  time

# Reduction to batch deletion

Theorem [Thorup, 2005]

- Given a batch deletion algorithm, dynamic APSP can be solved with worst-case update time  $T_{\text{prep}}/|B| + T_{\text{batch}} + |B|n^2$

- **Batch**( $B$ )

Remove a subset  $B \subseteq V$  of vertices from graph  $G$

- **Query**( $u, v$ )

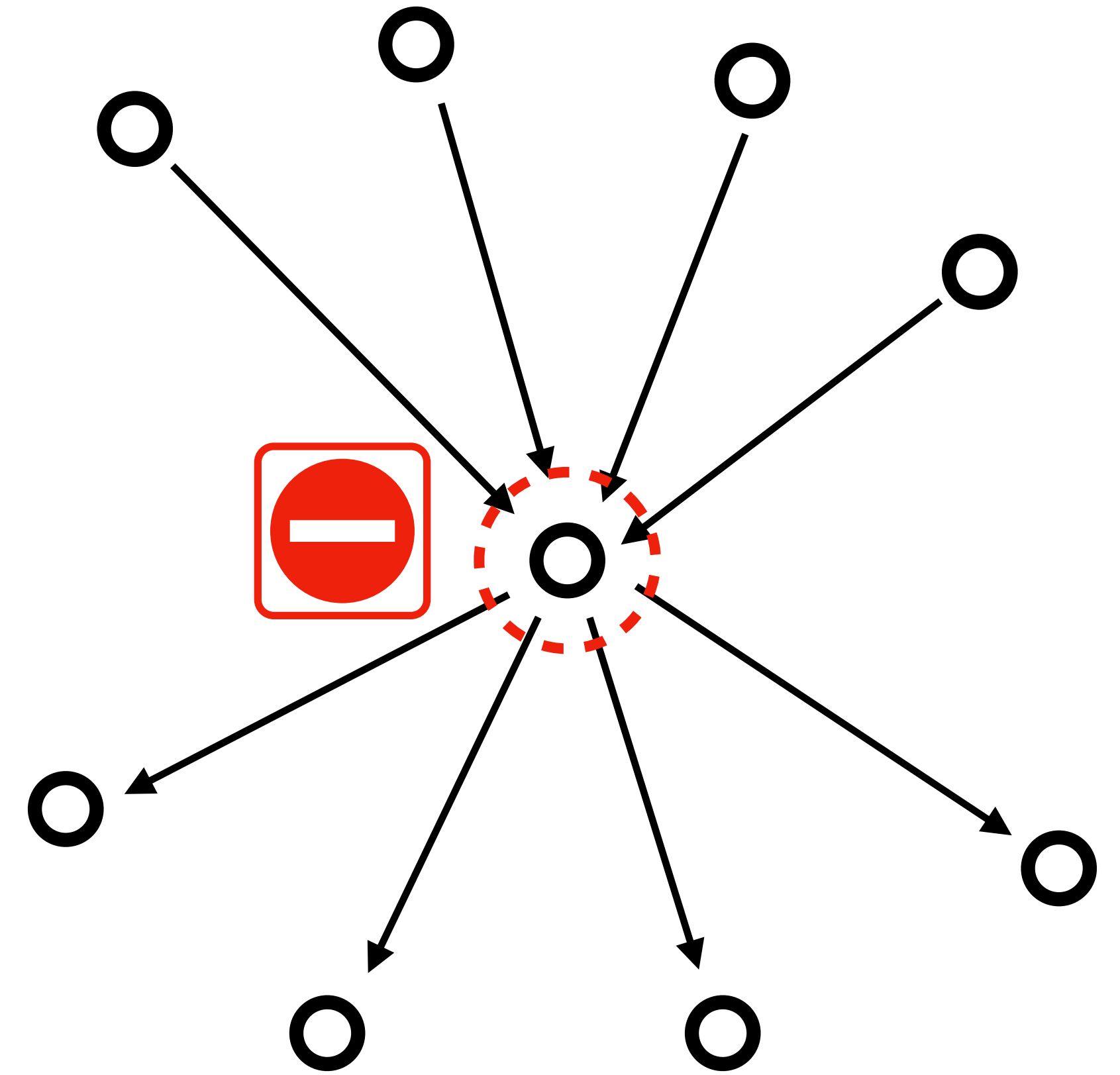
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# Batch Deletion

## Main difficulty:

Precompute shortest paths in  $G$

A single deletion can **destroy a lot**





# Batch Deletion

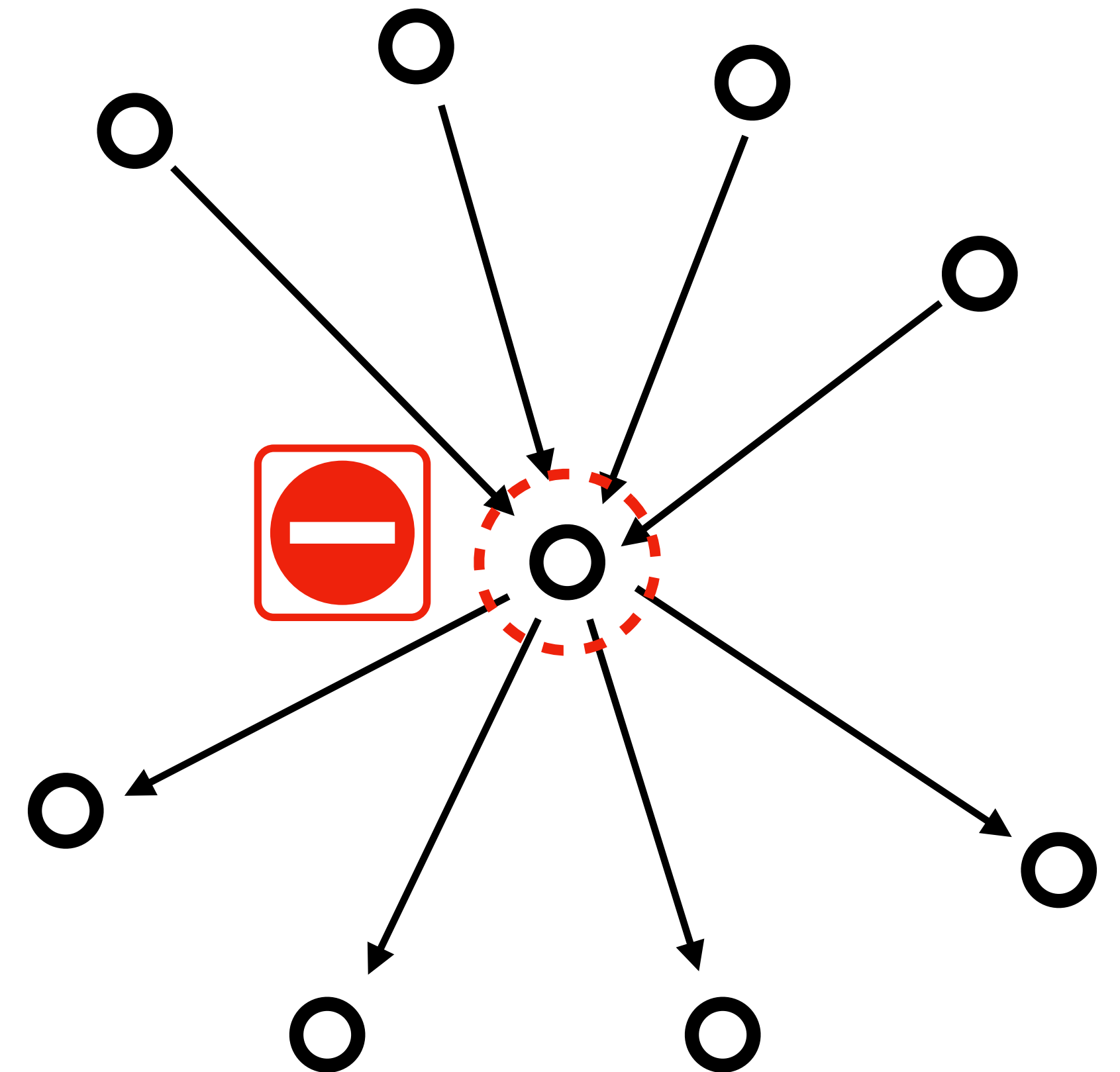
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Two basic ideas [Thorup, 2005]

- Shortest paths with small #hops  
Long-hop paths can be handled using hitting sets
- Prepare **low-congestion** shortest paths



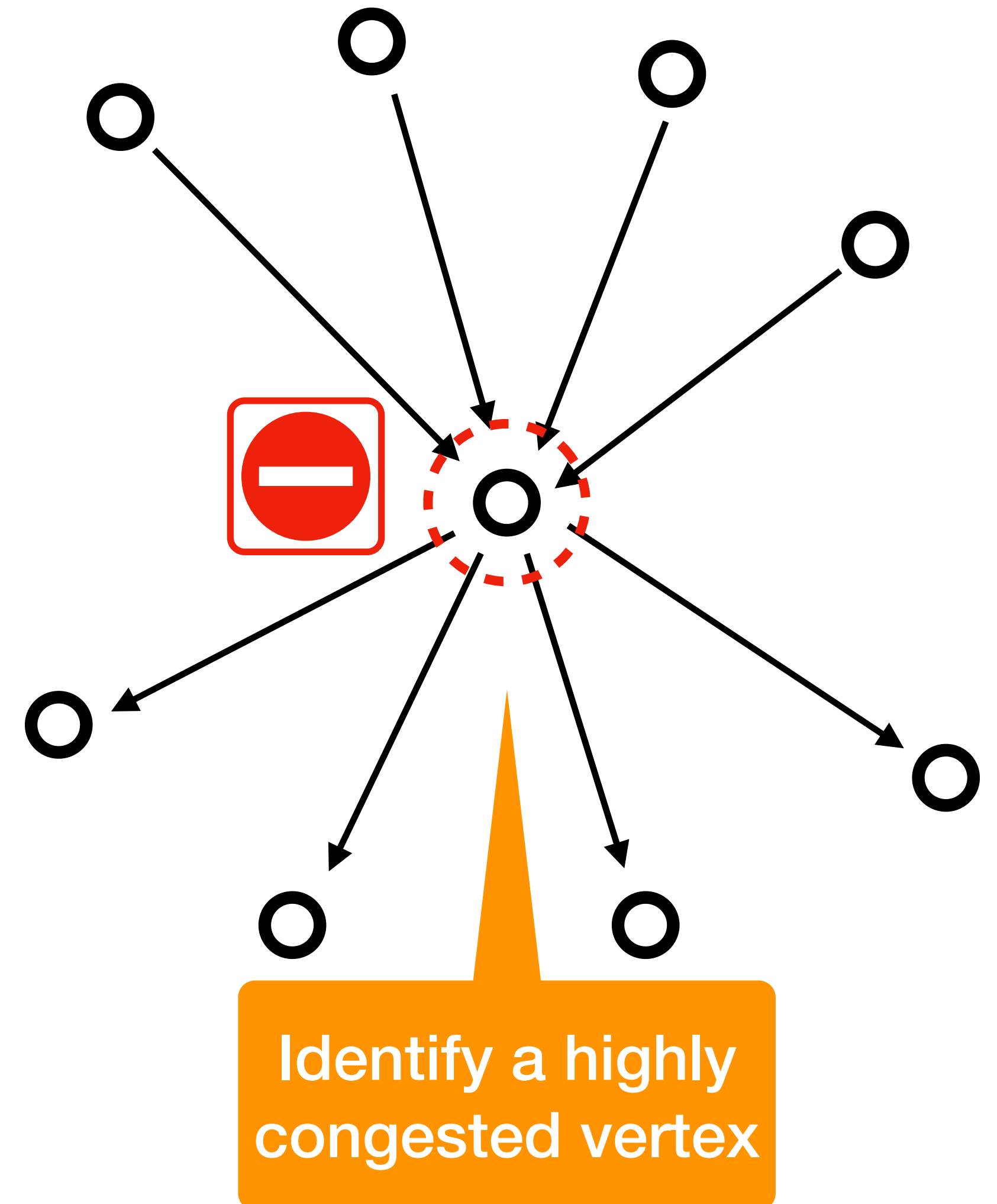
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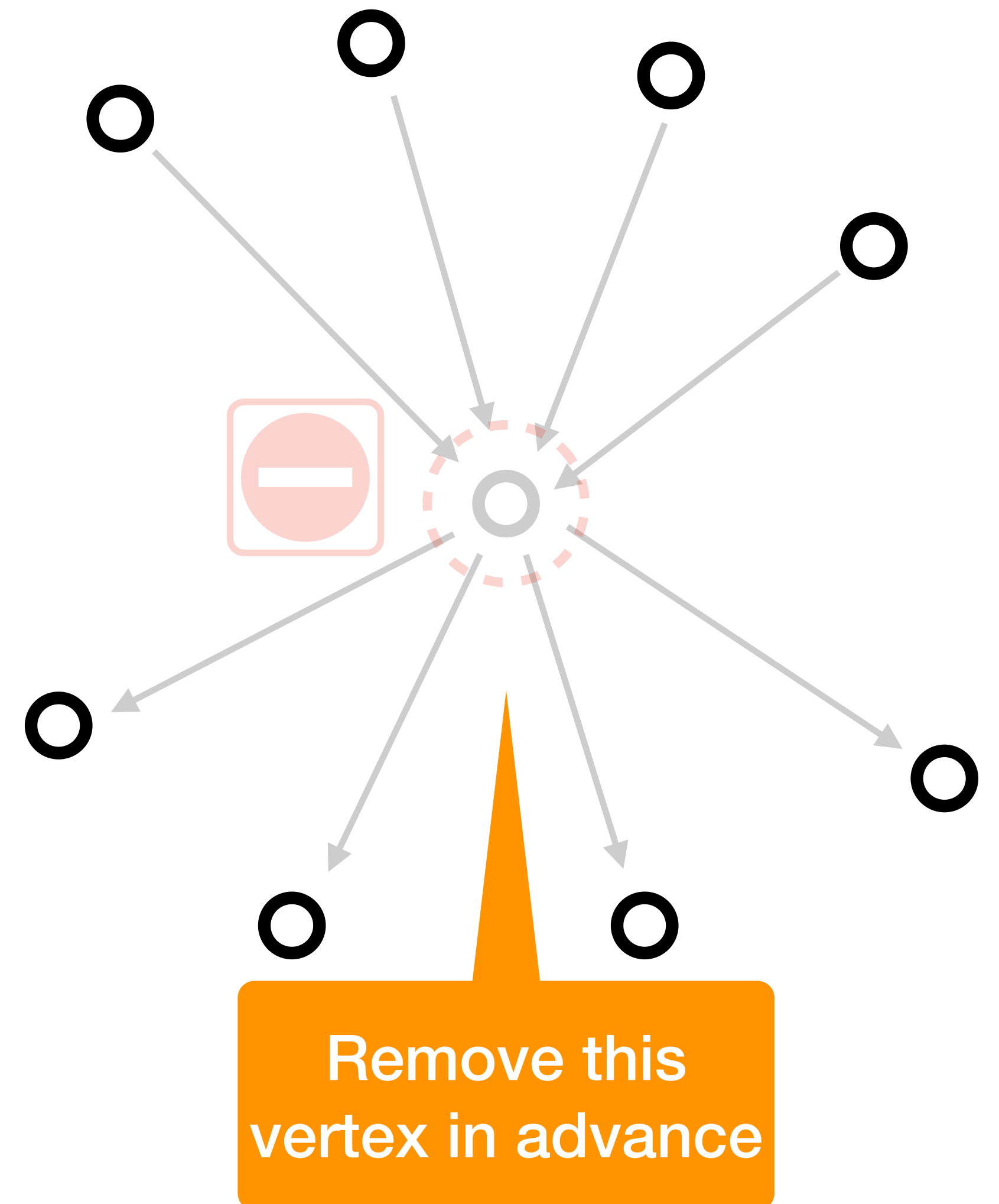
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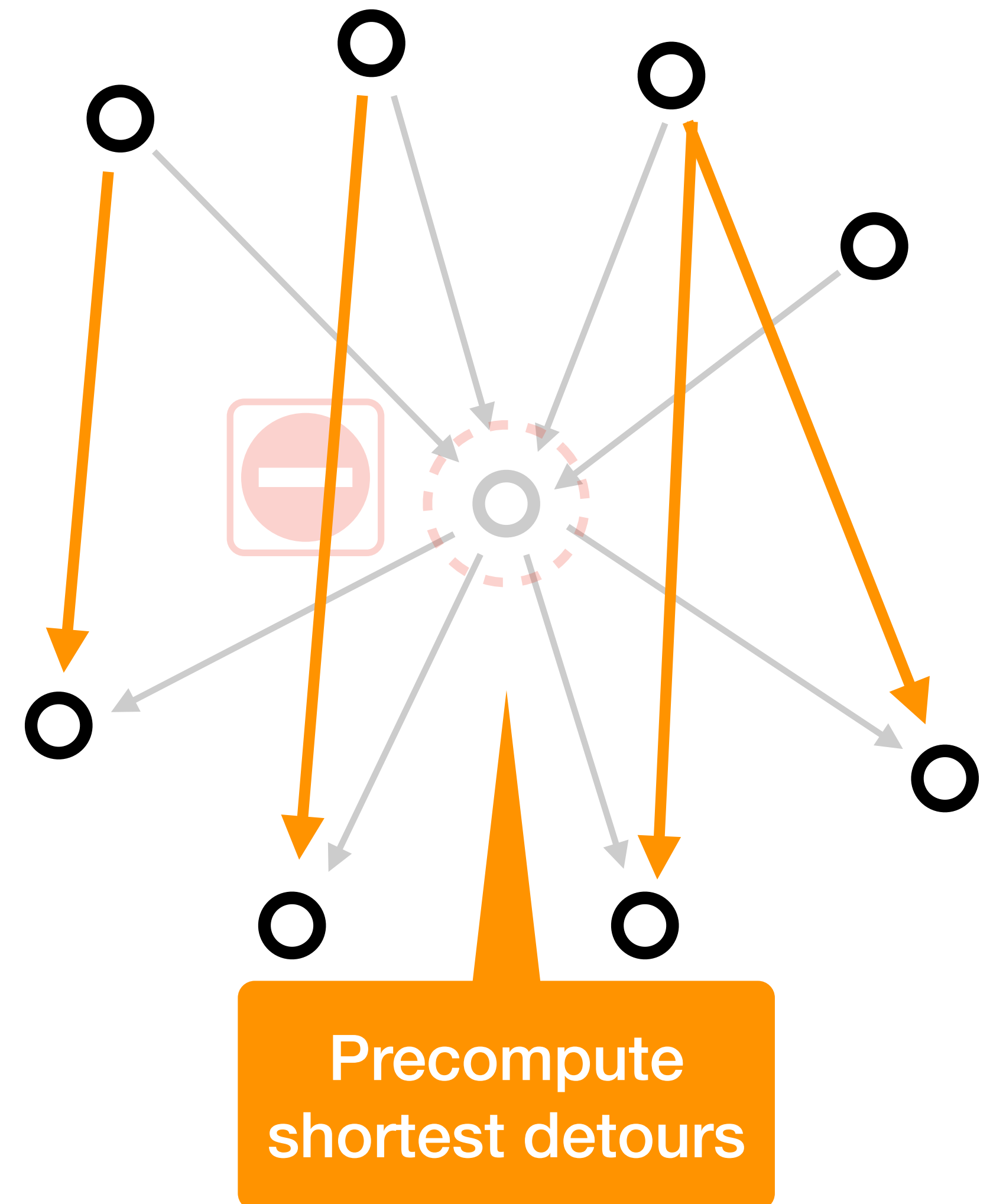
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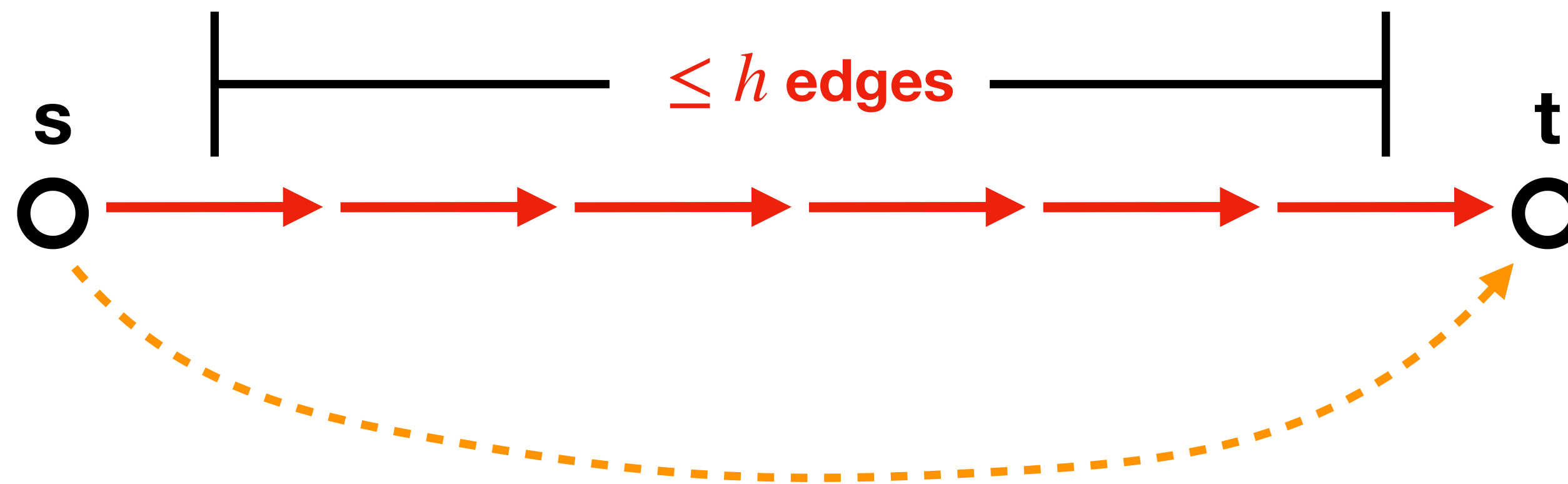
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# Hop-Restricted Shortest Paths

- An  $h$ -hop shortest path  $\pi_{s,t}$  is the **shortest path with at most  $h$  edges**
- Single-source  $h$ -hop paths  $\{\pi_{s,t}\}_{t \in V}$  can be computed using the **Bellman-Ford** algorithm in time  $n^2h$



**Ordinary shortest path** might contain  $\gg h$  edges

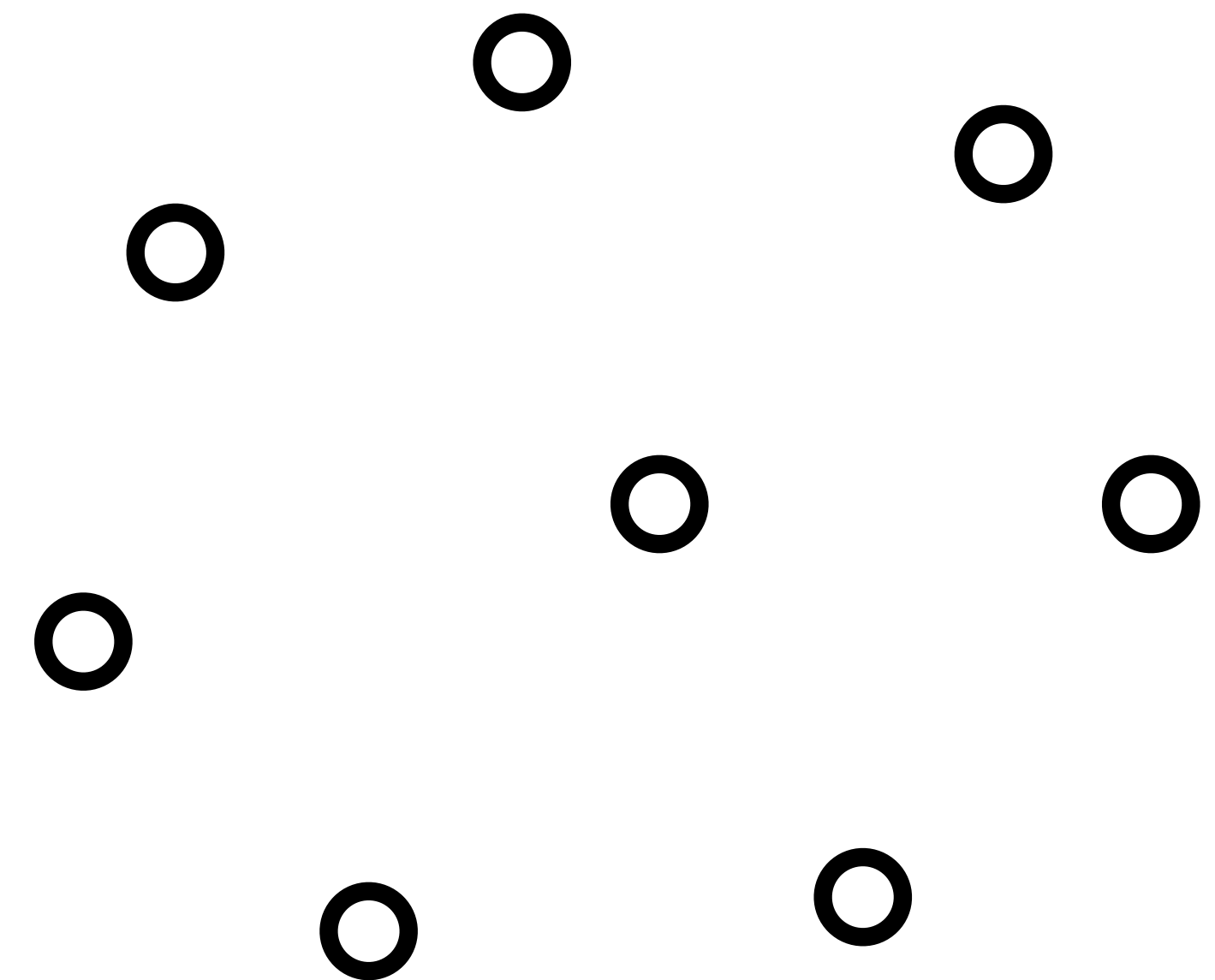
# The congestion technique

## Low-congestion shortest paths [Thorup'05]

1. Pick a vertex  $v$  that maximizes  $cg(v)$
2. Compute  $h$ -hop shortest paths at  $v$  using Bellman-Ford
3. Add  $h$ -hop paths to  $\Pi$ , update  $cg(\cdot)$
4. Remove  $v$  from graph, go to Step 1

$\Pi = \{\pi_{s,t} \mid s, t \in V\}$  a set of short paths

$cg(v) = \#\text{paths in } \Pi \text{ containing } v$



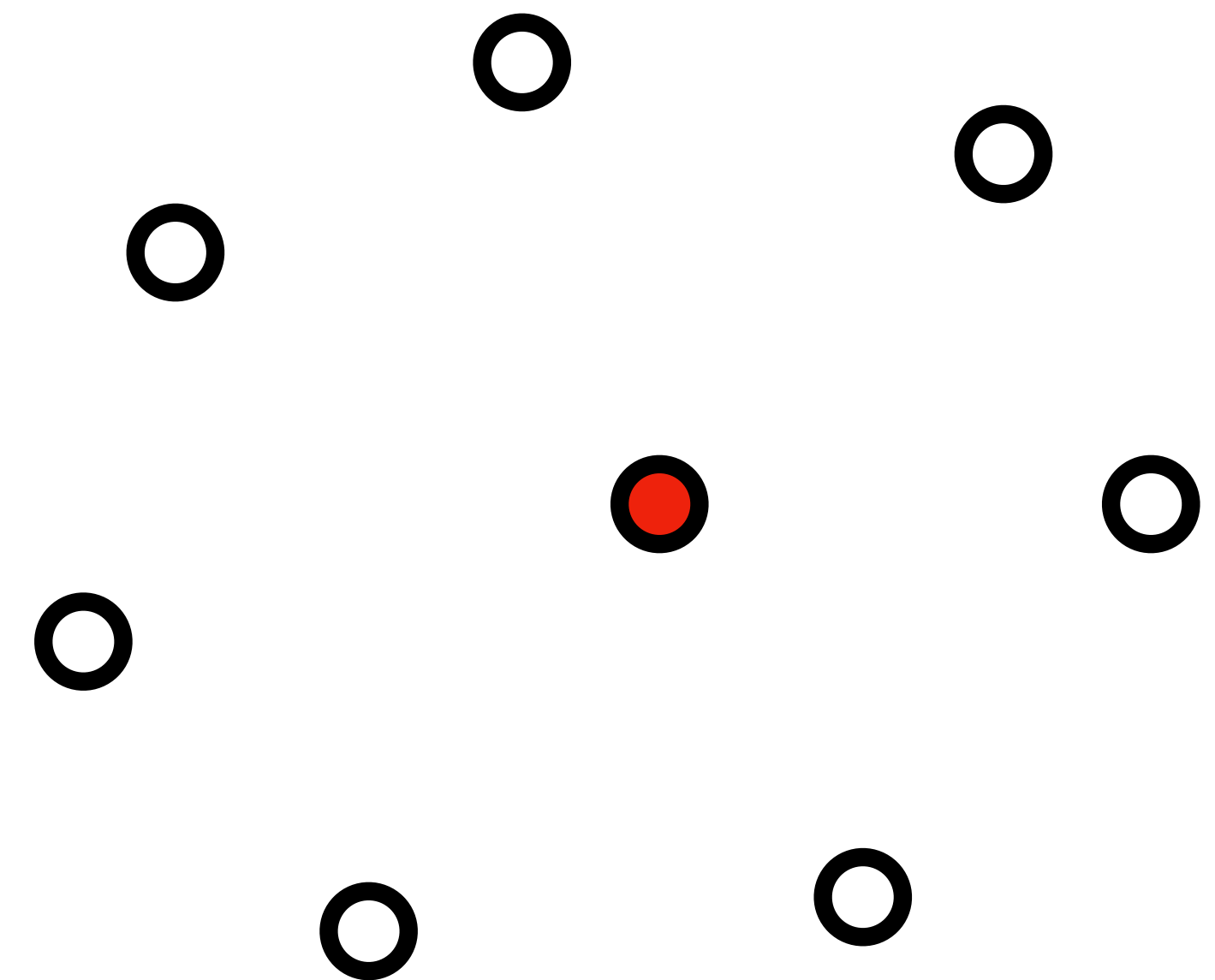
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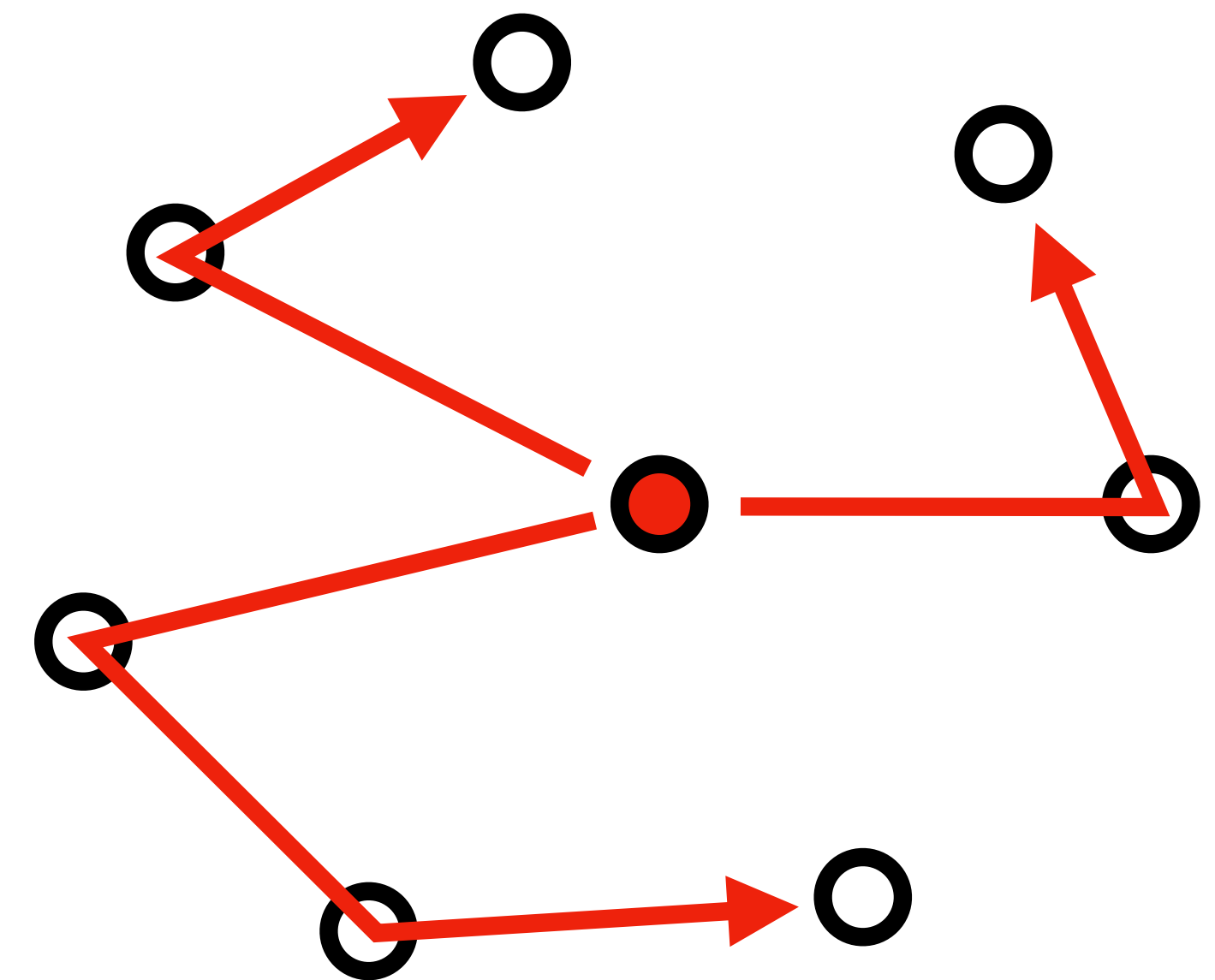
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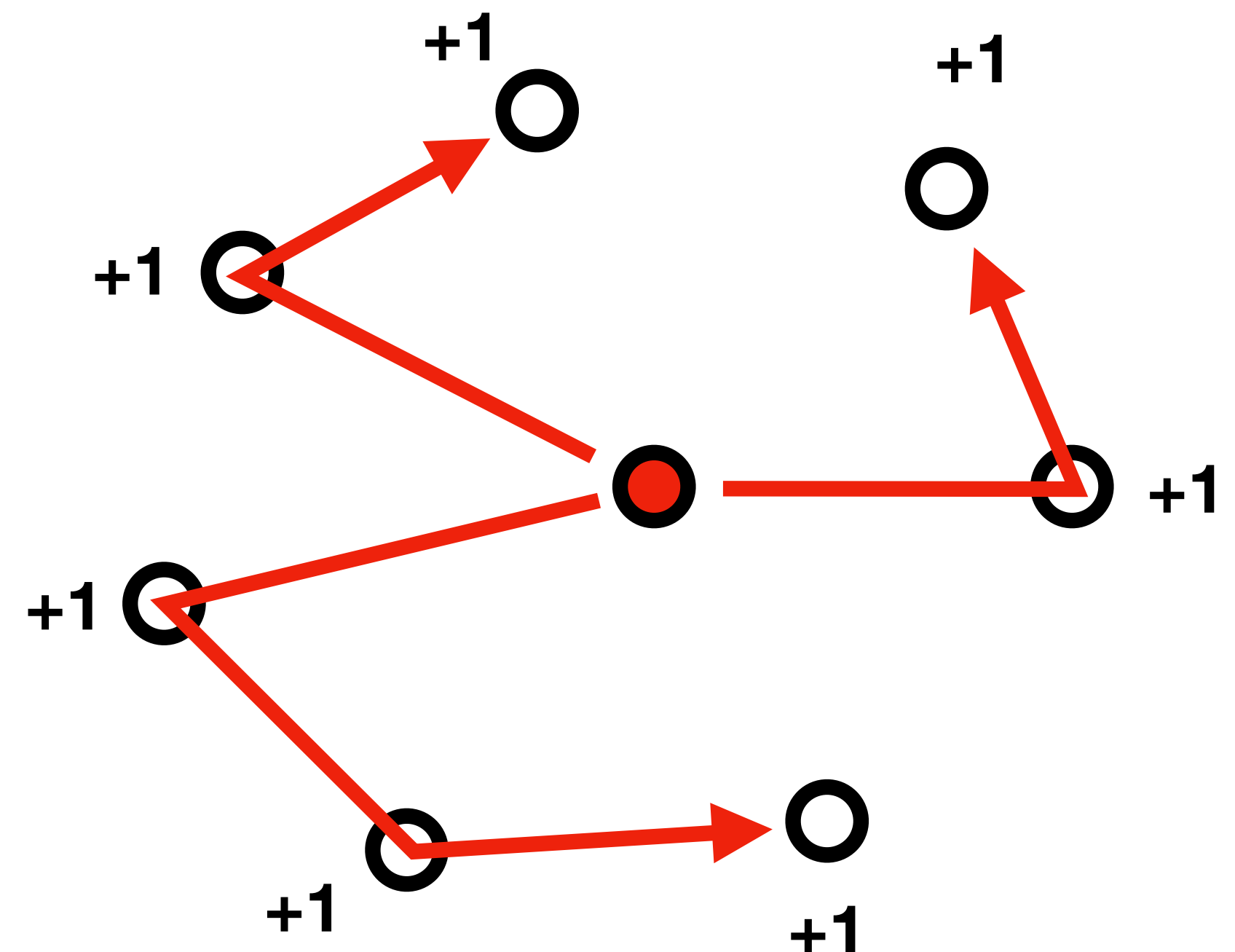
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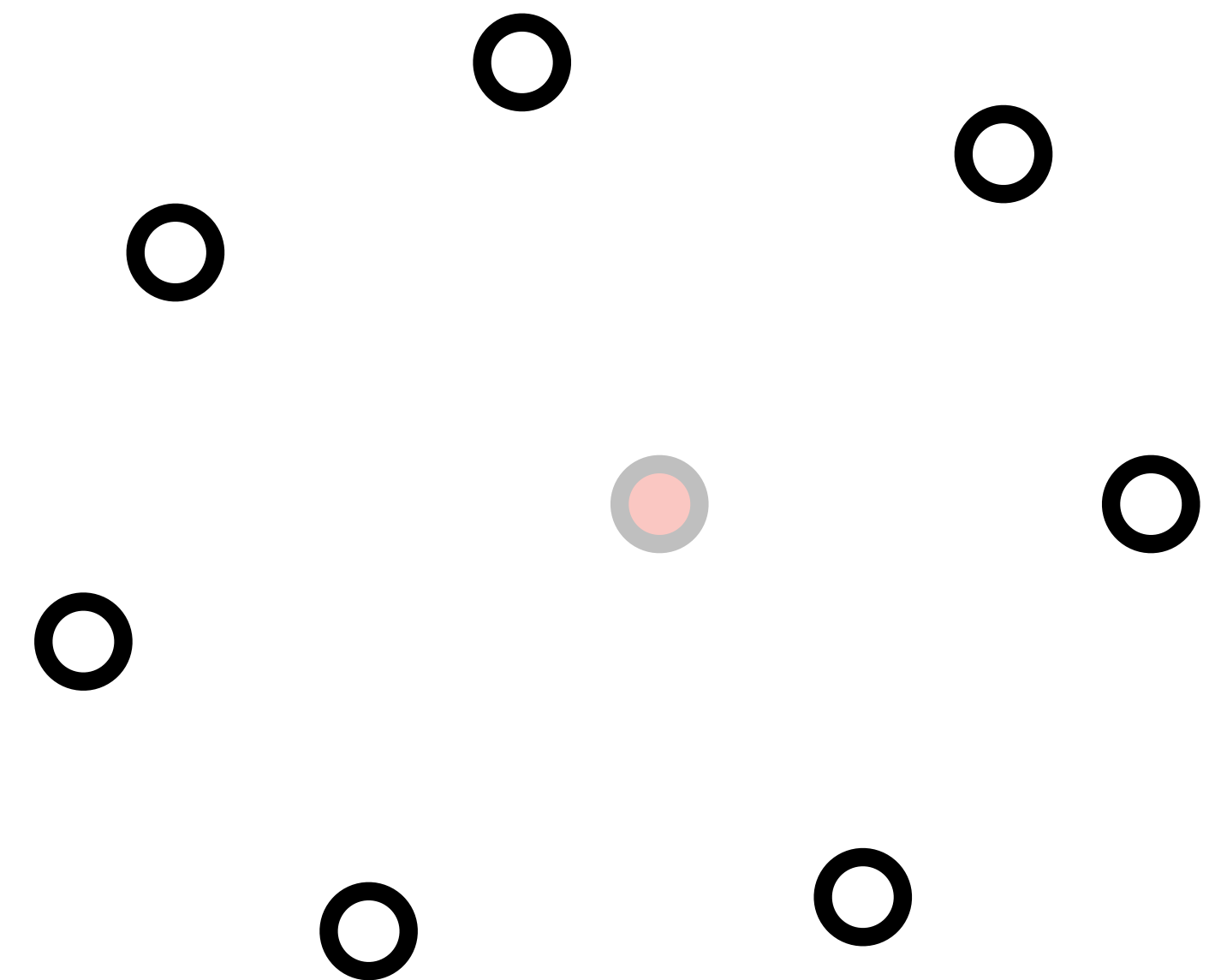
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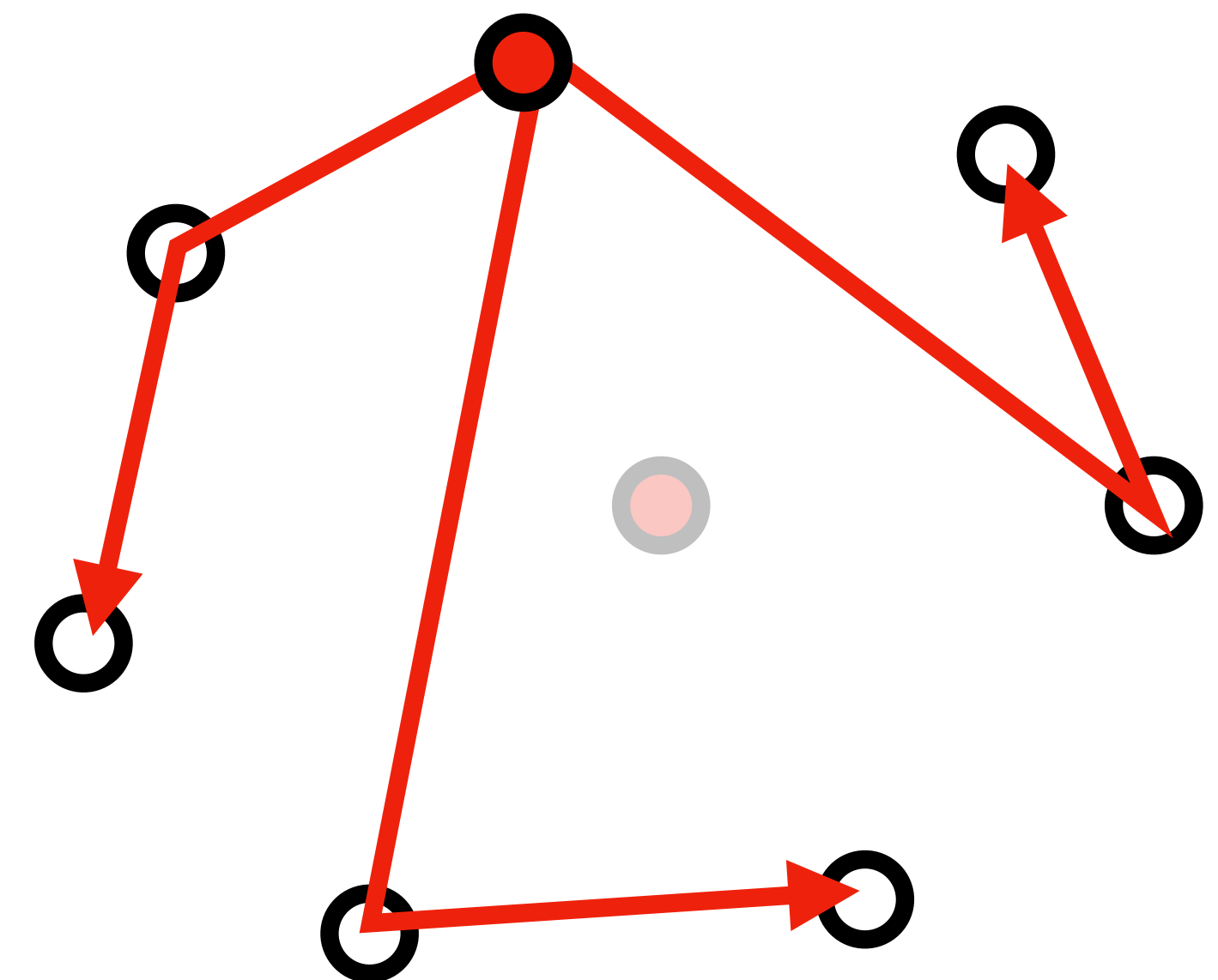
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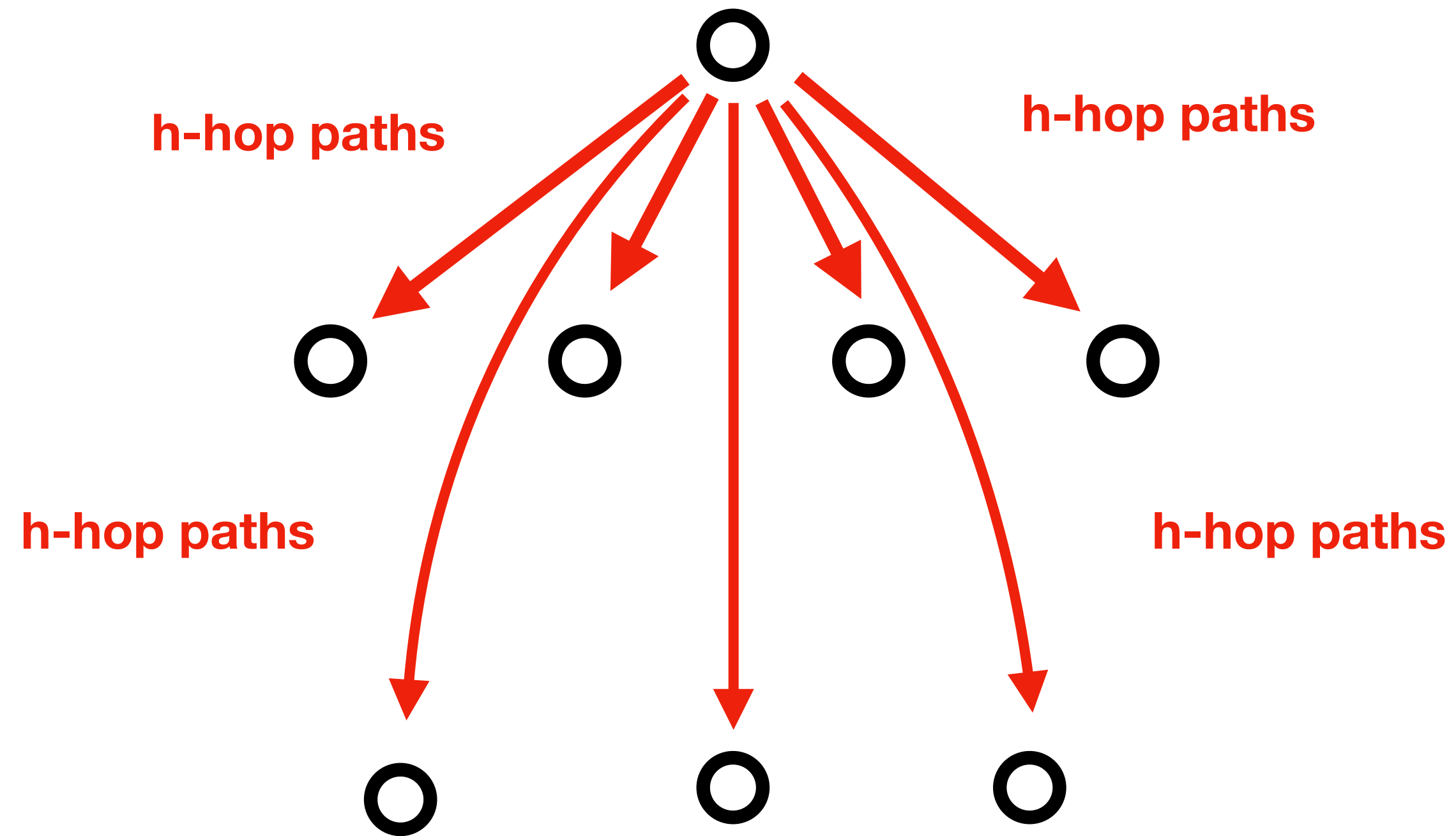
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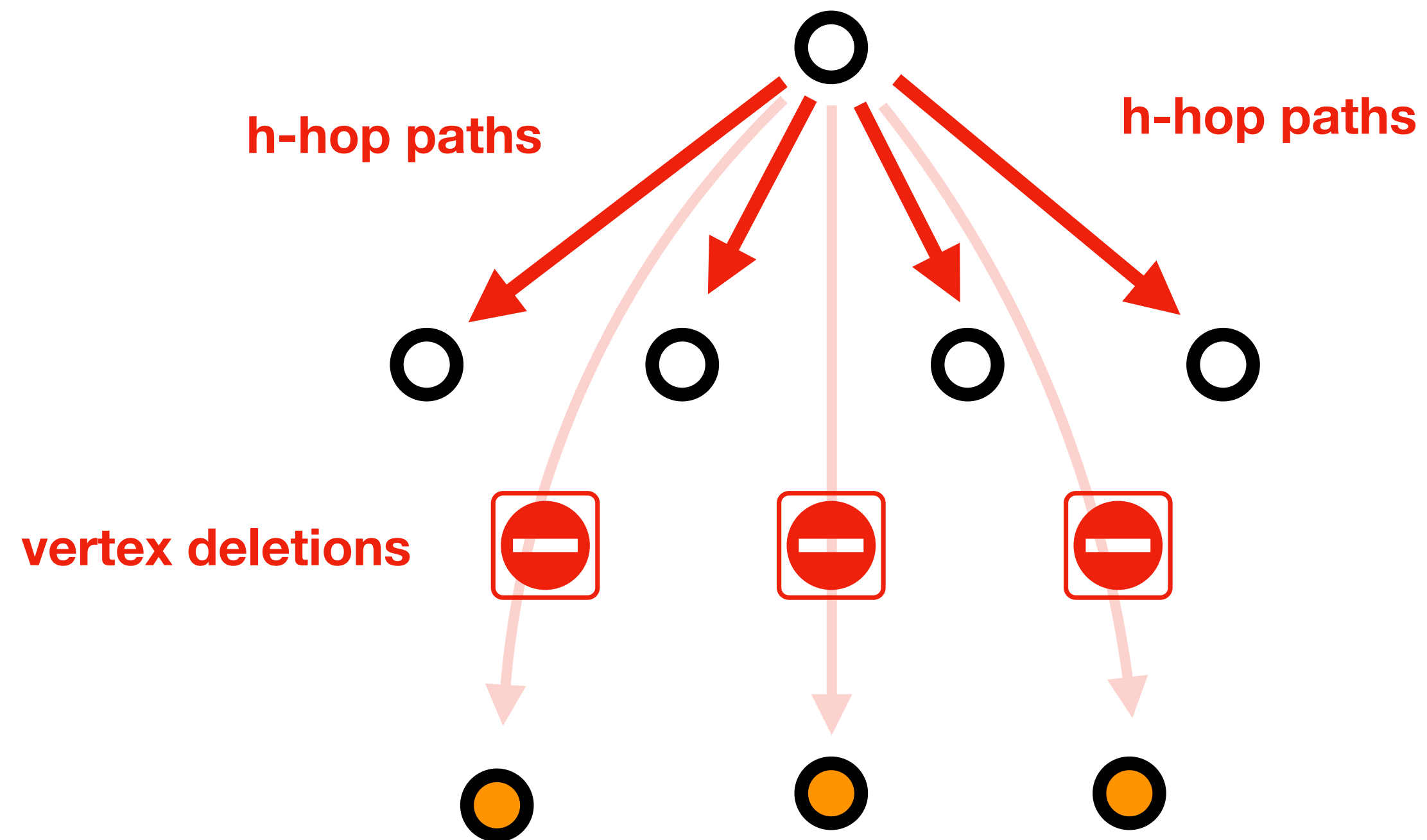
# Recovery from batch deletion

Recovery by Dijkstra's algorithm [ACK'17]



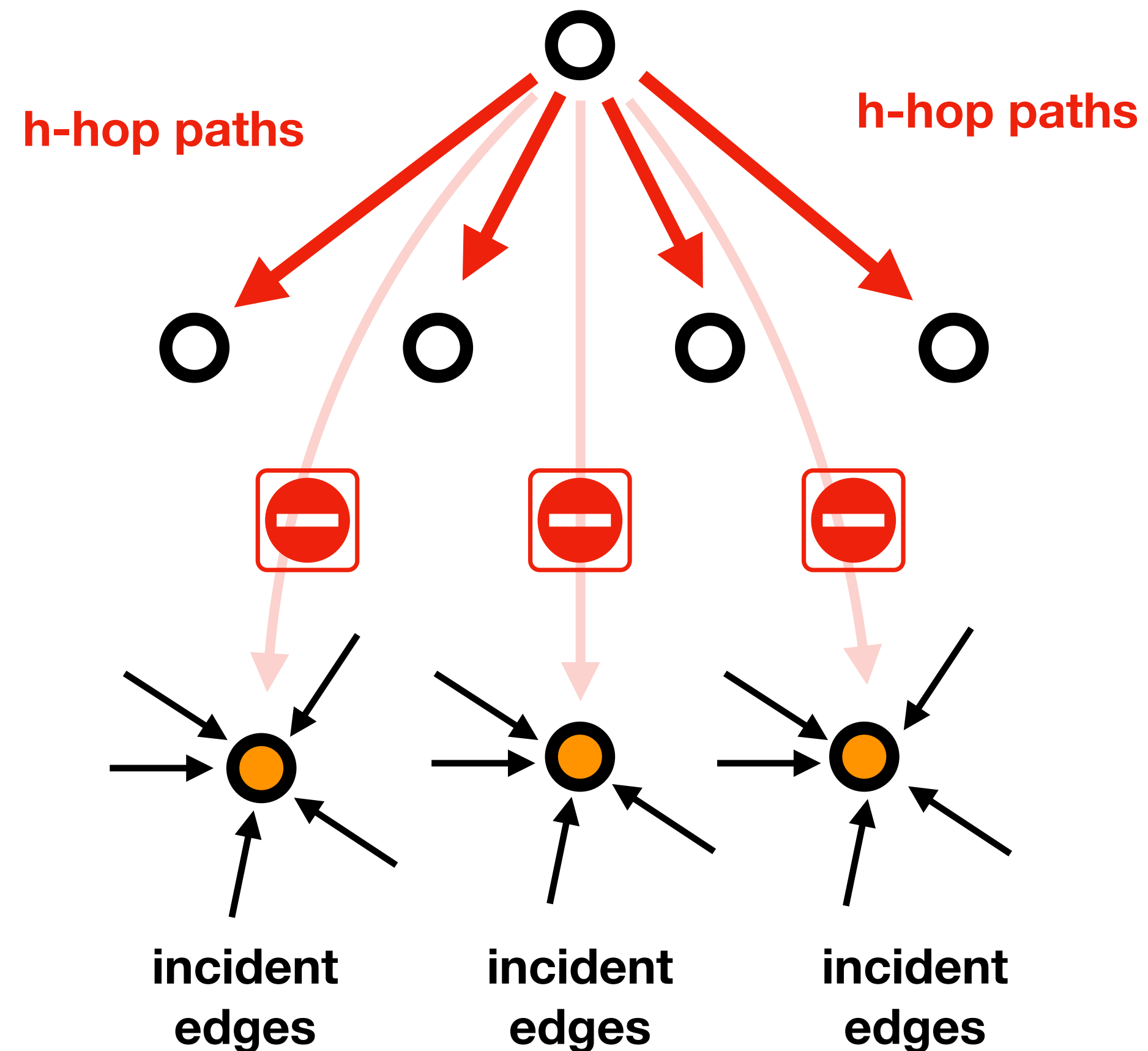
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# Recovery from batch deletion

## Recovery by Dijkstra's algorithm [ACK'17]



### Recovery algorithm:

1. View red paths as **shortcuts**
2. Run Dijkstra on red / black edges

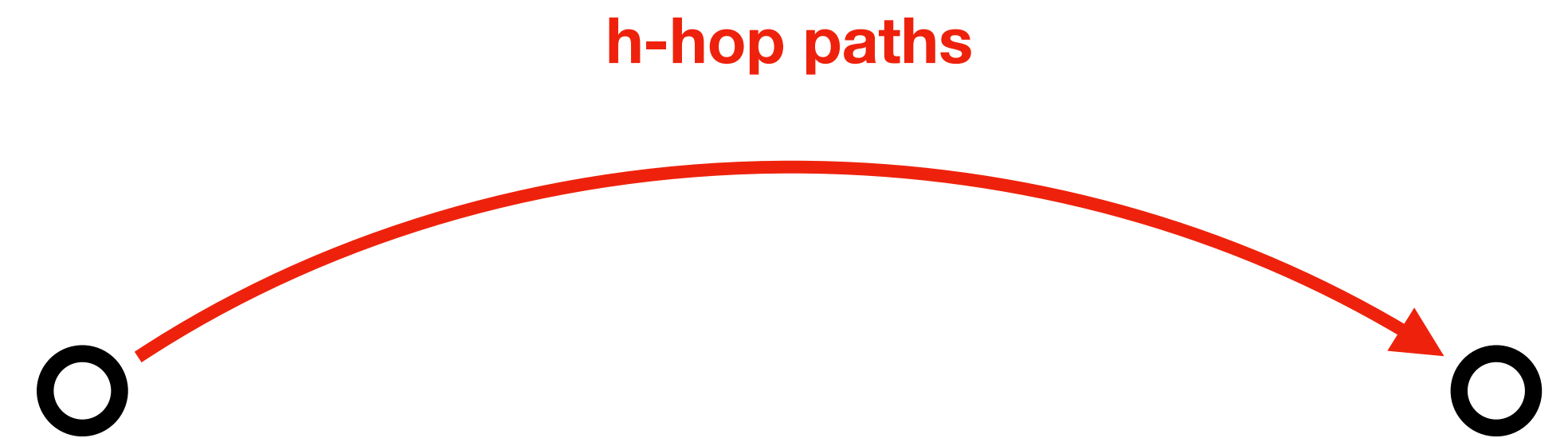
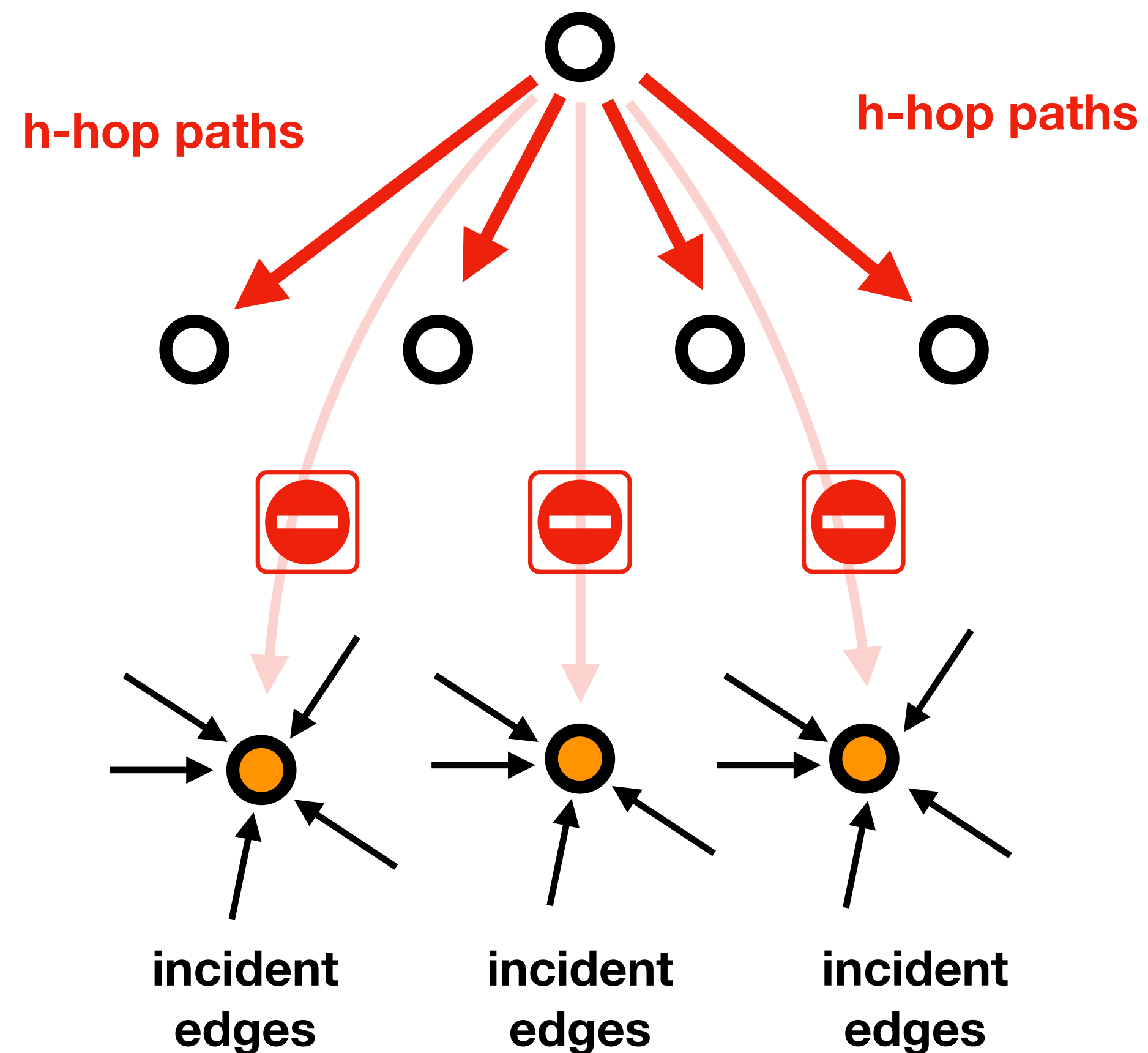
Runtime =  $n \cdot$  **#destroyed** h-hop paths

**#destroyed** is small by the congestion technique

# Recovery from batch deletion

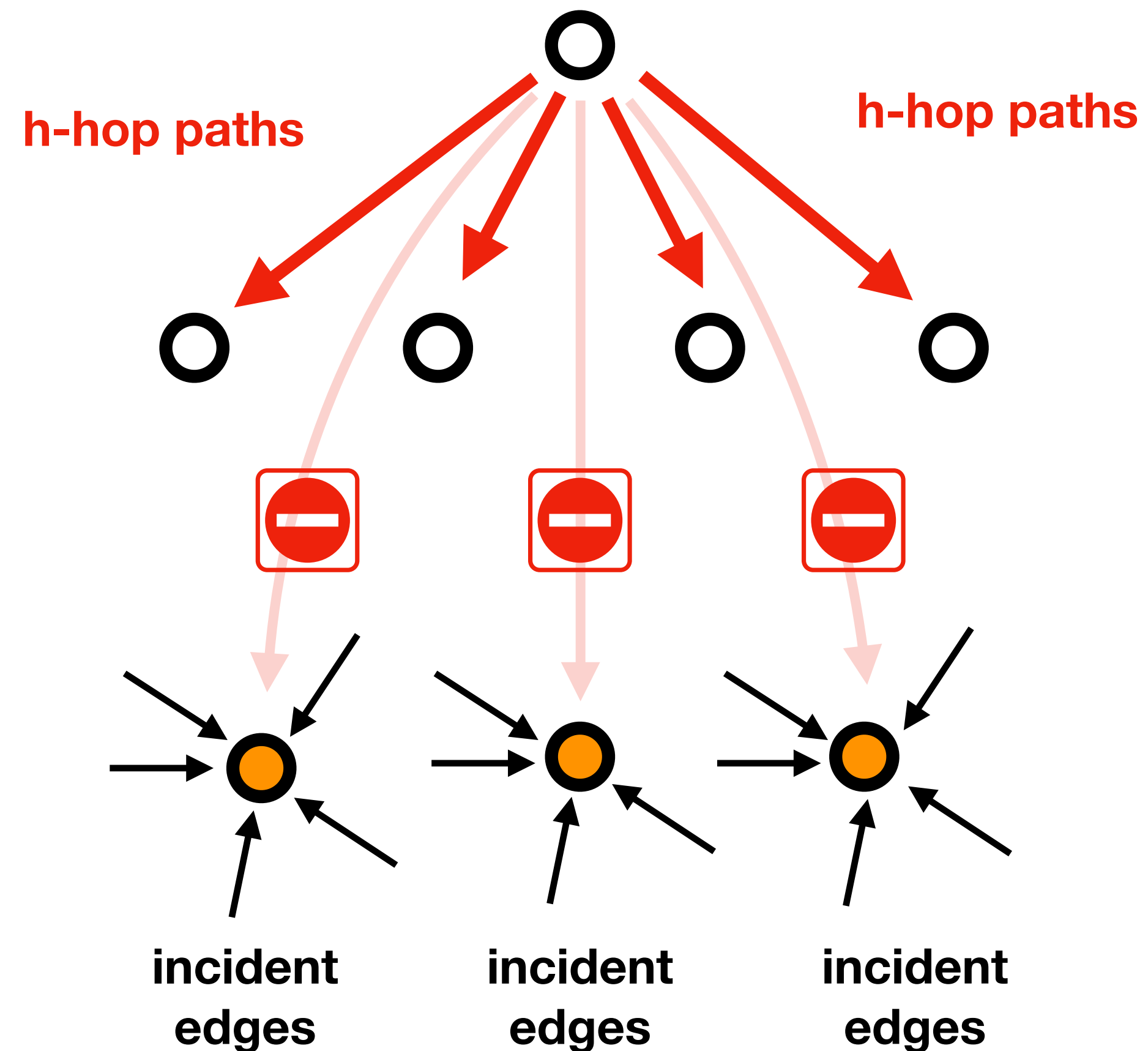
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Recovery by path concat [P-WN'20]



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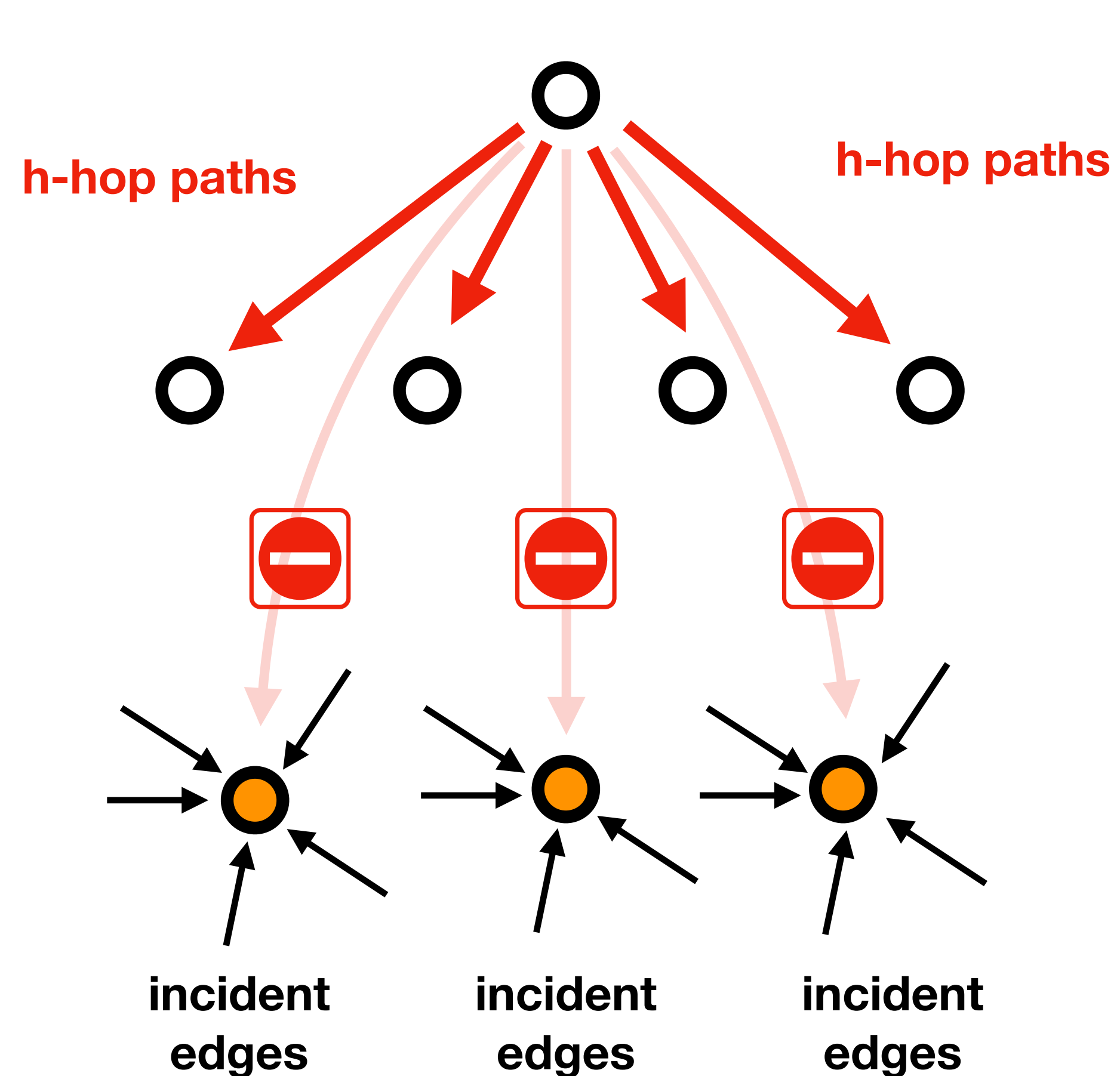




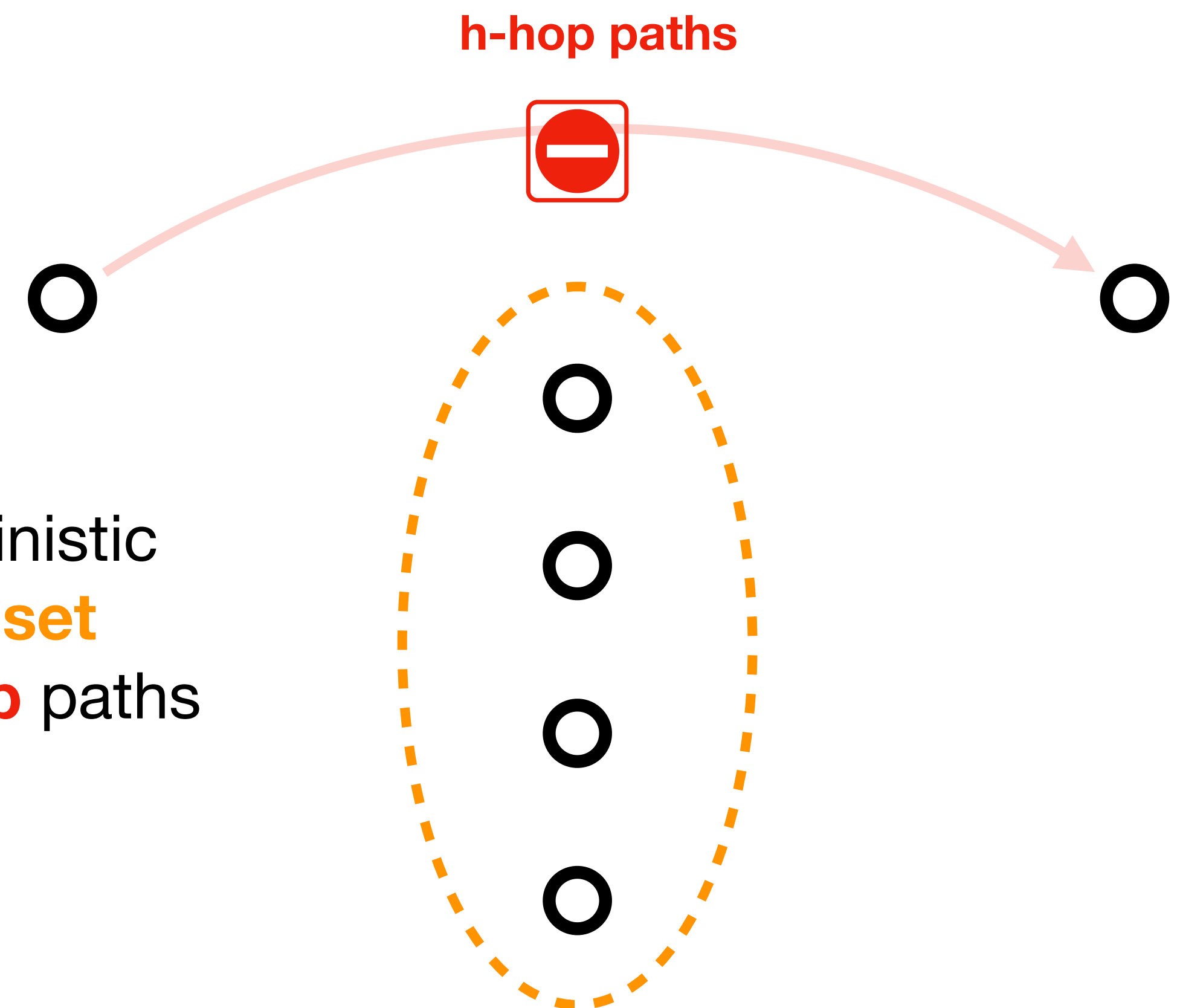
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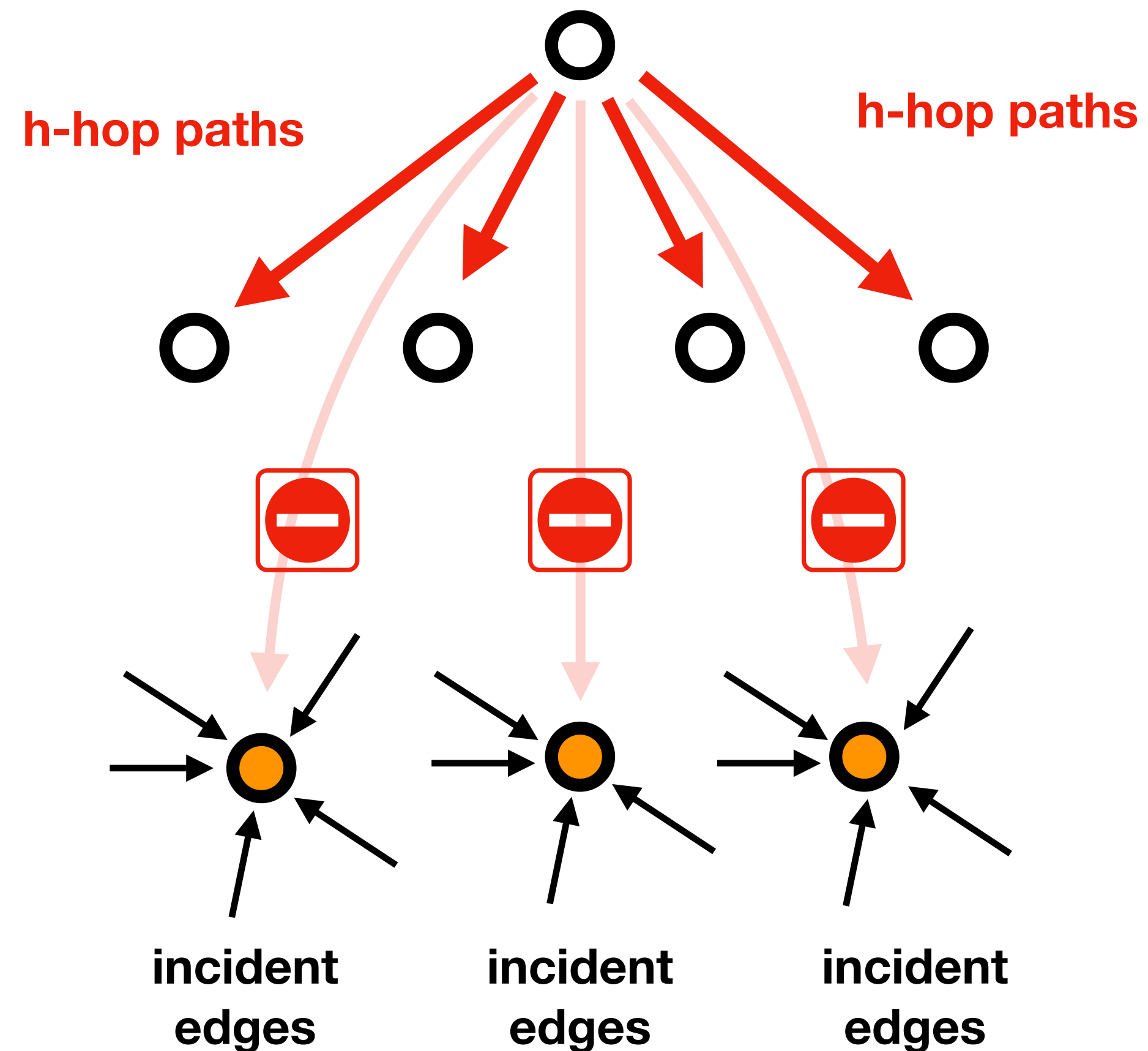


A deterministic  
**hitting set**  
for  **$h/2$ -hop** paths

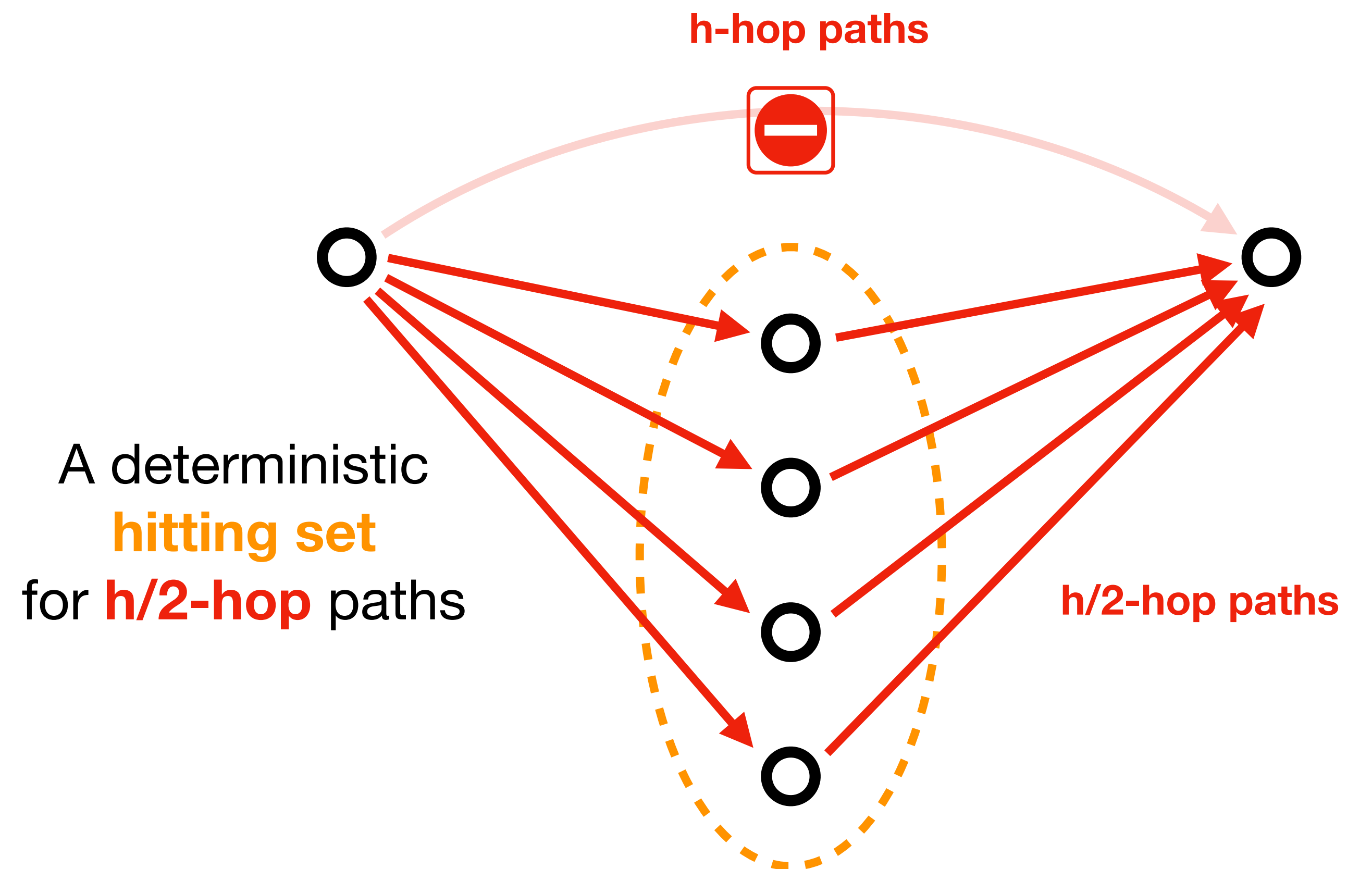


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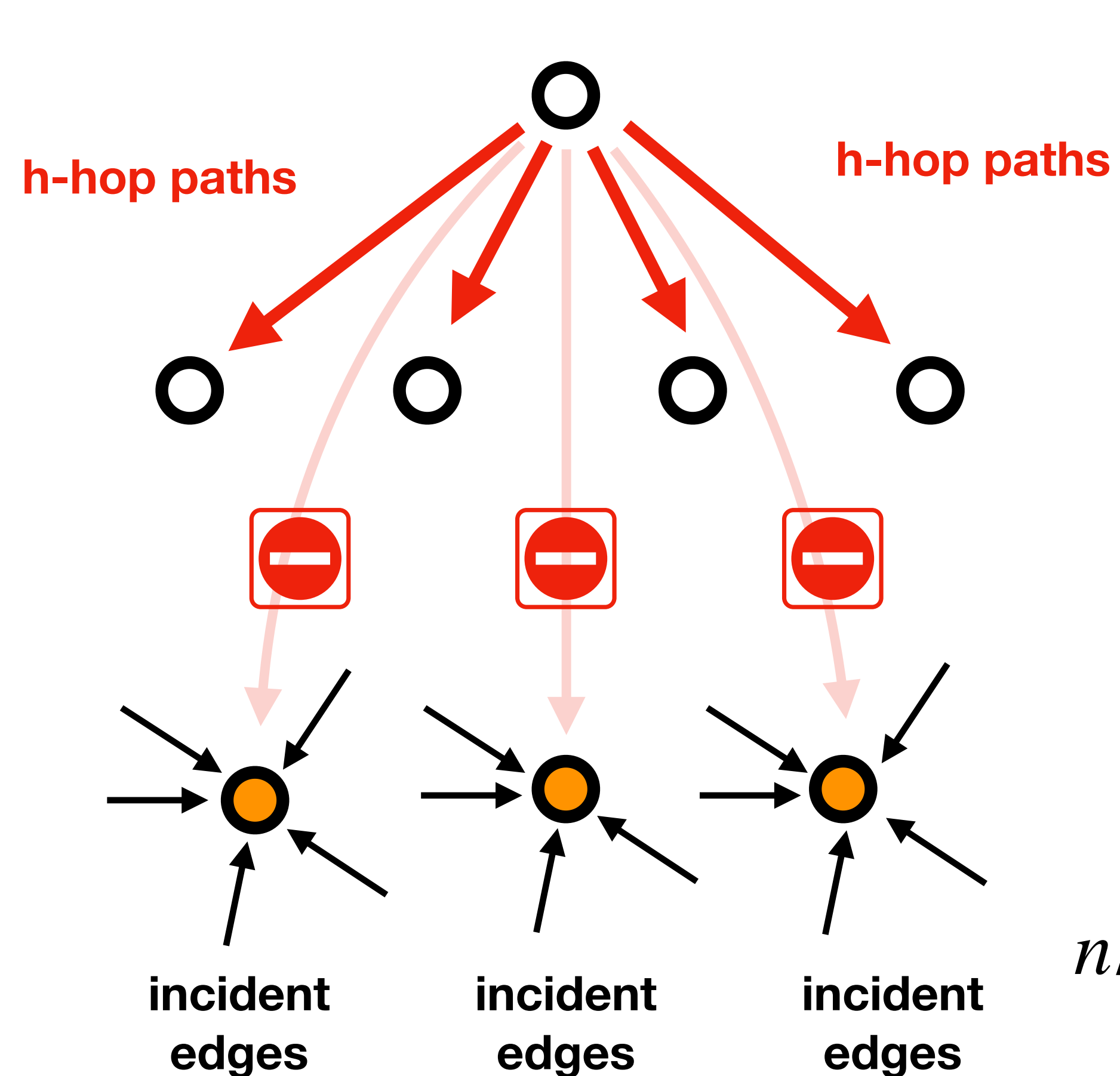
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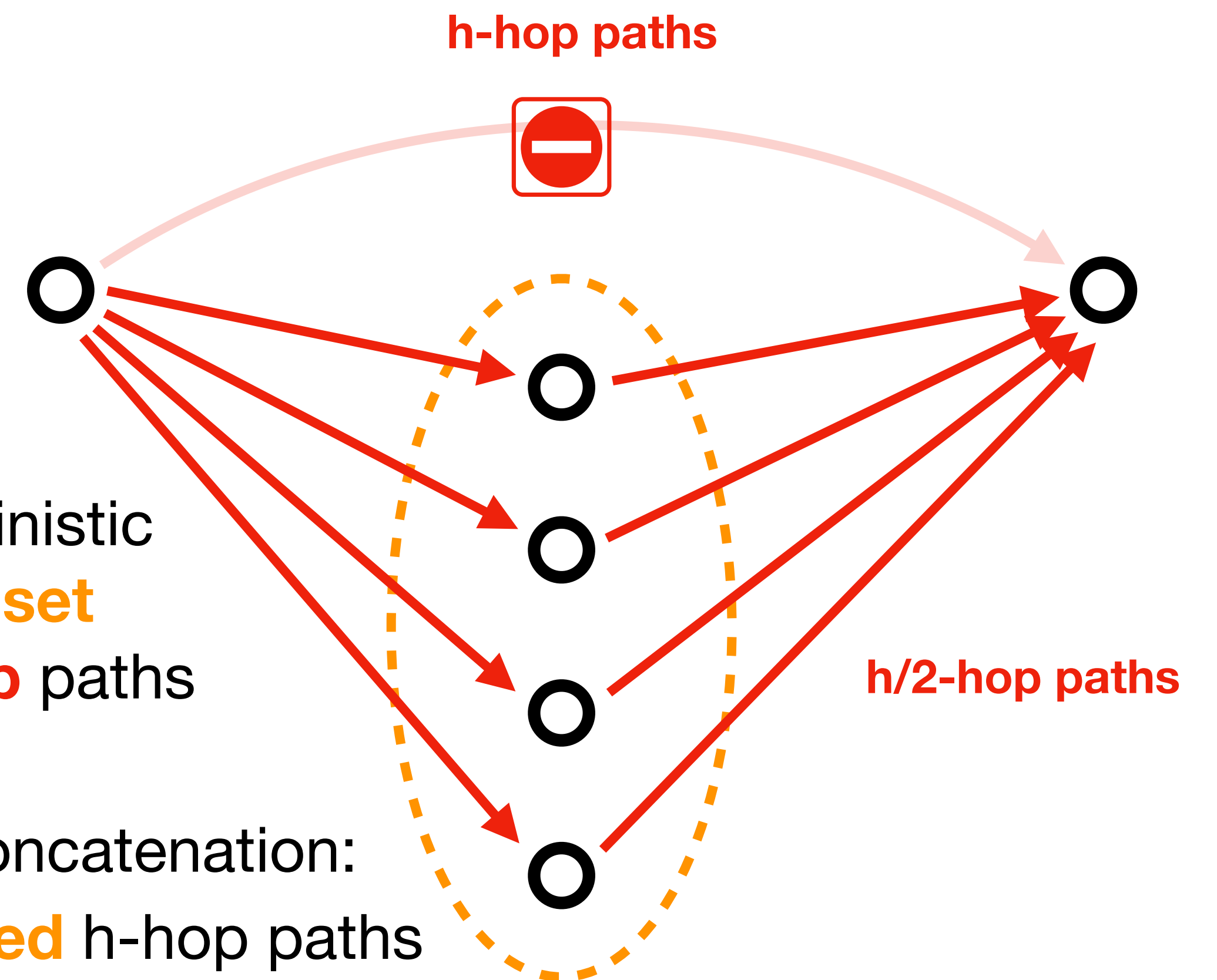
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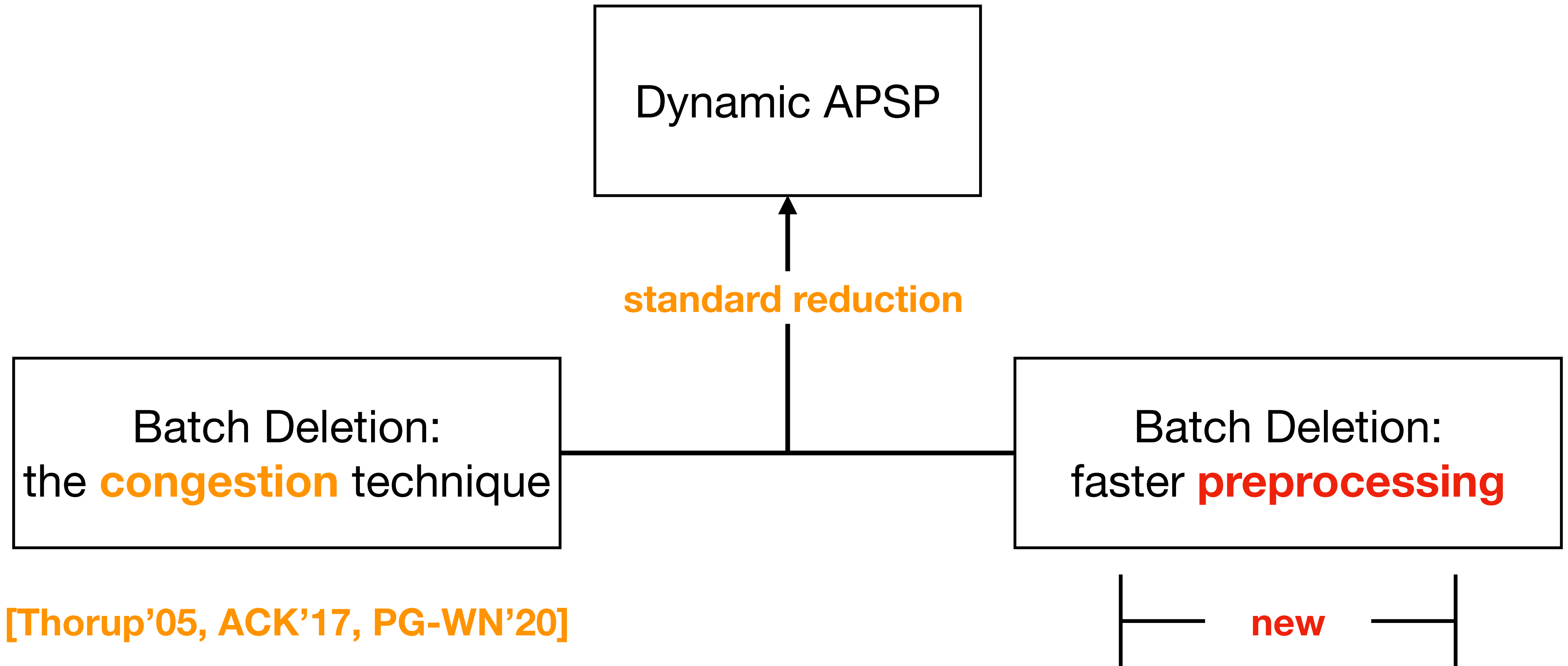
A deterministic  
**hitting set**  
for  **$h/2$ -hop paths**

Runtime of concatenation:  
 $n/h \cdot$  **#destroyed** h-hop paths



**Our improvement**

# Outline



# Decremental hop-restricted shortest paths

Low-congestion shortest paths [Thorup'05]

1. Pick a vertex  $v$  that maximizes  $cg(v)$
2. Compute  $h$ -hop shortest paths at  $v$  using Bellman-Ford
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Decremental  $h$ -hop shortest paths:

1. Adversary picks a vertex  $v$
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3. Adversary **deletes an arbitrary** vertex
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# Decremental hop-restricted shortest paths

Trivial algorithm:

- Apply Bellman-Ford for h-hop SSSP
- Total time =  $n^2h \cdot \text{\#deletions}$

Faster runtime?

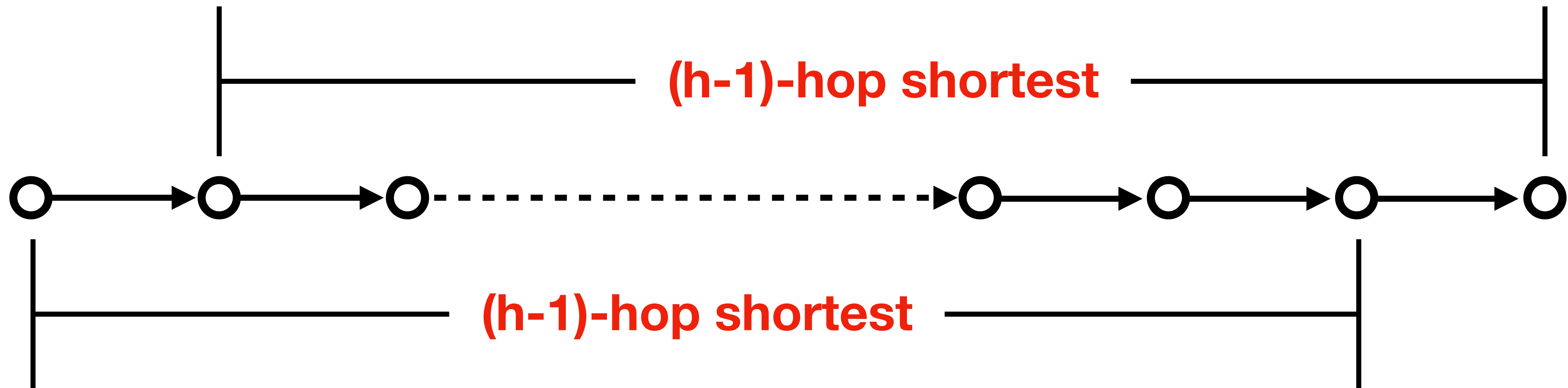
- Try to **maintain all h-hop paths** under vertex deletions

Decremental h-hop shortest paths:

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# Locally h-hop shortest paths

- Adapt the idea of **locally shortest paths** in [Demetrescu and Italiano, 2004]
- A path  $\langle u_0, u_1, \dots, u_k \rangle$  is **locally h-hop shortest**, if both of the sub-paths  $\langle u_0, u_1, \dots, u_{k-1} \rangle$  and  $\langle u_1, \dots, u_k \rangle$  are **(h-1)-hop shortest paths**

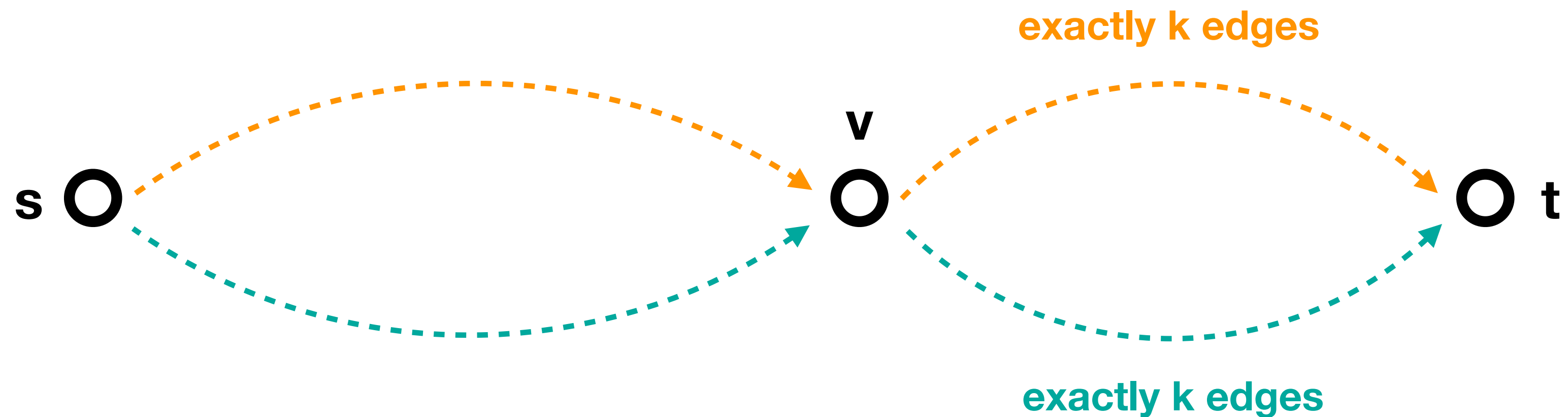


# Locally h-hop shortest paths

**Shortest** locally h-hop shortest paths = h-hop shortest paths

#(locally h-hop) can be bounded

- Each vertex  $v$  is on **at most  $h$**  different **locally h-hop paths** from  $s$  to  $t$
- At most  $n^3 \log n$  **(all-pairs locally h-hop) in total**

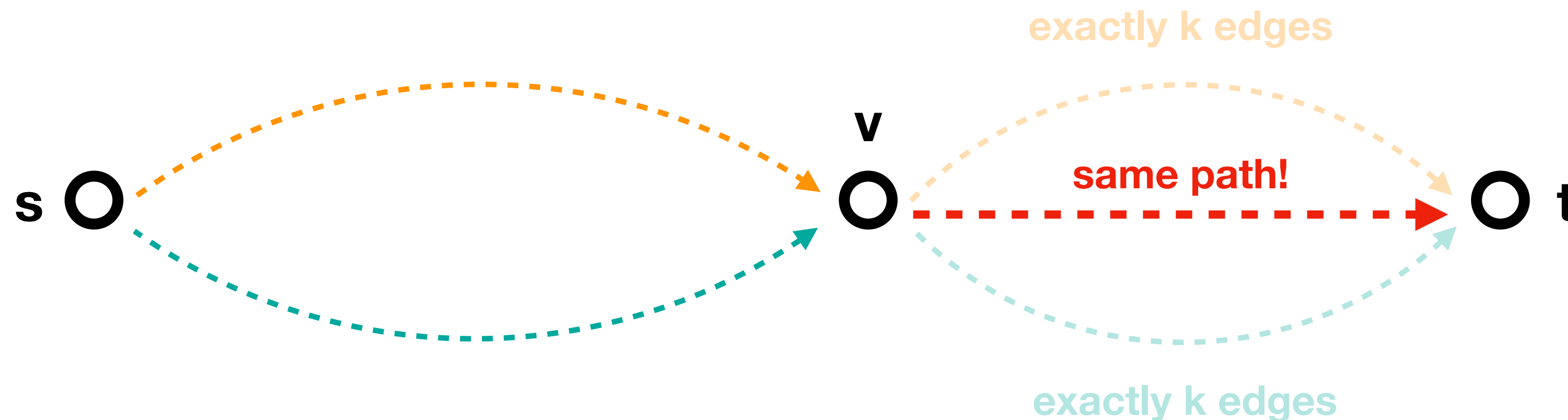


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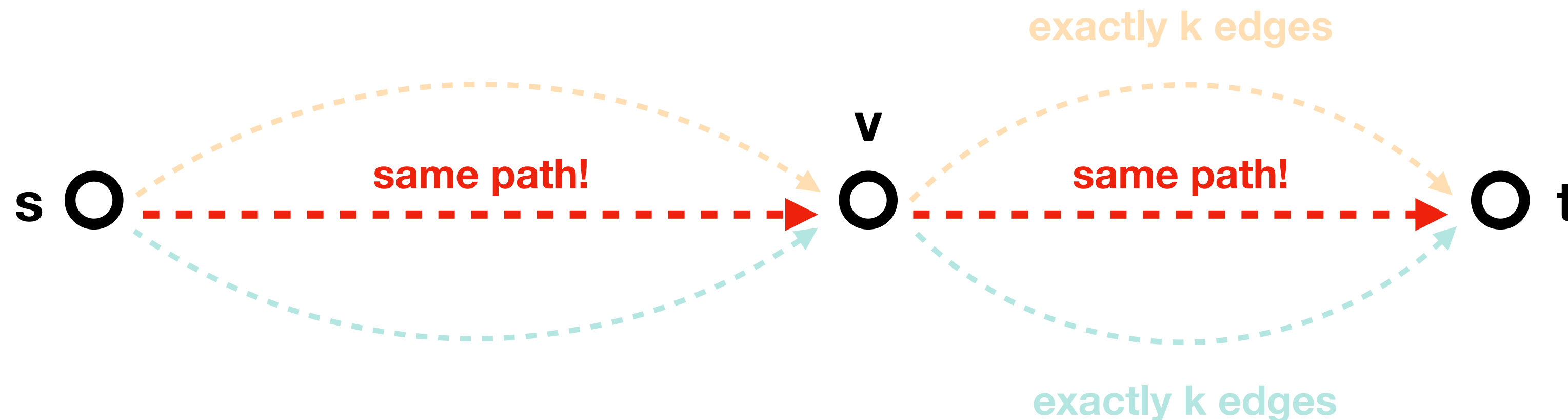


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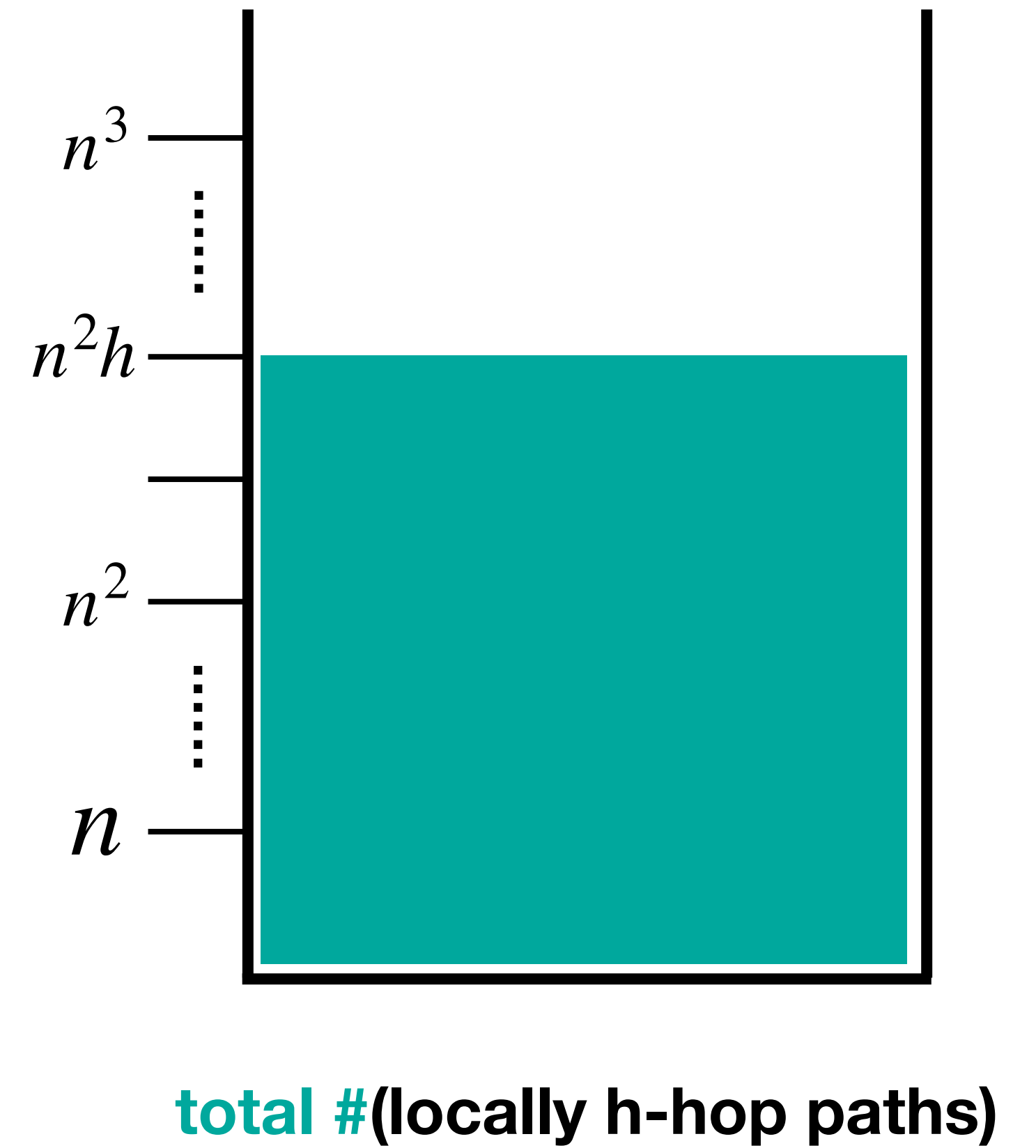
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# Decremental locally h-hop shortest paths

Maintain all locally h-hop shortest paths under **vertex deletions**

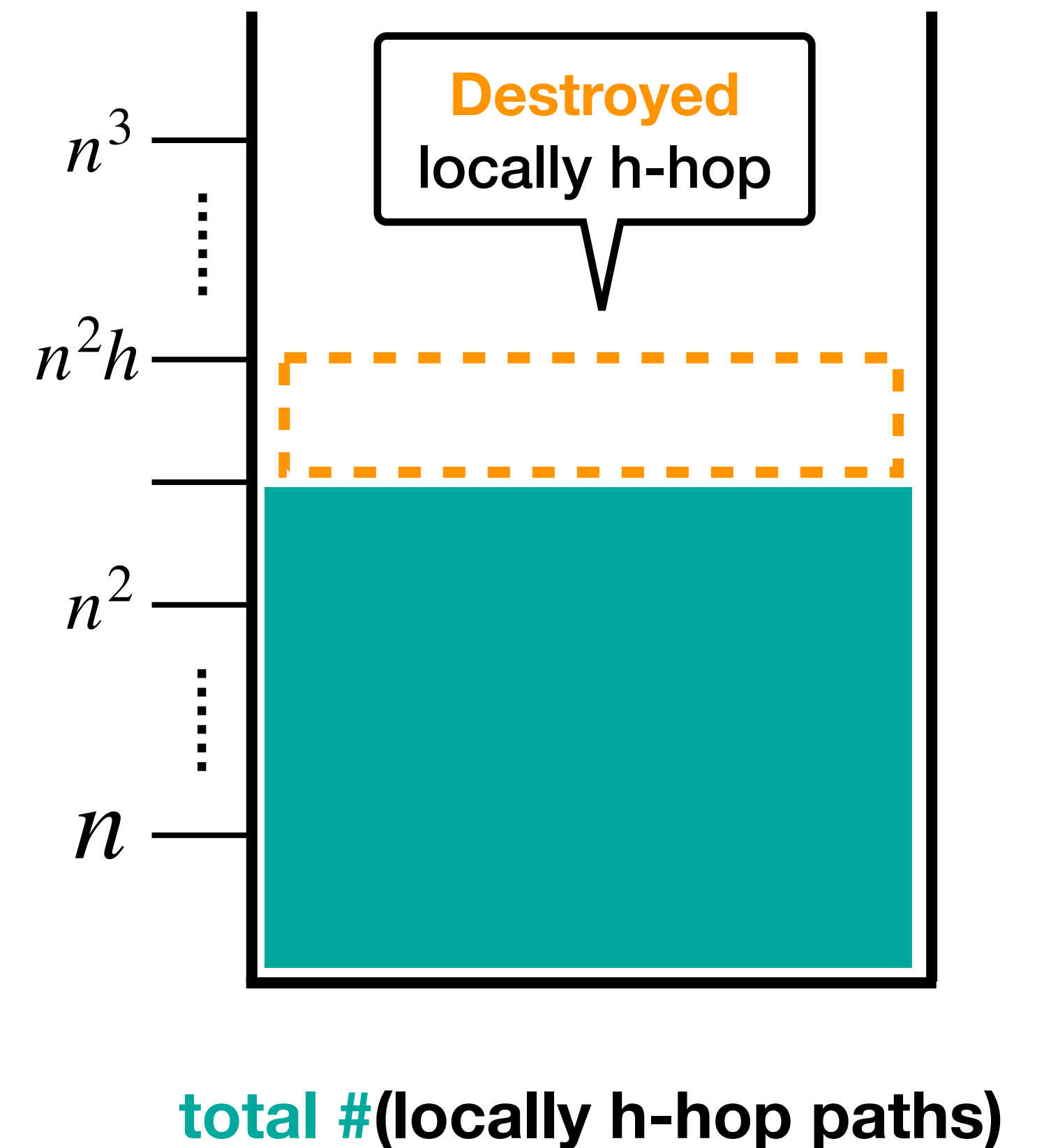


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$$\text{Runtime} = \#(\text{destroyed}) \leq n^2h$$



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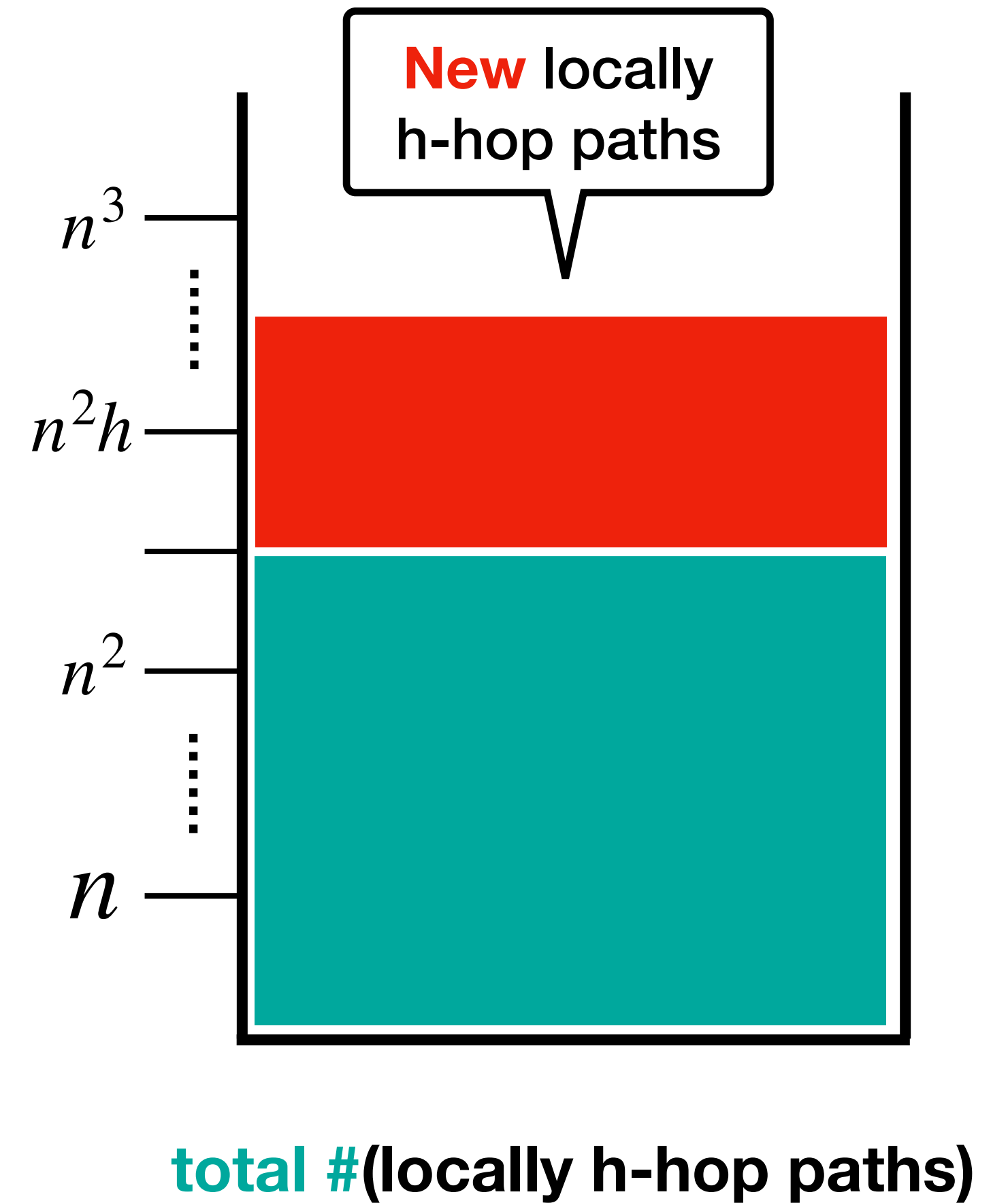
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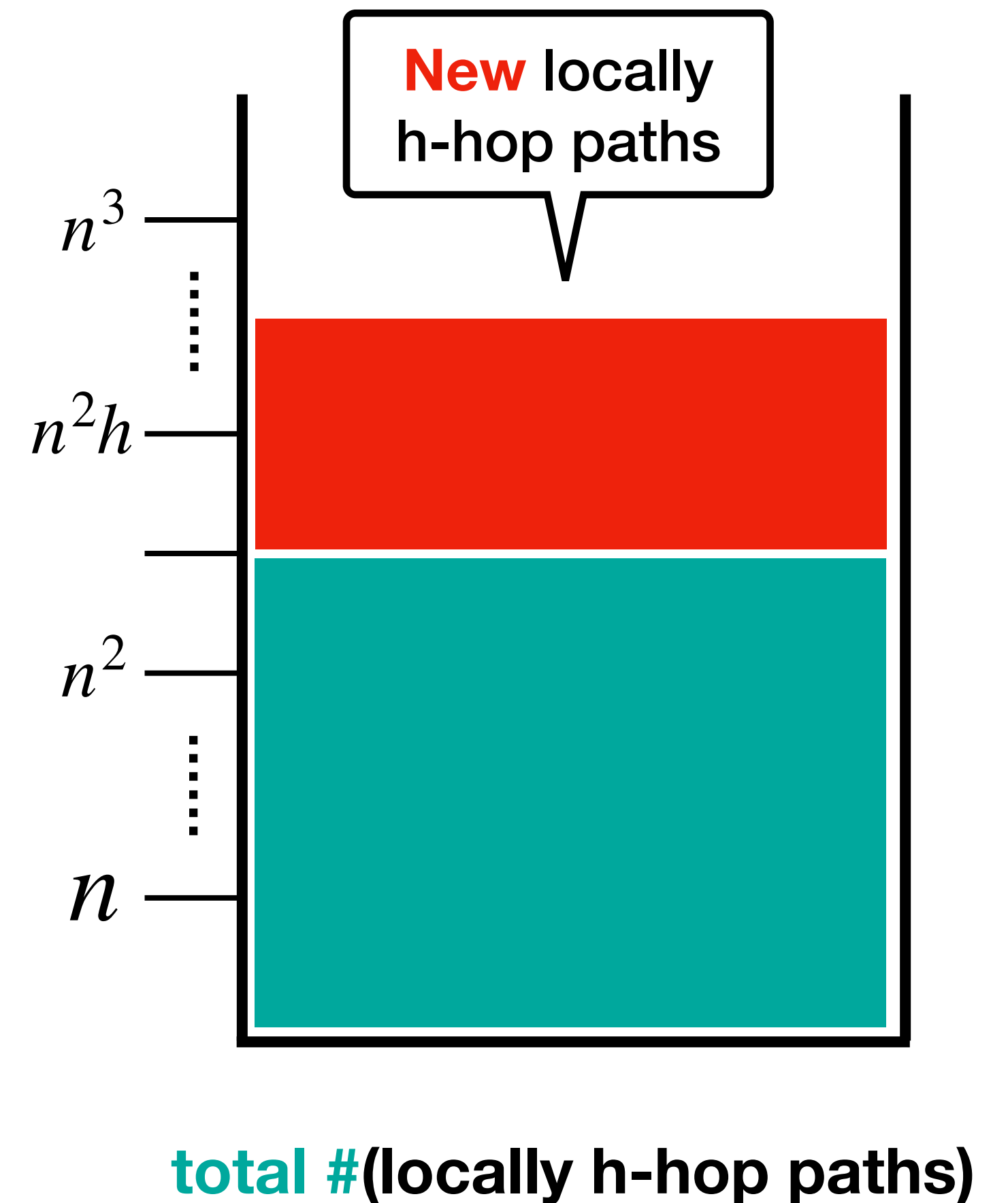
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3. Output-sensitive  $\rightarrow$  total time =  $n^3h$



# Accelerating Bellman-Ford

## Original goal:

1. Adversary picks a vertex  $v$
2. Compute  **$h$ -hop SSSP at  $v$**
3. Adversary **deletes an arbitrary** vertex
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Trivial algorithm:

- Apply Bellman-Ford for  $h$ -hop SSSP
- Total time =  $n^2h \cdot \text{\#deletions}$

Faster runtime?

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## Solution:

- Apply decremental  $g$ -hop paths
- Bellman-Ford runs in time  $n^2h/g$
- Total time =  $n^3(g + h/g) < n^3h$

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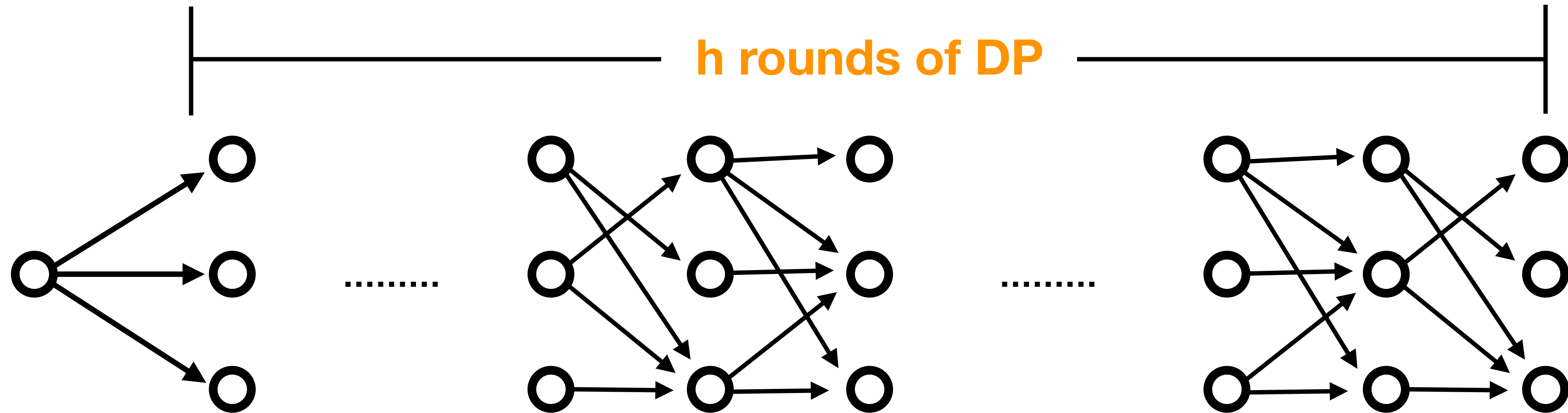
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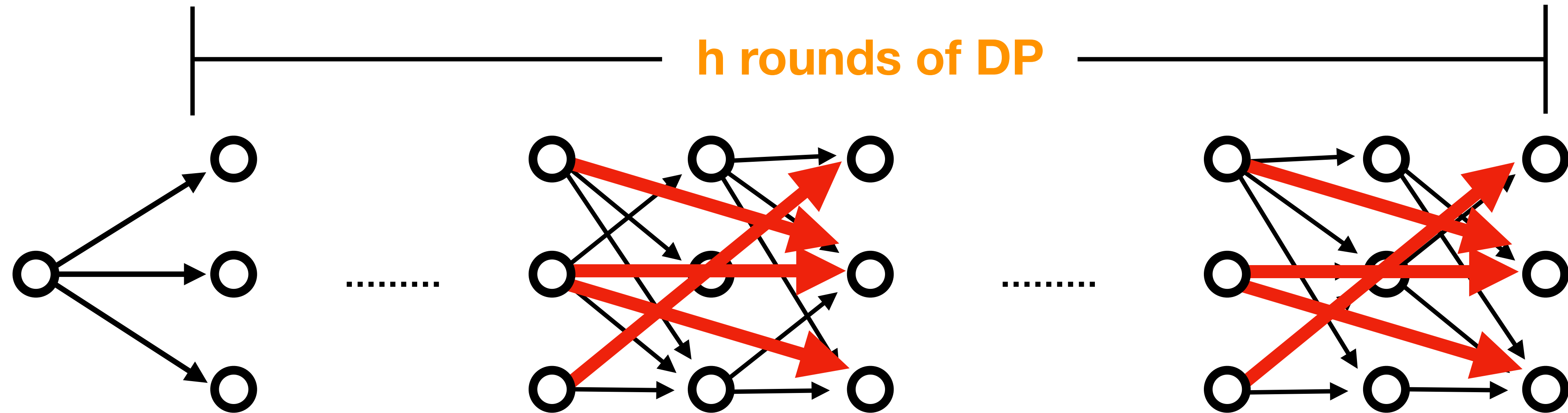
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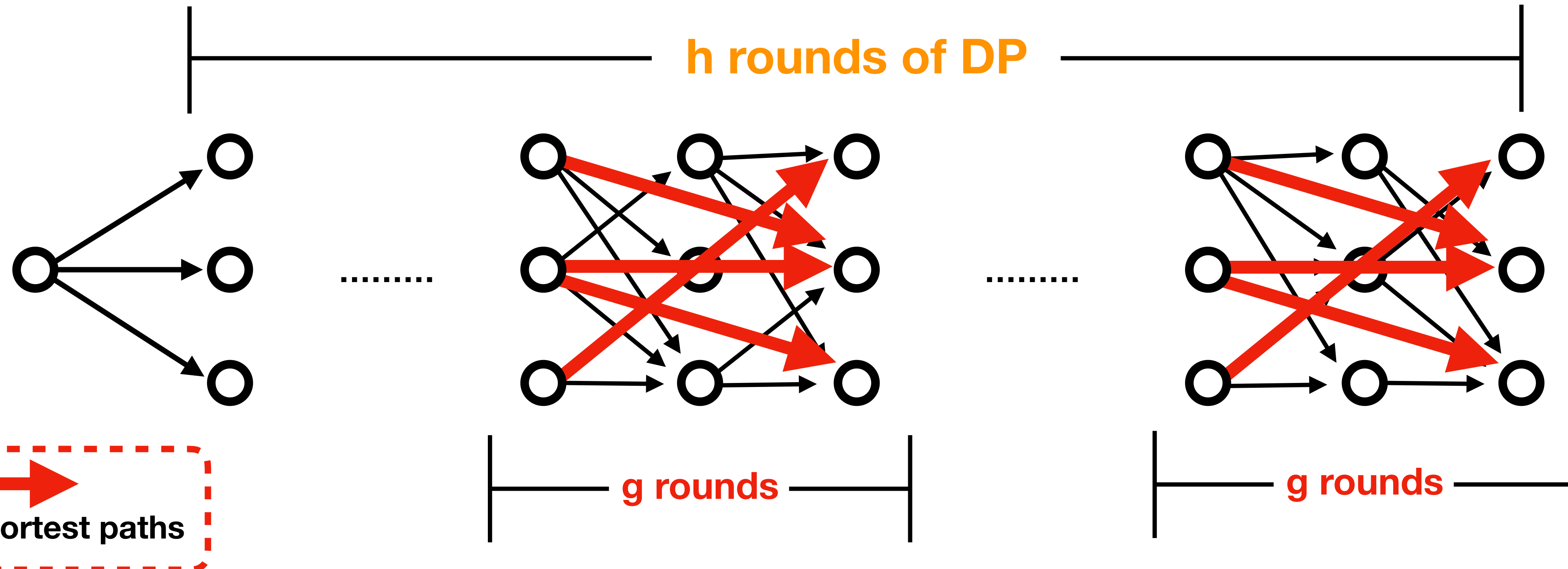


  
= **g-hop** shortest paths

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# Further Questions

- Faster randomized worst-case update time  $n^{3-1/3-\epsilon}$  ?
- Faster deterministic worst-case update time  $n^{3-1/3}$  ?

Thank you !