

# **Part 4: Structural Decompositions and Algorithms**

Friedrich Slivovsky

# Dynamic Programming on Tree Decompositions

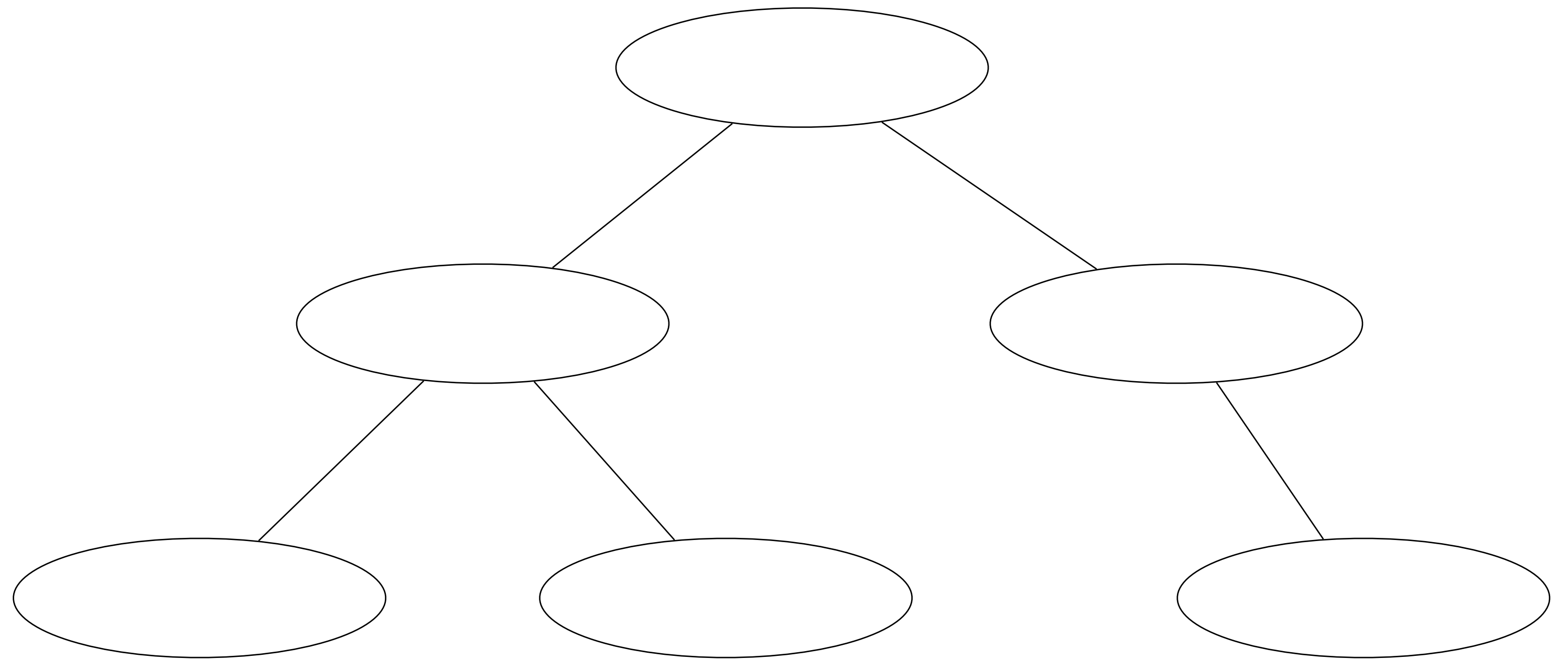
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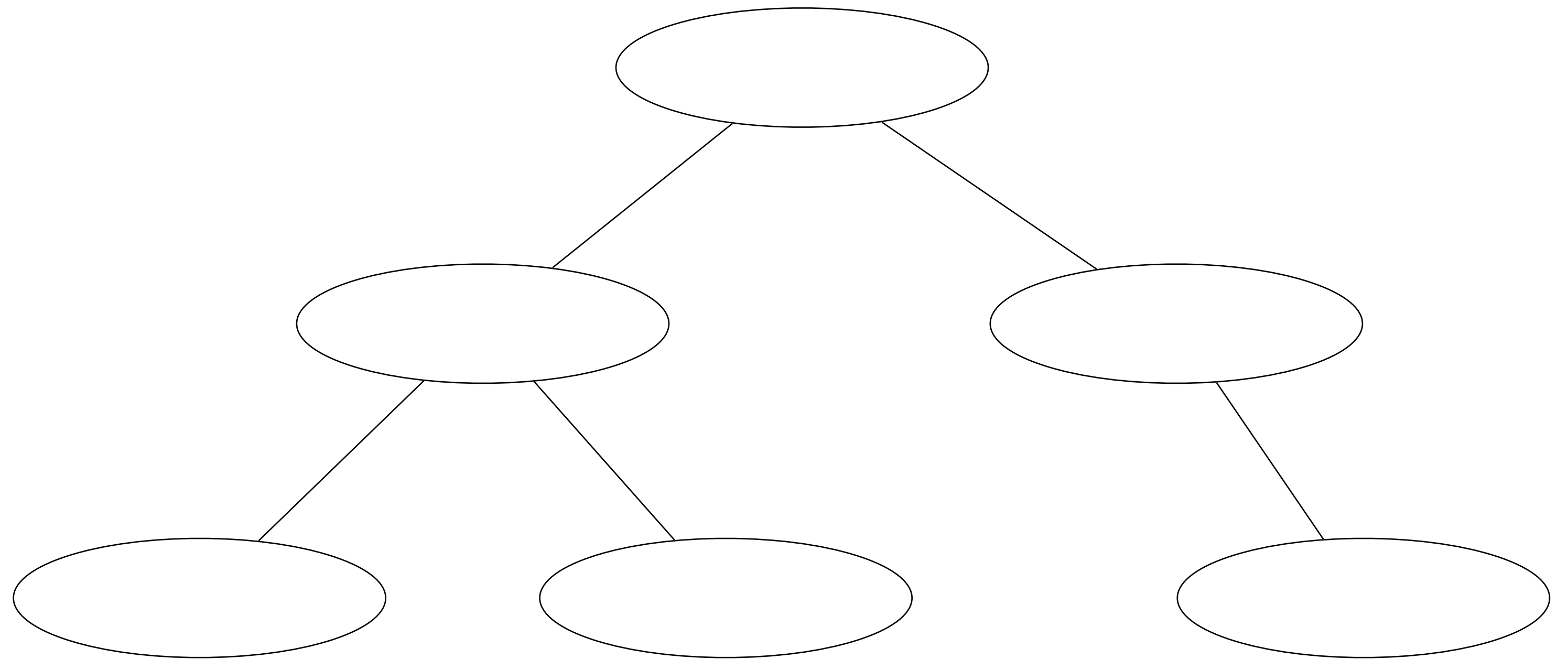
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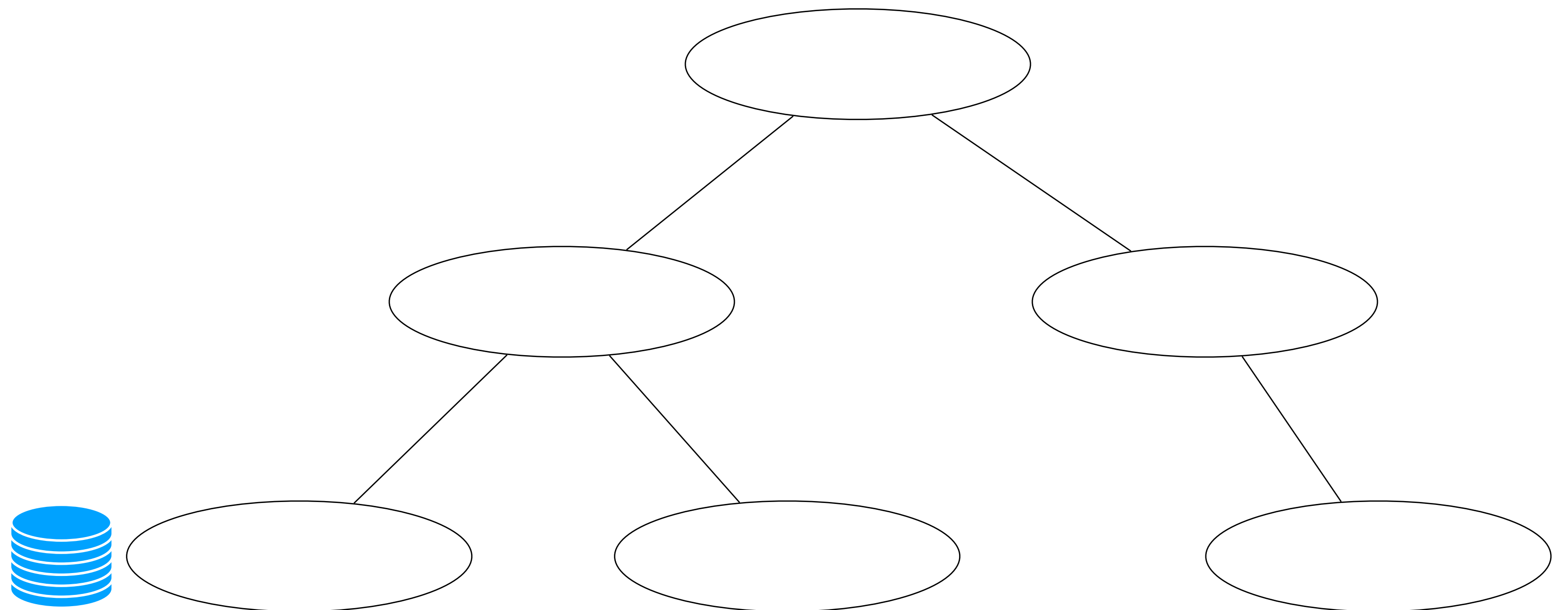
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2. Go from leaves to root and compute table for each node.



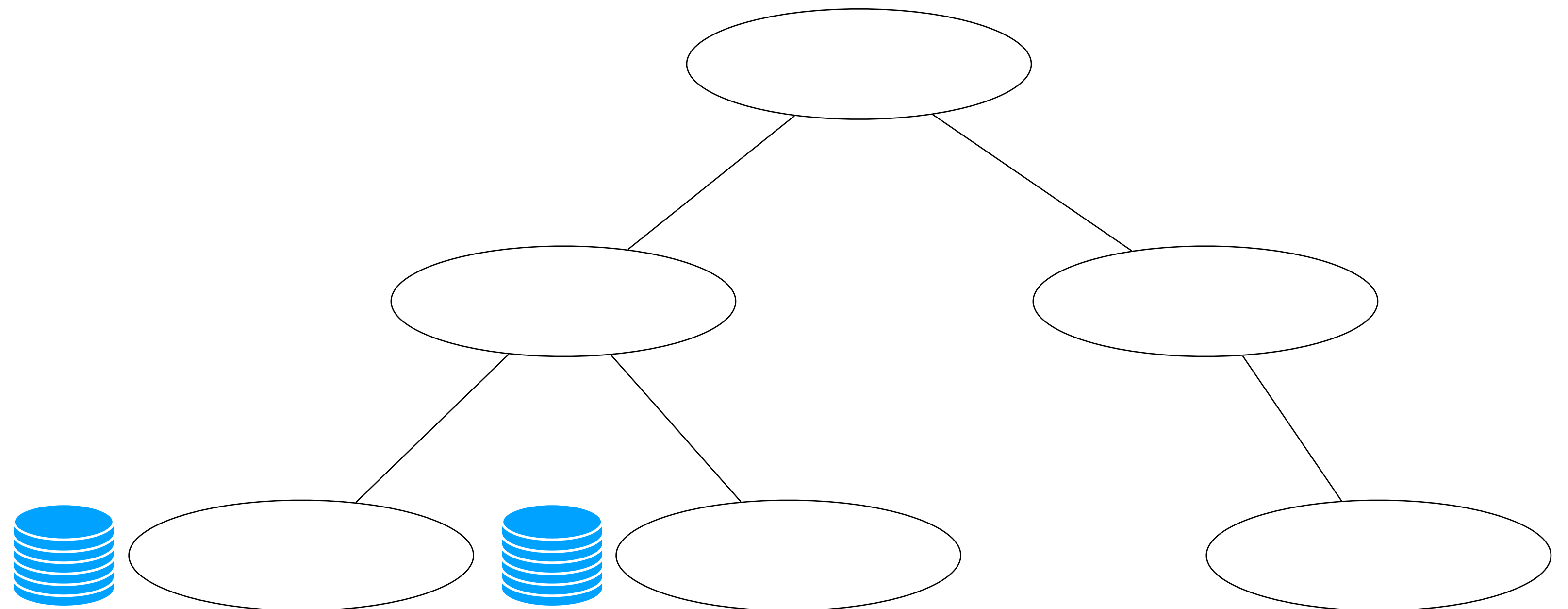
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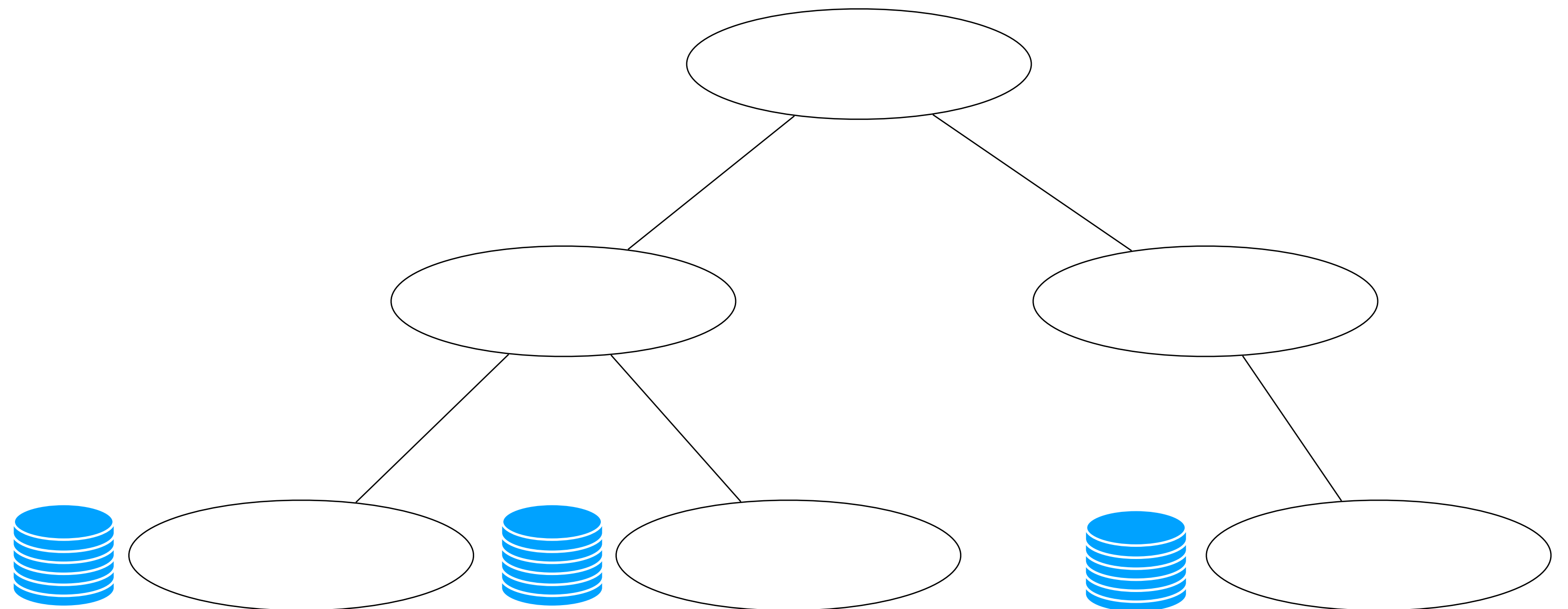
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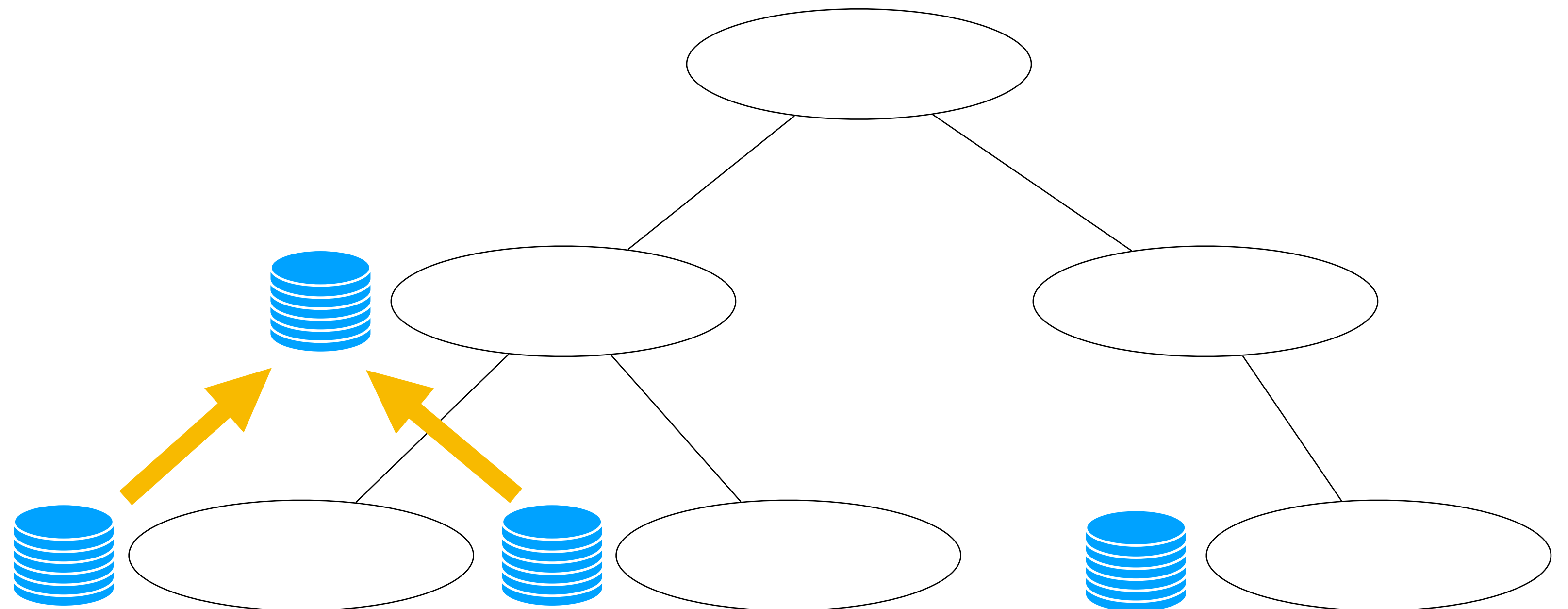
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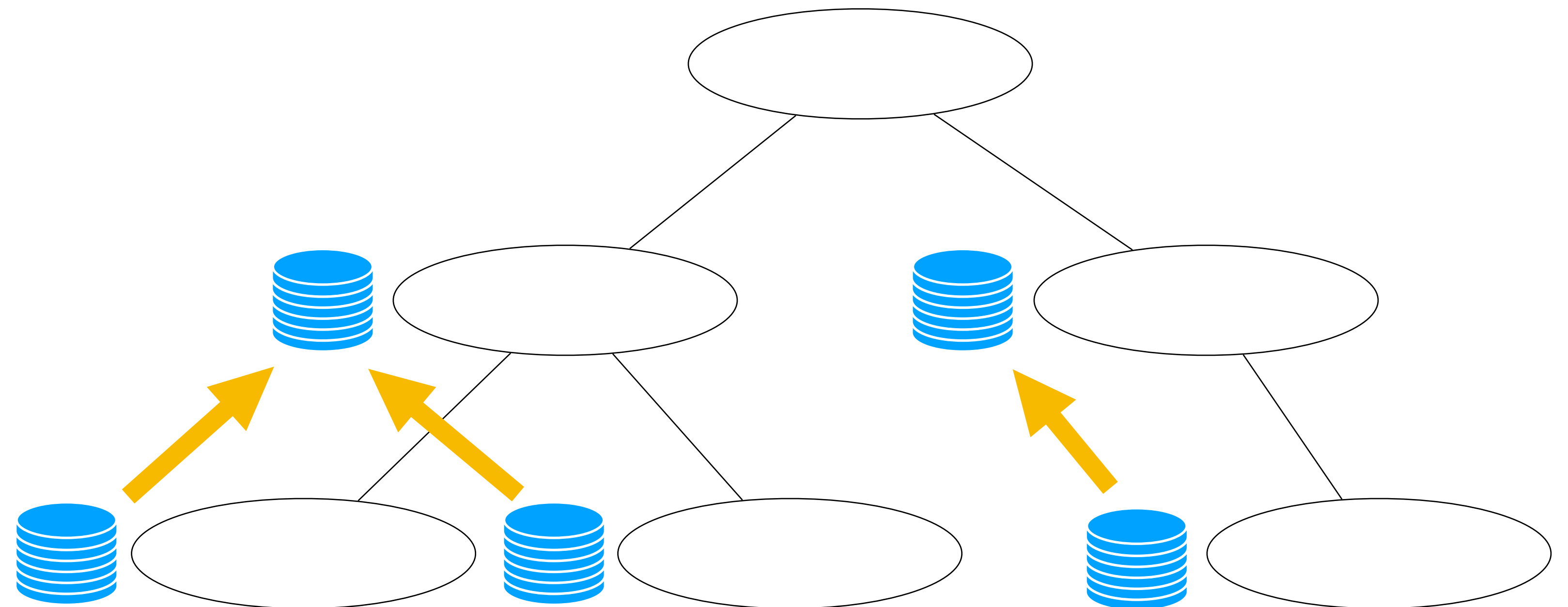
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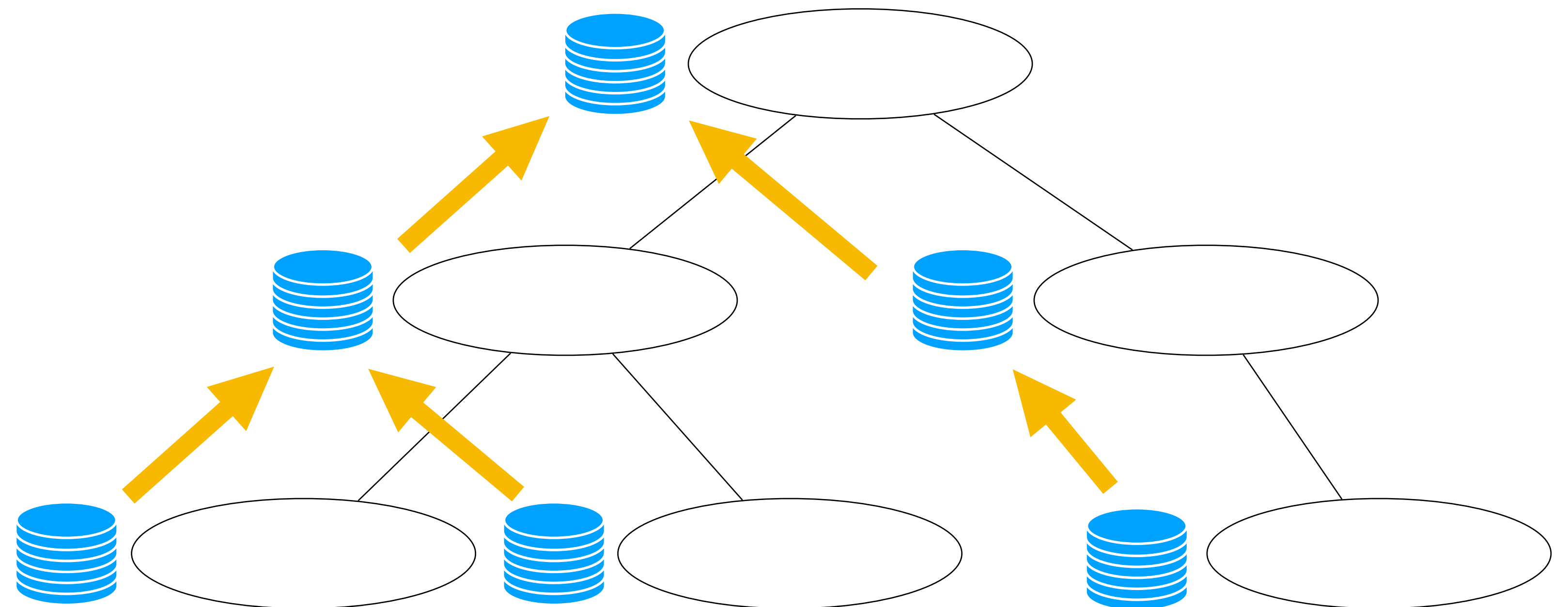
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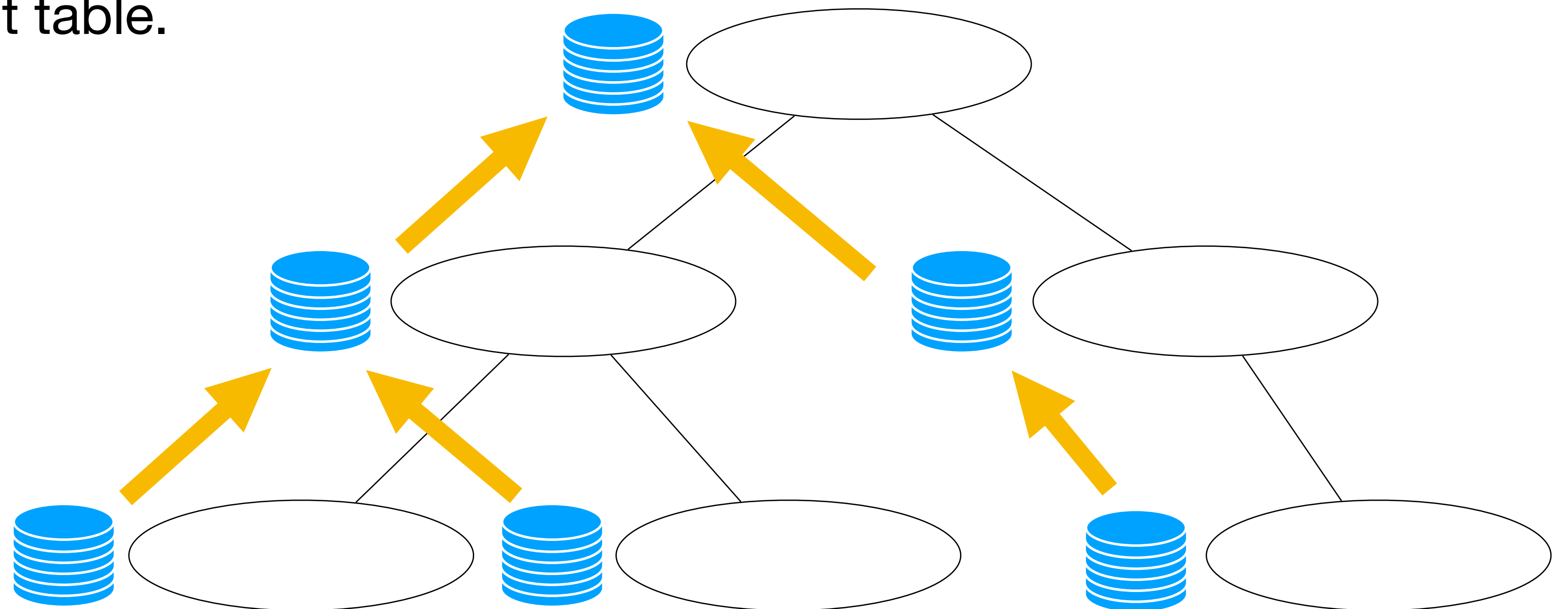
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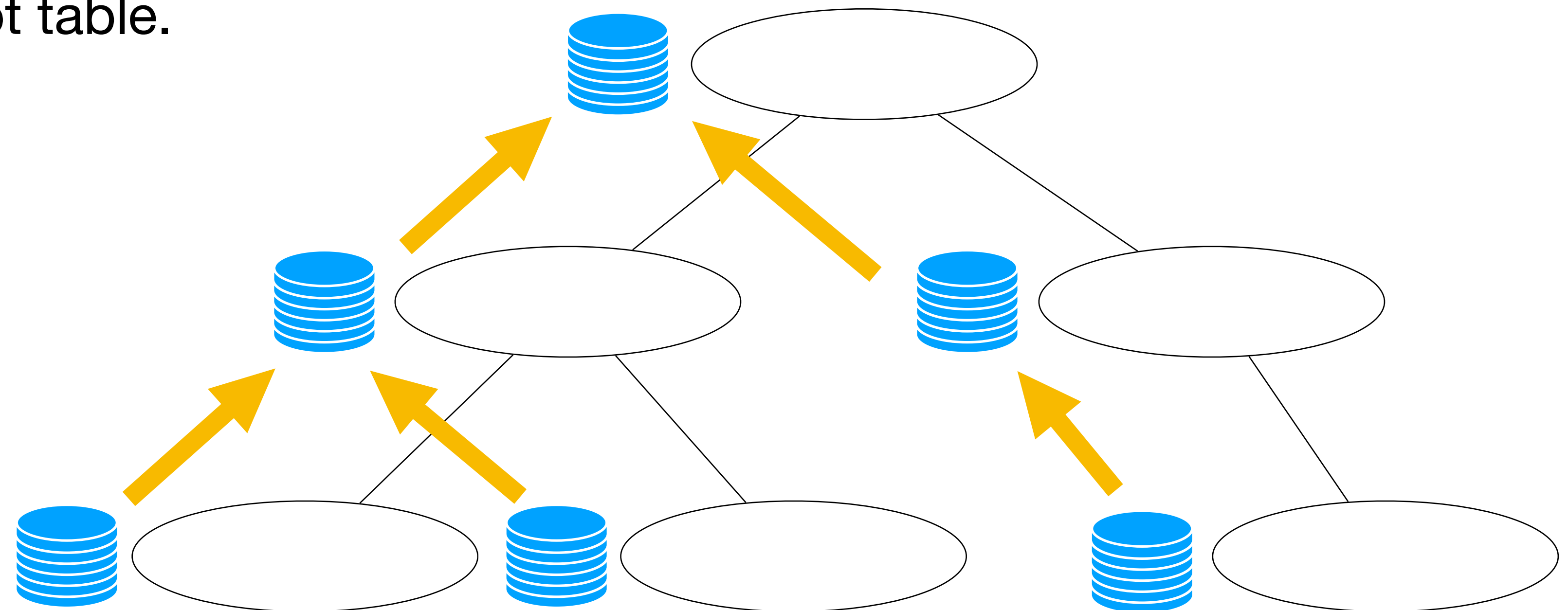
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3. Retrieve solution from root table.



# Dynamic Programming on Tree Decompositions

Treewidth  $k$

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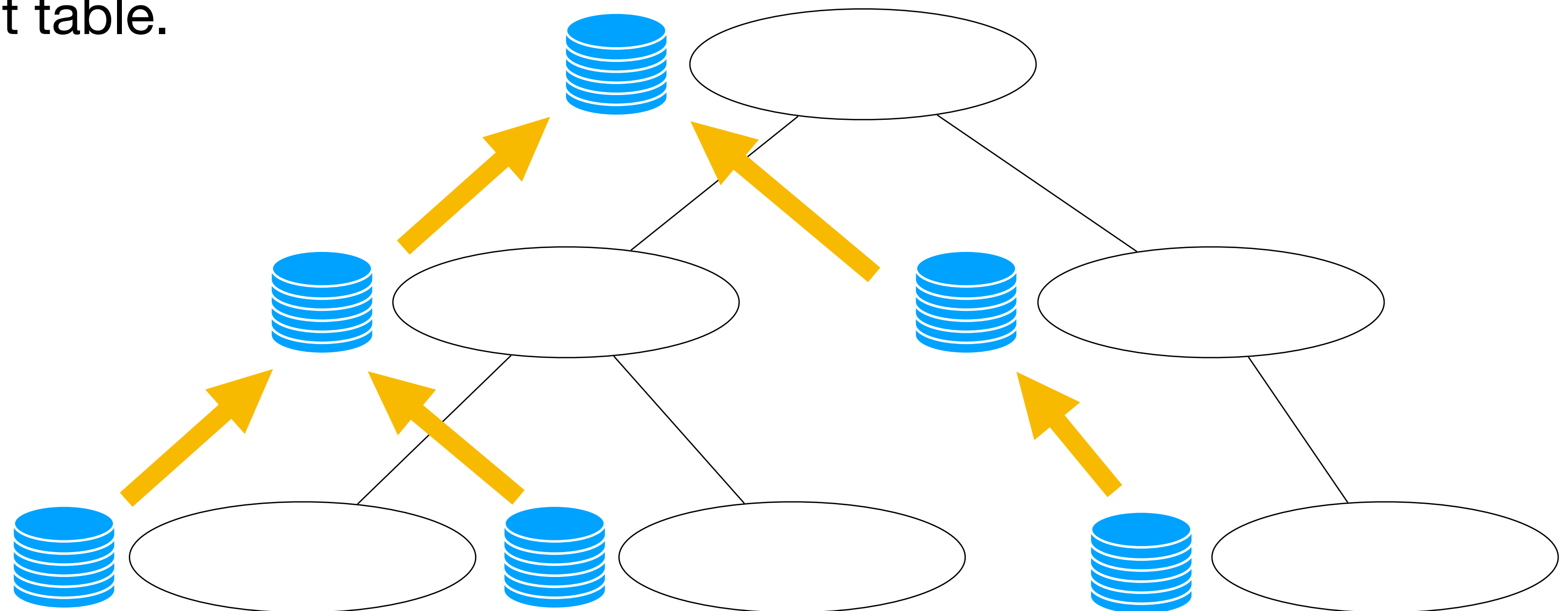


# Dynamic Programming on Tree Decompositions

$$f(k) p(n)$$

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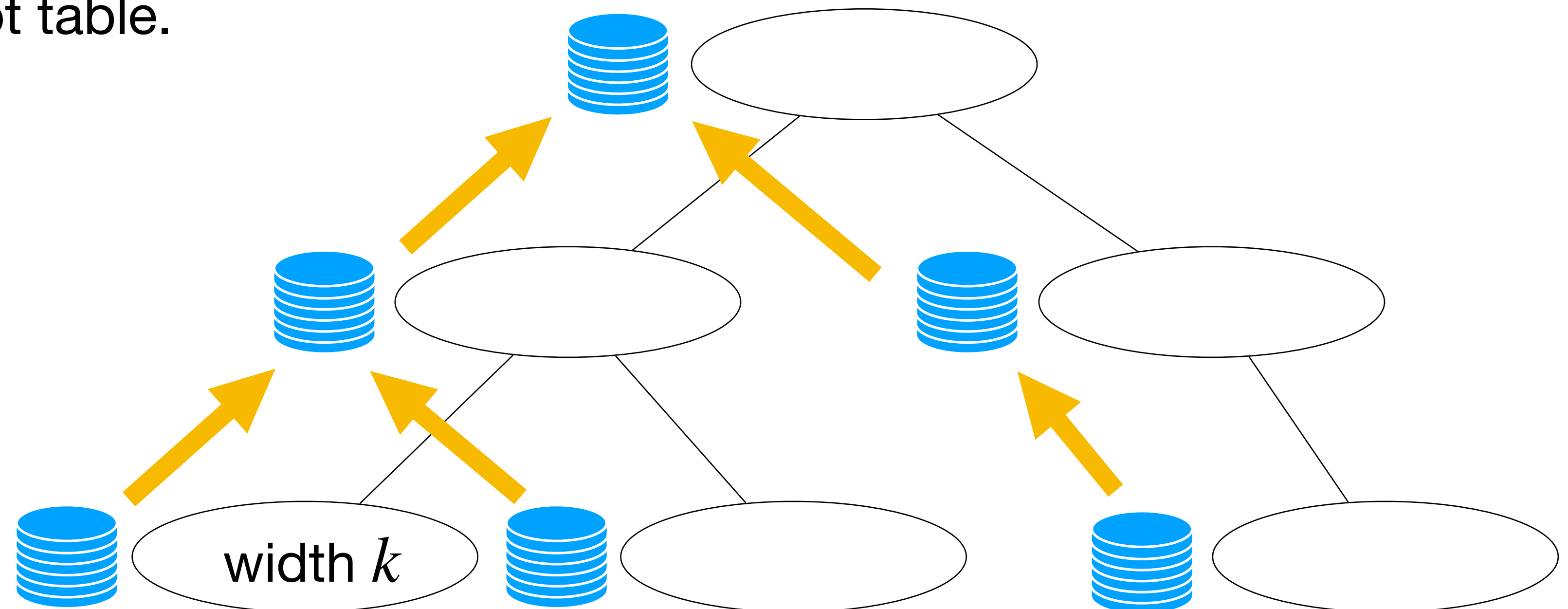


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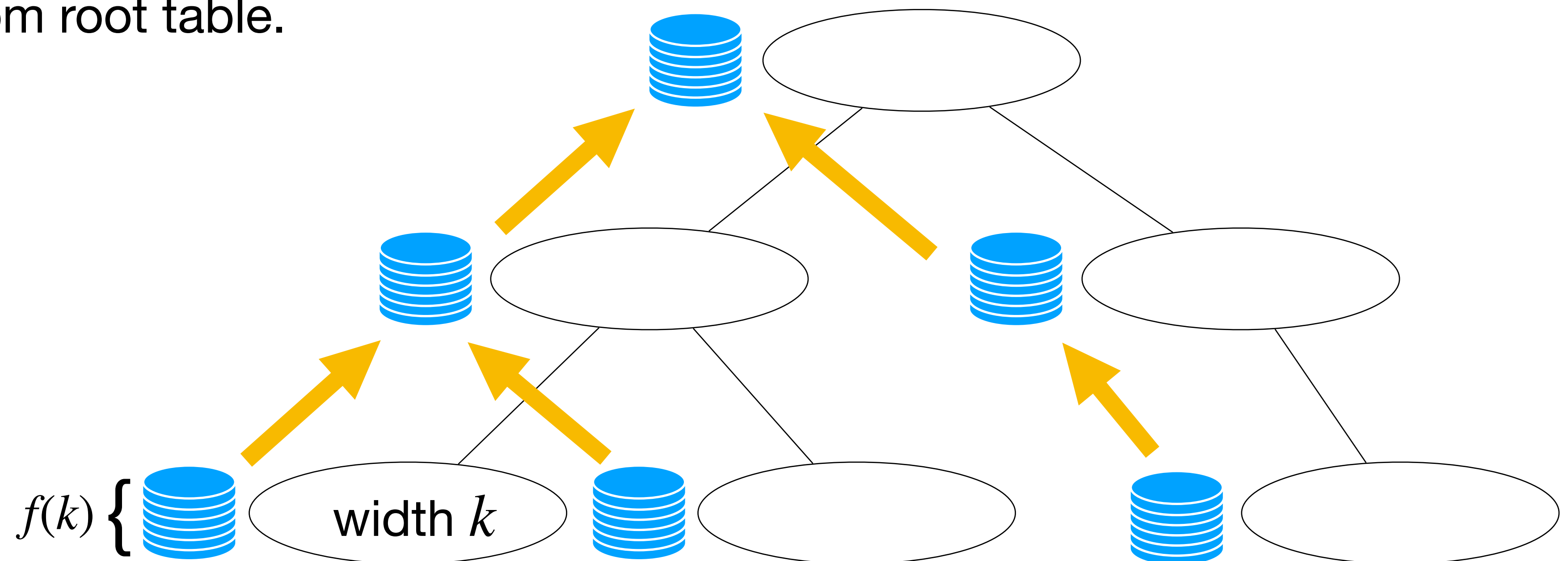


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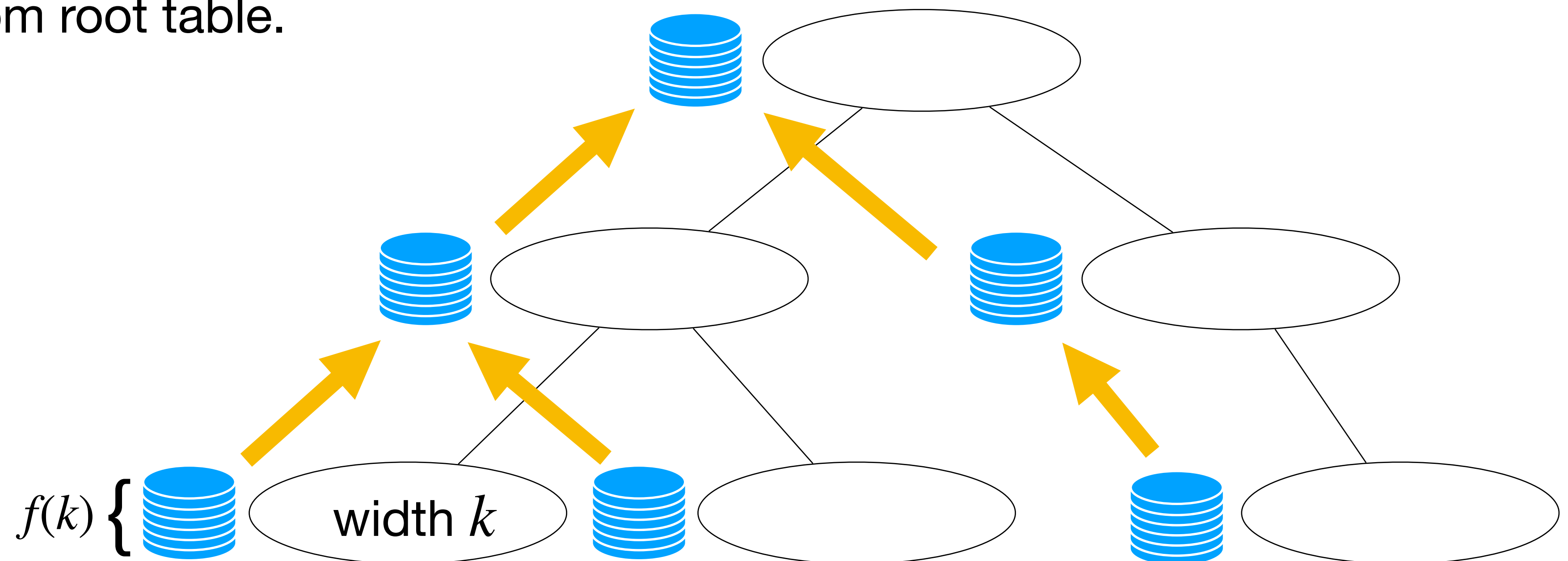
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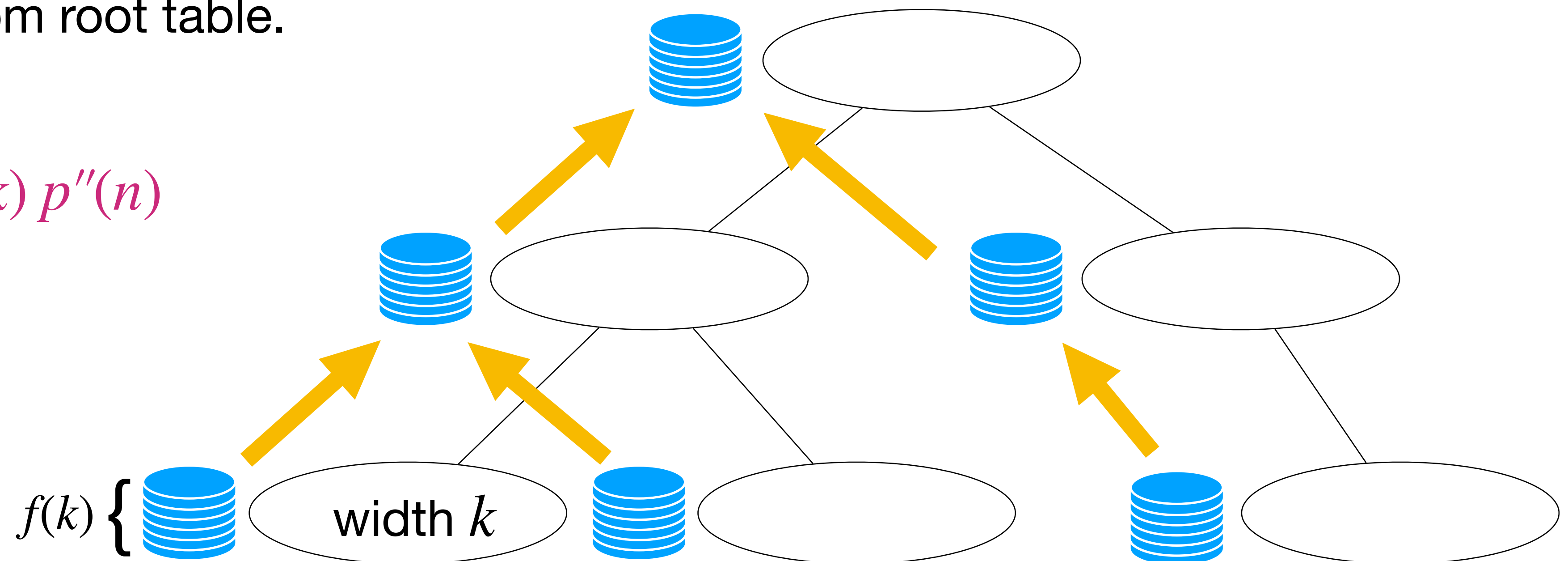
# Dynamic Programming on Tree Decompositions

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# Independent Set

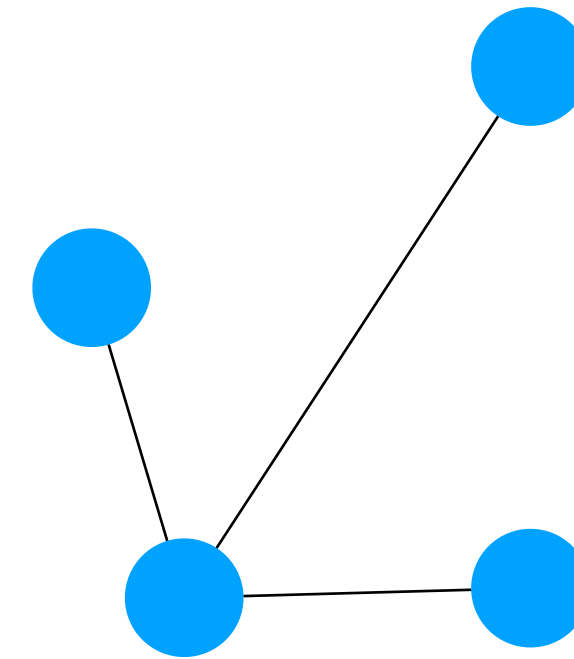
## Definition

An **independent set** of a graph is a subset of vertices such that no two vertices are adjacent.

# Independent Set

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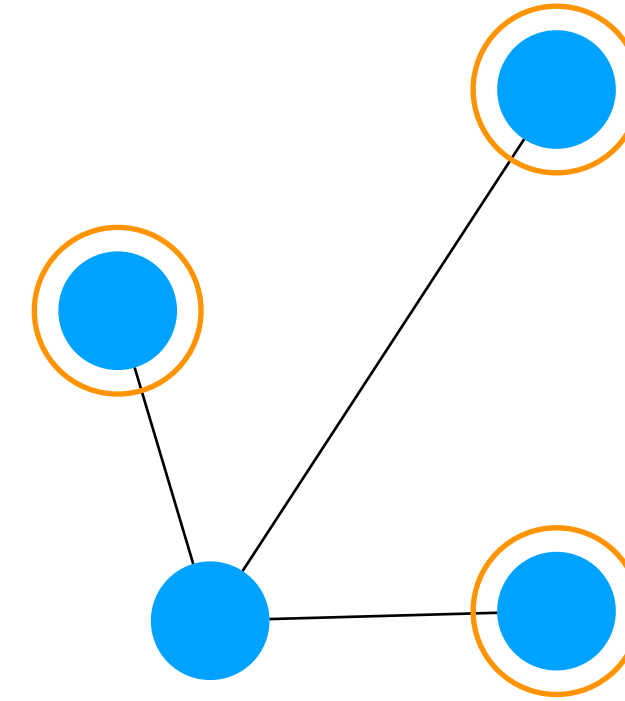
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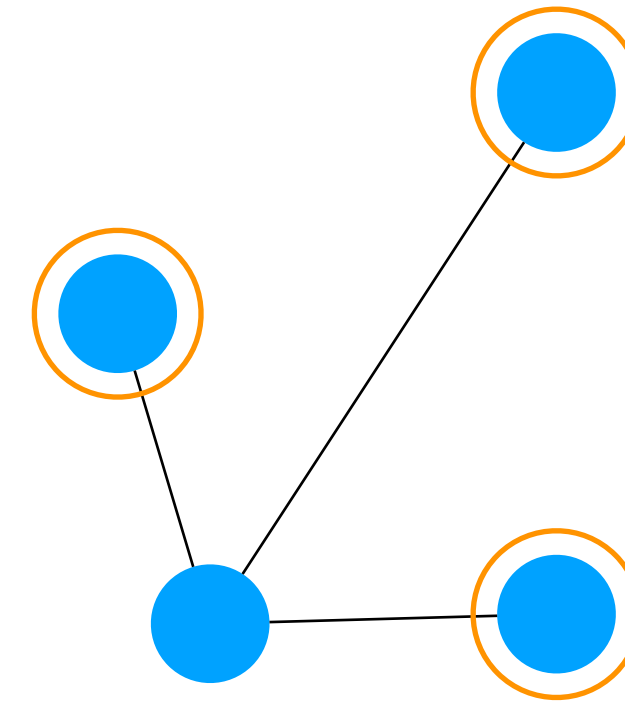
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## INDEPENDENT SET

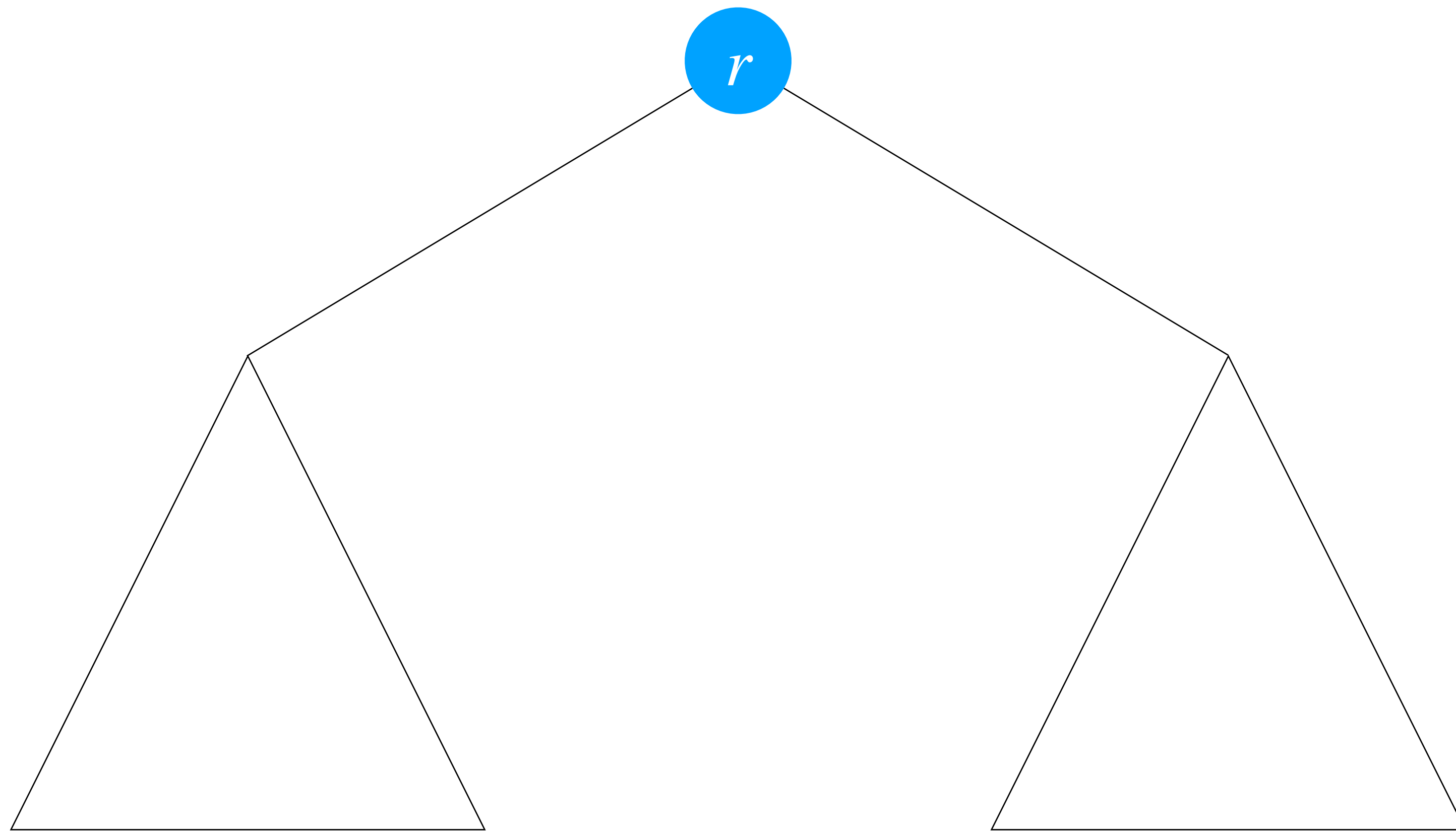
**Input:** A graph  $G$  and an integer  $k$ .

**Question:** Does  $G$  have an **IS** of size  $k$ ?

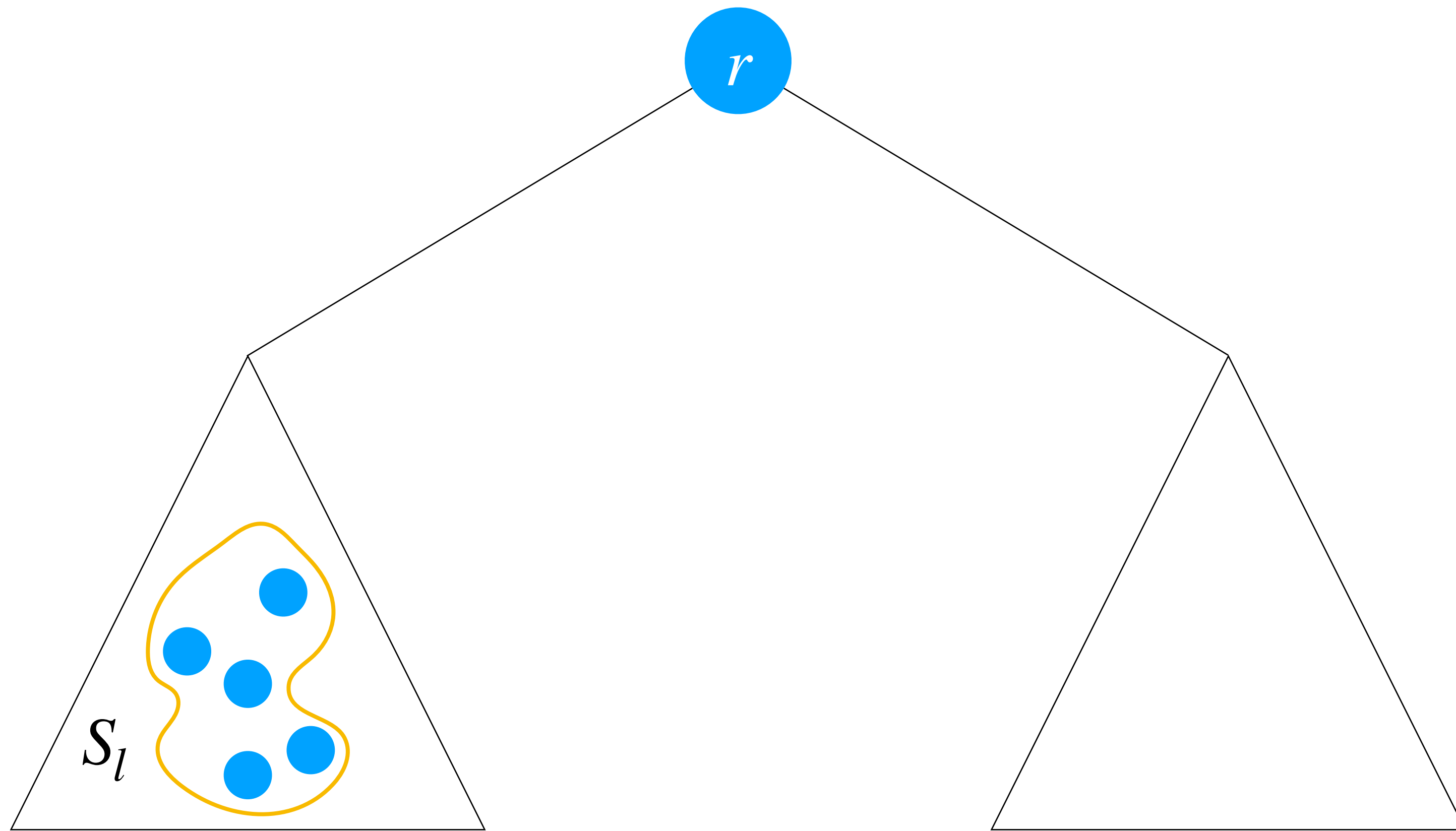
# Independent Set on Trees



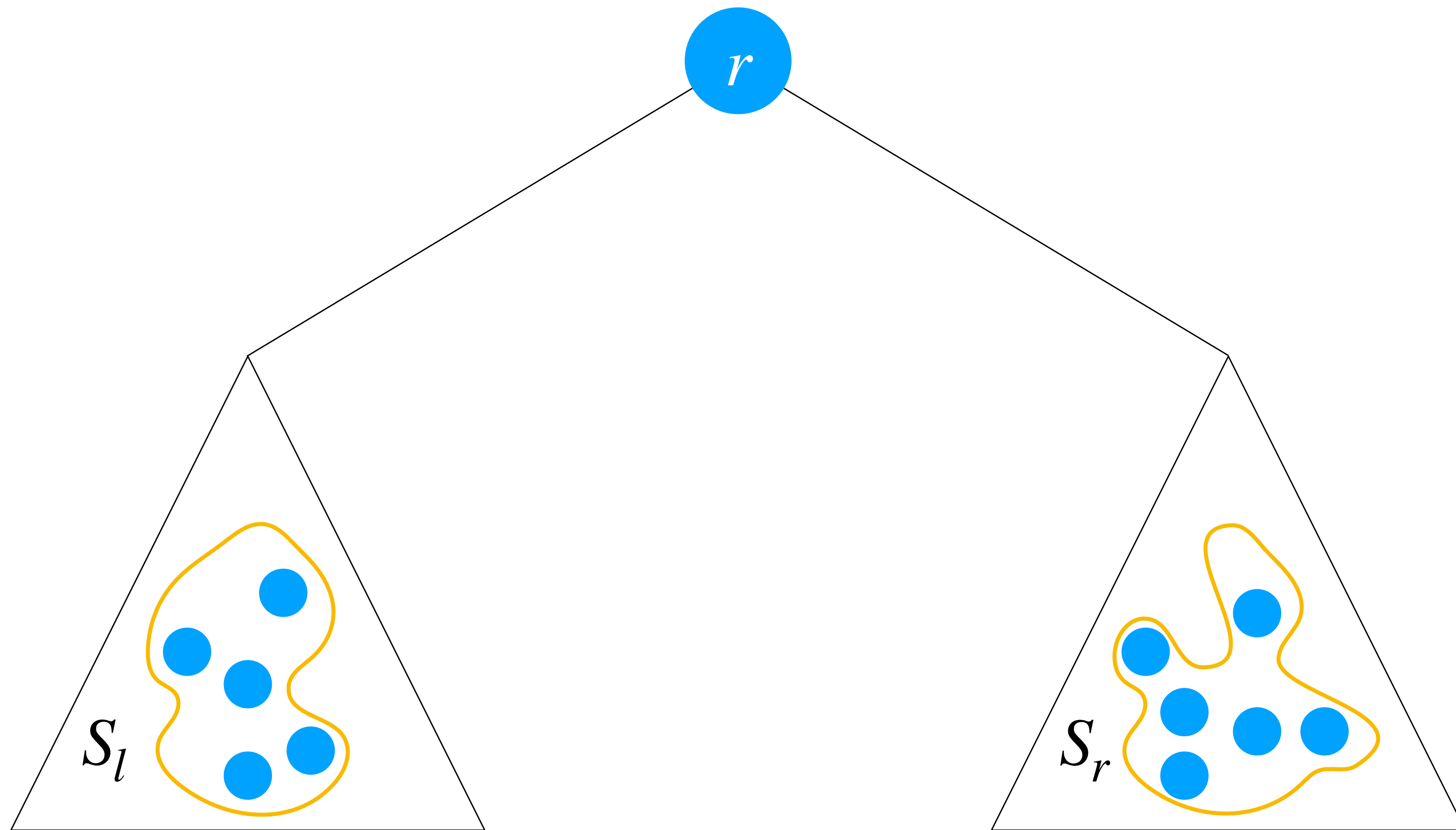
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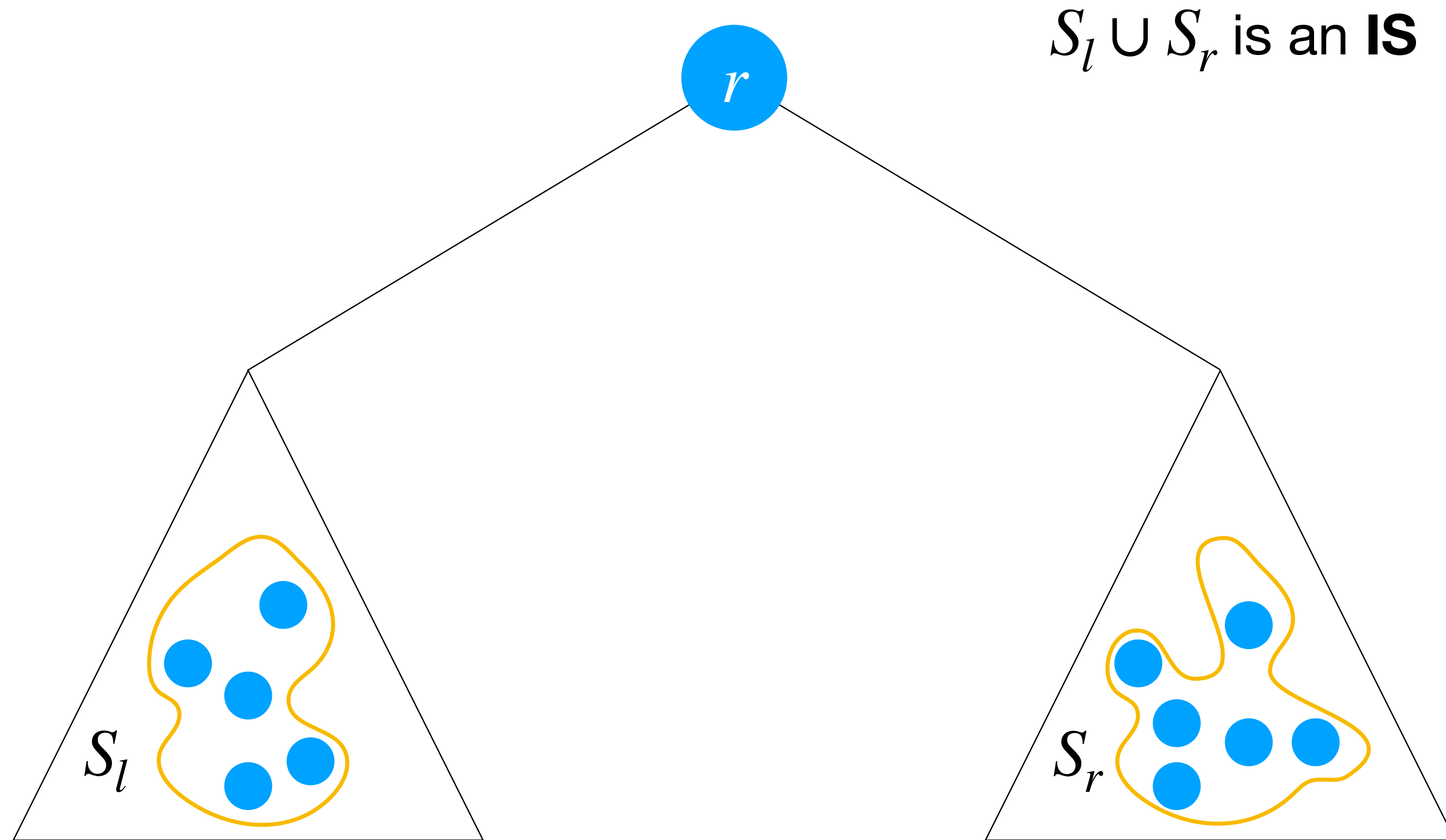
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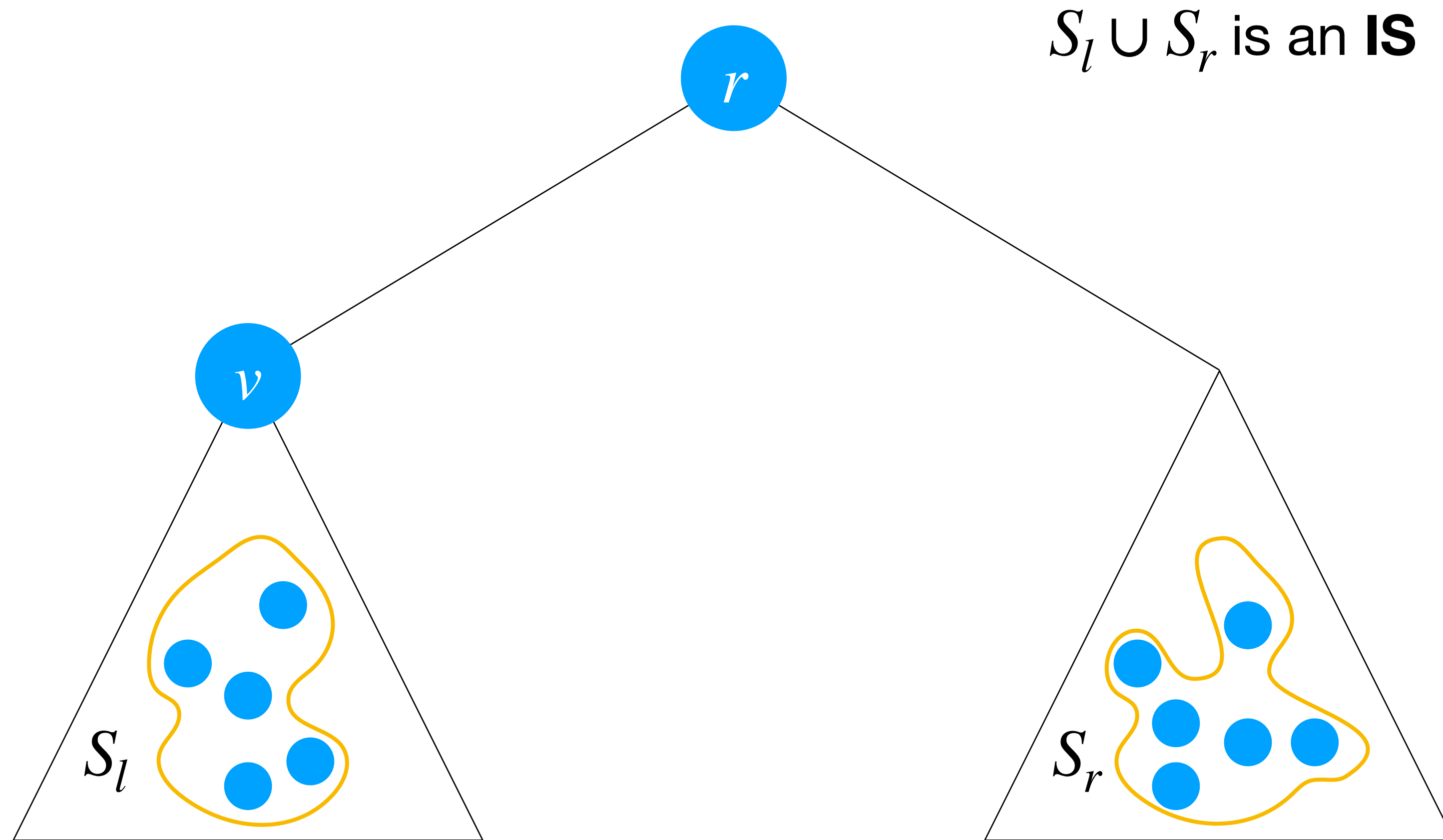
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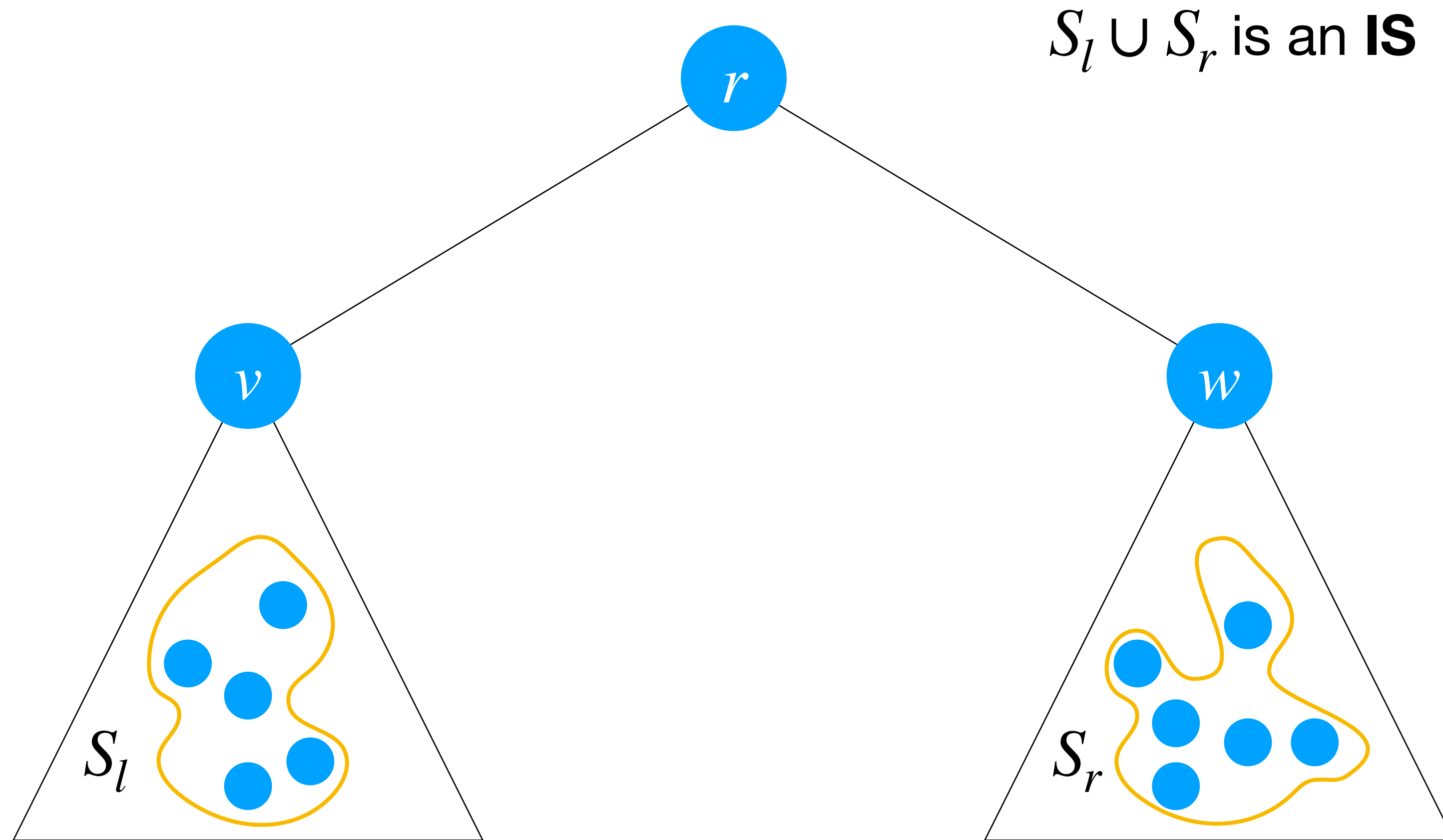
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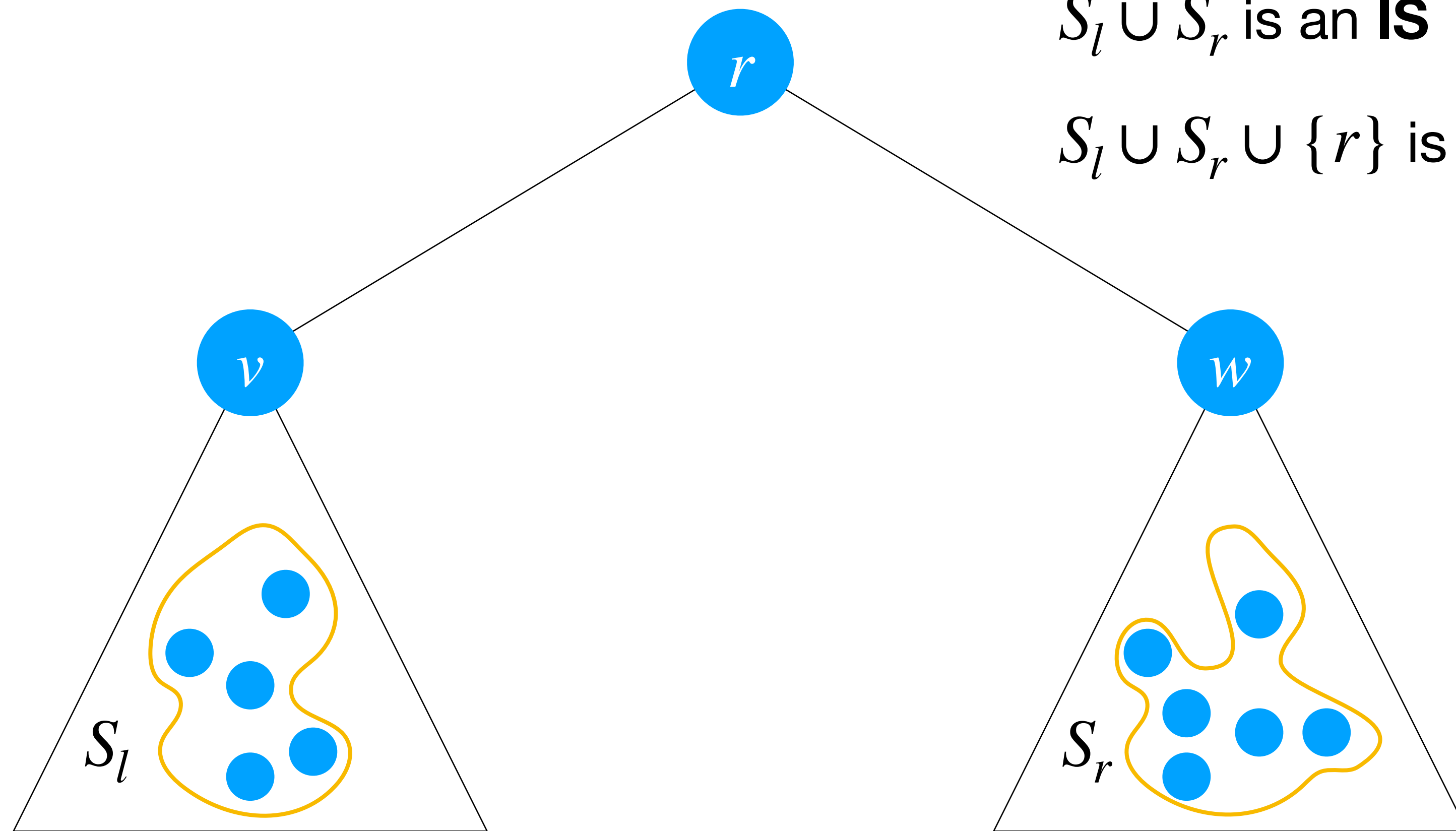
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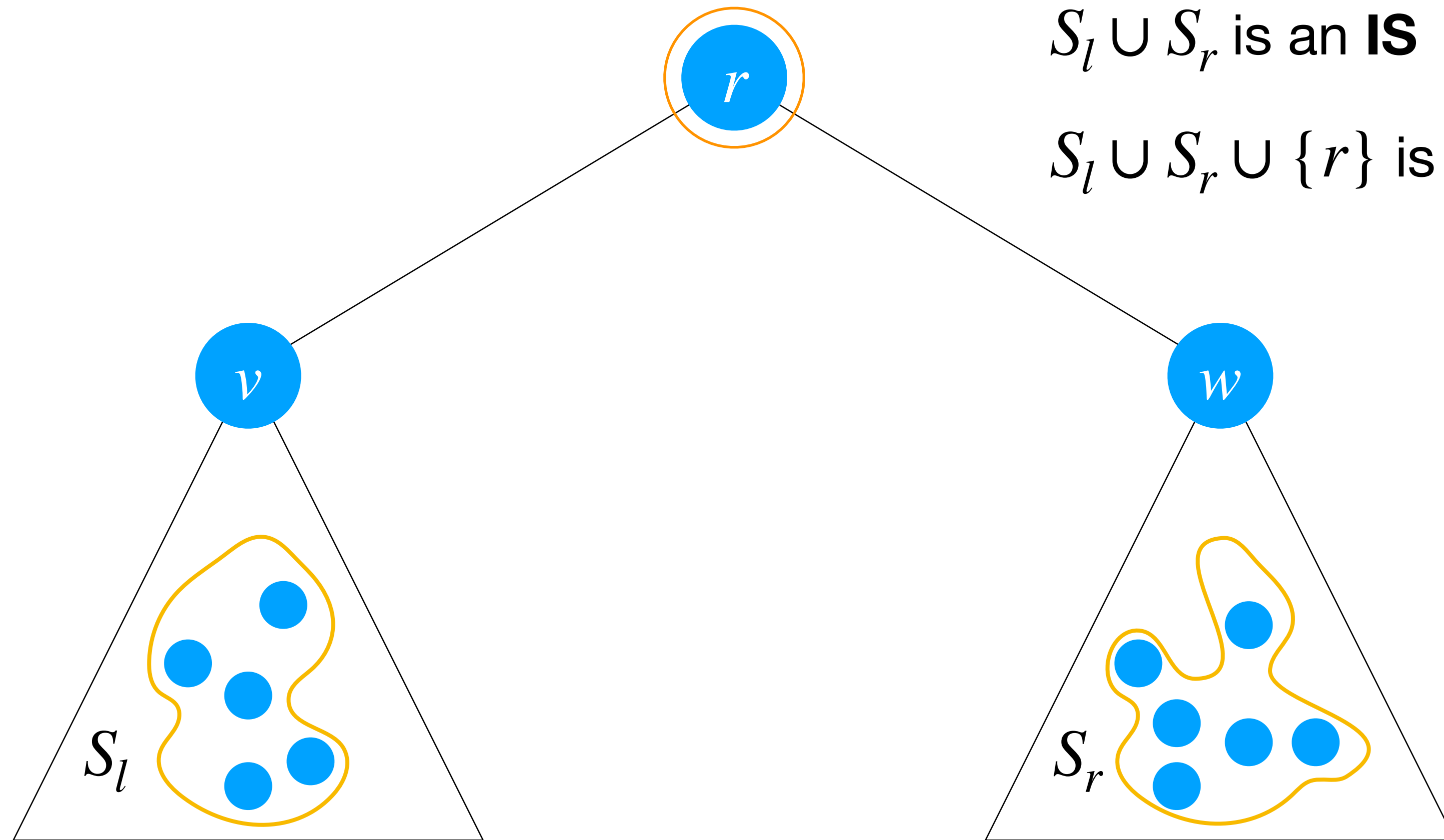
# Independent Set on Trees



$S_l \cup S_r$  is an **IS**

$S_l \cup S_r \cup \{r\}$  is an **IS** if  $v \notin S_l$  and  $w \notin S_r$

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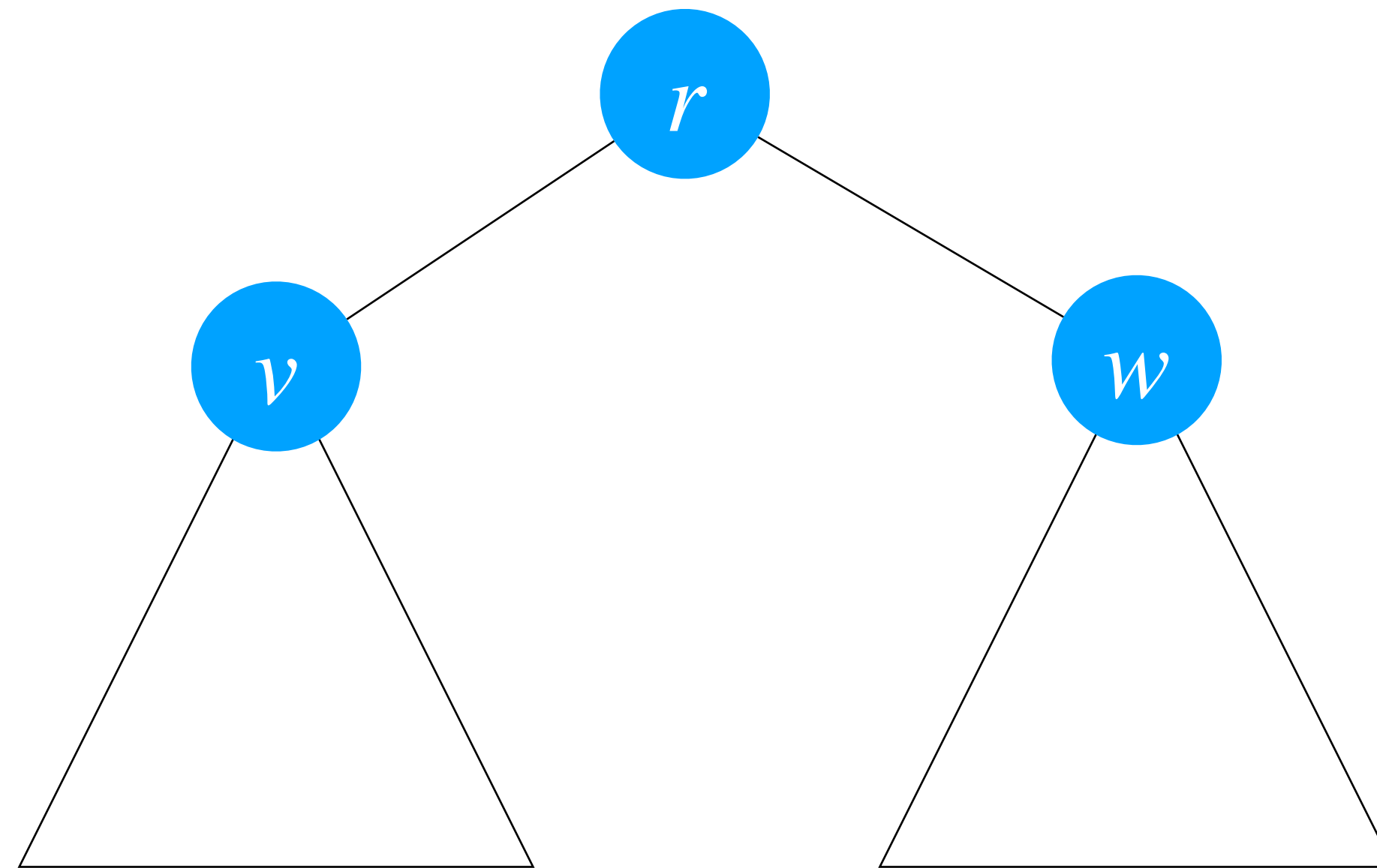


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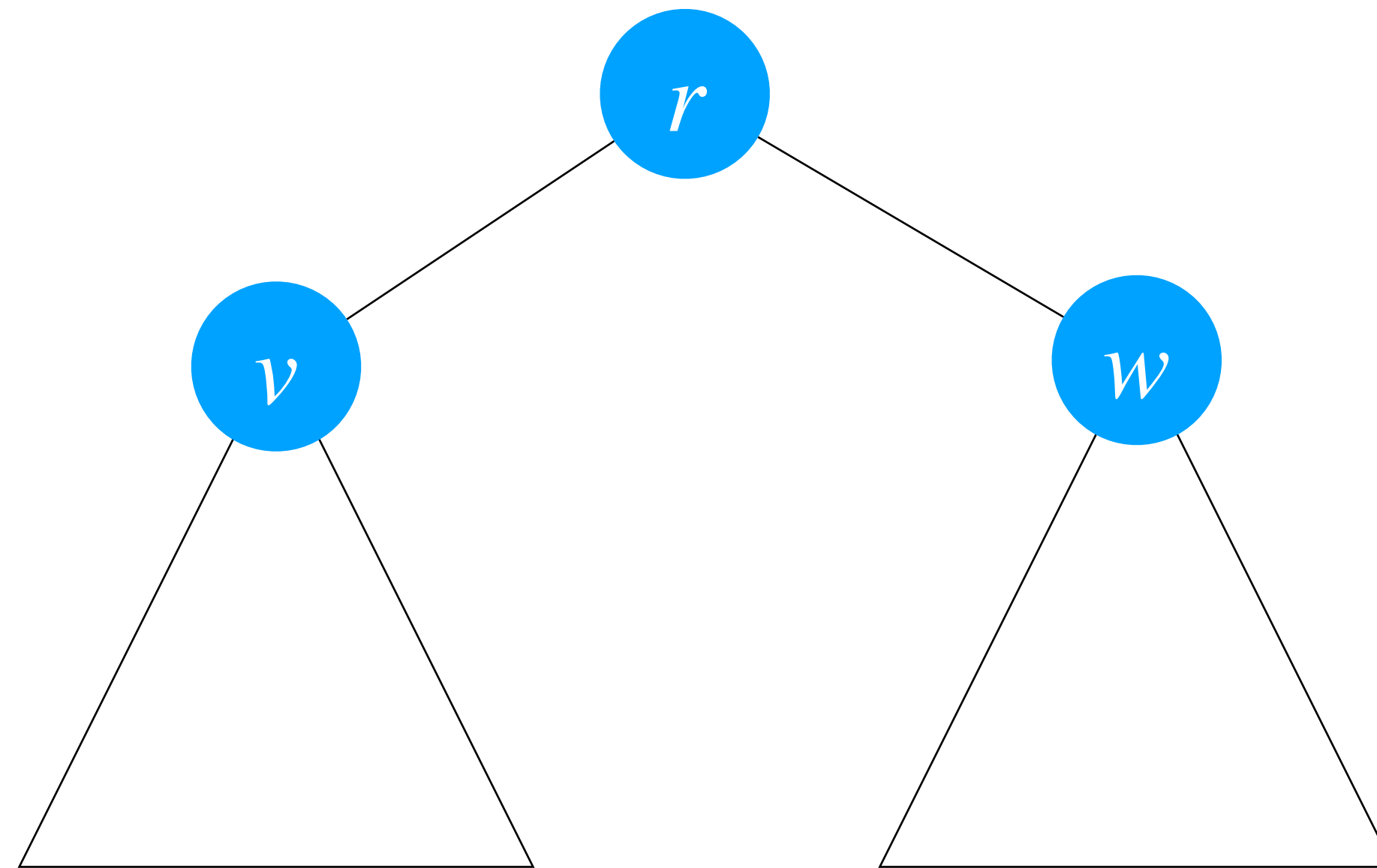


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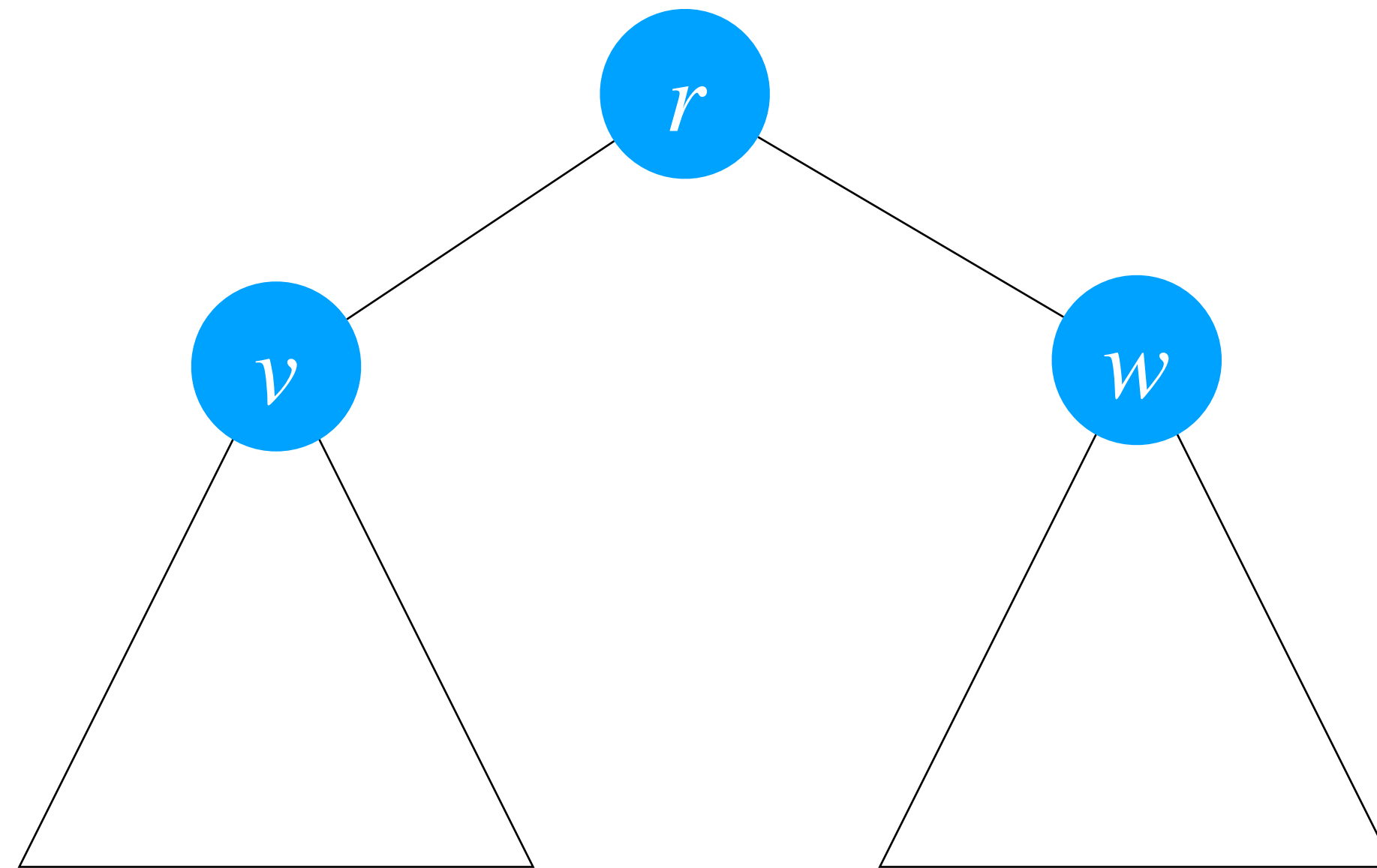
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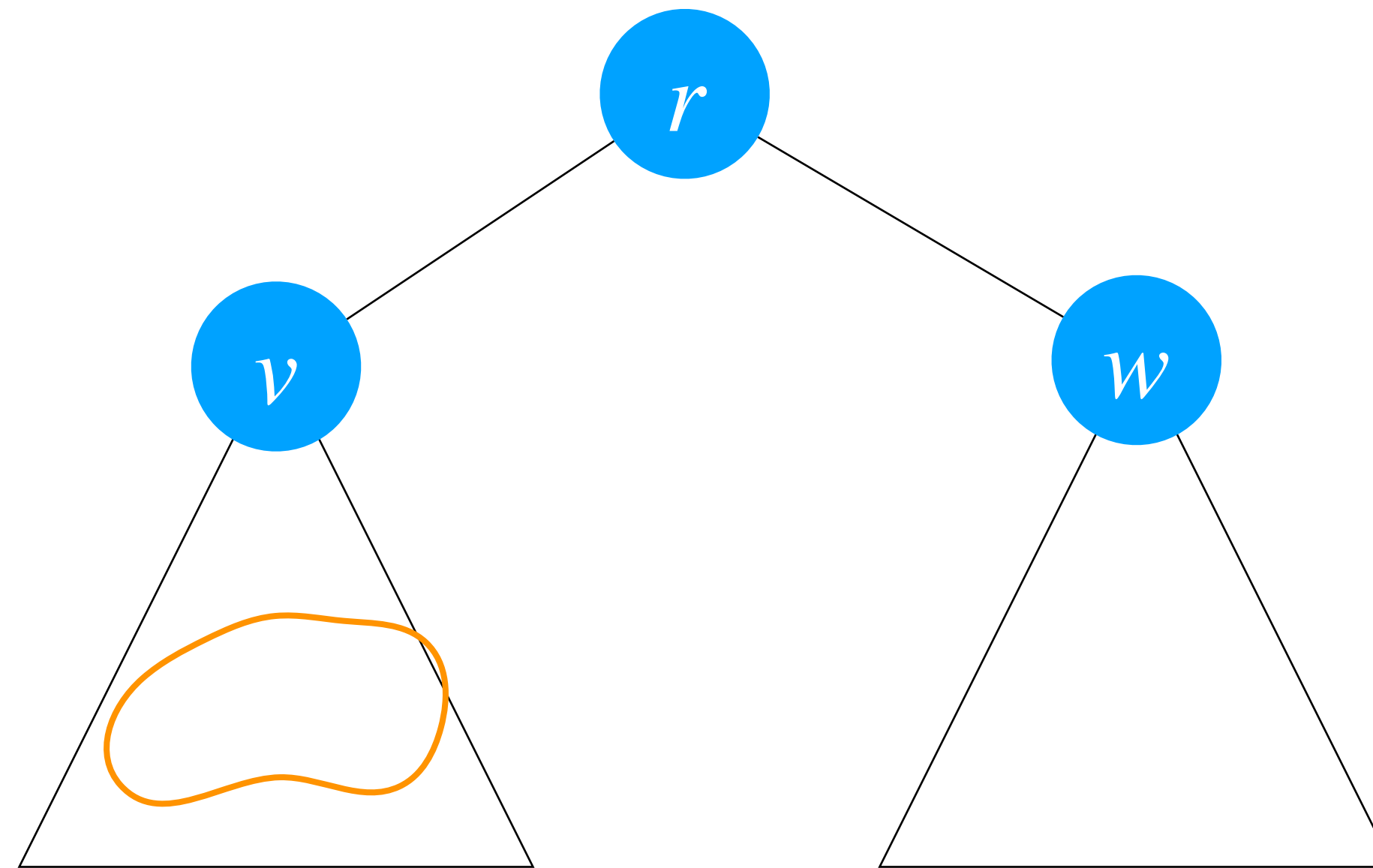
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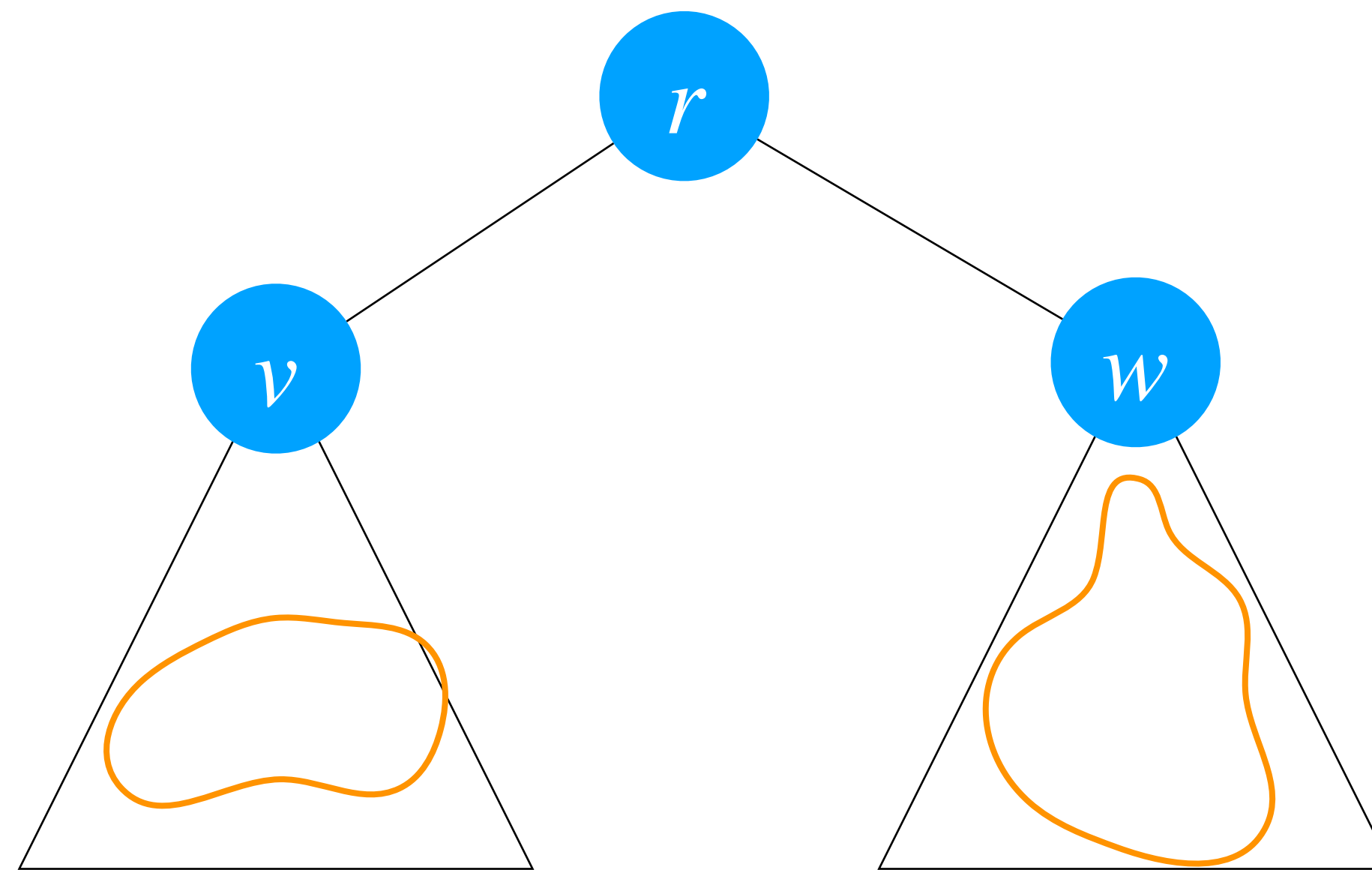
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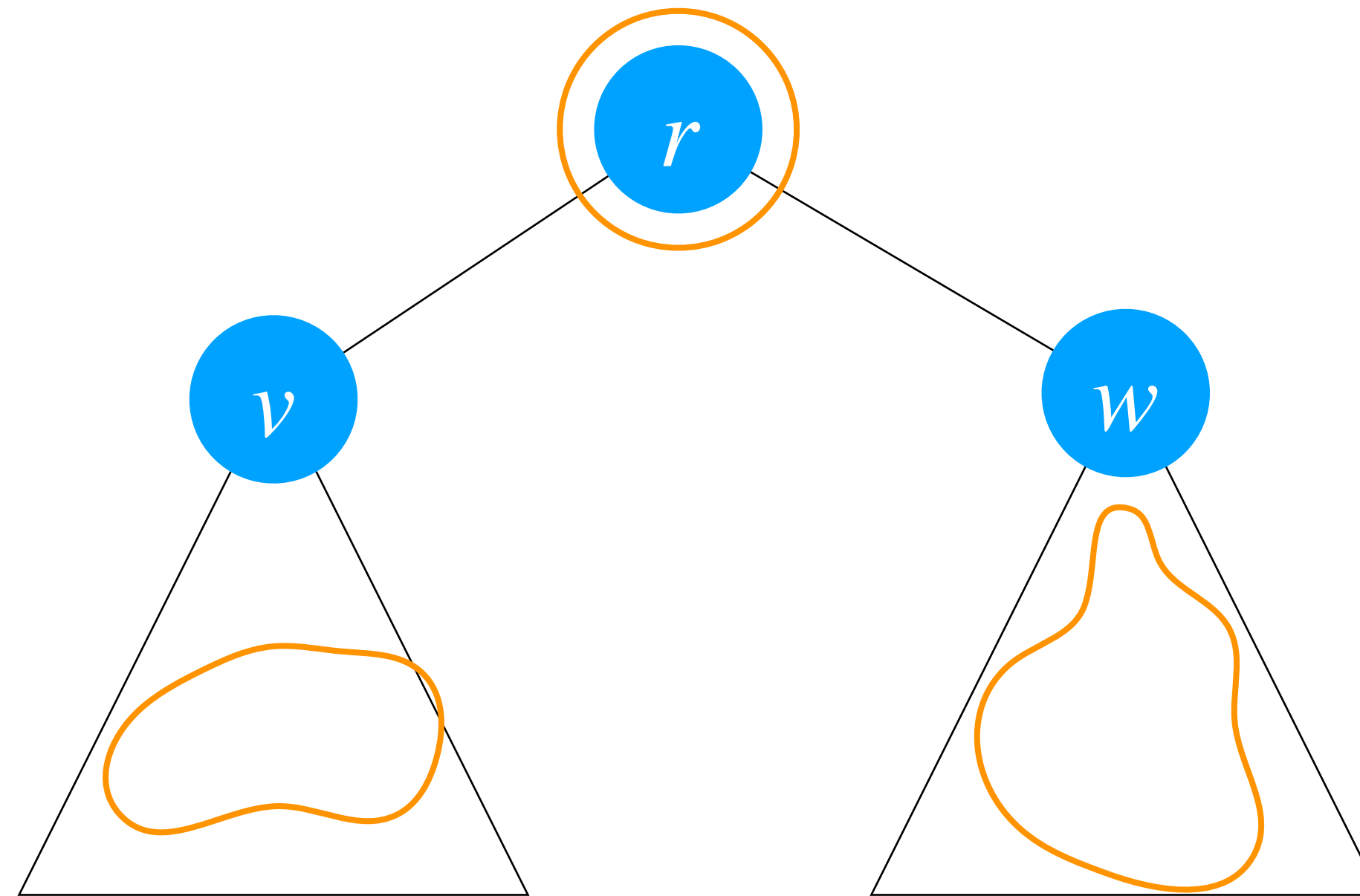
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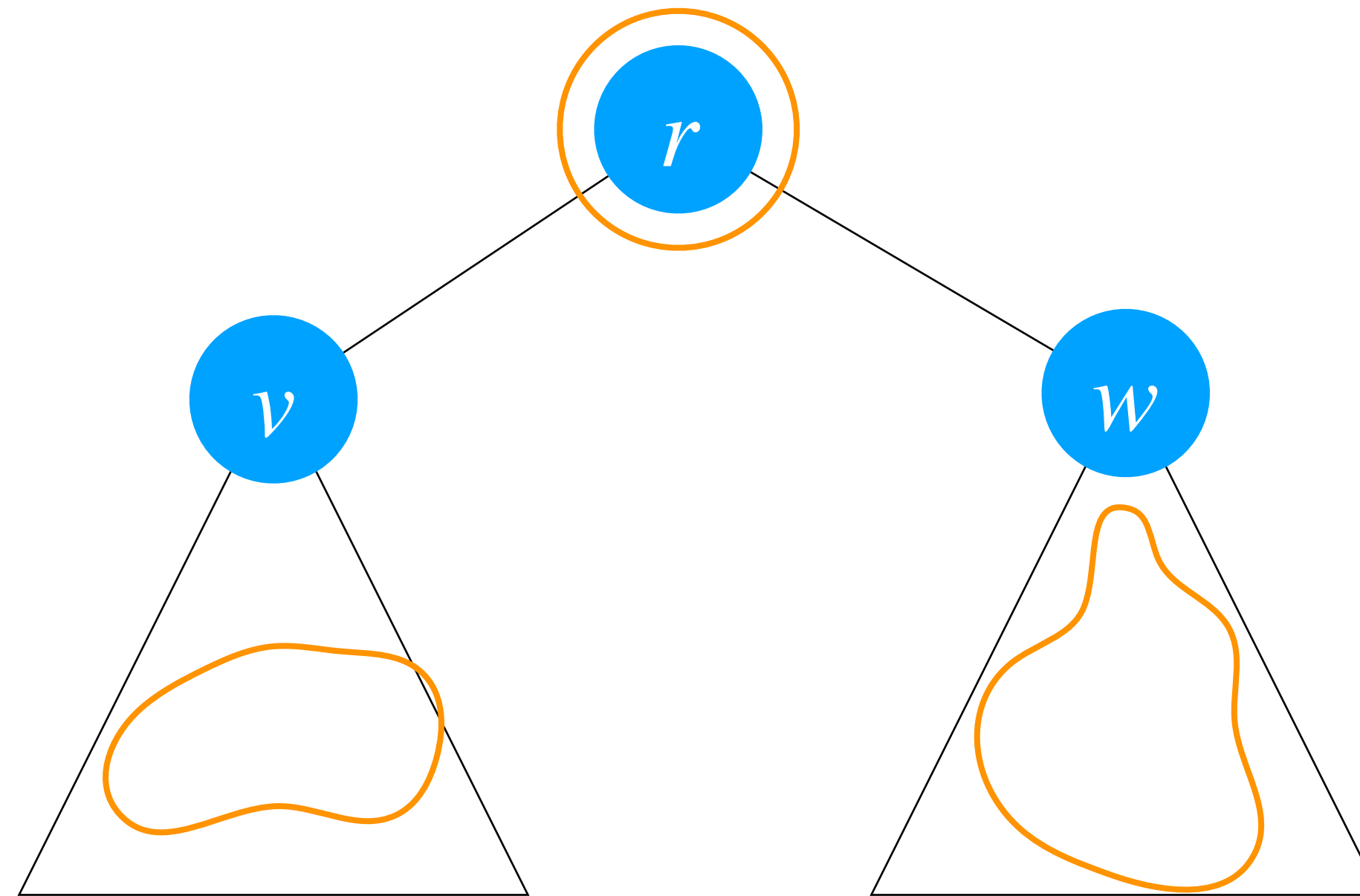


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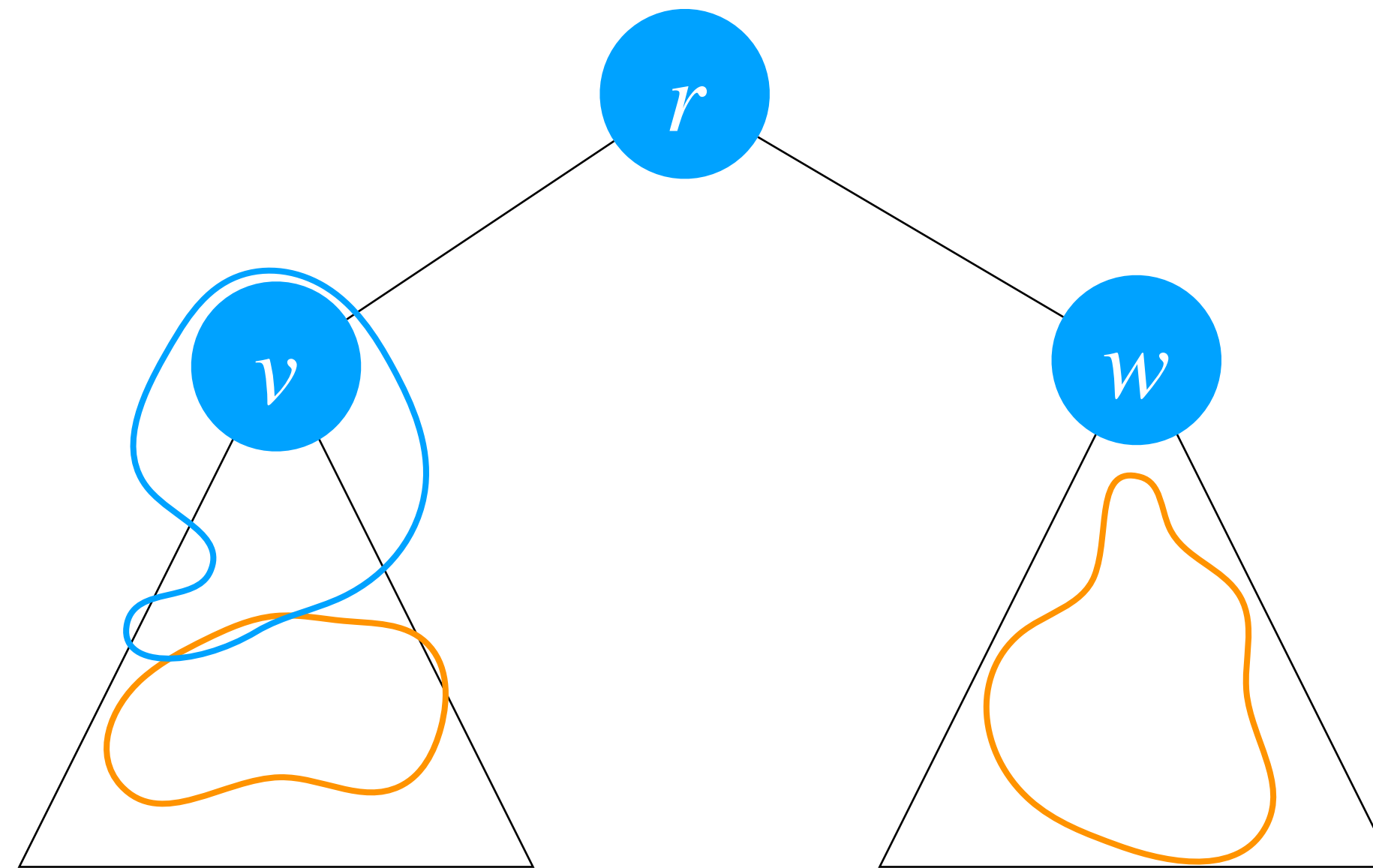


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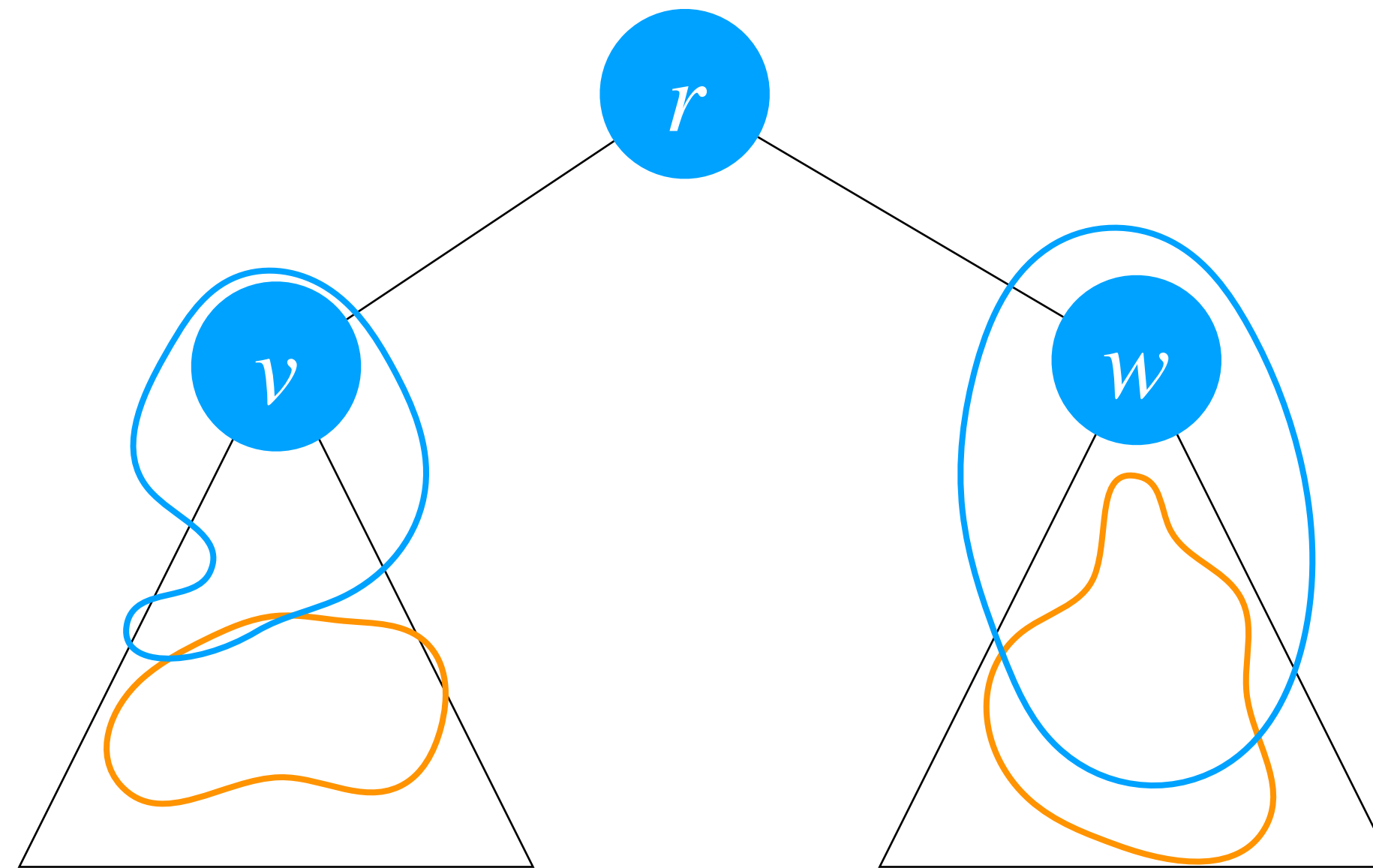


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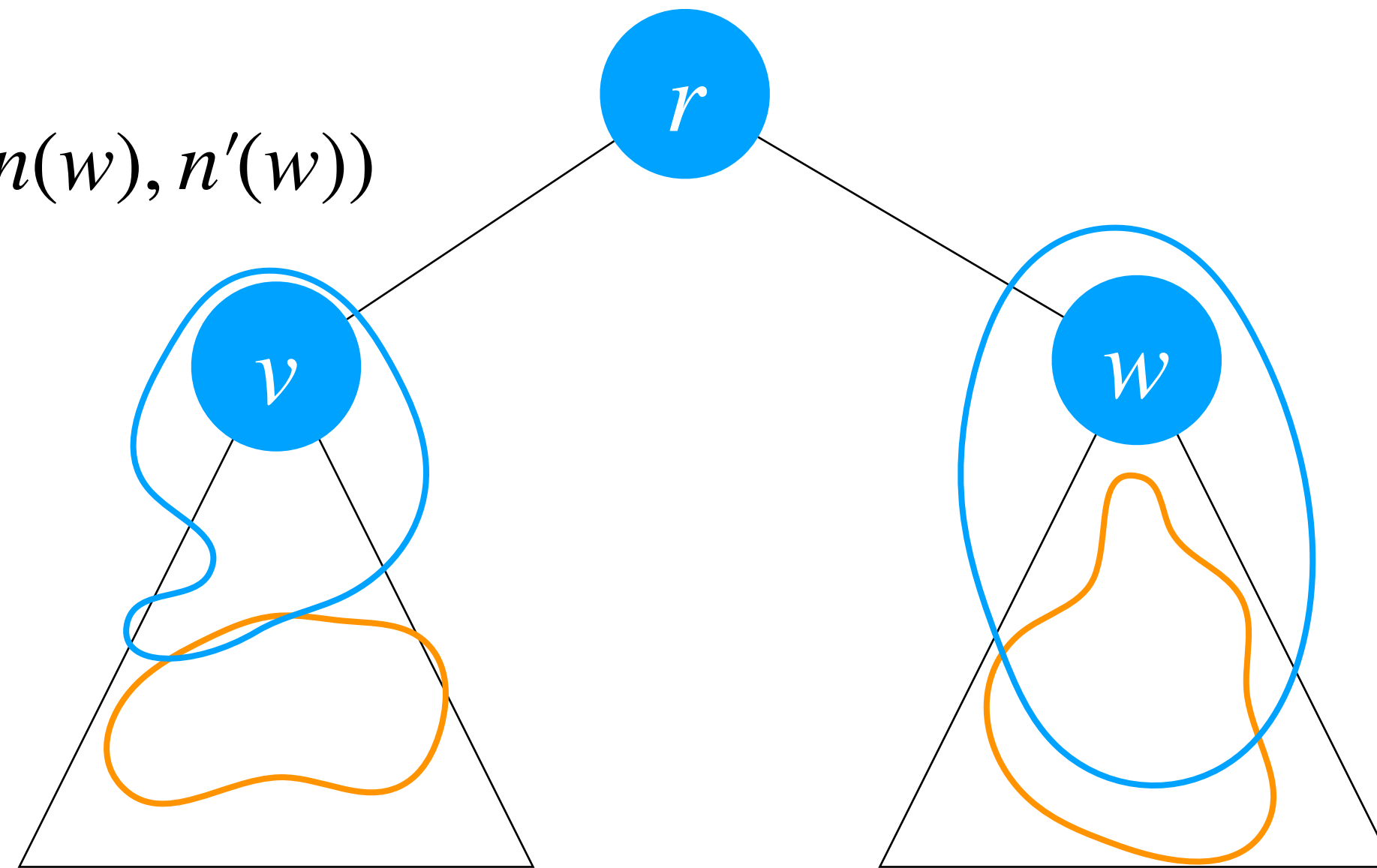
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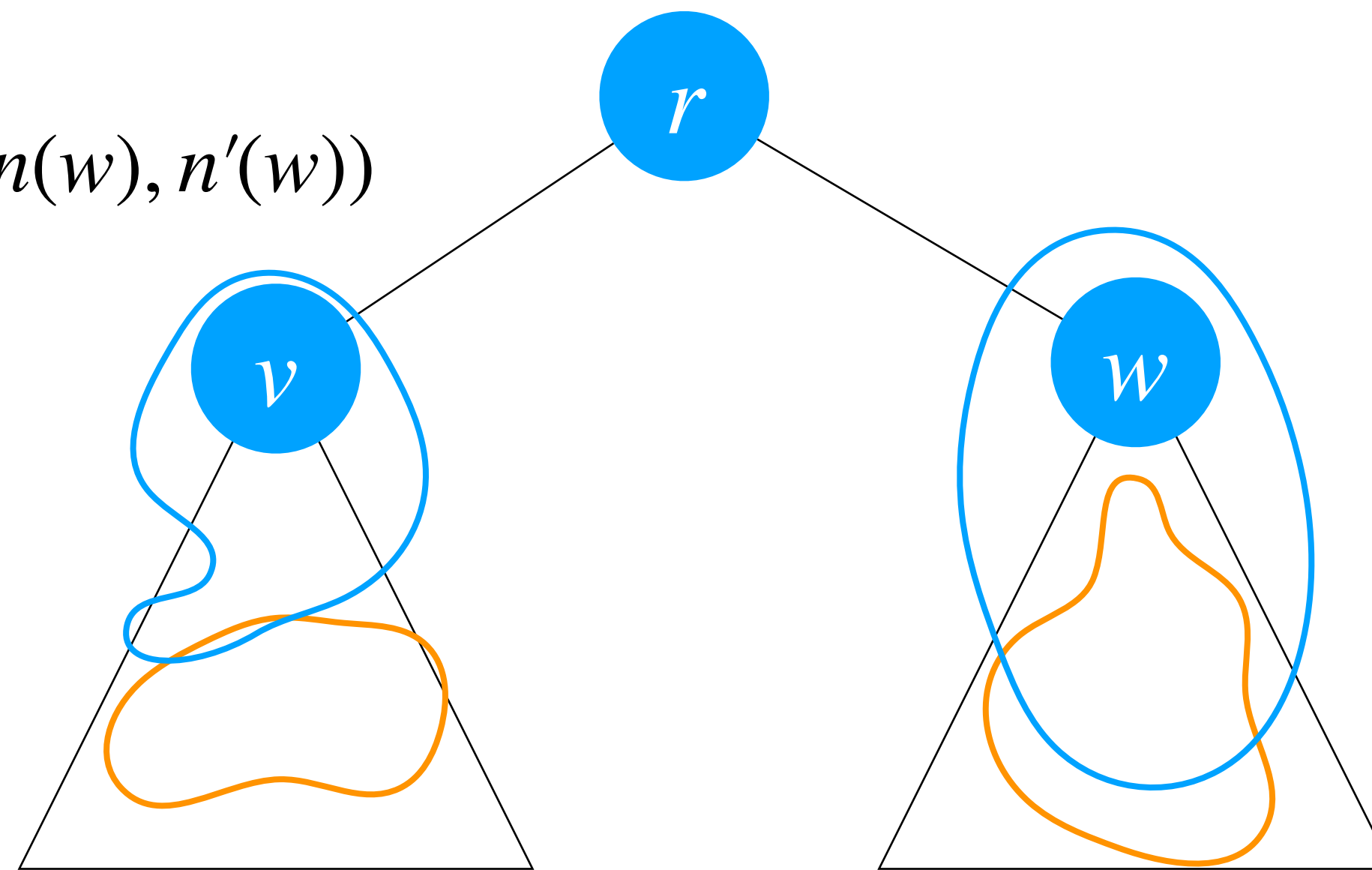
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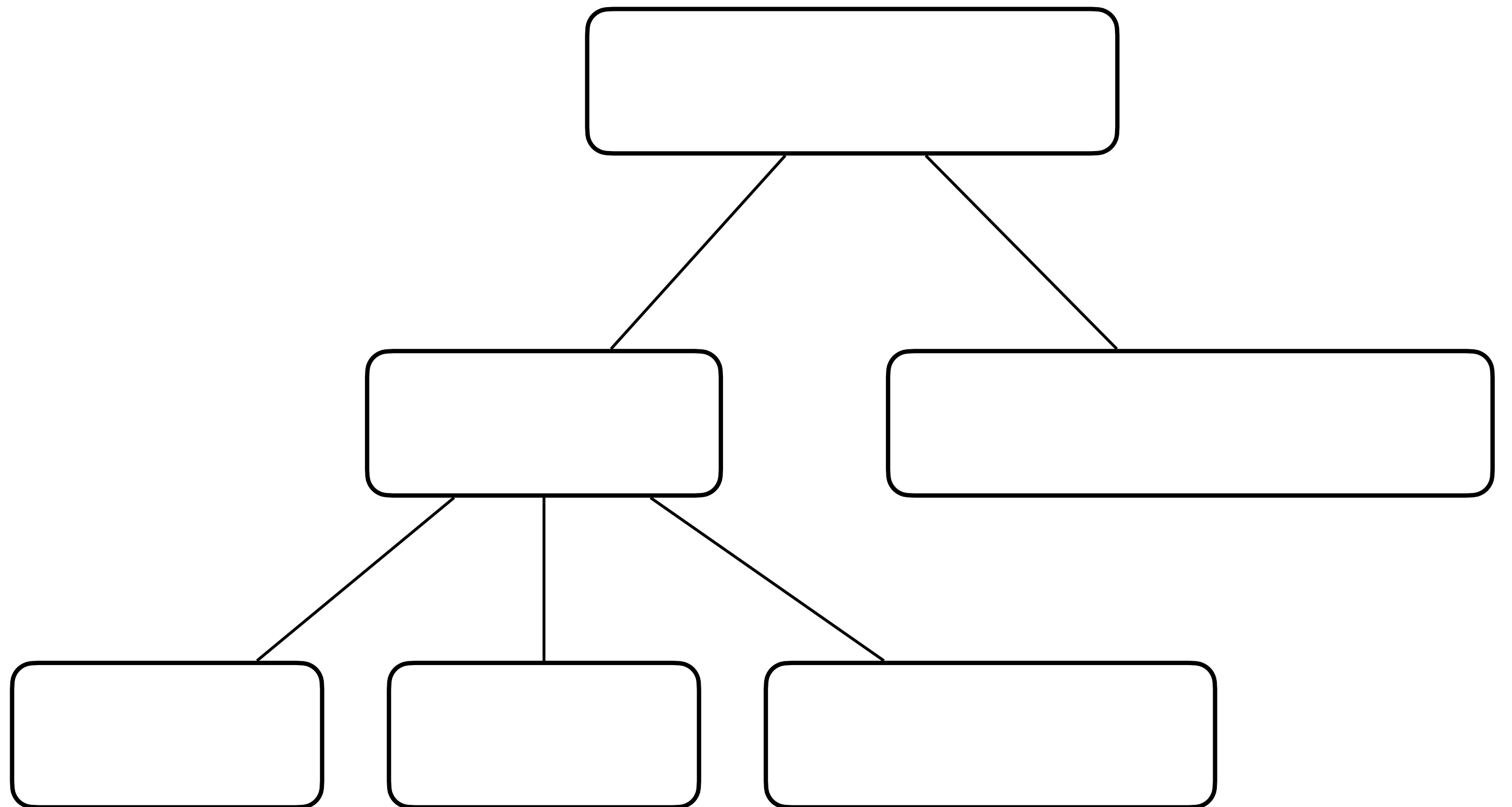
$$n'(r) = \max(n(v), n'(v)) + \max(n(w), n'(w))$$



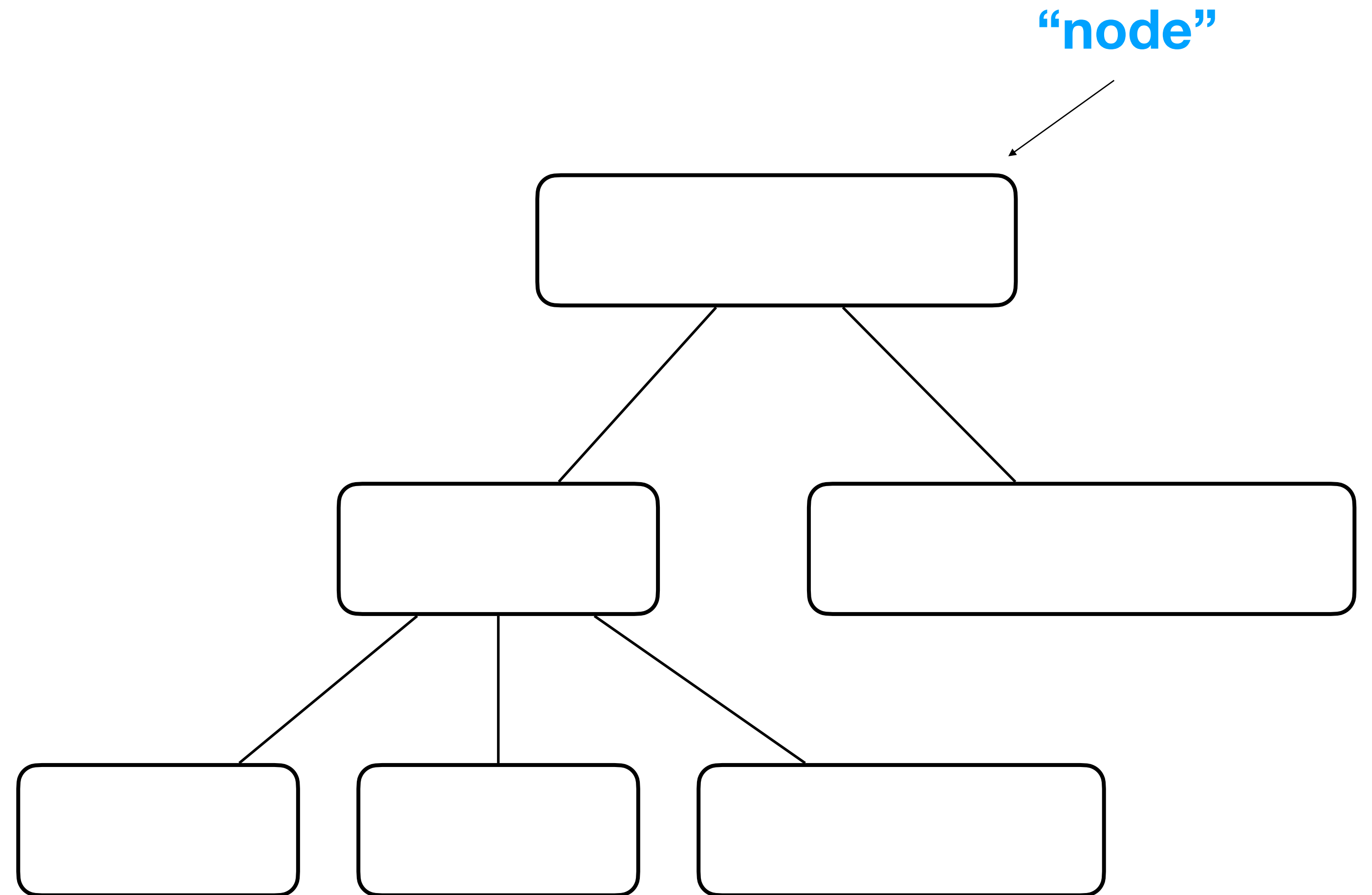
$\max(n(r), n'(r))$  is the size of the largest **IS**.

# Tree Decomposition

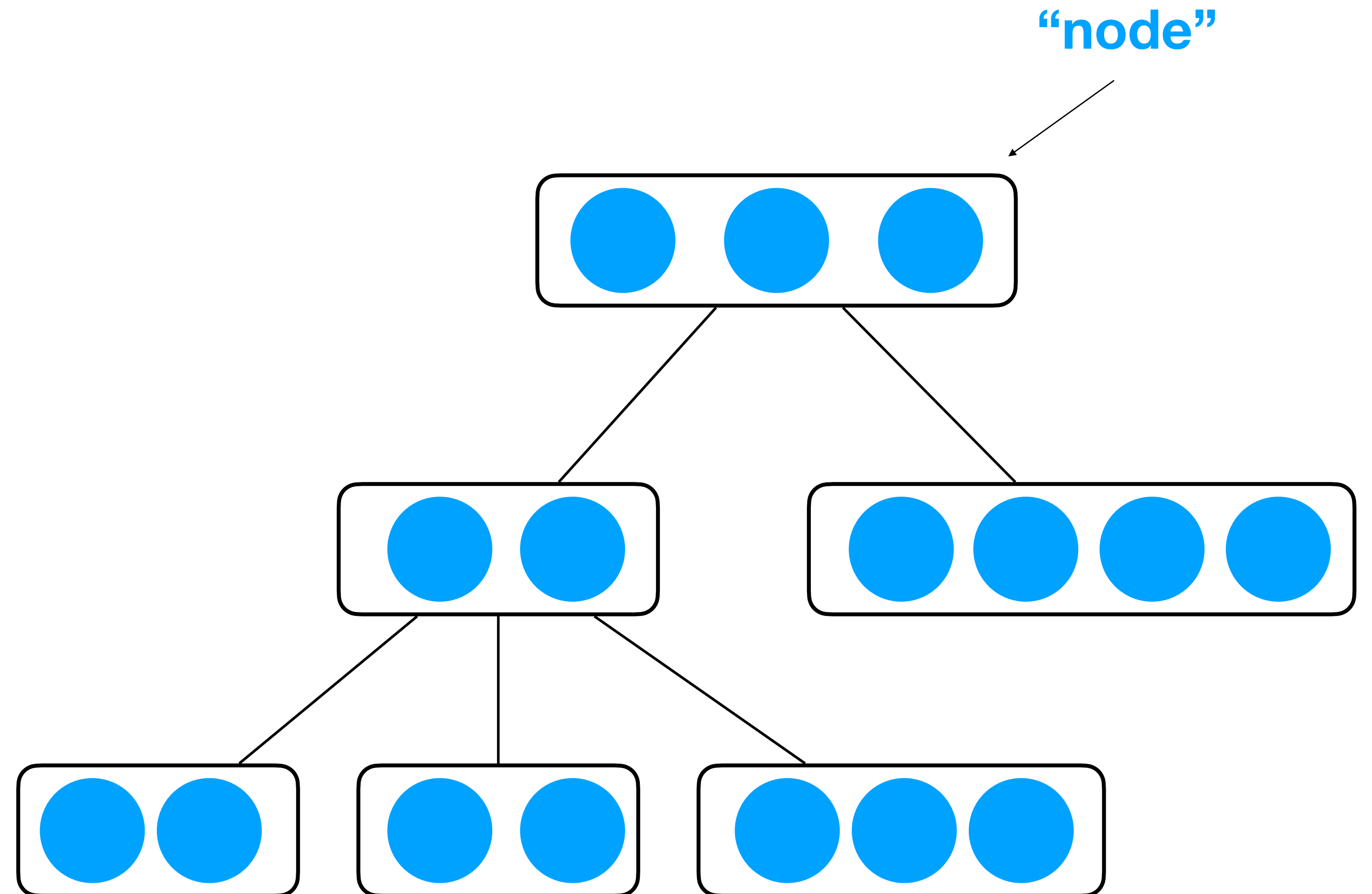
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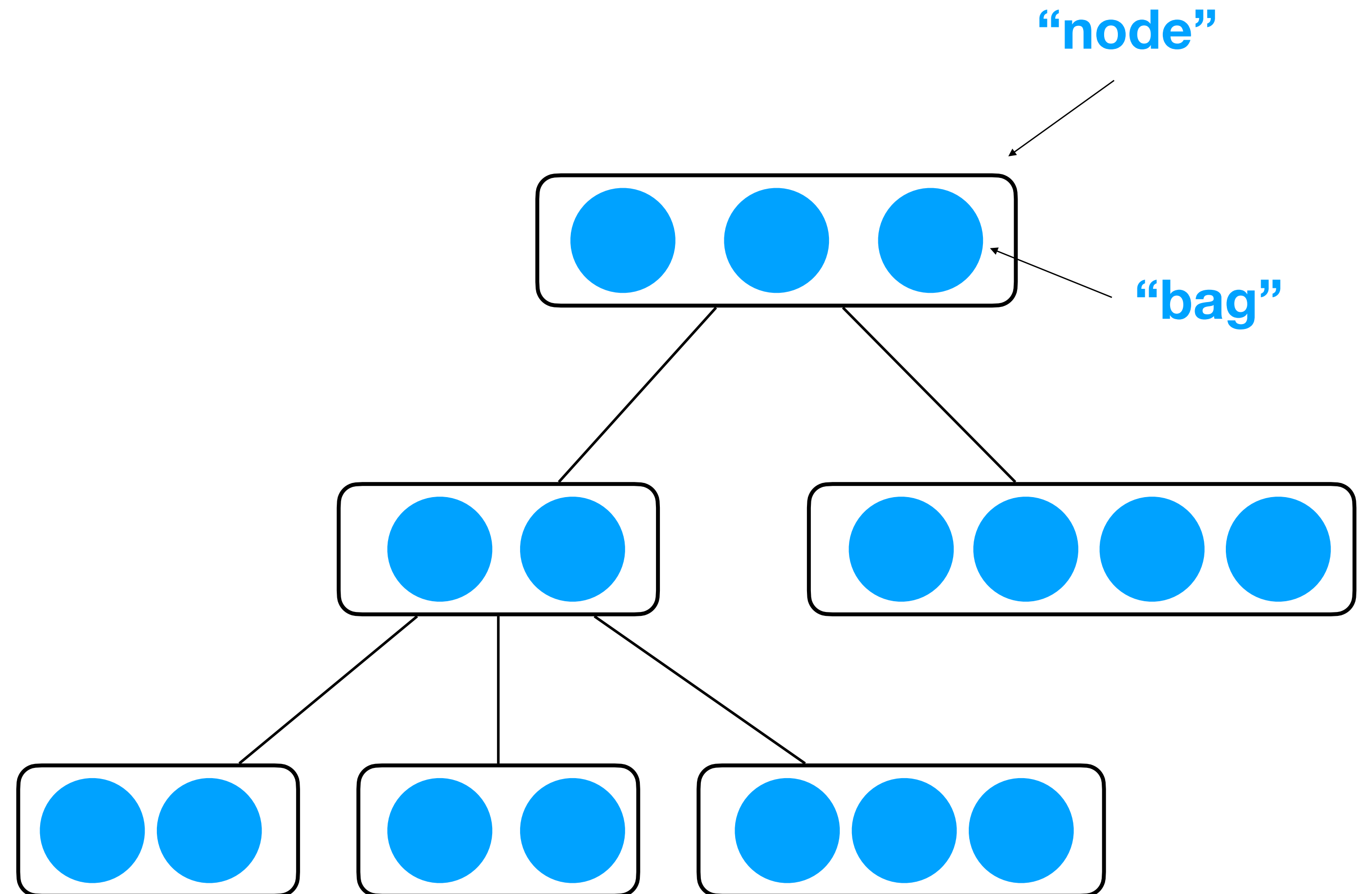
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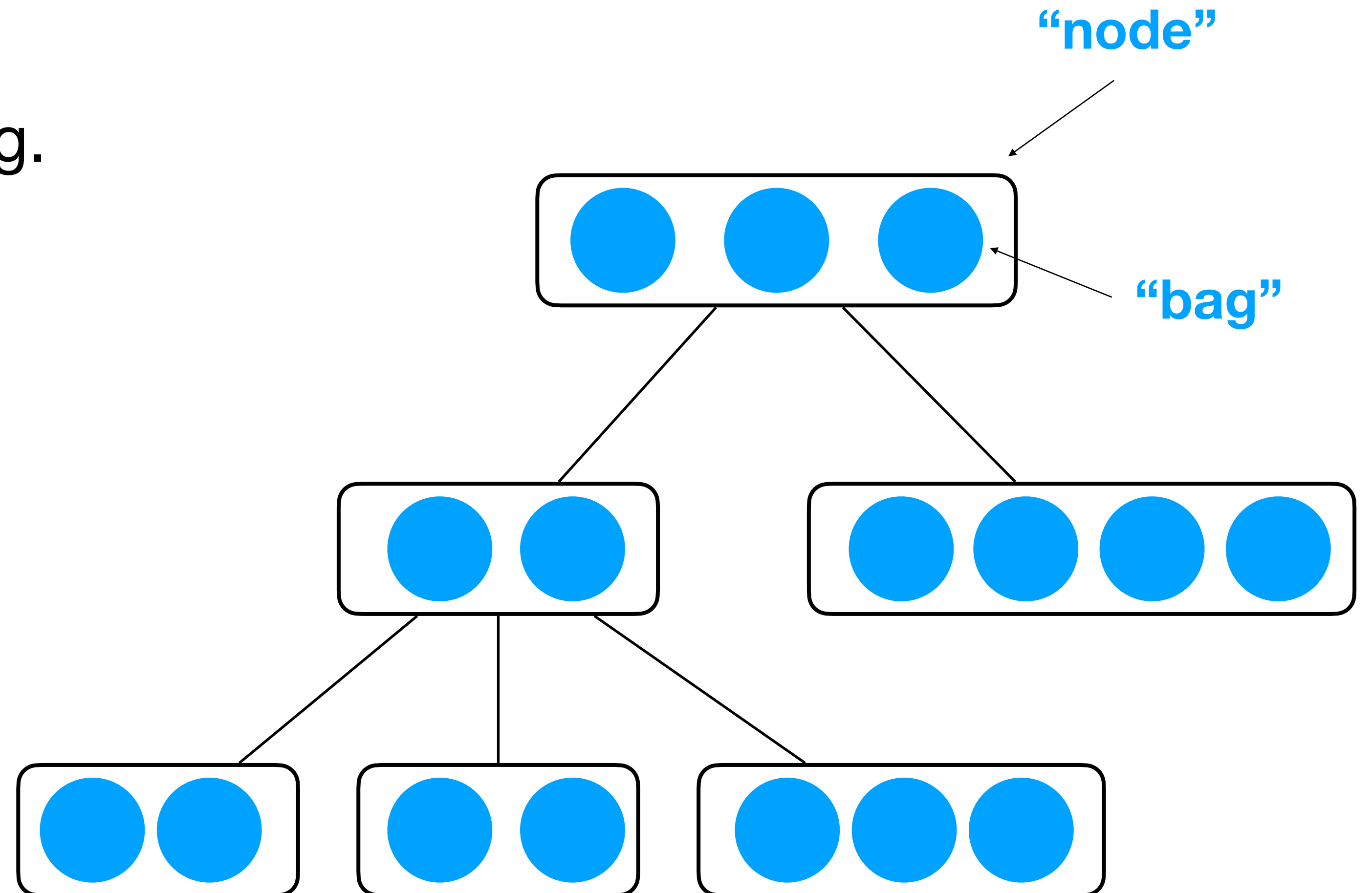
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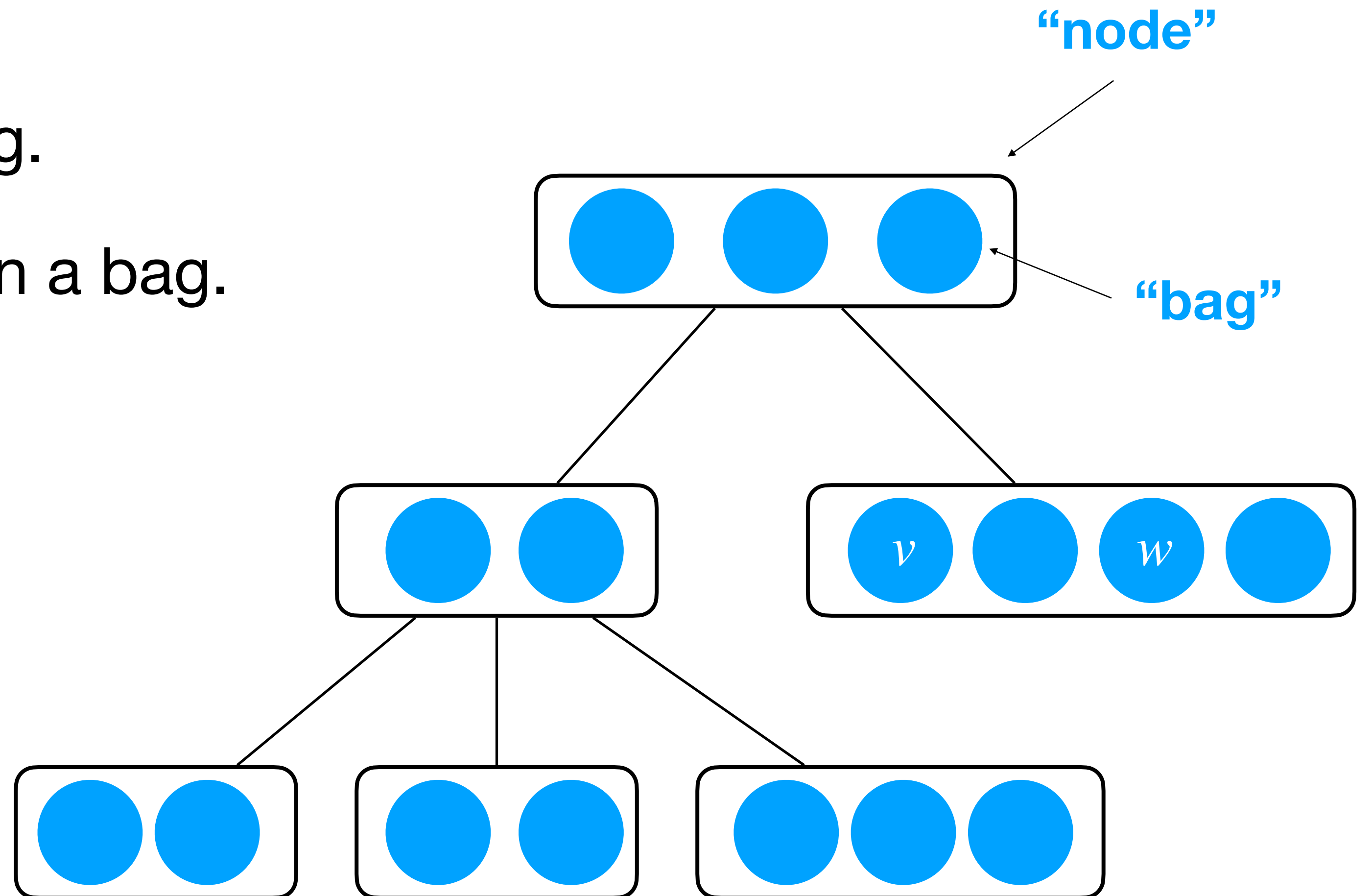
# Tree Decomposition

1. Each **vertex** appears in a bag.



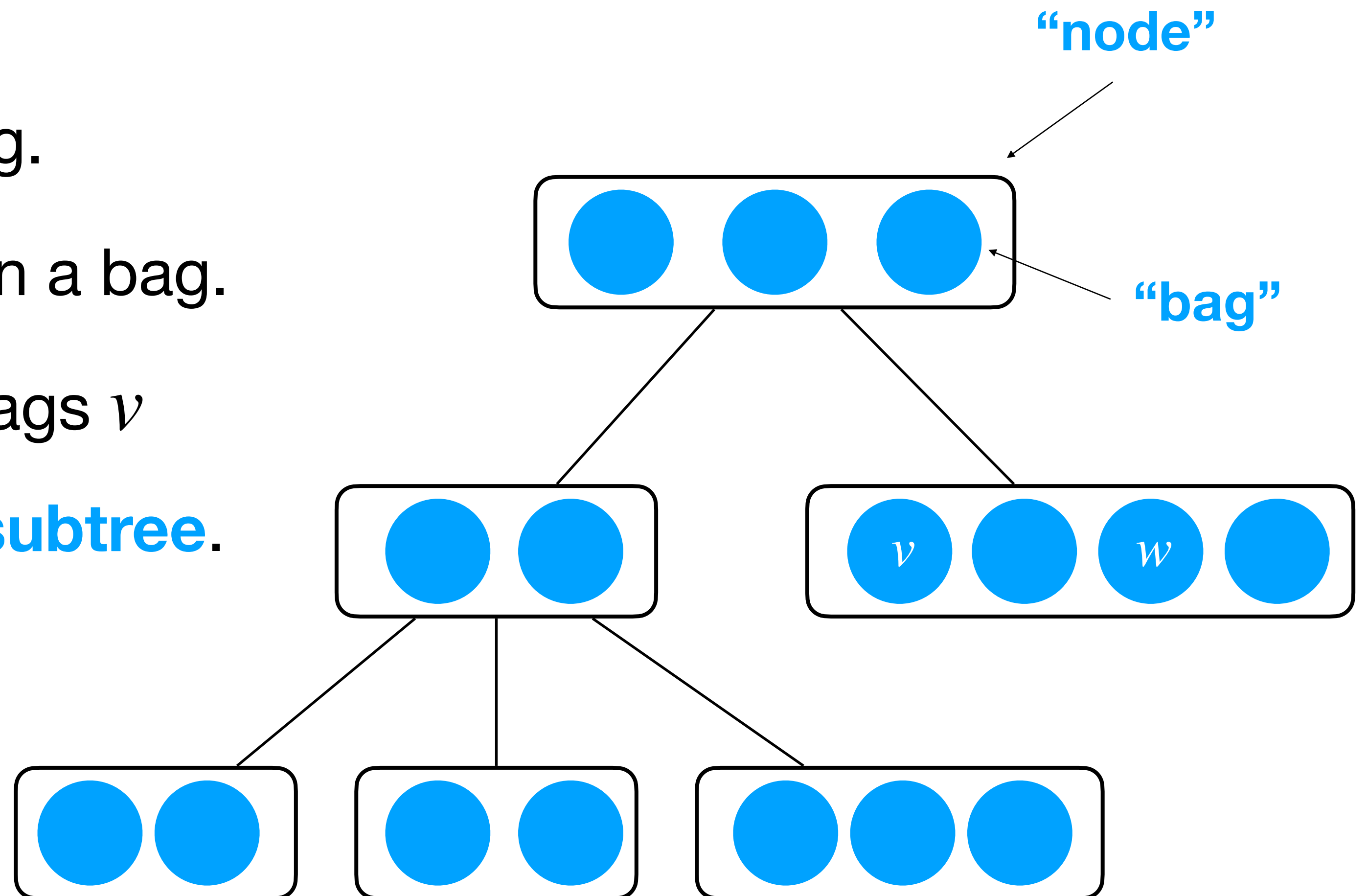
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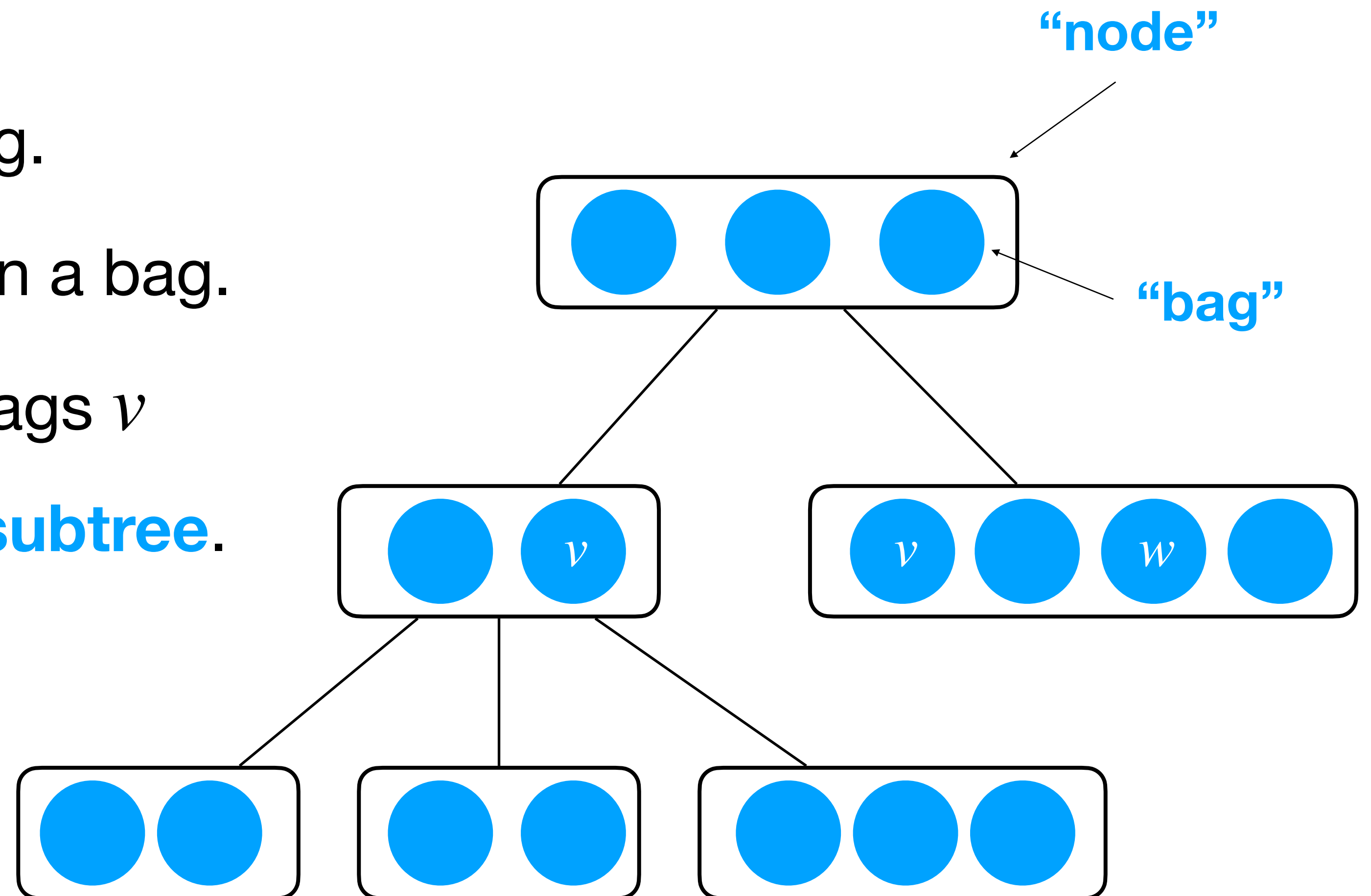
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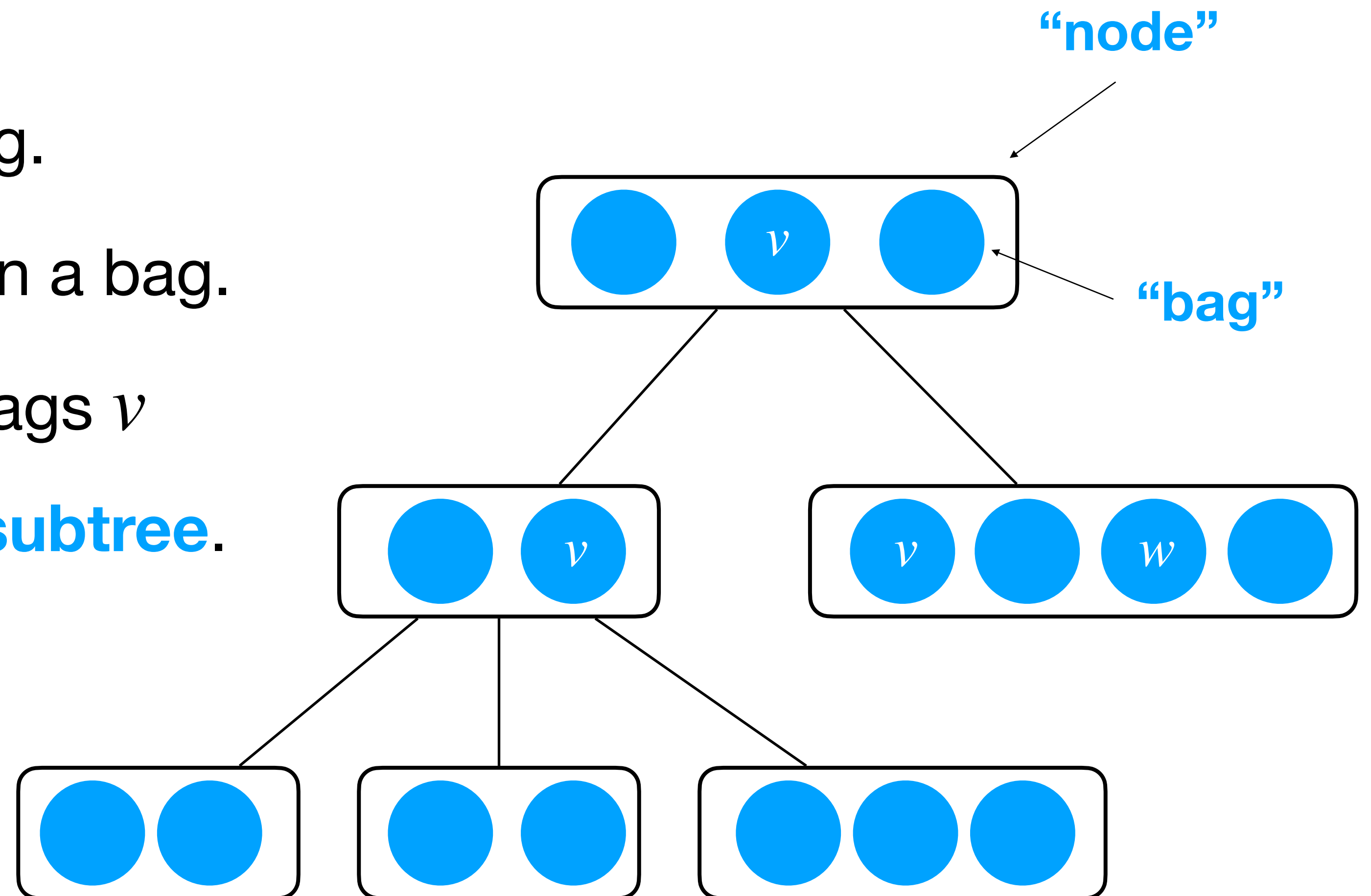
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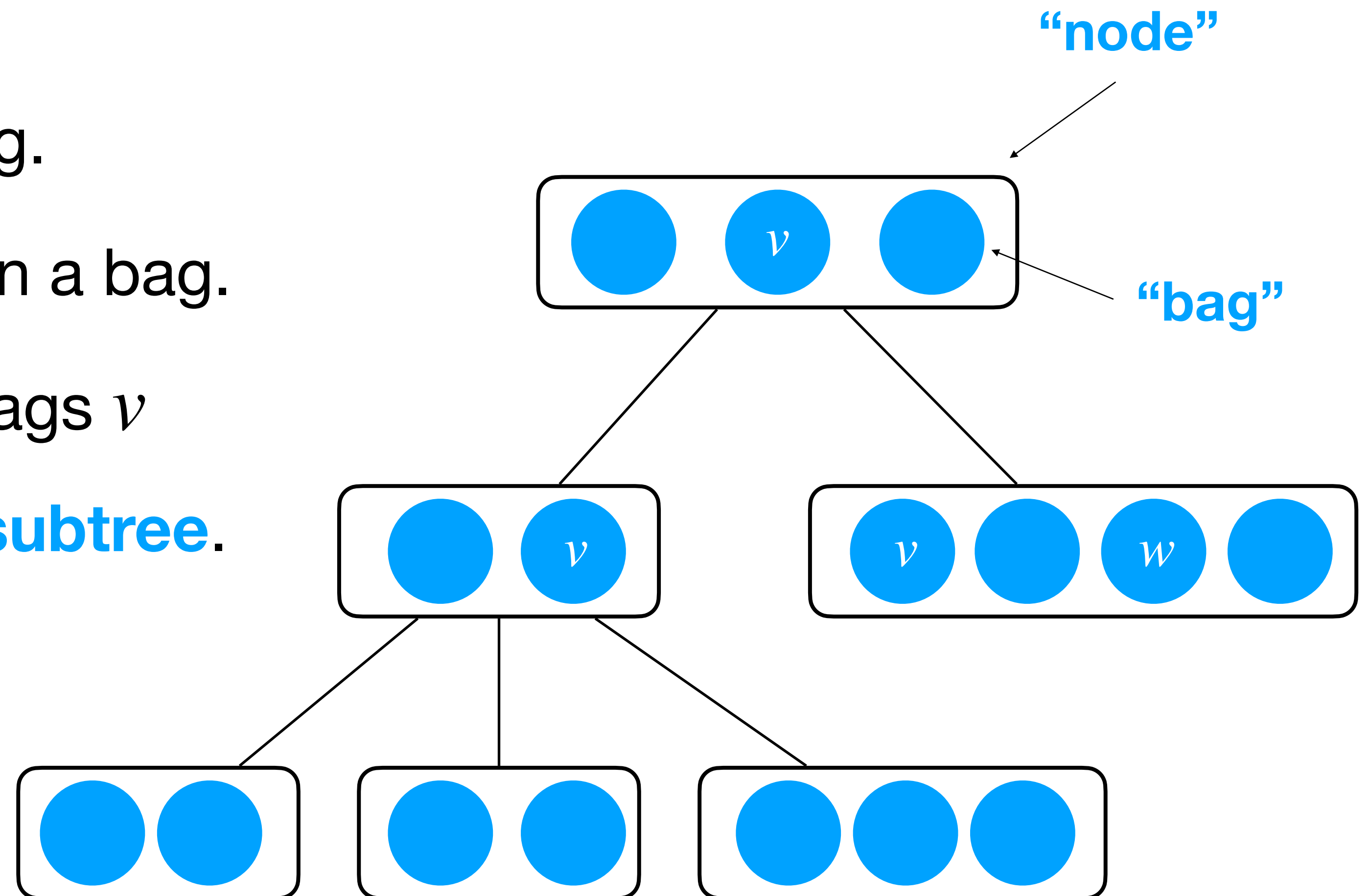
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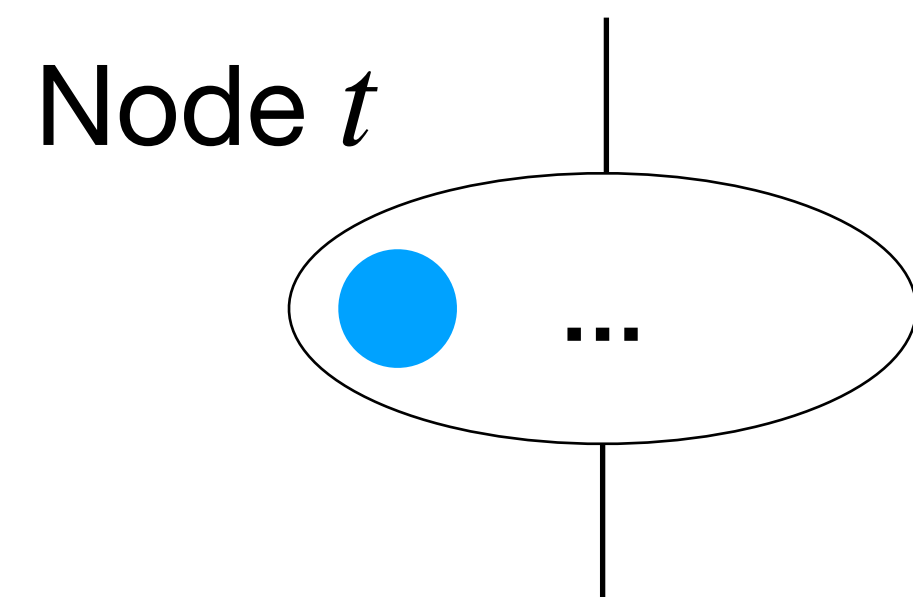
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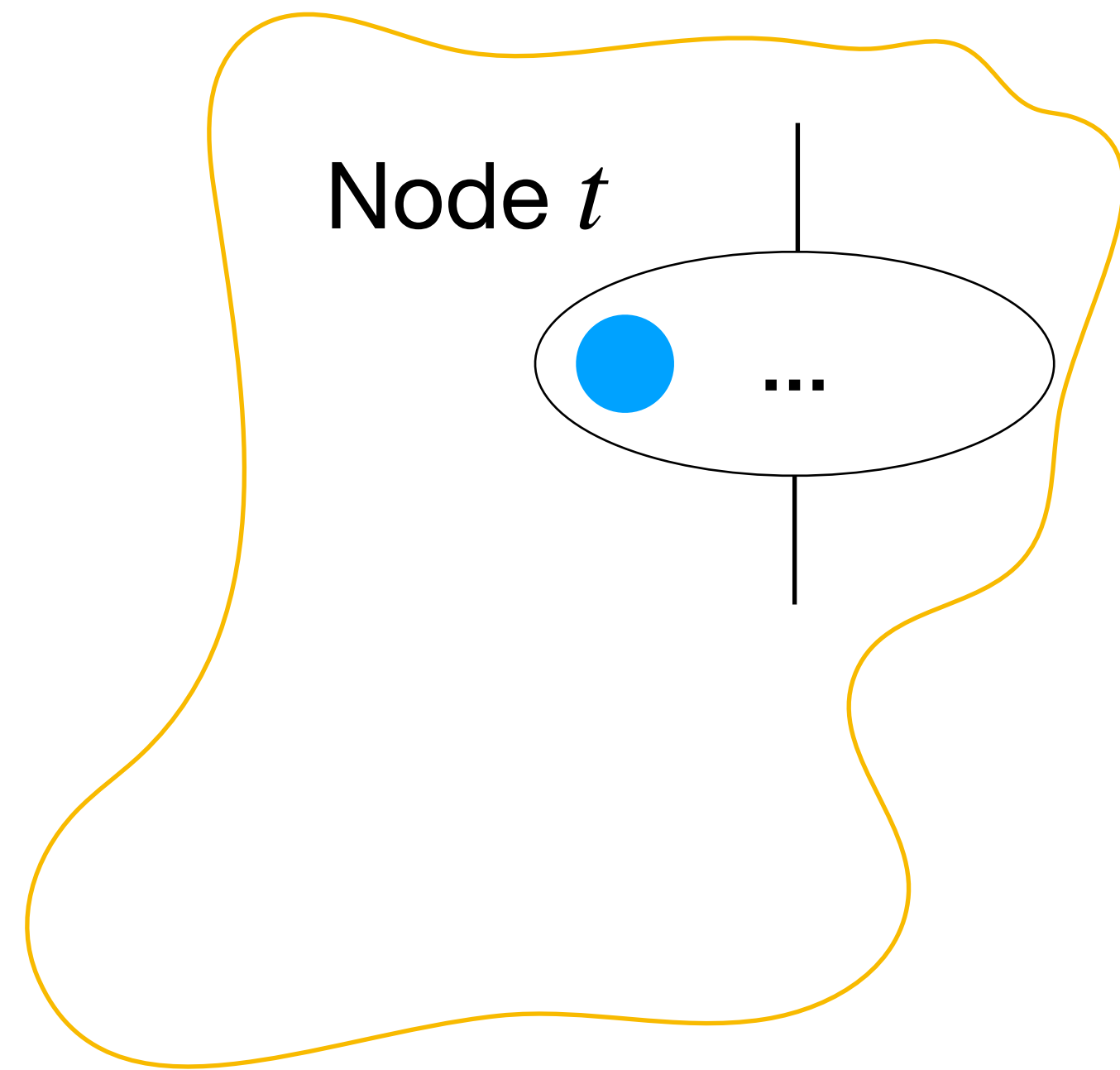


The **width** of a tree decomposition is the size of its largest bag - 1.

# Independent Set with Tree Decompositions

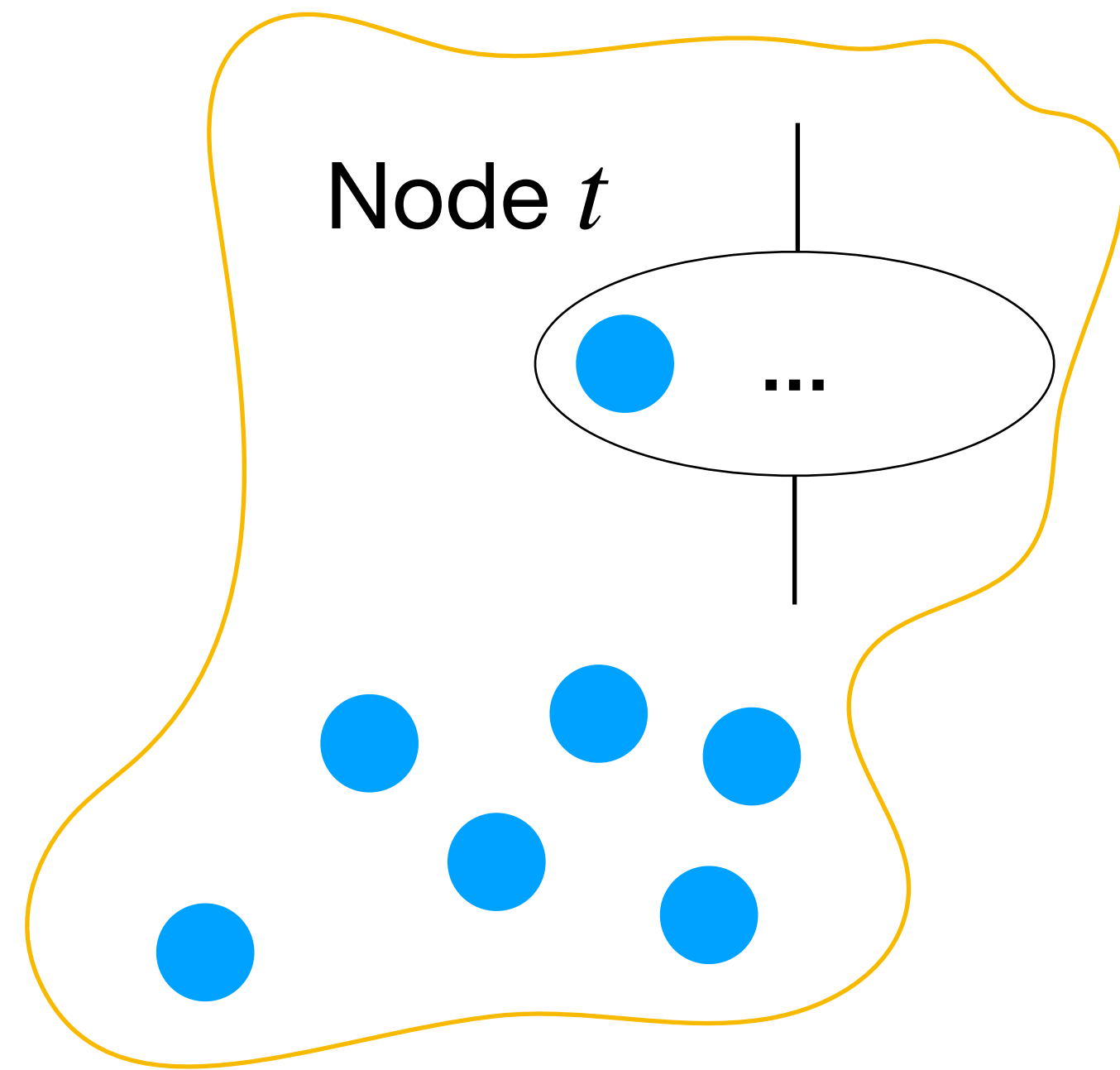


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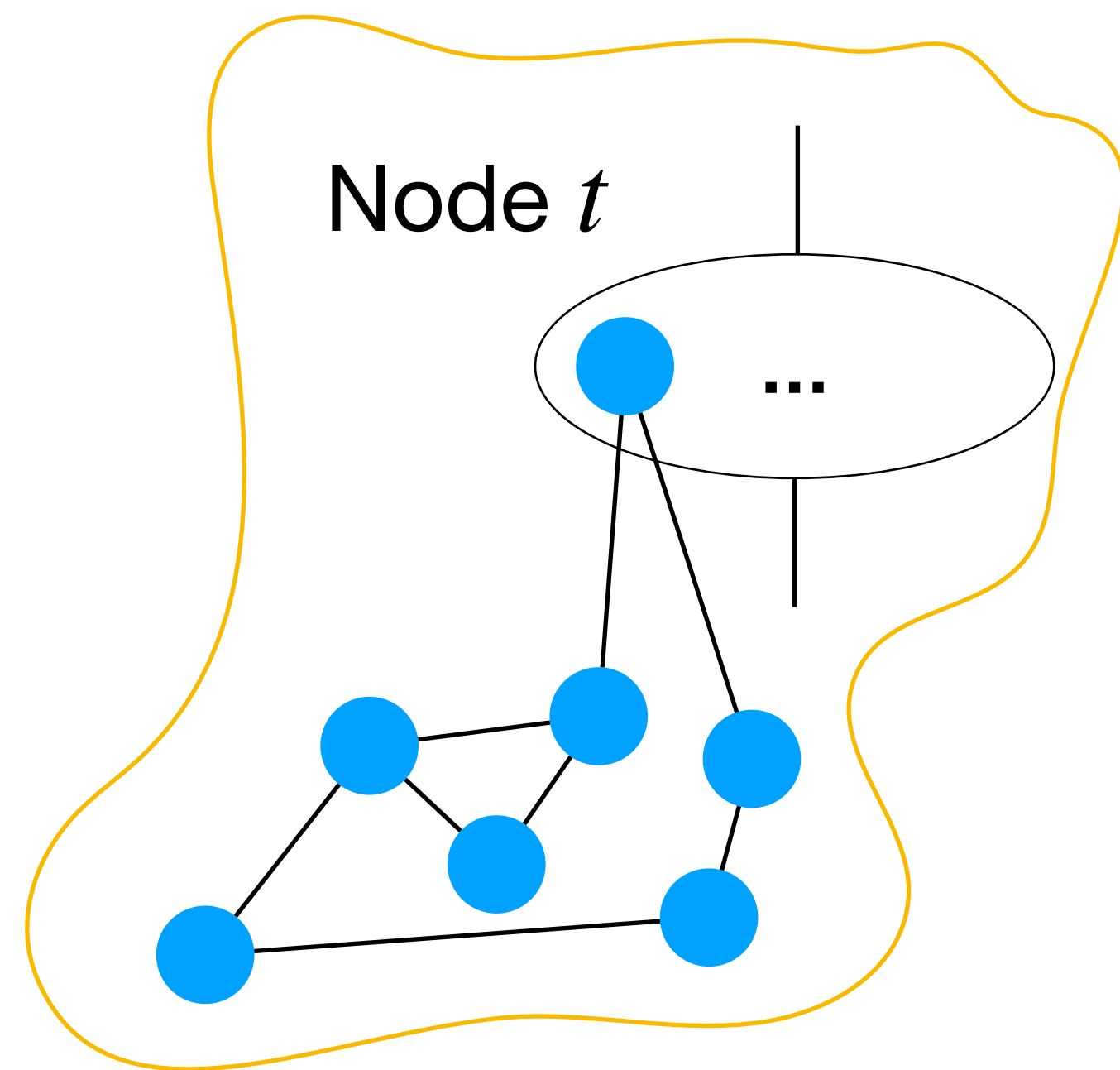




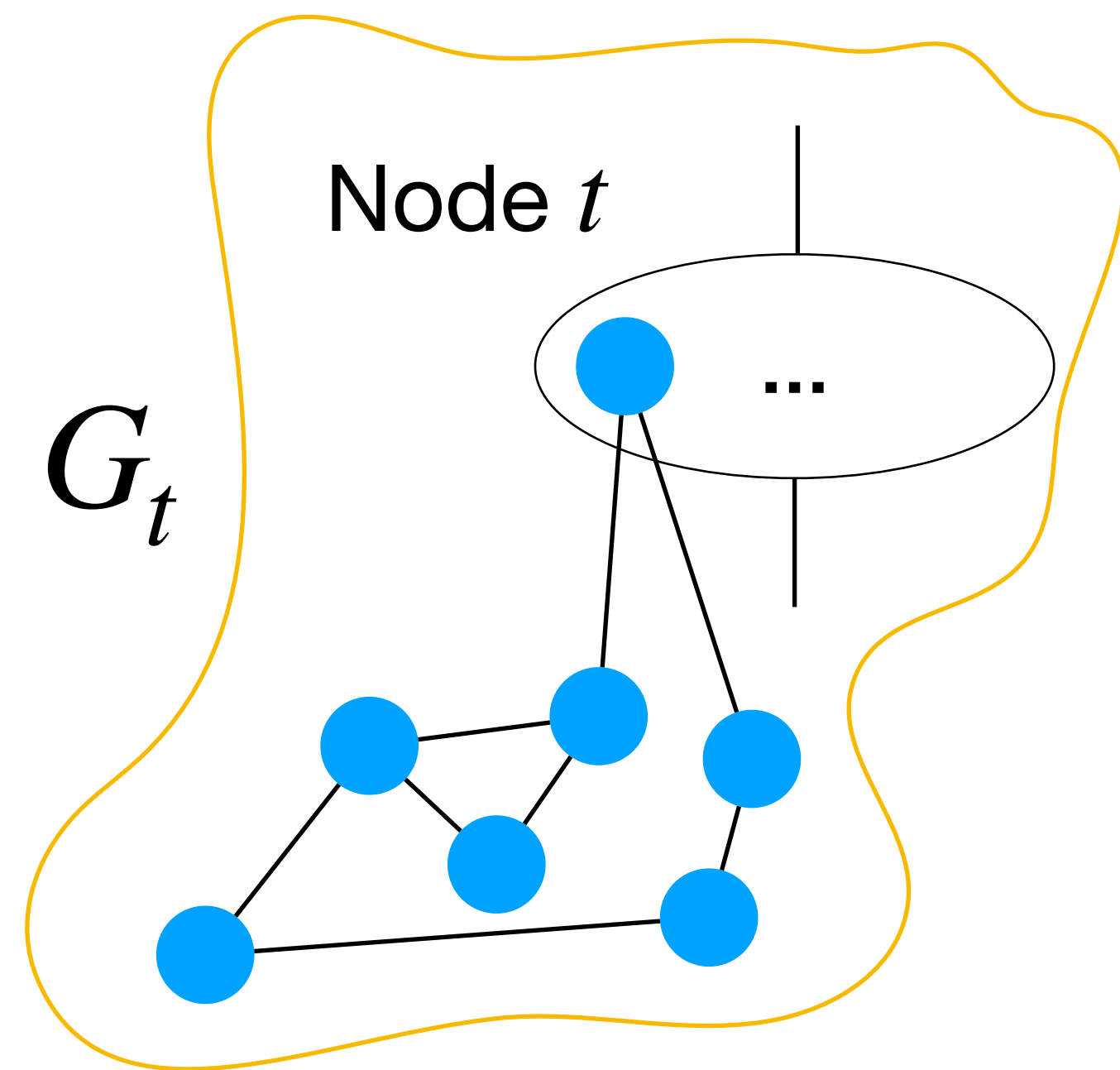
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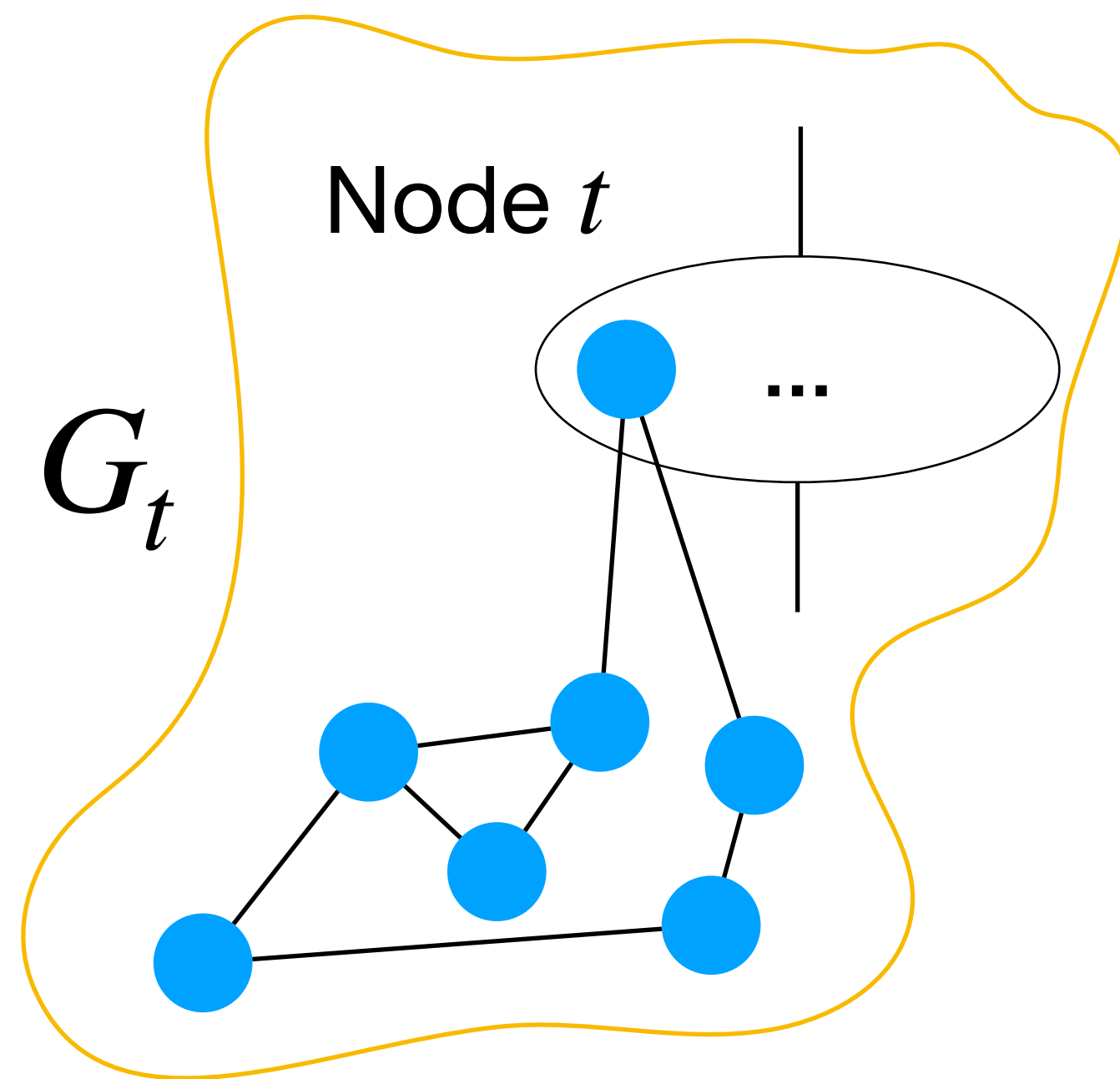
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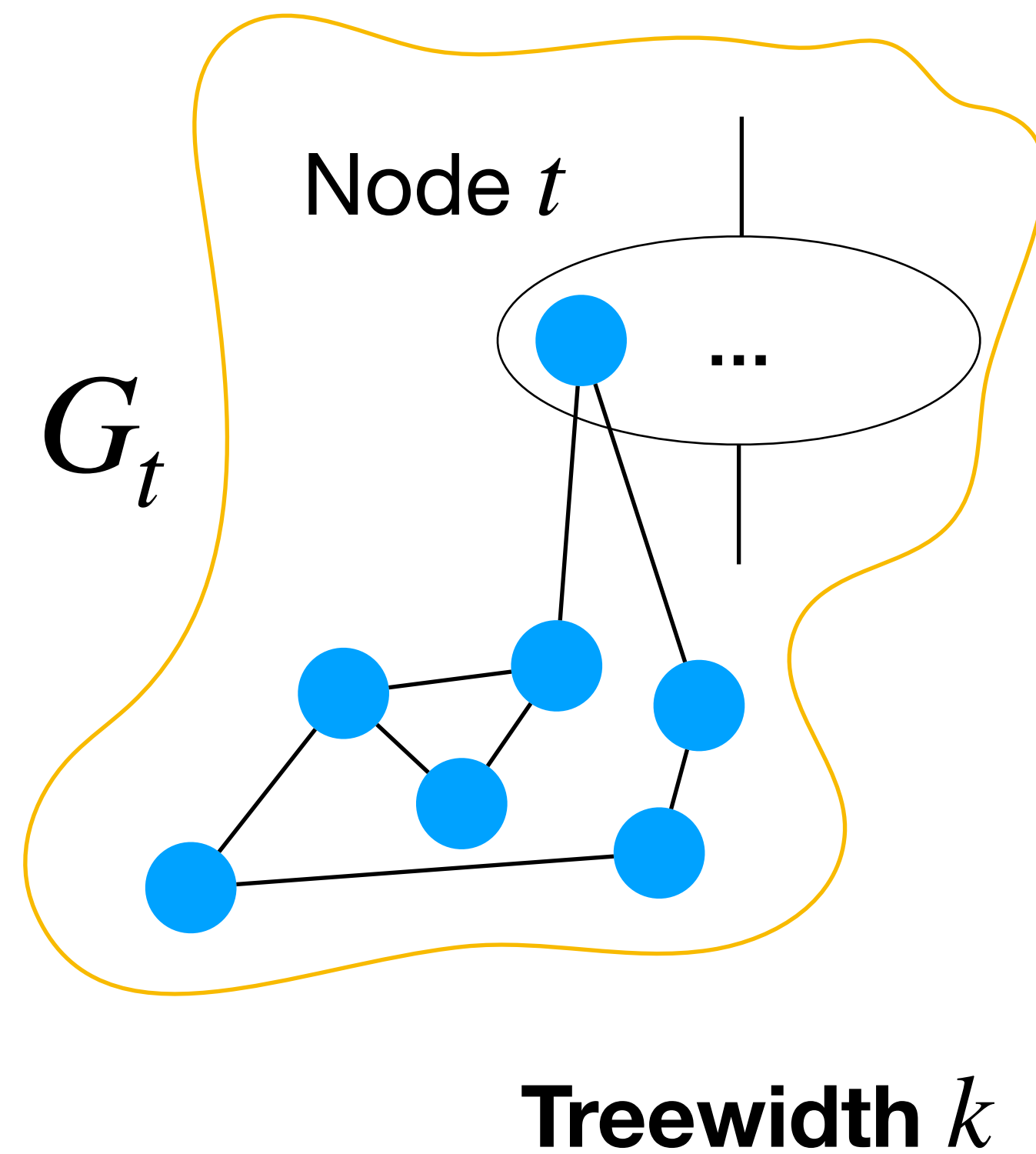


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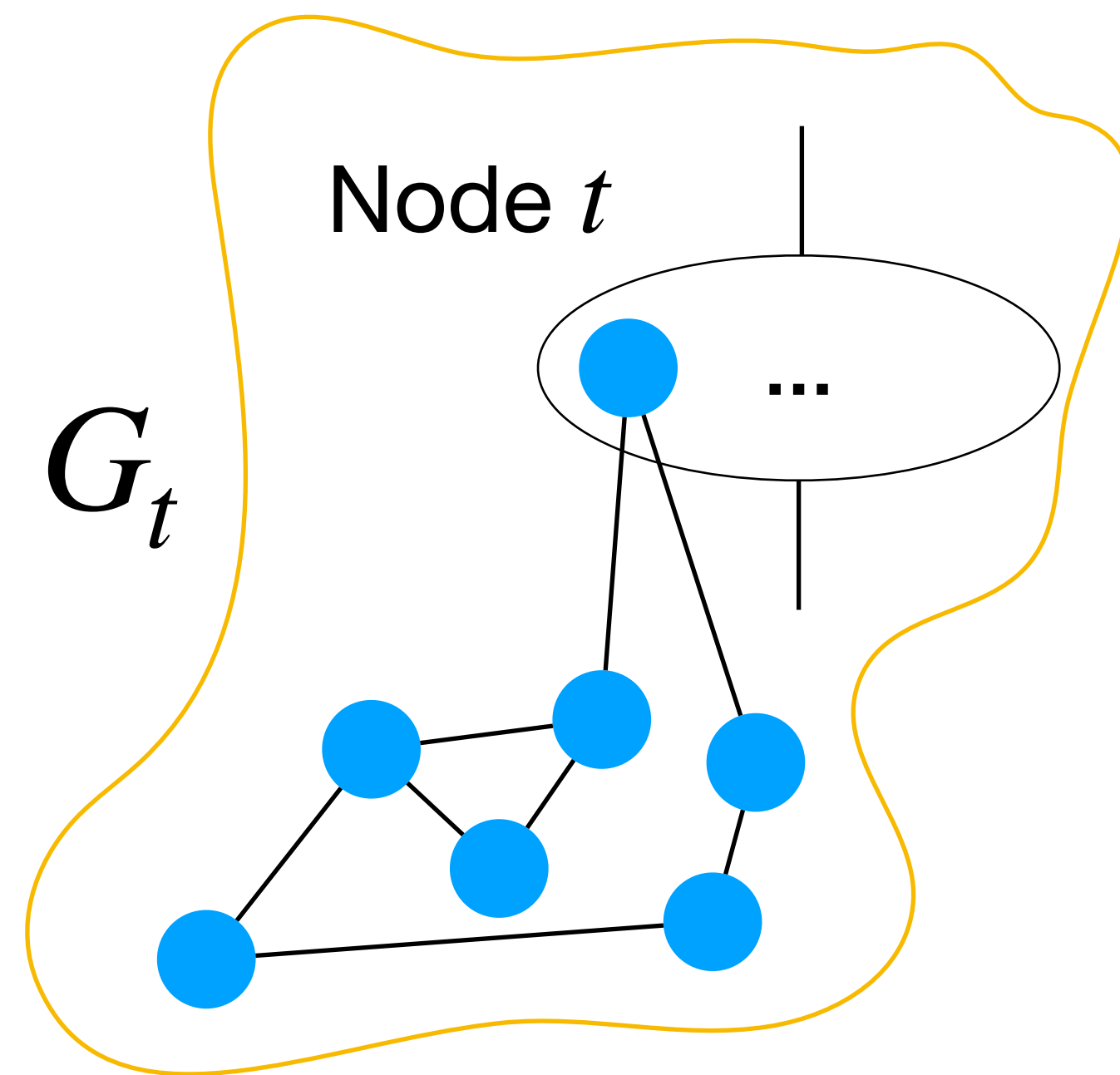
For each subset  $S \subseteq \chi(t)$  of bag vertices,  
 $n_t(S) = | \text{max. IS of } G_t \text{ containing } S |$ .

# Independent Set with Tree Decompositions



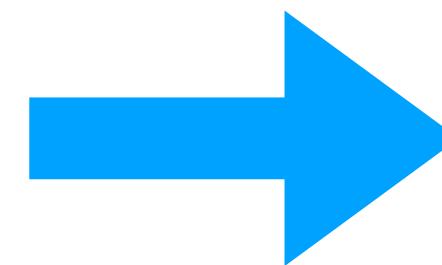
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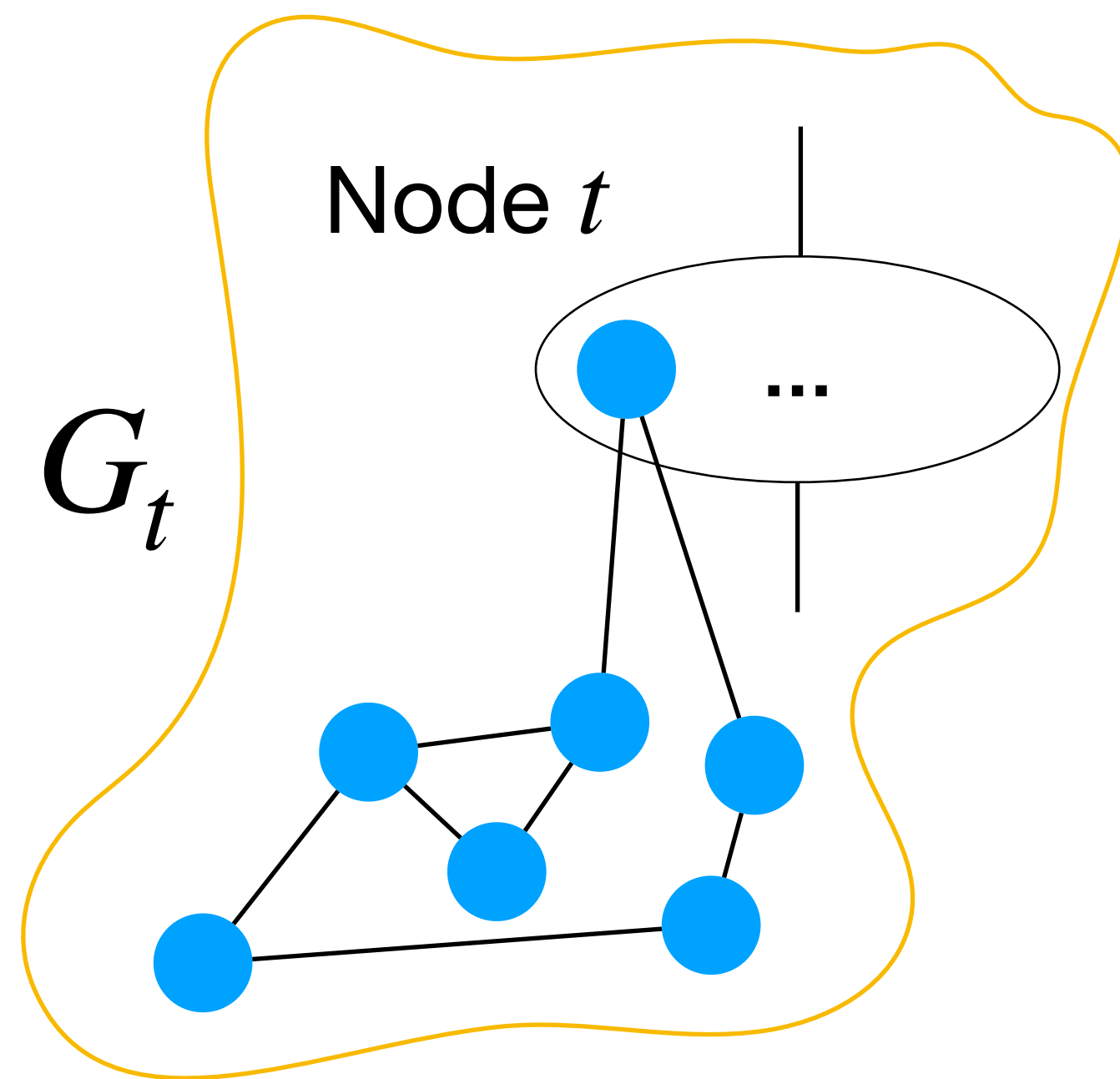


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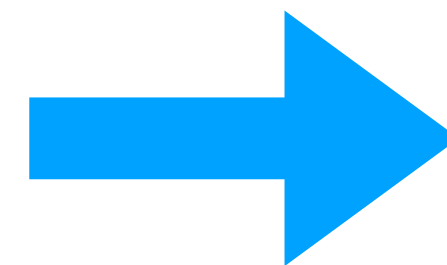


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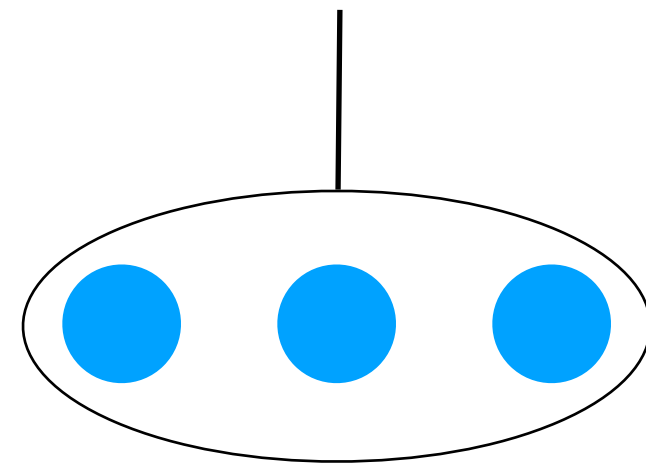
$\leq 2^{k+1}$  subsets

# Nice Tree Decompositions



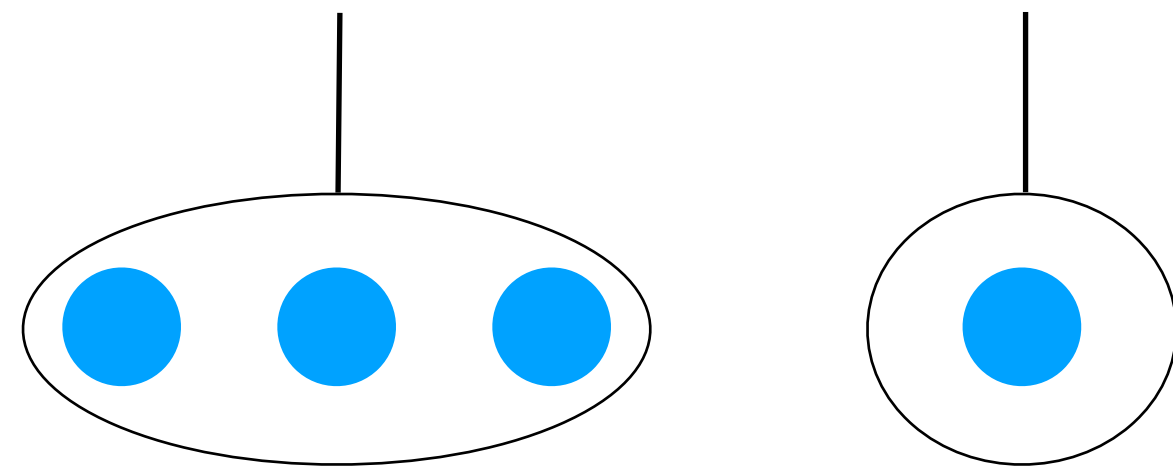
# Nice Tree Decompositions

“Leaf”



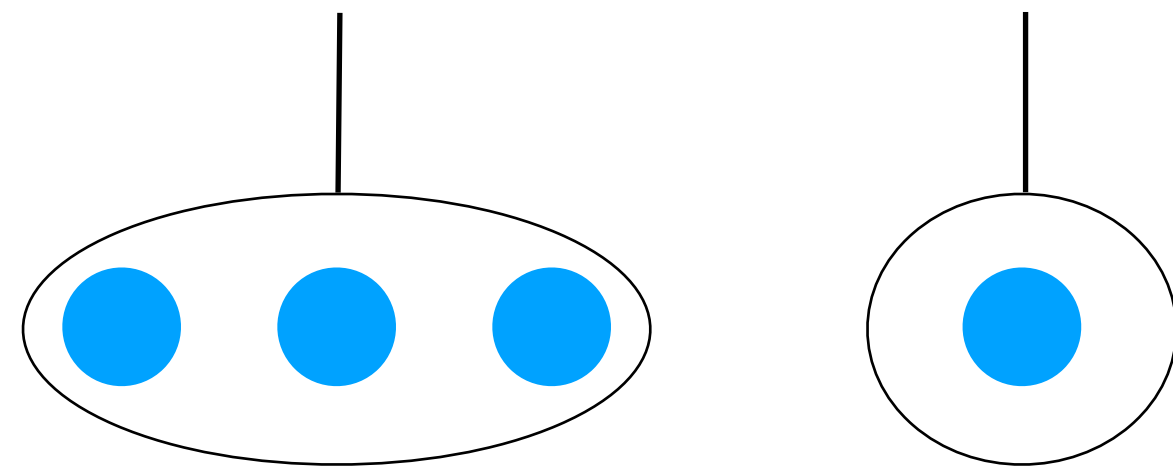
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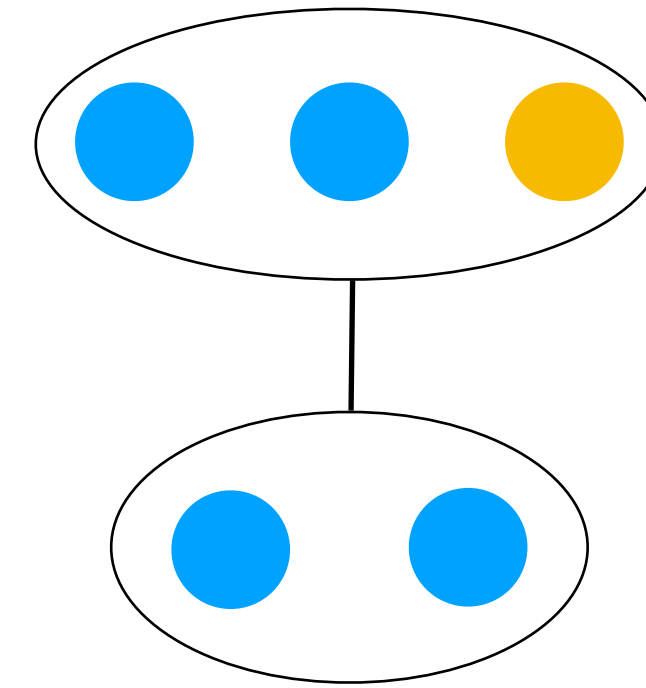


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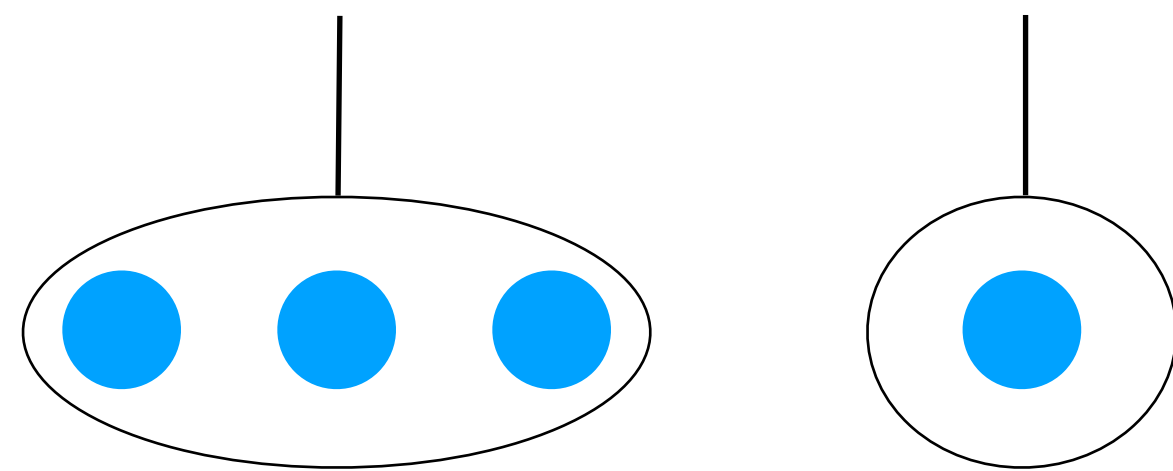


“Introduce”

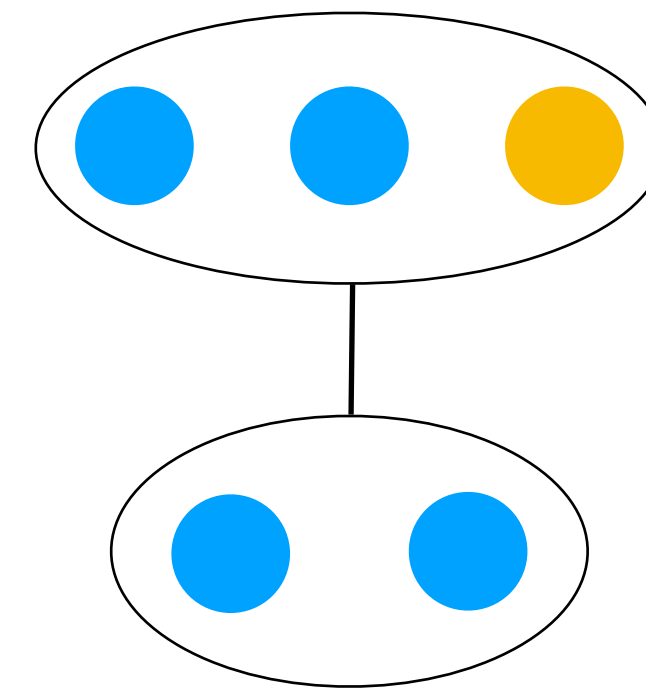


# Nice Tree Decompositions

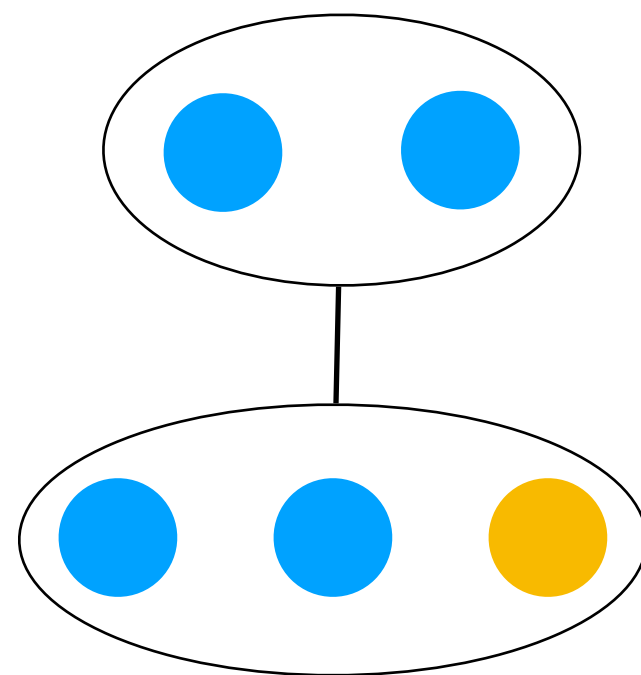
**“Leaf”**



**“Introduce”**

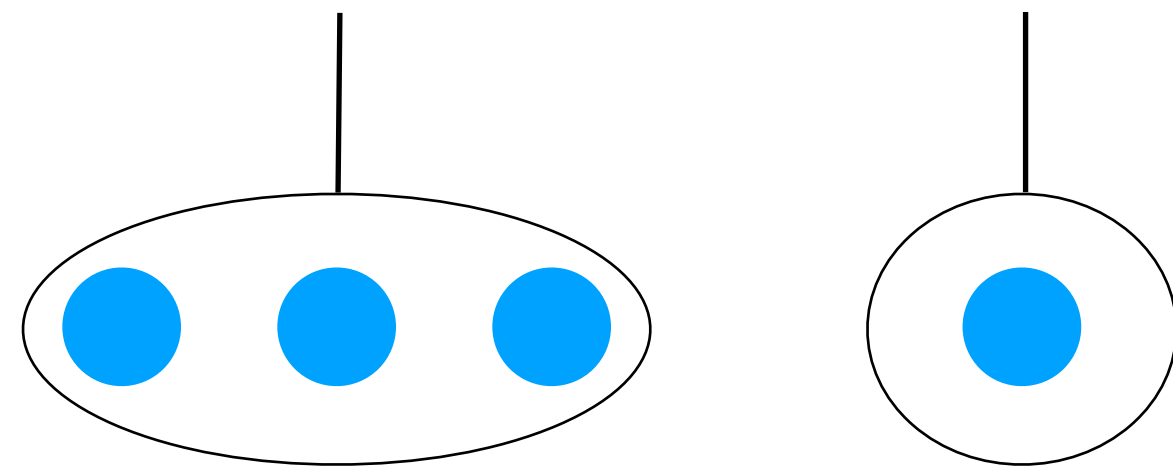


**“Forget”**

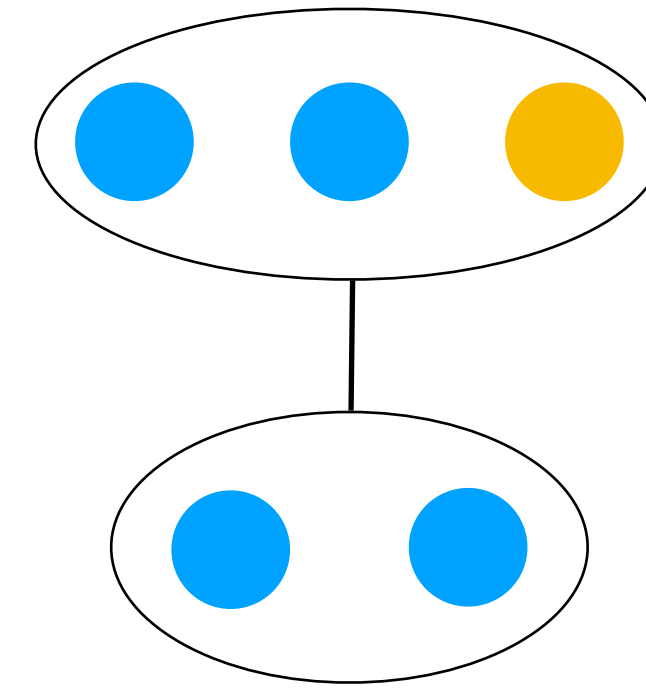


# Nice Tree Decompositions

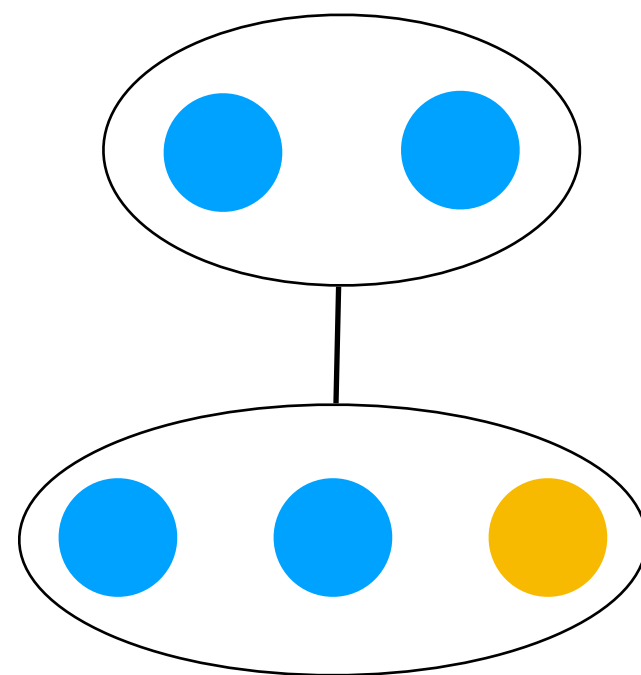
**“Leaf”**



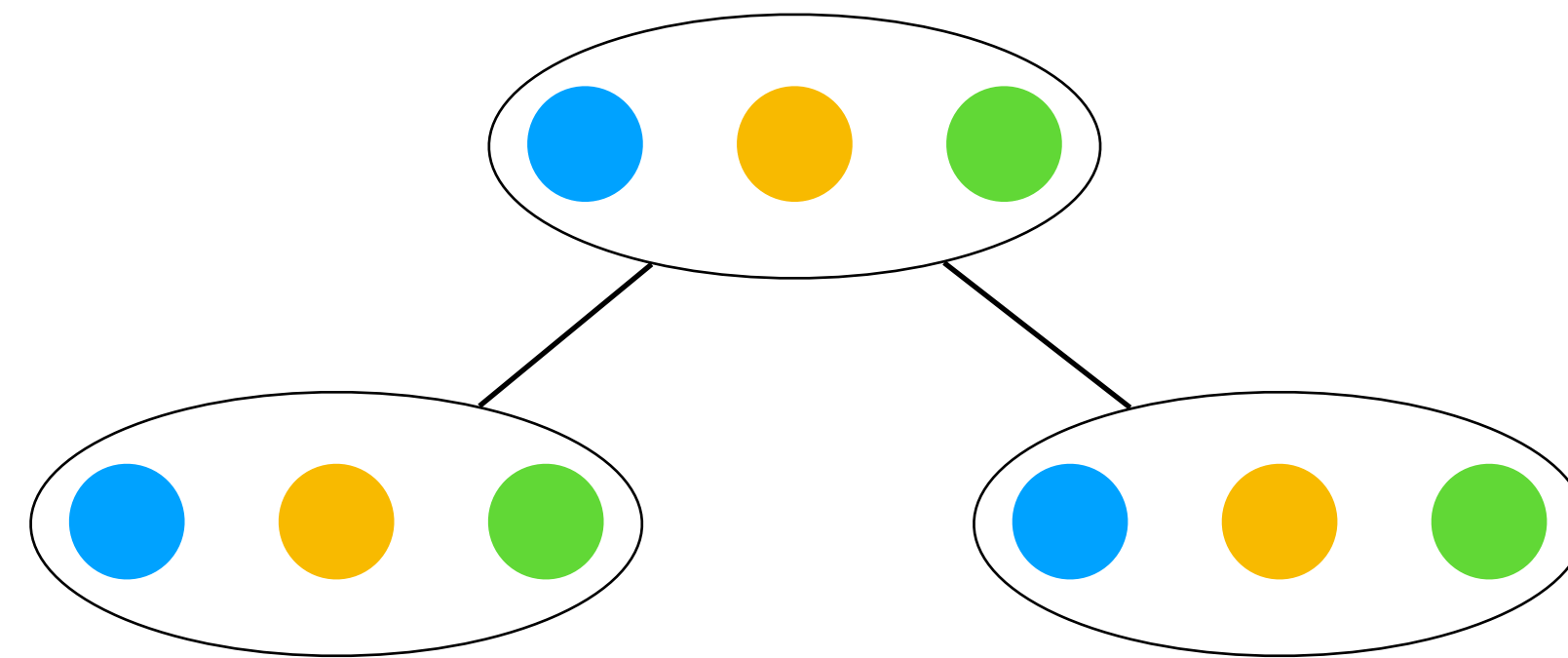
**“Introduce”**



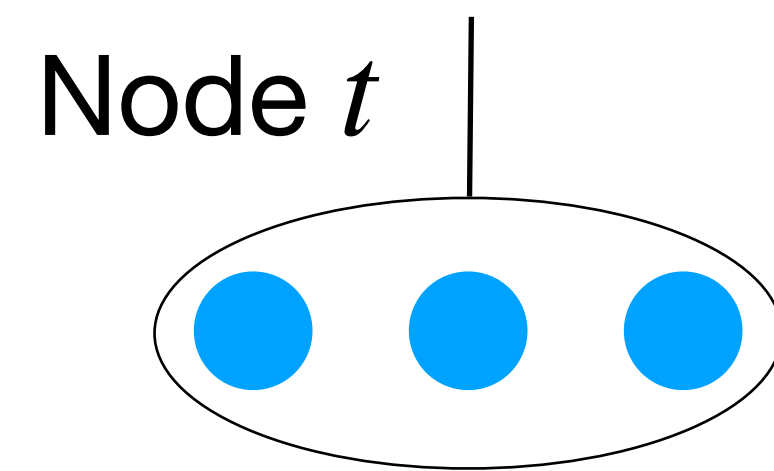
**“Forget”**



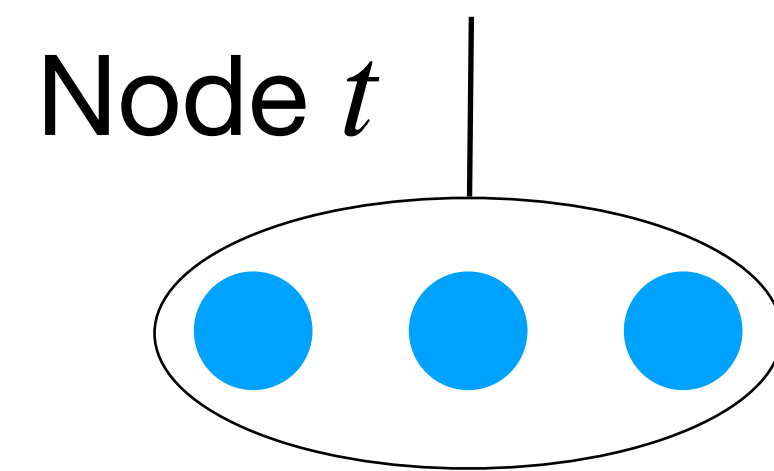
**“Join”**



# Leaf Nodes

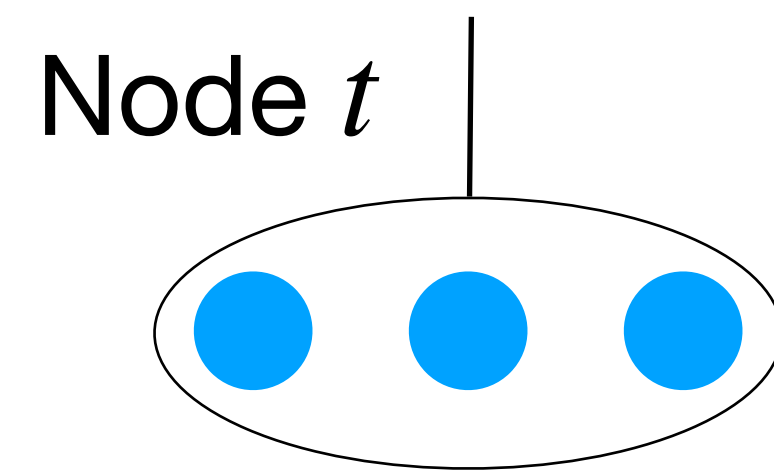


# Leaf Nodes



**for**  $S \subseteq \chi(t)$ :

# Leaf Nodes

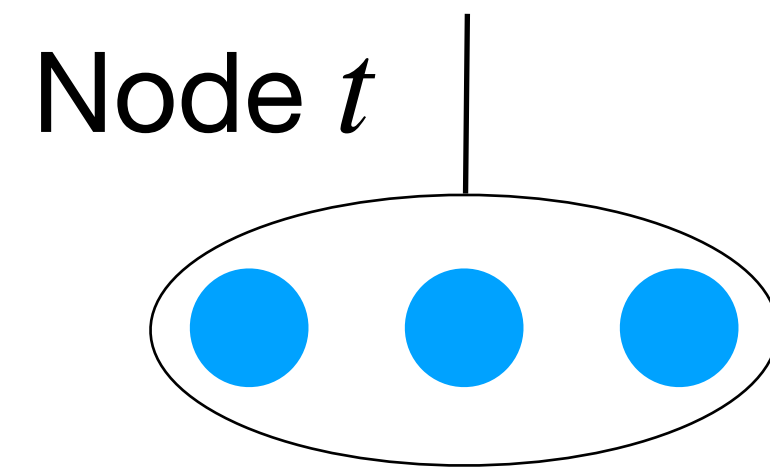


**for**  $S \subseteq \chi(t)$ :

**if**  $S$  is an independent set



# Leaf Nodes

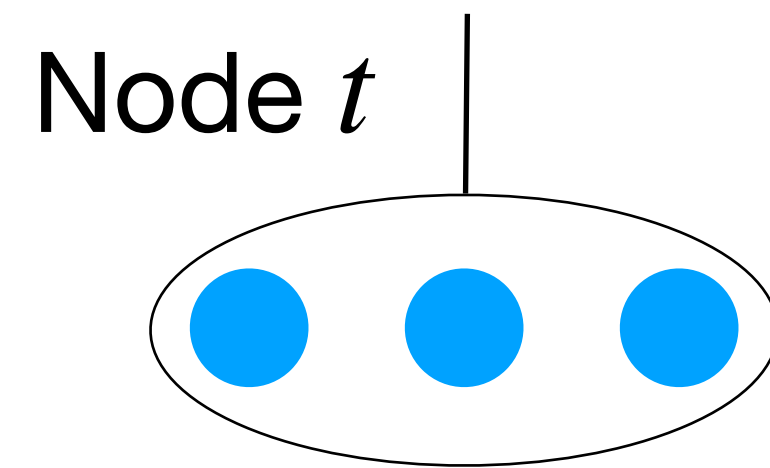


**for**  $S \subseteq \chi(t)$ :

**if**  $S$  is an independent set

**then**  $n_t(S) := |S|$

# Leaf Nodes



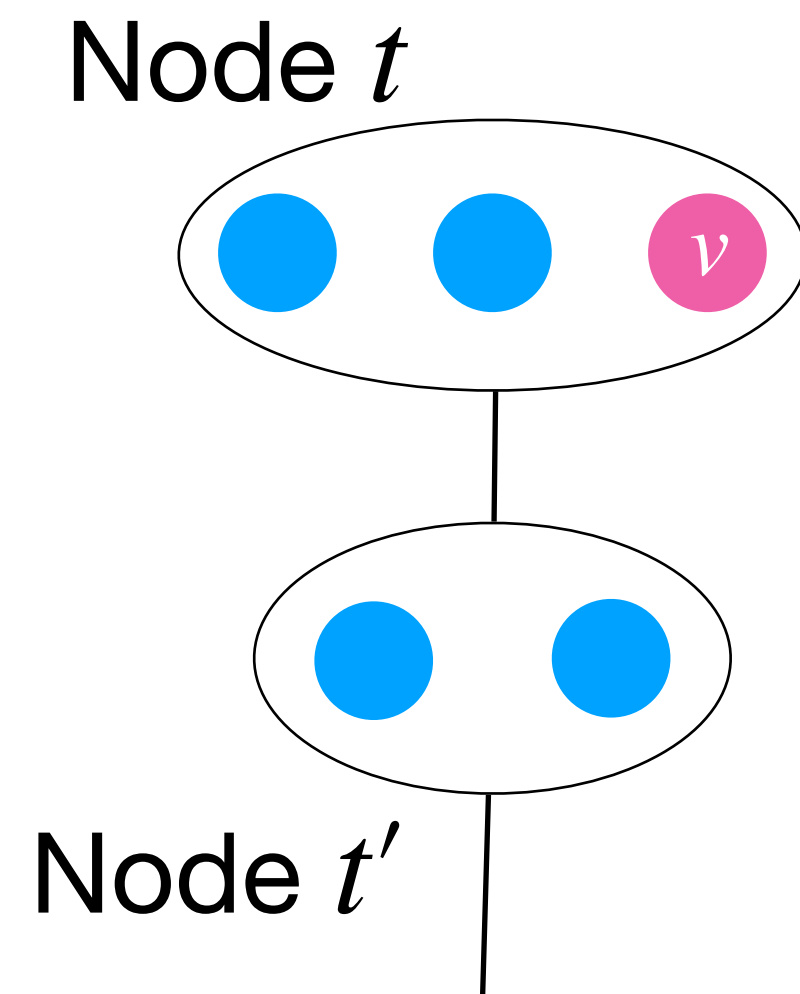
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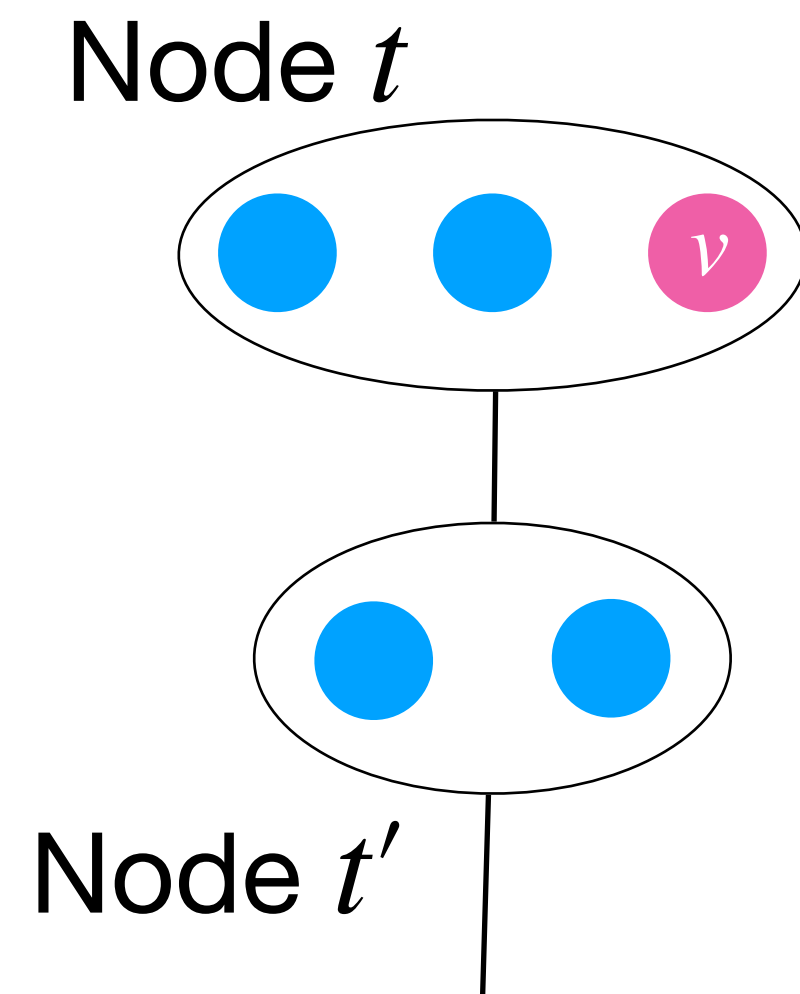
**then**  $n_t(S) := |S|$

**else**  $n_t(S) := 0$

# Introduce Nodes

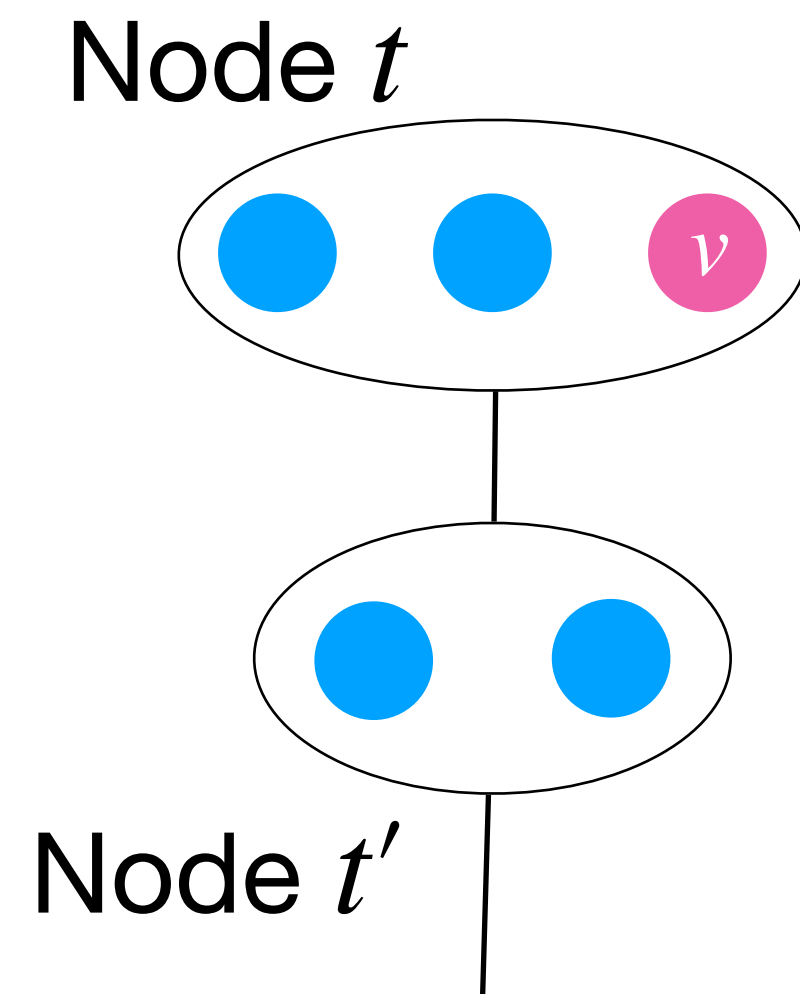


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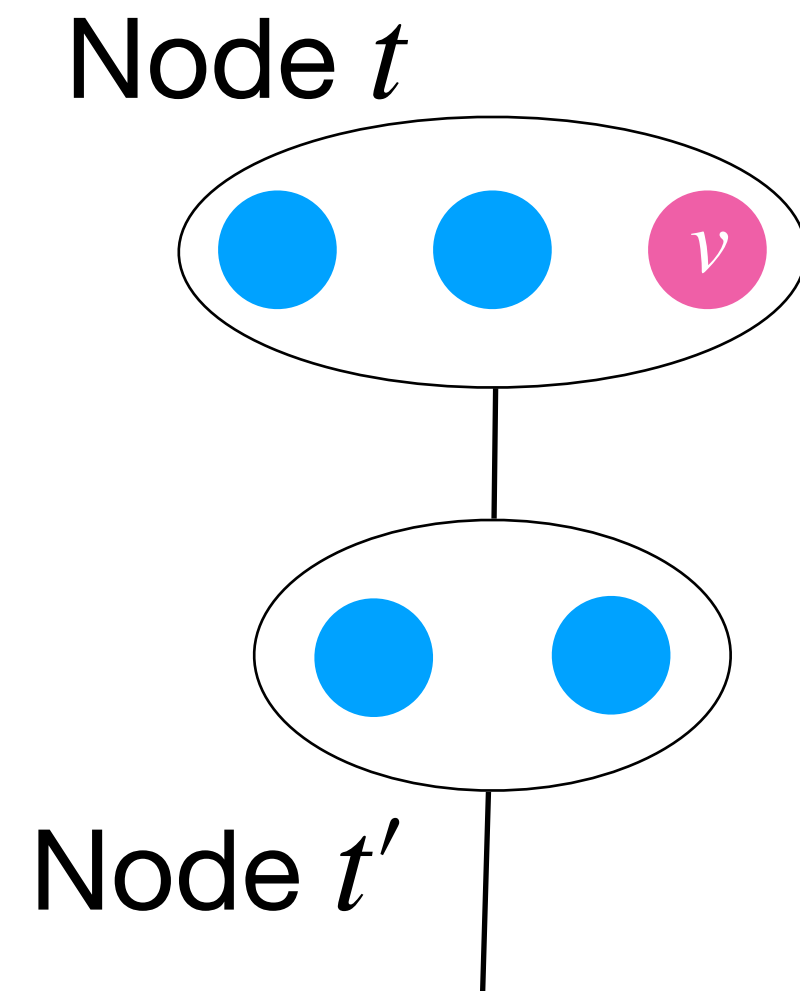
for  $S \subseteq \chi(t)$ :

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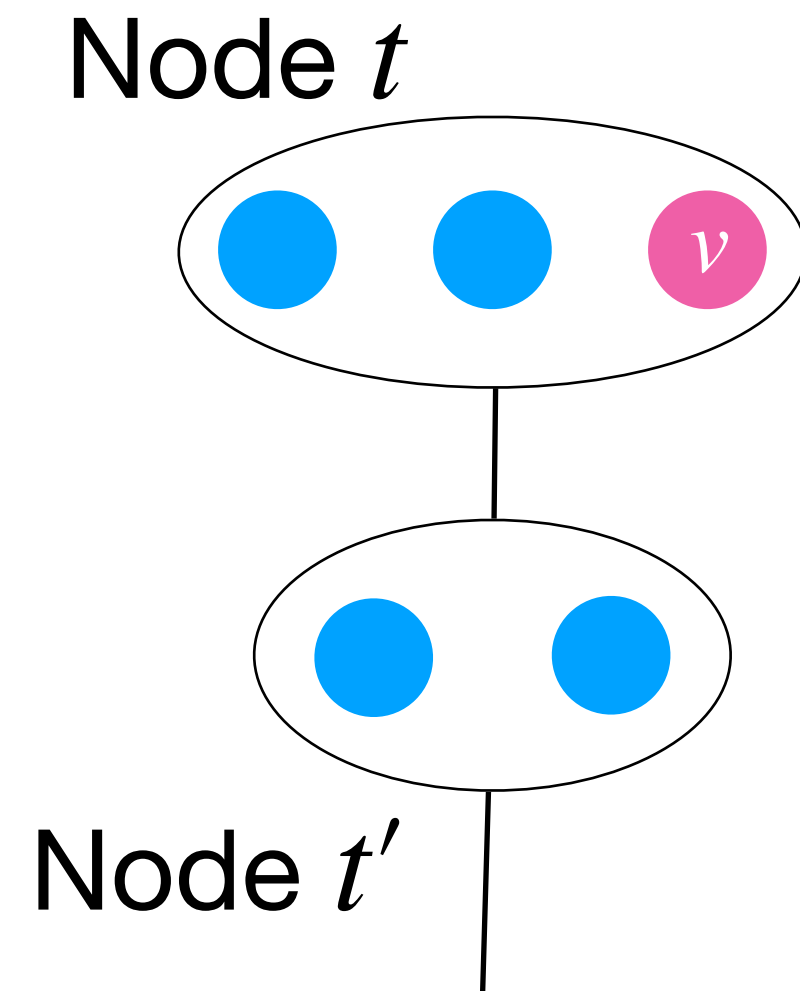
**for**  $S \subseteq \chi(t)$ :  
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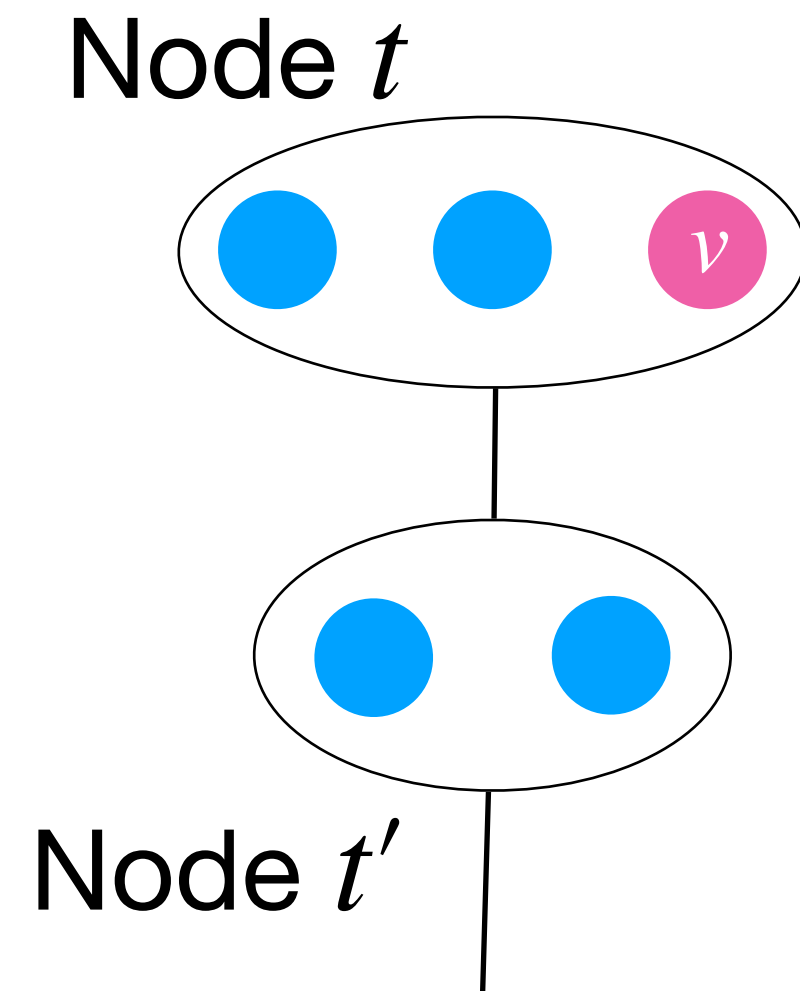
**for**  $S \subseteq \chi(t)$ :  
**if**  $v \notin S$   
**then**  $n_t(S) := n_{t'}(S)$

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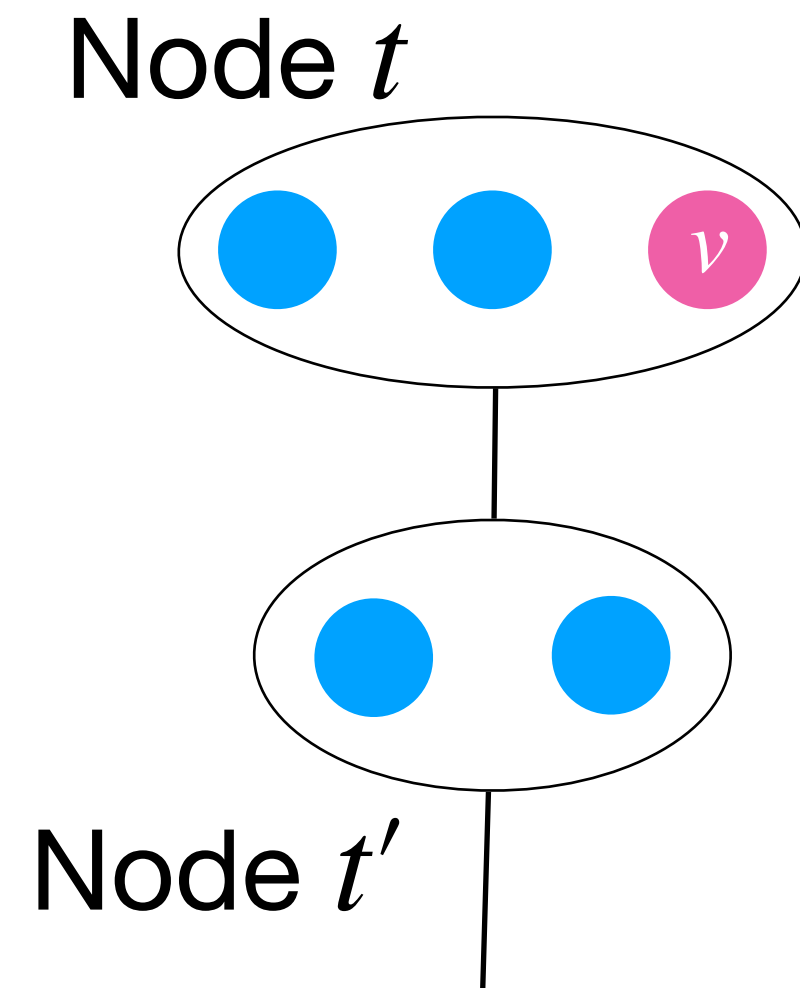
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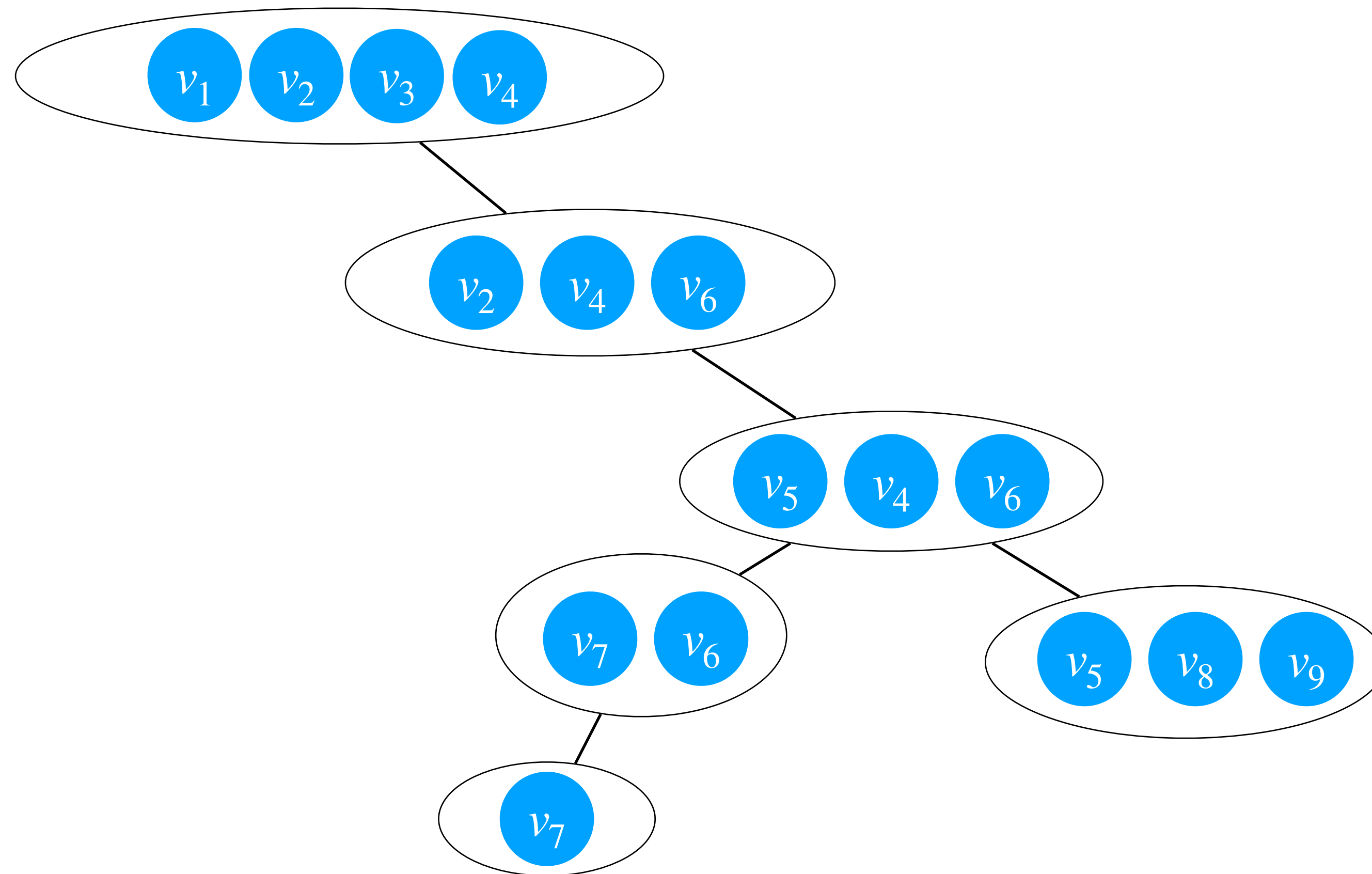


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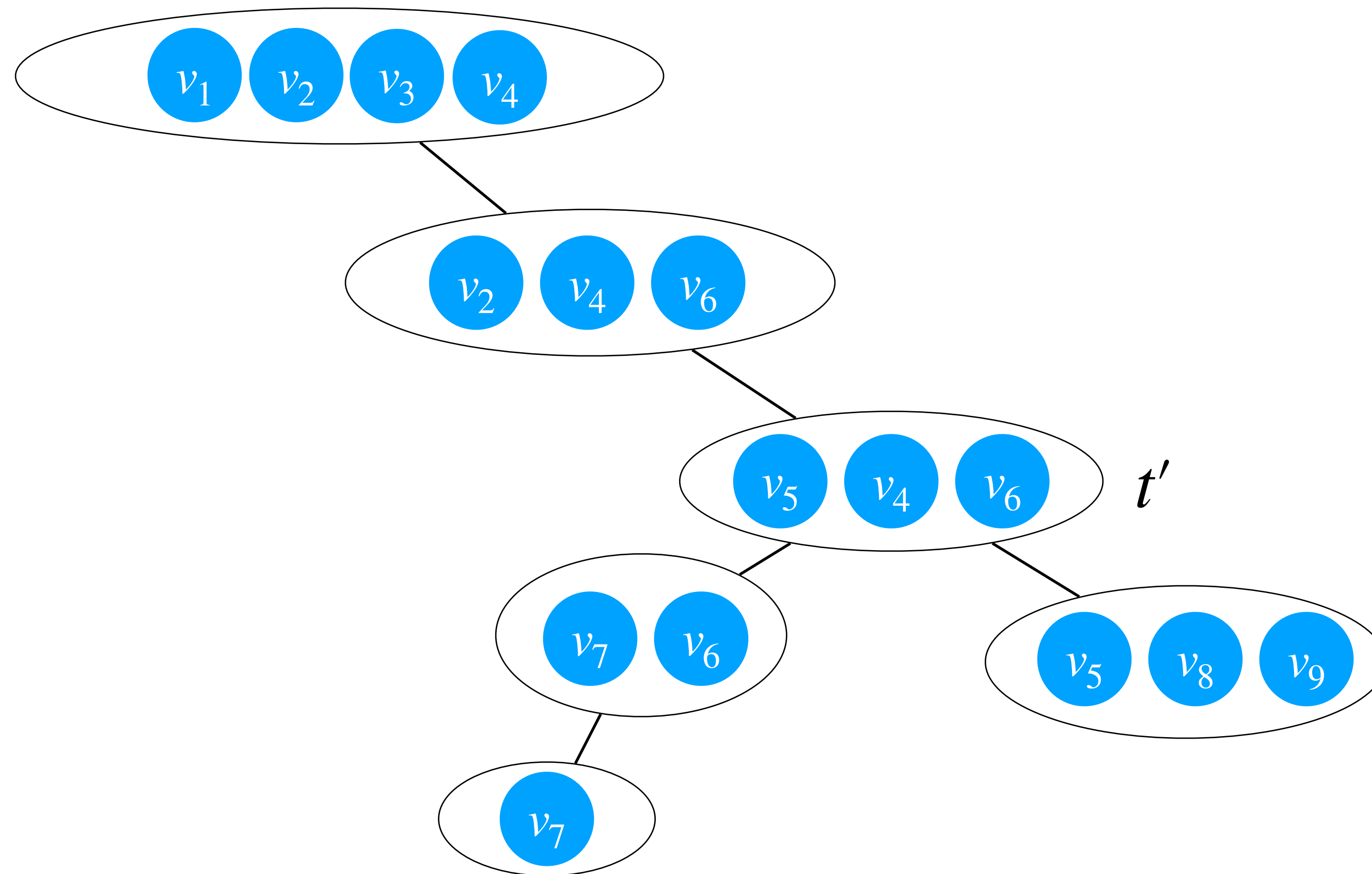


```
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    then  $n_t(S) := 0$   
  else  $n_t(S) := n_{t'}(S \setminus v) + 1$ 
```

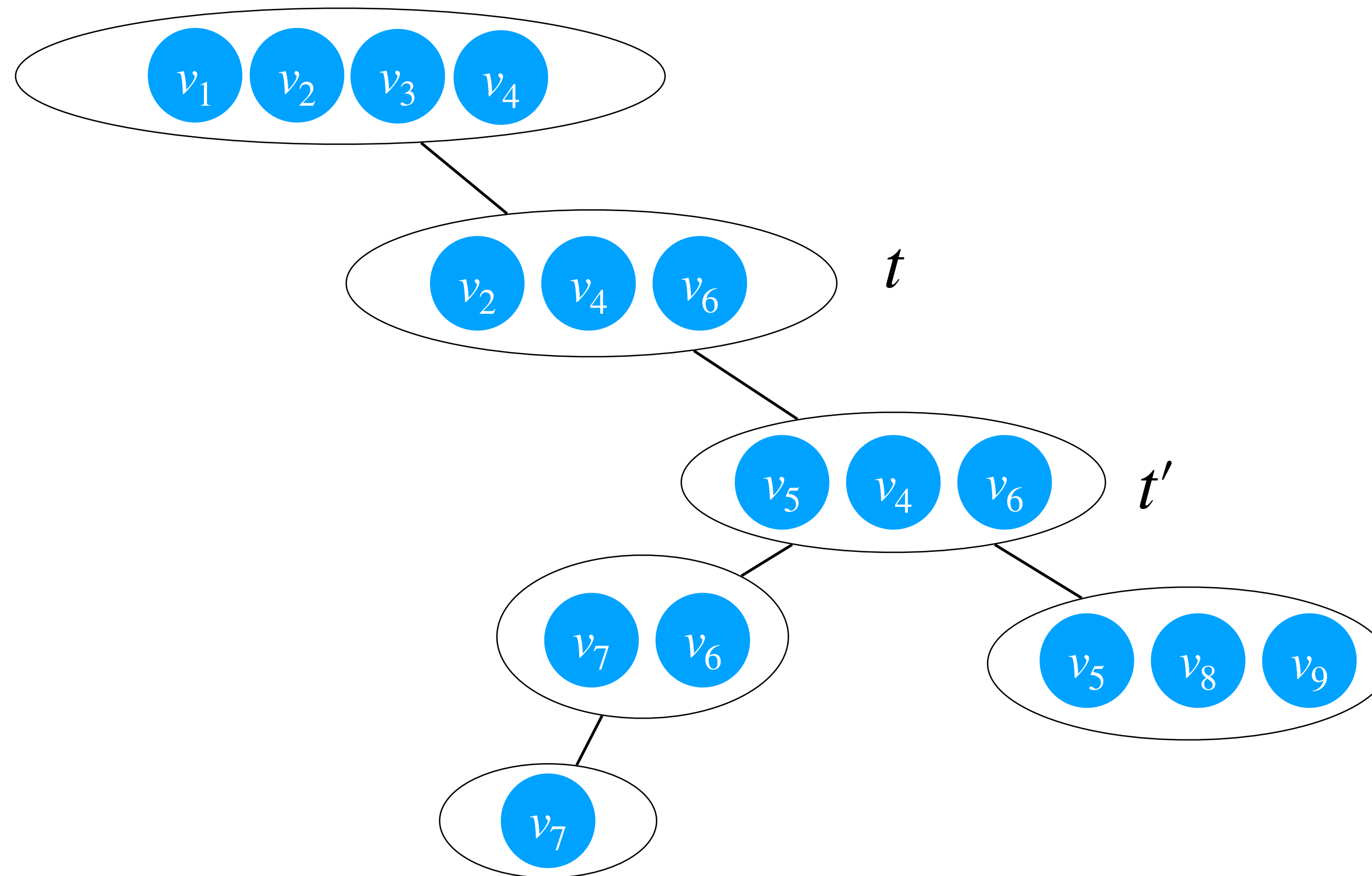
# Bags are Separators



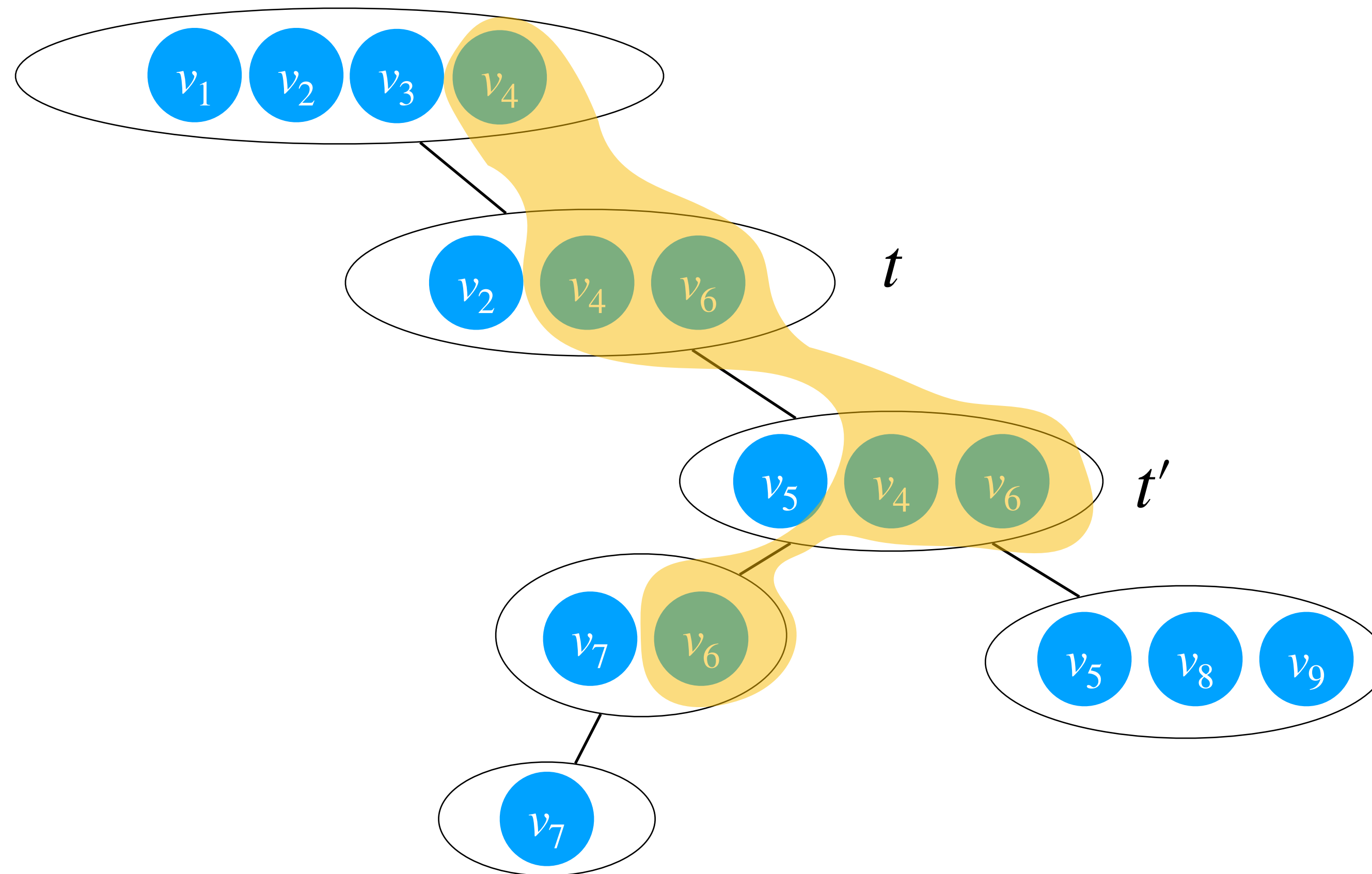
# Bags are Separators



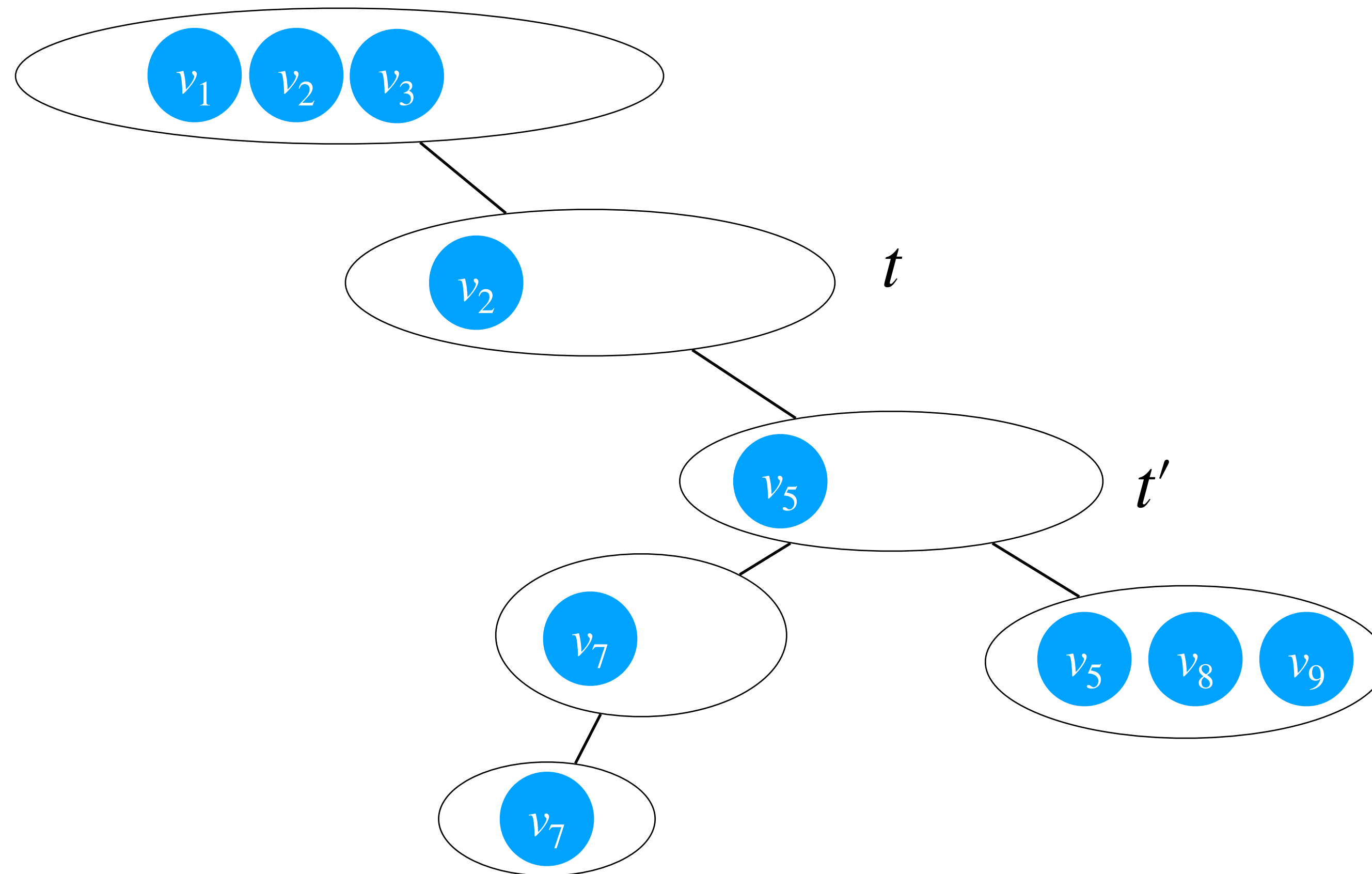
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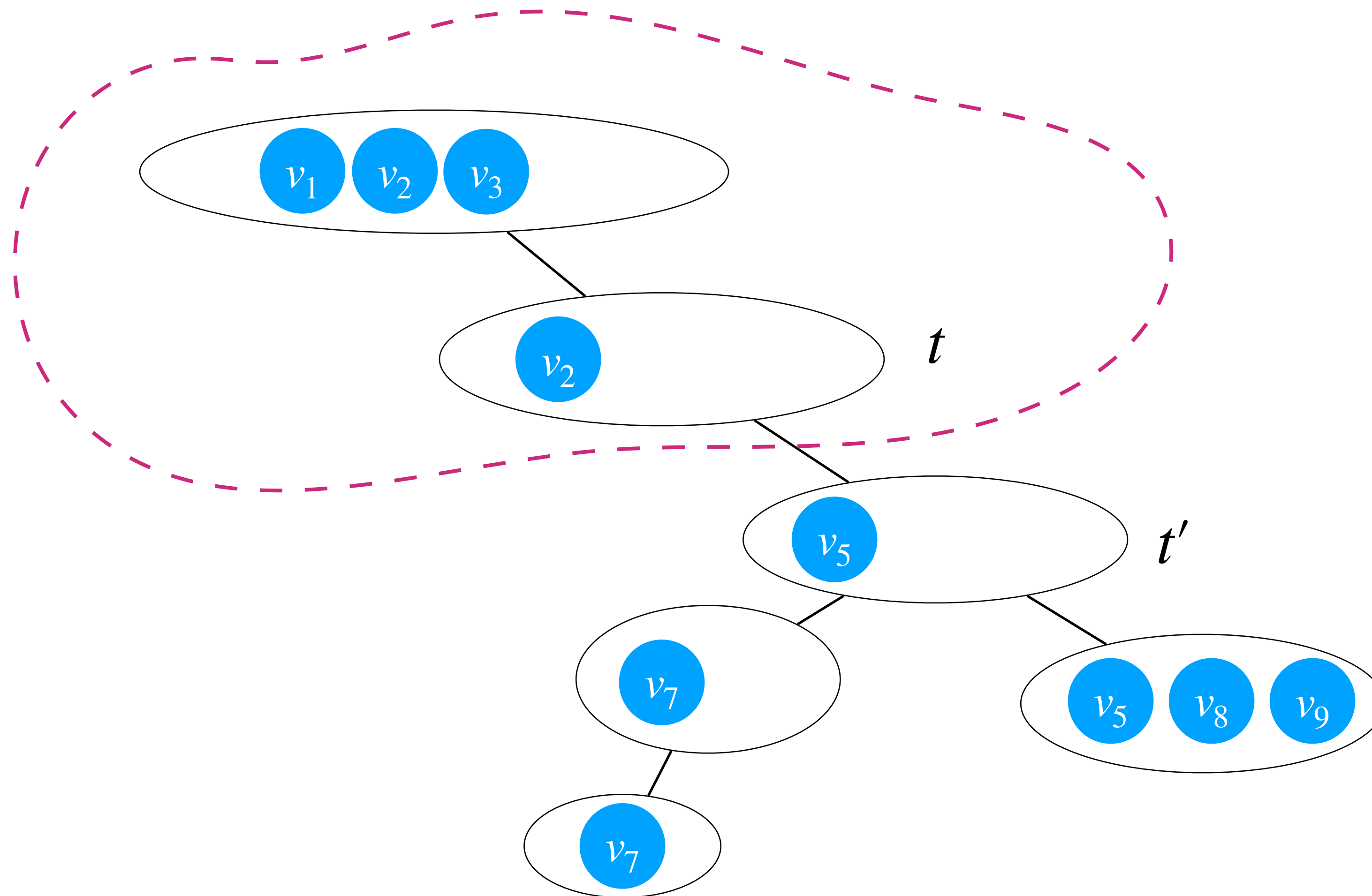
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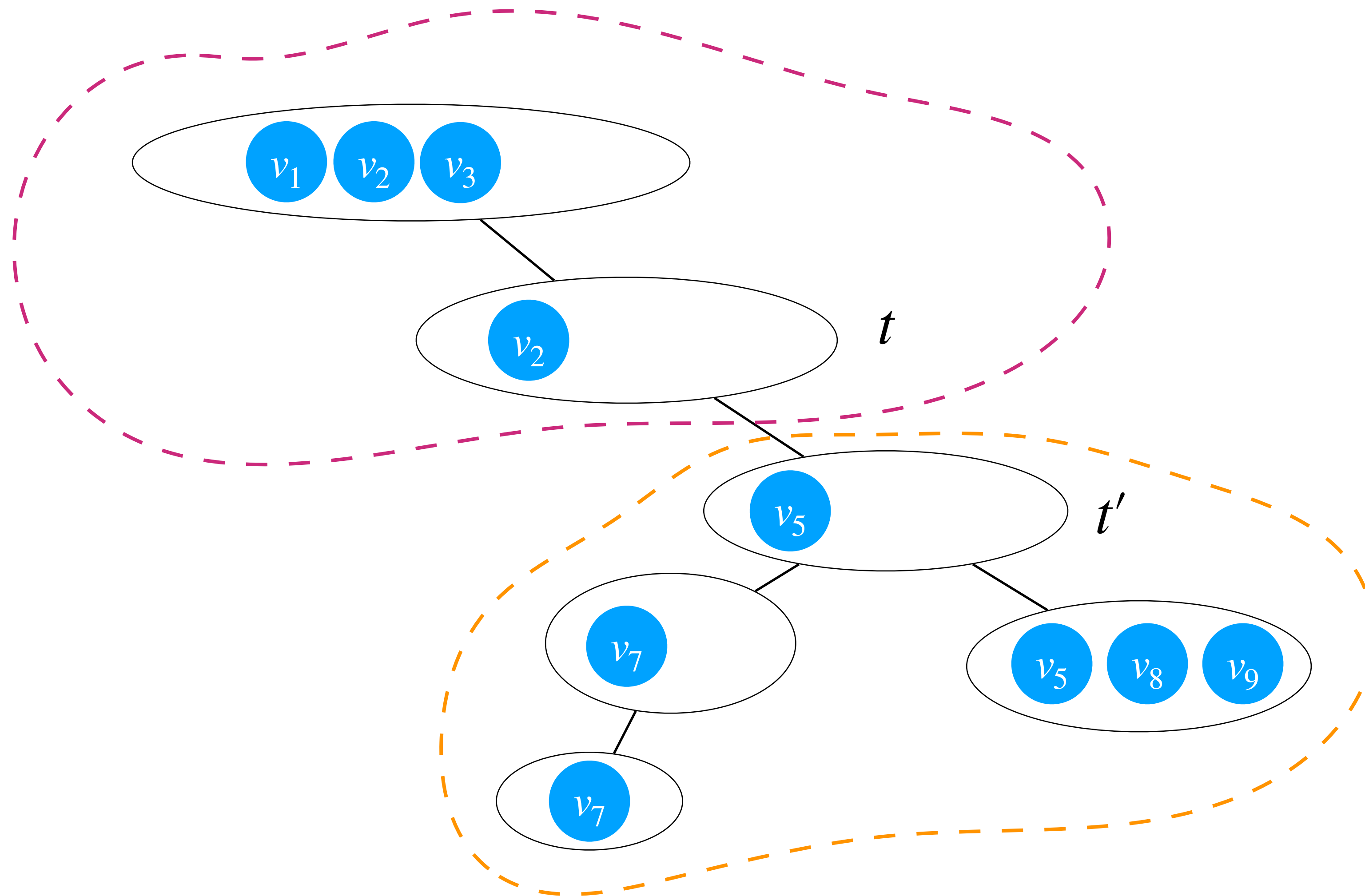
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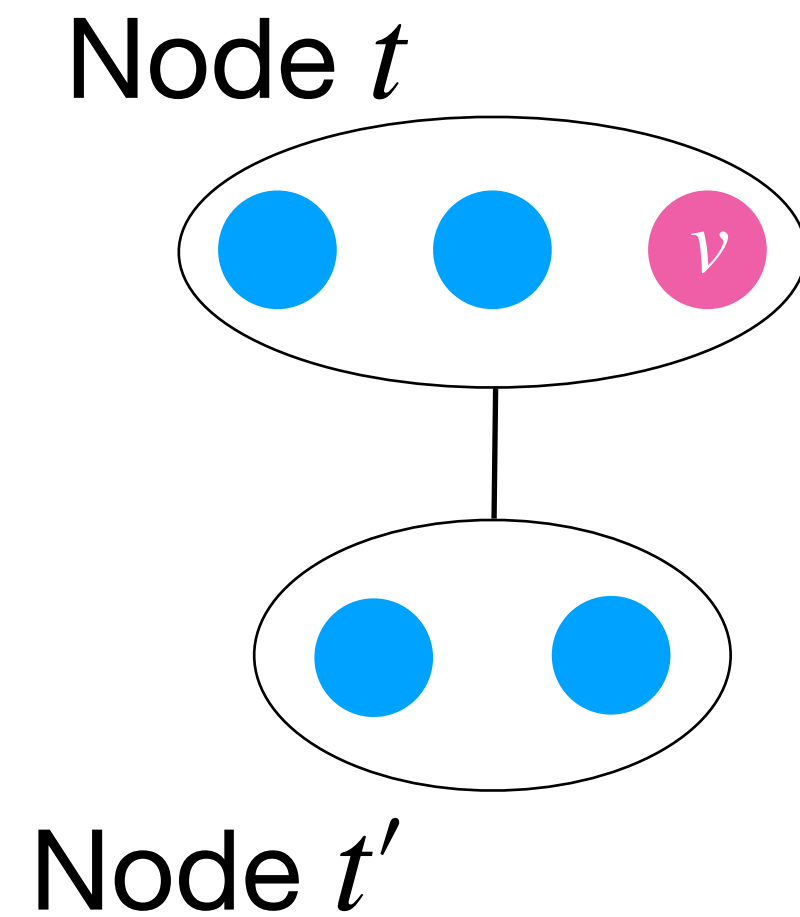


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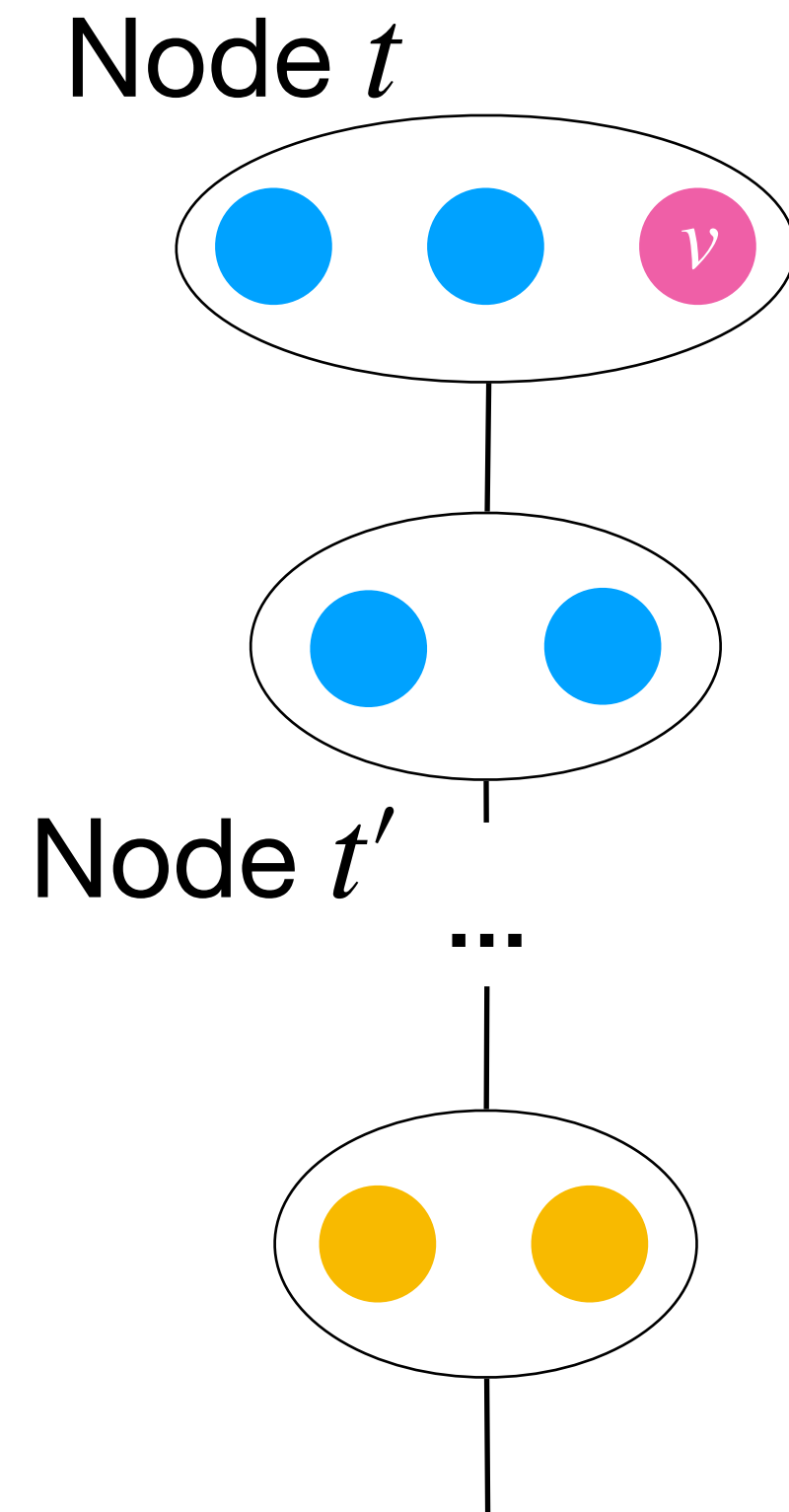


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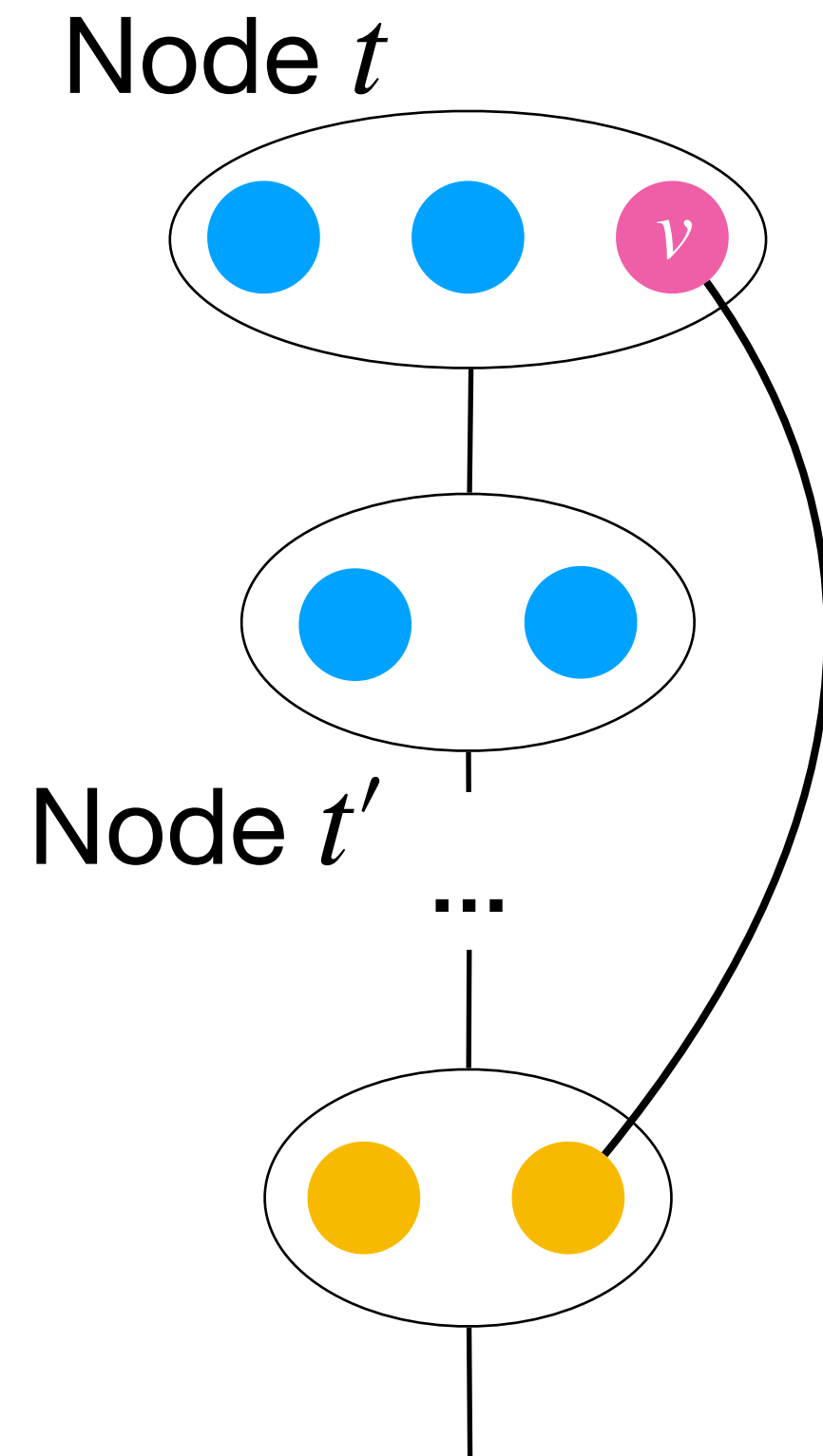
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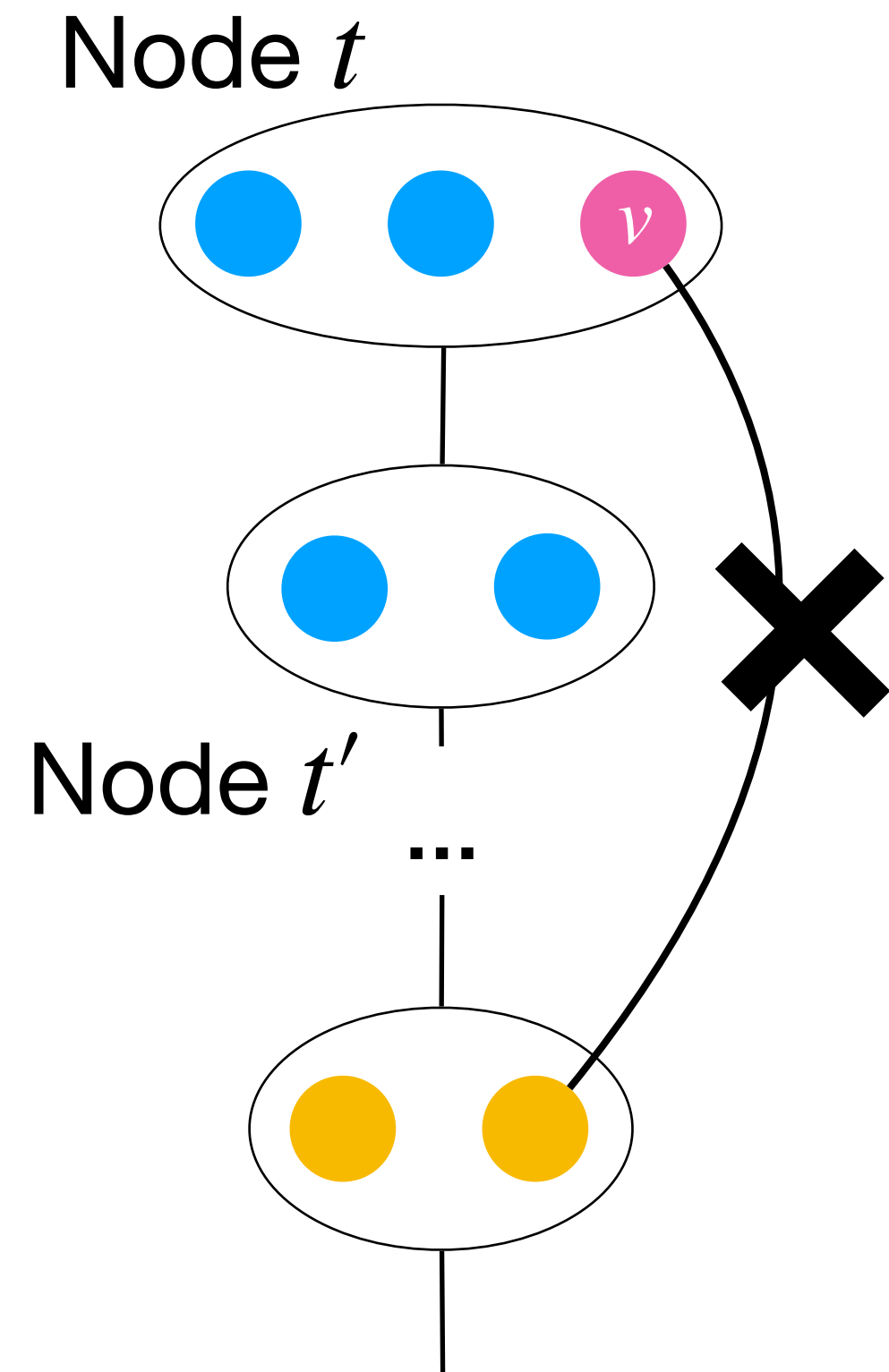
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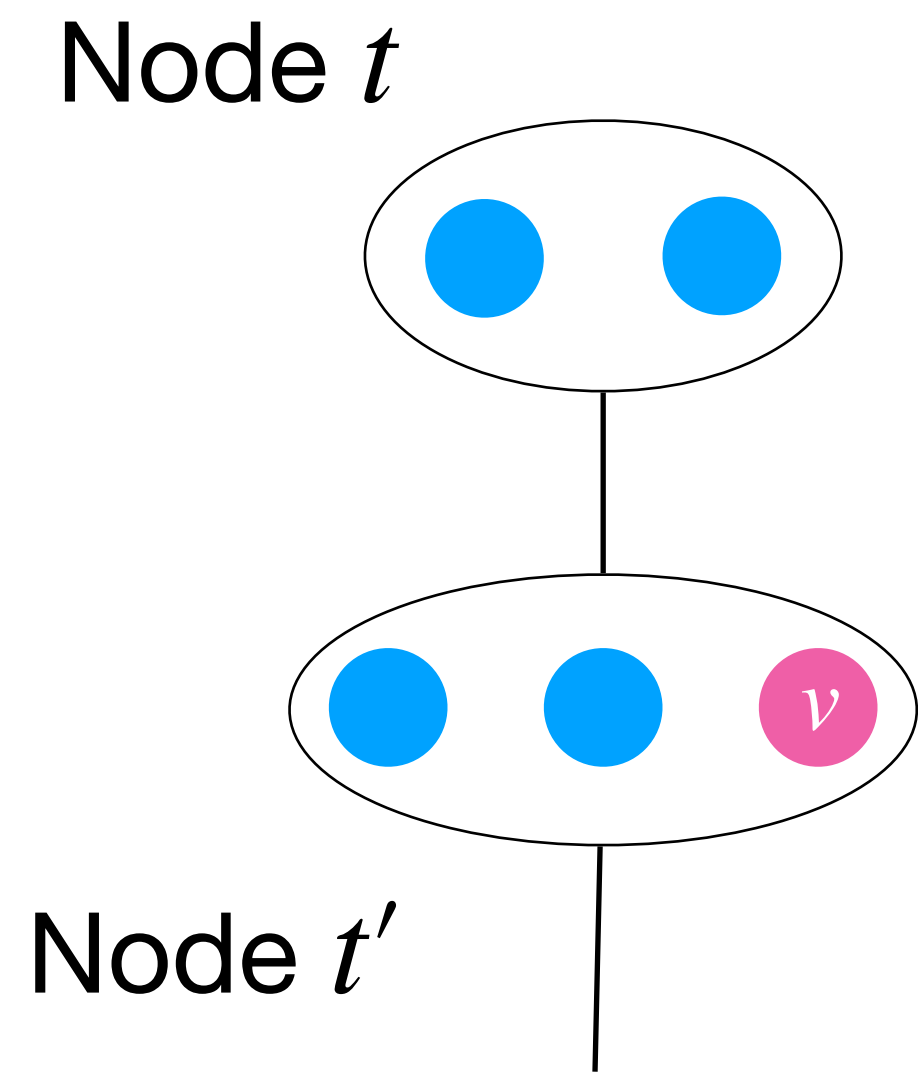
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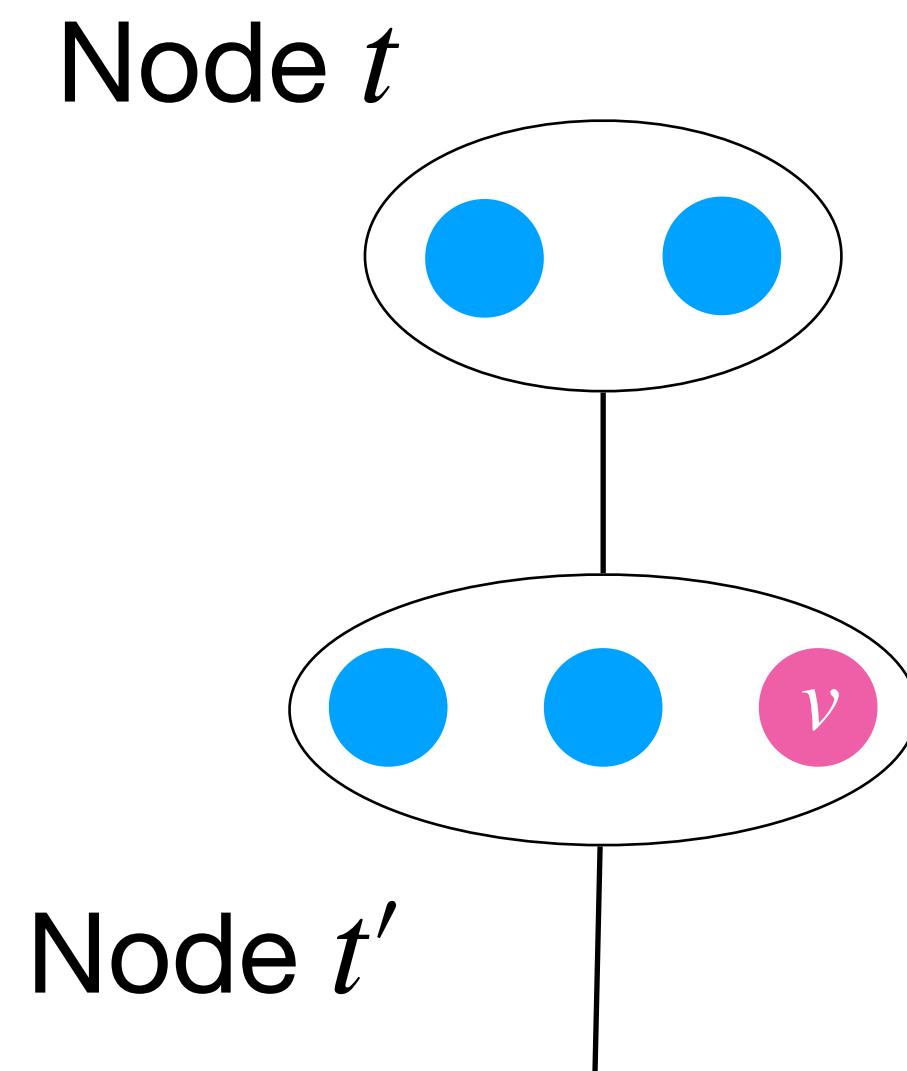


**for**  $S \subseteq \chi(t)$ :  
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# Forget Nodes

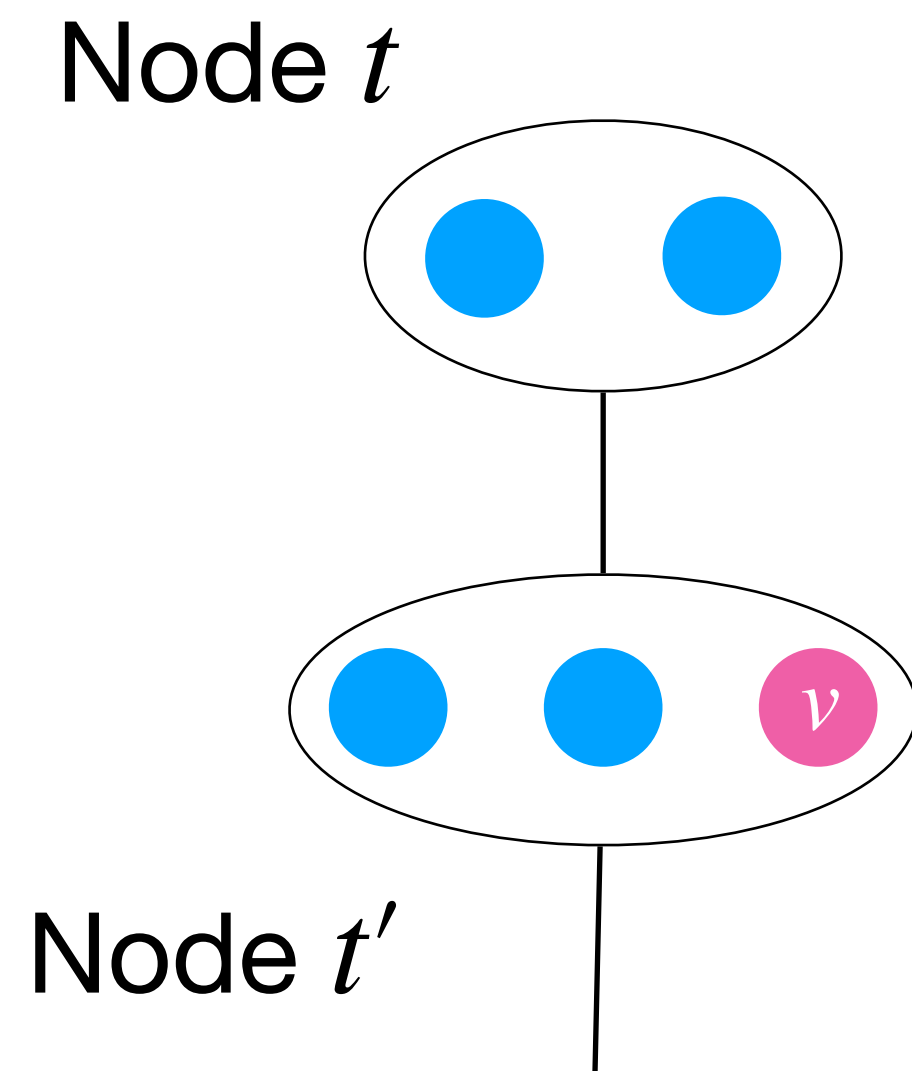


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for  $S \subseteq \chi(t)$

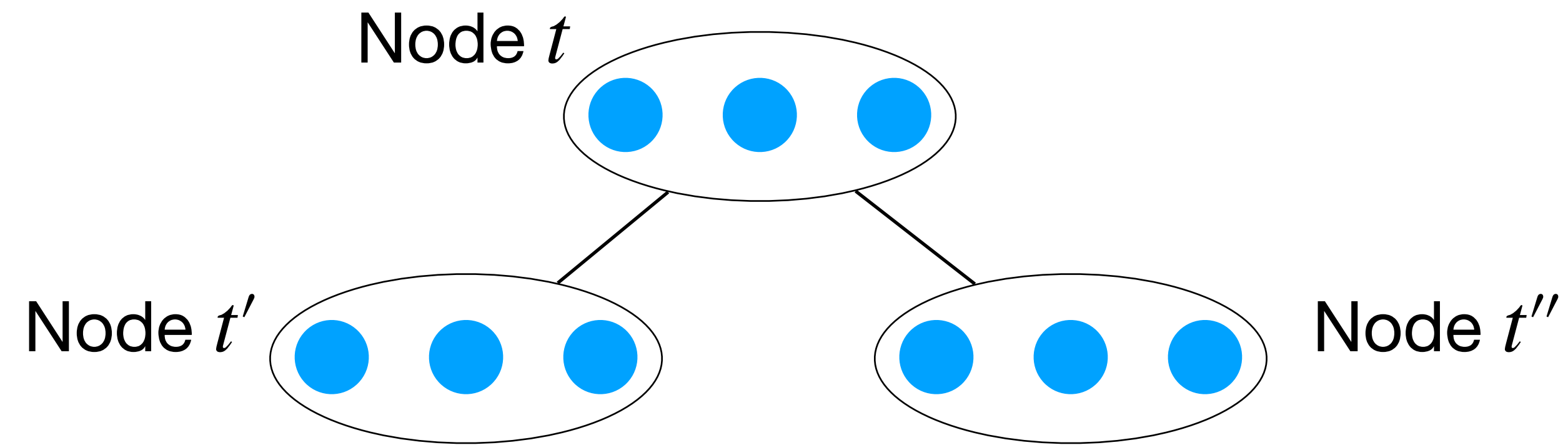
# Forget Nodes



**for**  $S \subseteq \chi(t)$

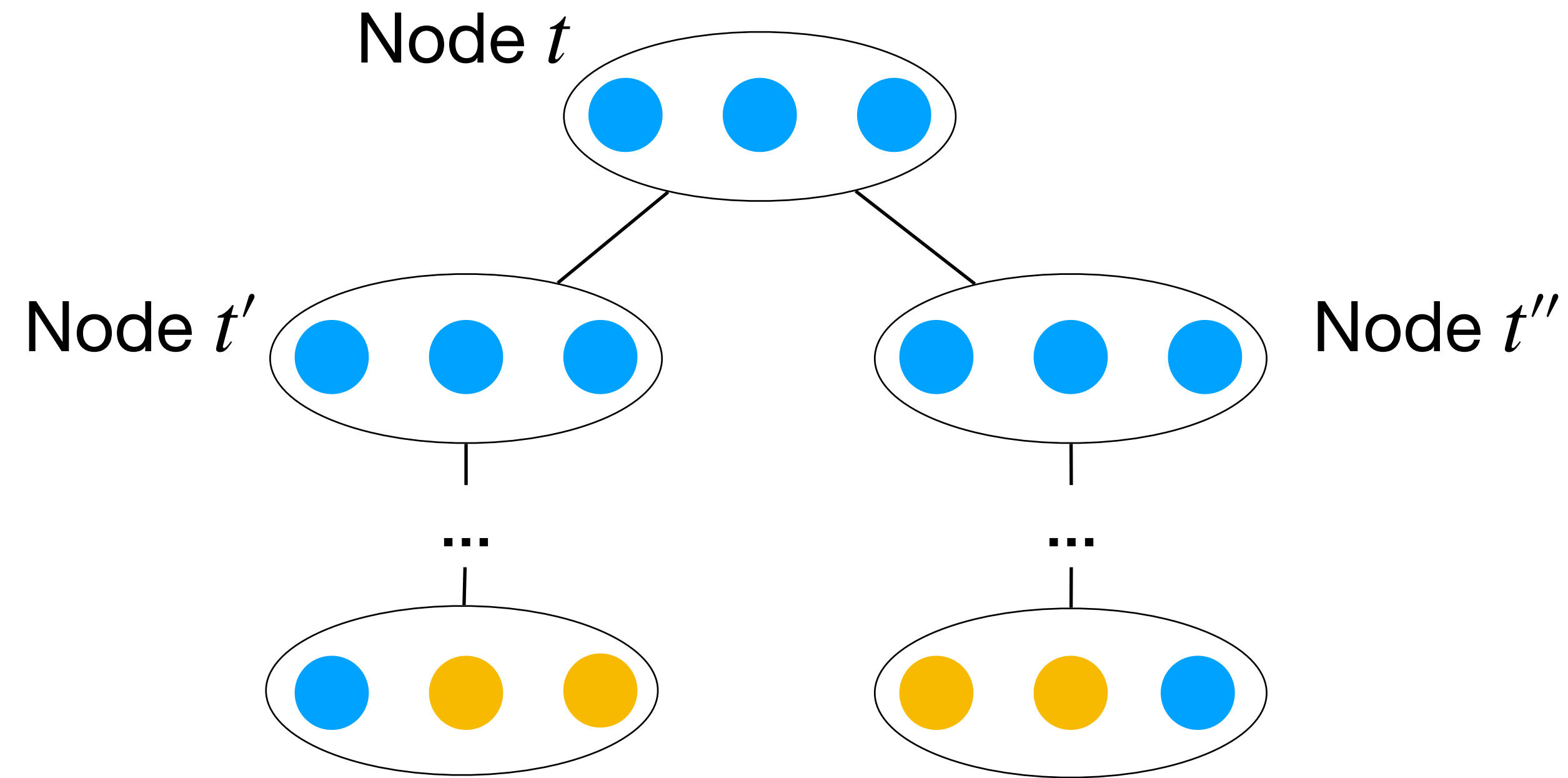
$$n_t(S) := \max(n_{t'}(S), n_{t'}(S \cup \{v\}))$$

# Join Nodes

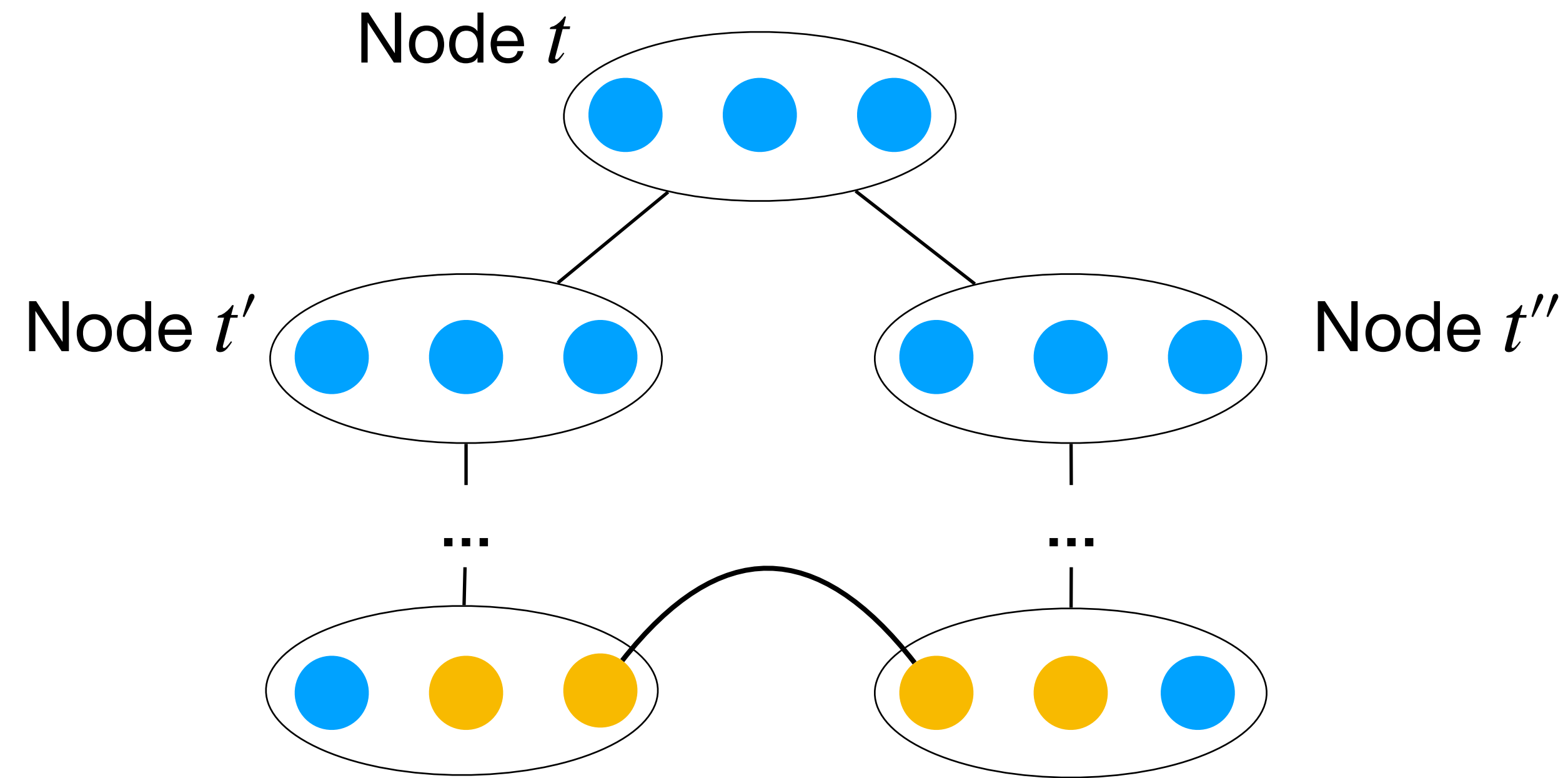




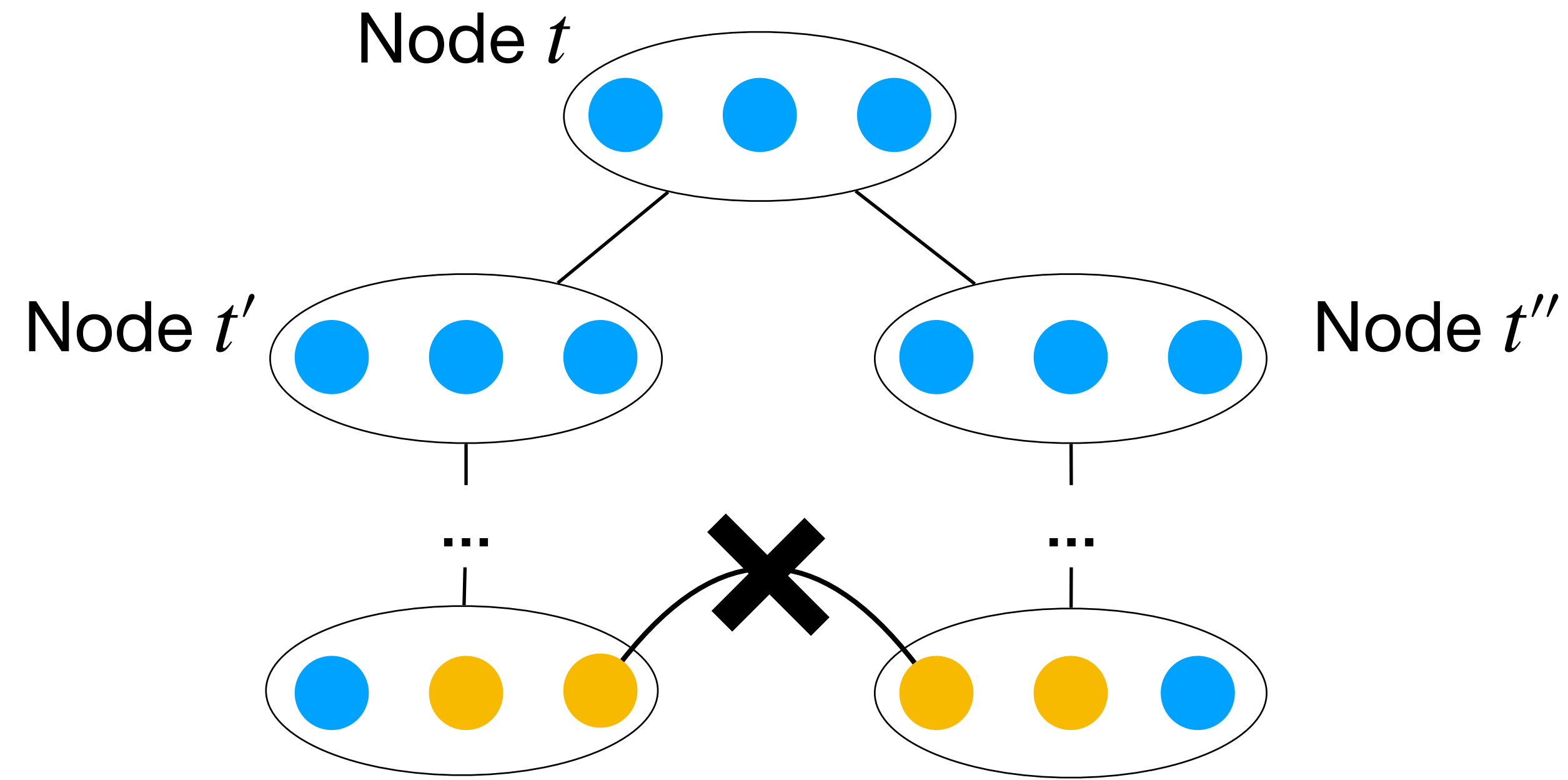
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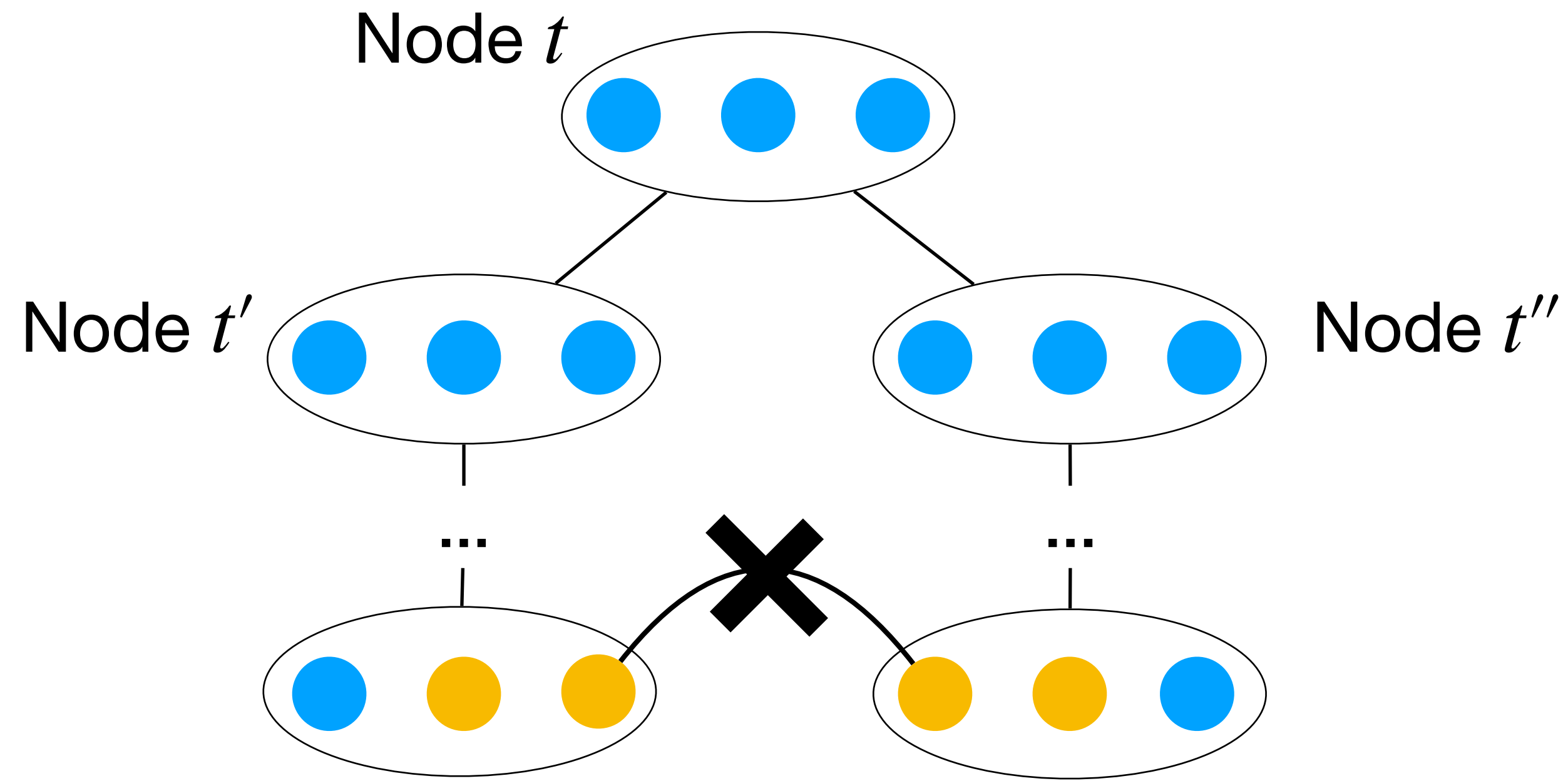
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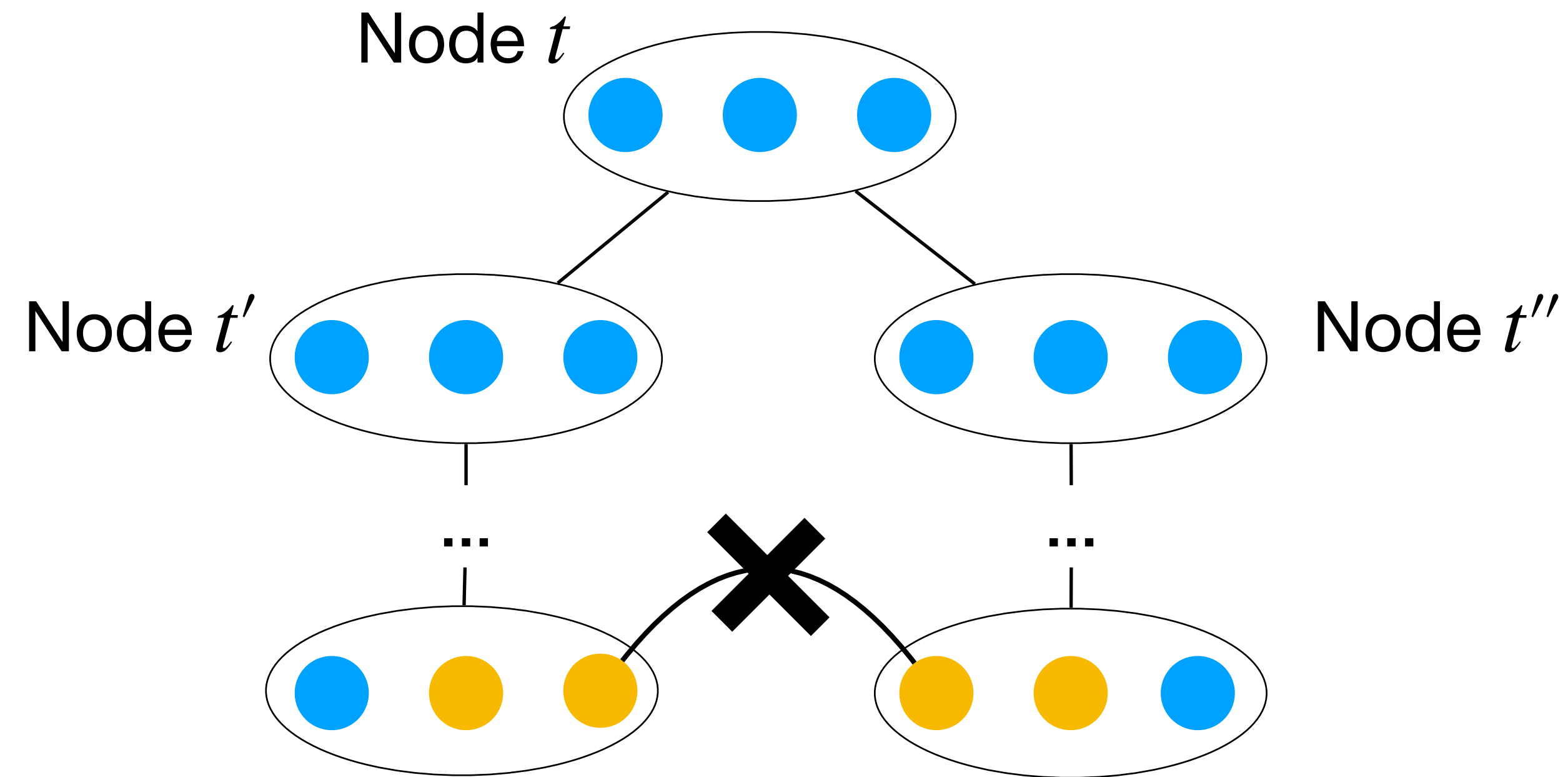


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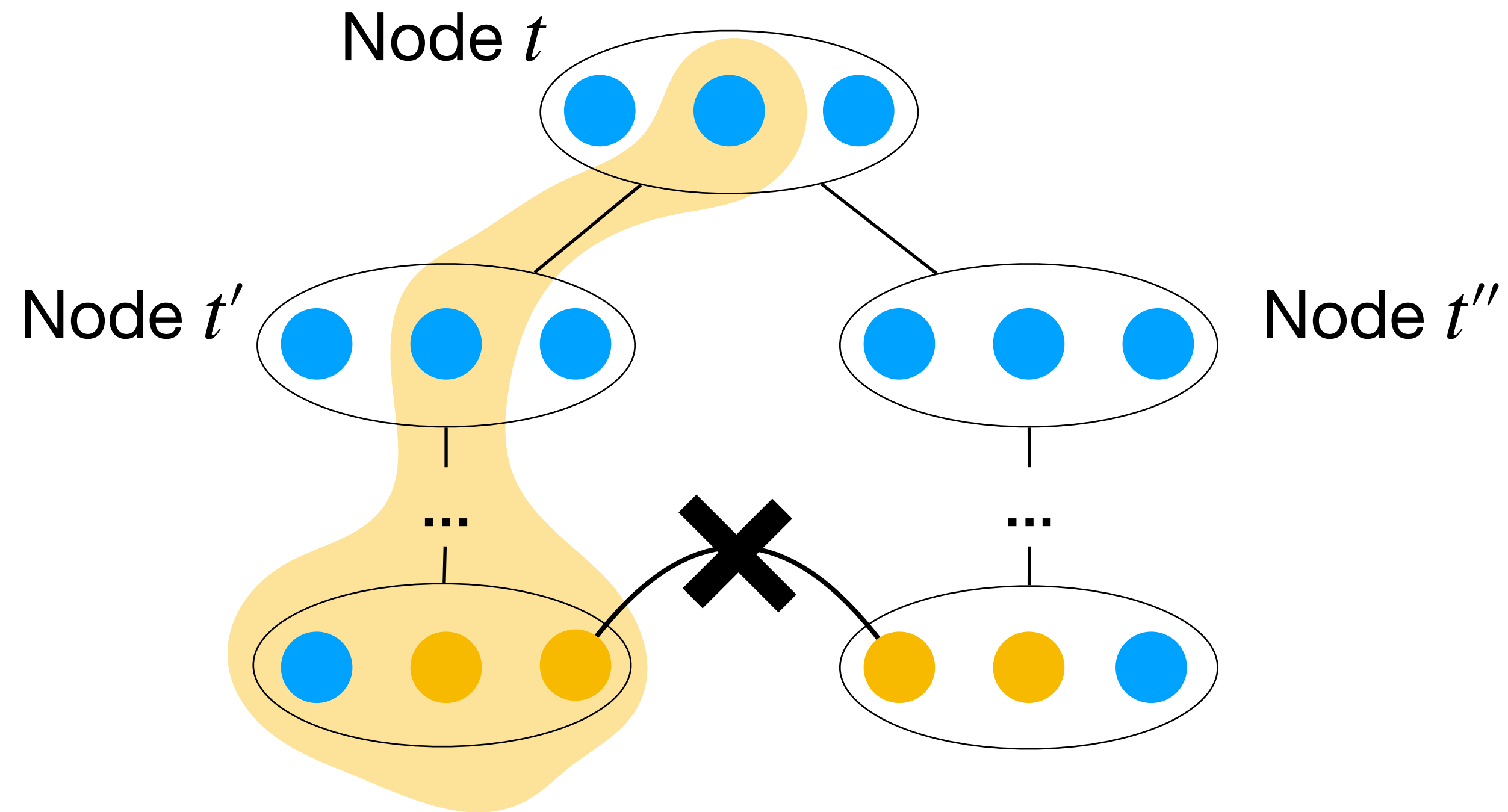
# Join Nodes



for  $S \subseteq \chi(t)$

$$n_t(S) := n_{t'}(S) + n_{t''}(S)$$

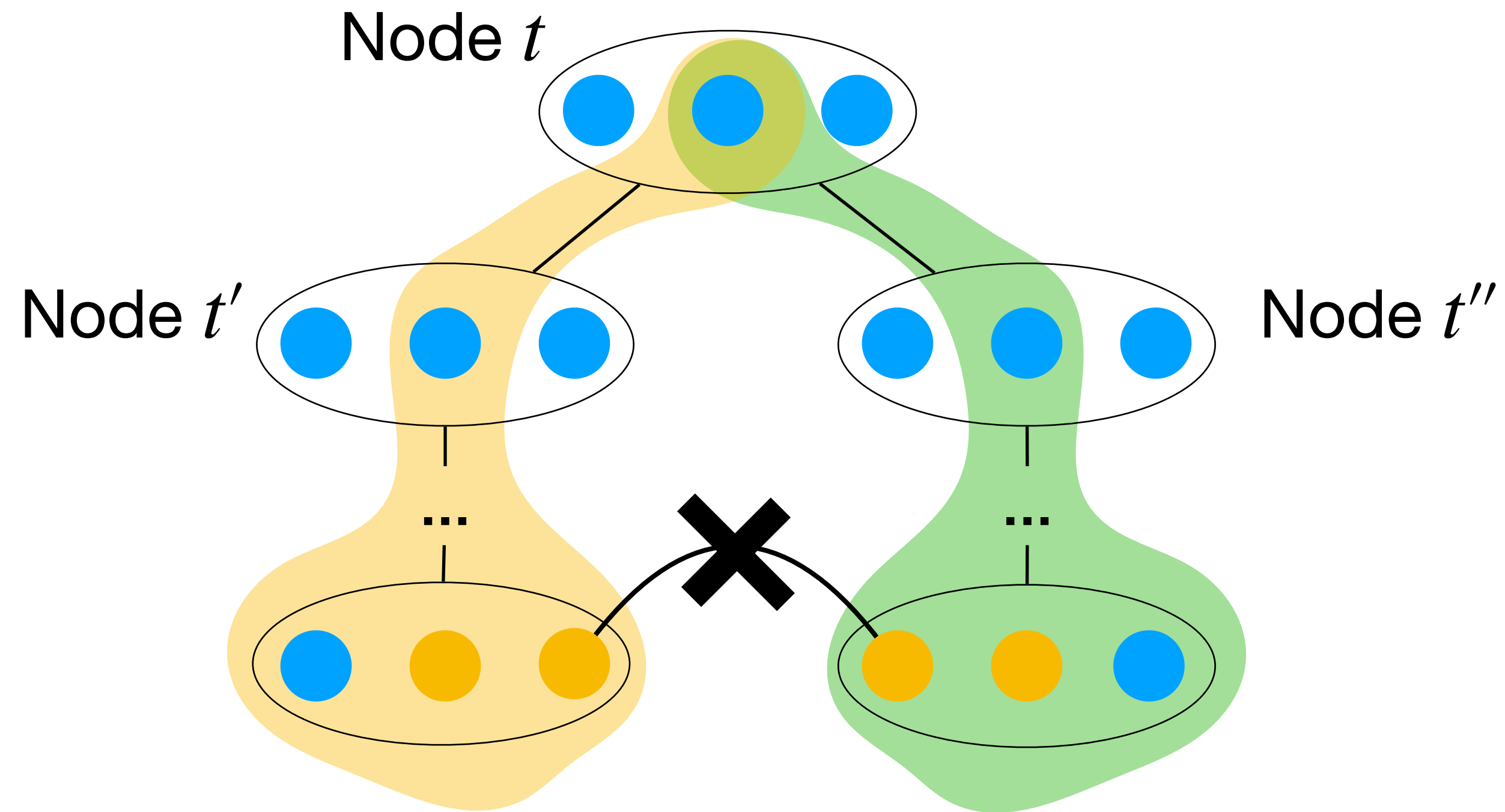
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for  $S \subseteq \chi(t)$

$$n_t(S) := n_{t'}(S) + n_{t''}(S)$$

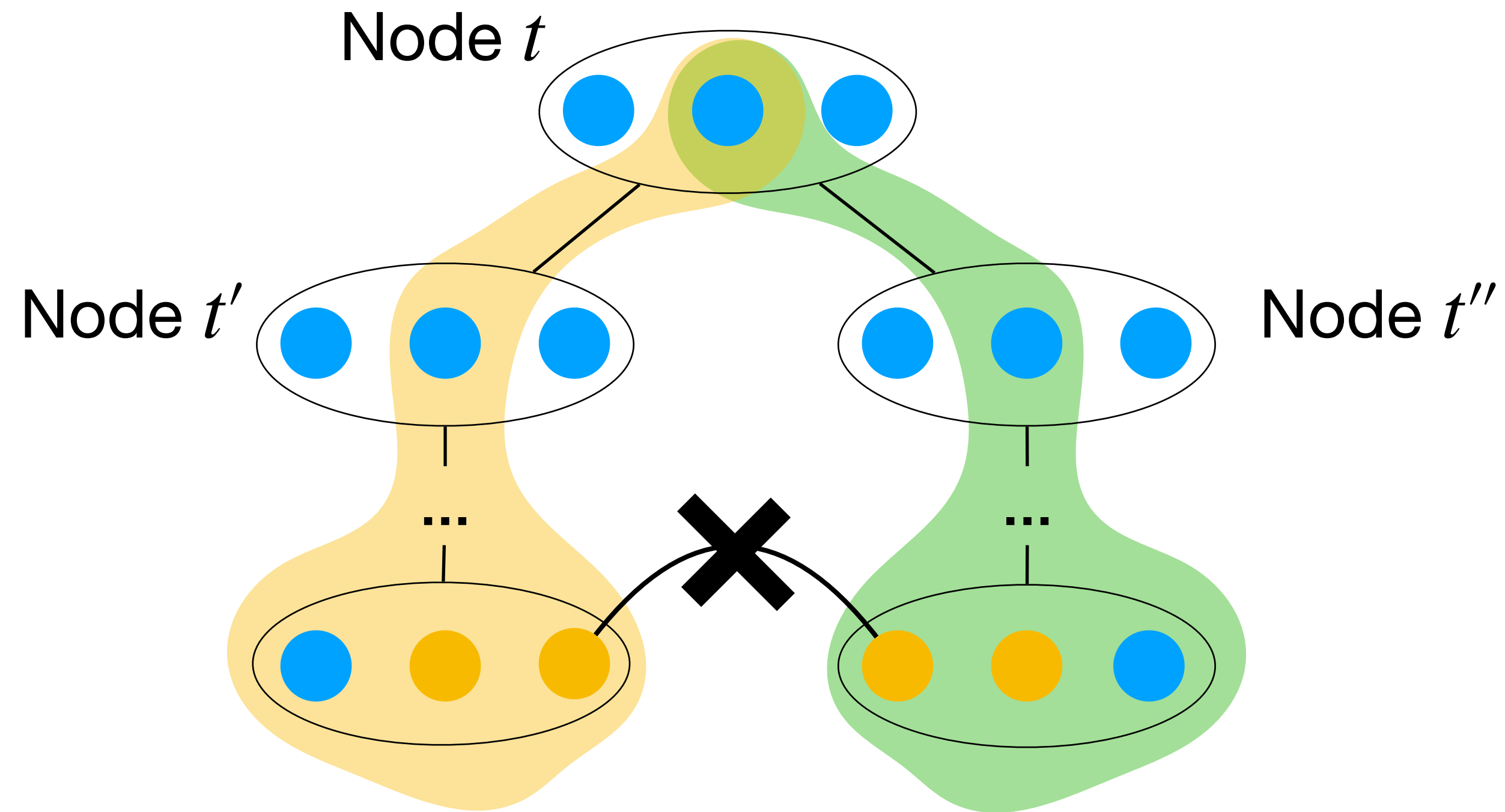
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for  $S \subseteq \chi(t)$

$$n_t(S) := n_{t'}(S) + n_{t''}(S)$$

# Join Nodes



for  $S \subseteq \chi(t)$

$$n_t(S) := n_{t'}(S) + n_{t''}(S) - |S|$$



# The Algorithm

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1. Compute a nice tree decomposition  $(T, \chi)$  of  $G$

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for each  $U \subseteq \chi(t)$

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for each  $U \subseteq \chi(t)$

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## Introduce Nodes

for each  $U \subseteq \chi(t)$

if  $v \notin U$   $n_t(U) := n_{t'}(U)$

if  $v \in U$   $n_t(U) := n_{t'}(U \setminus \{v\}) + 1$  if  $U$  is an IS  
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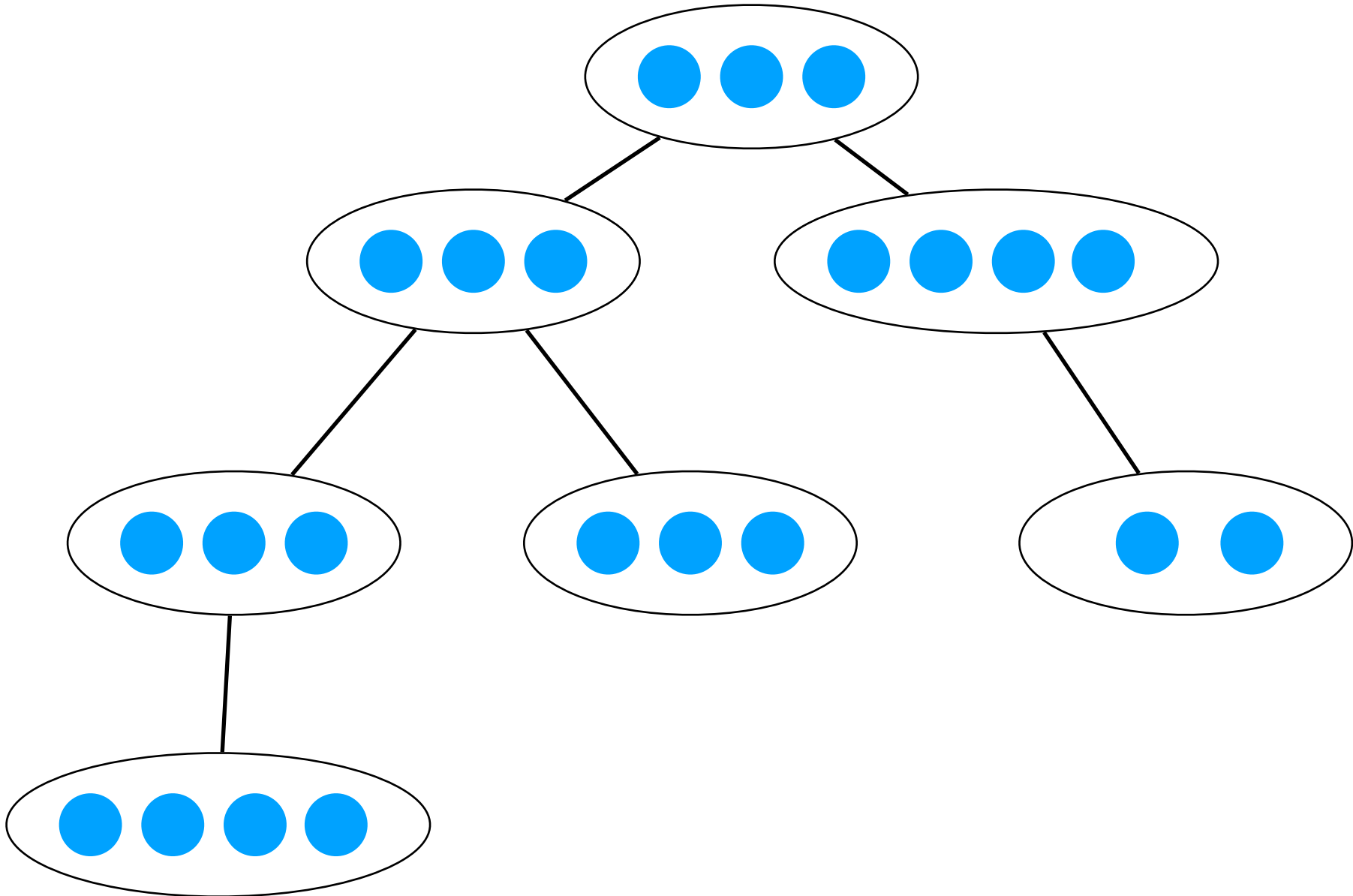
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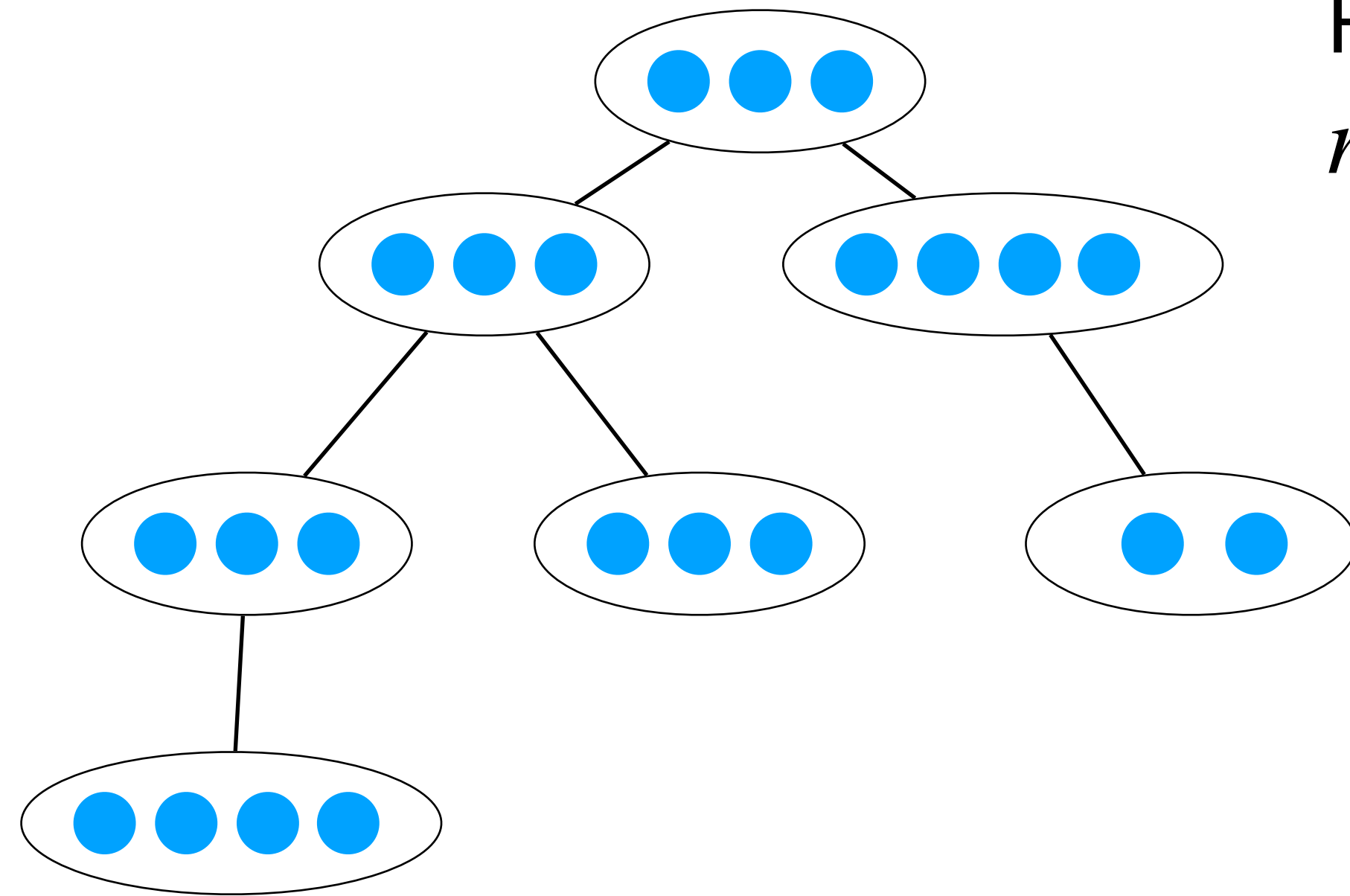
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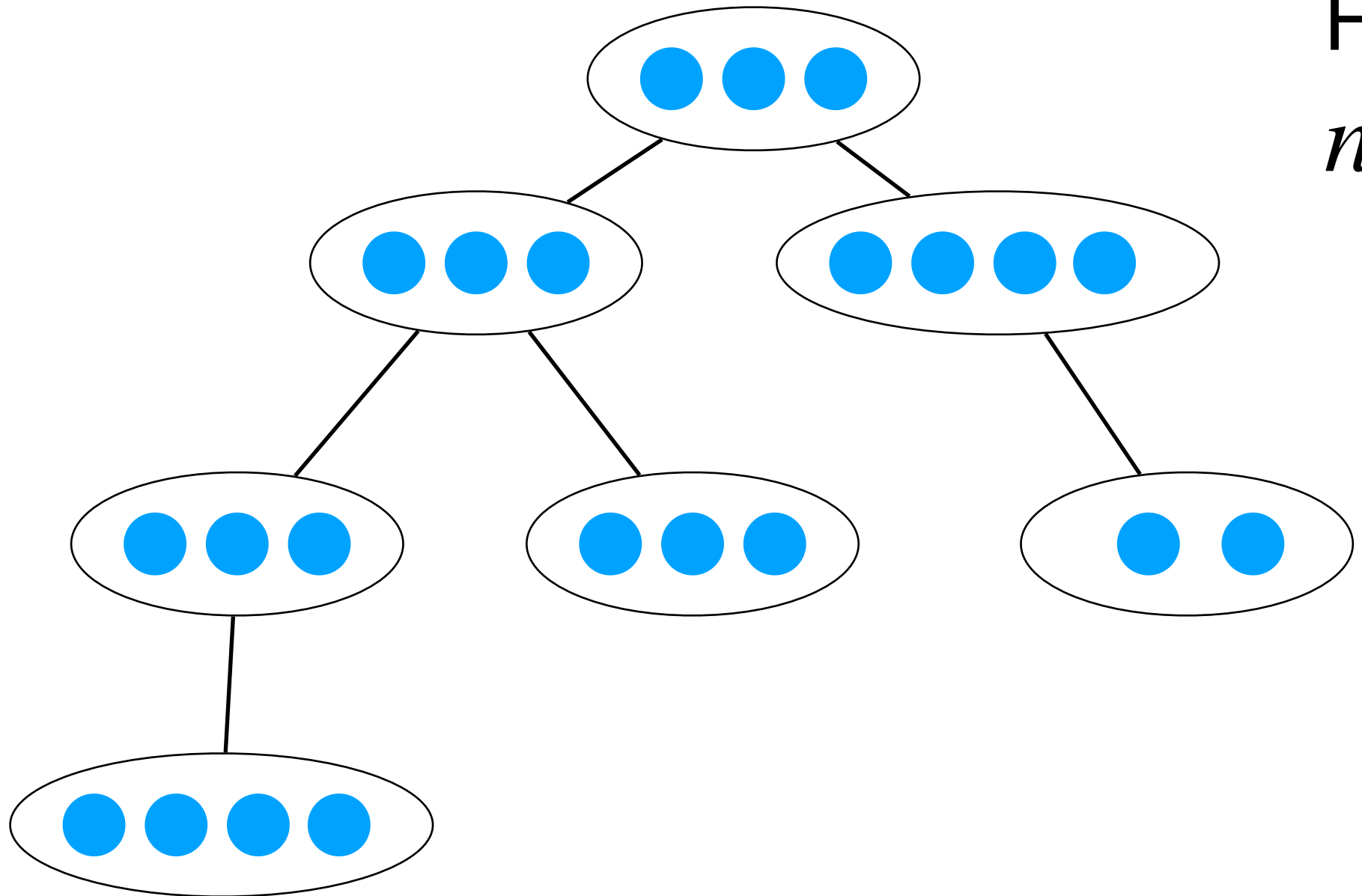
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For each subset  $S \subseteq \chi(t)$  of bag vertices,  
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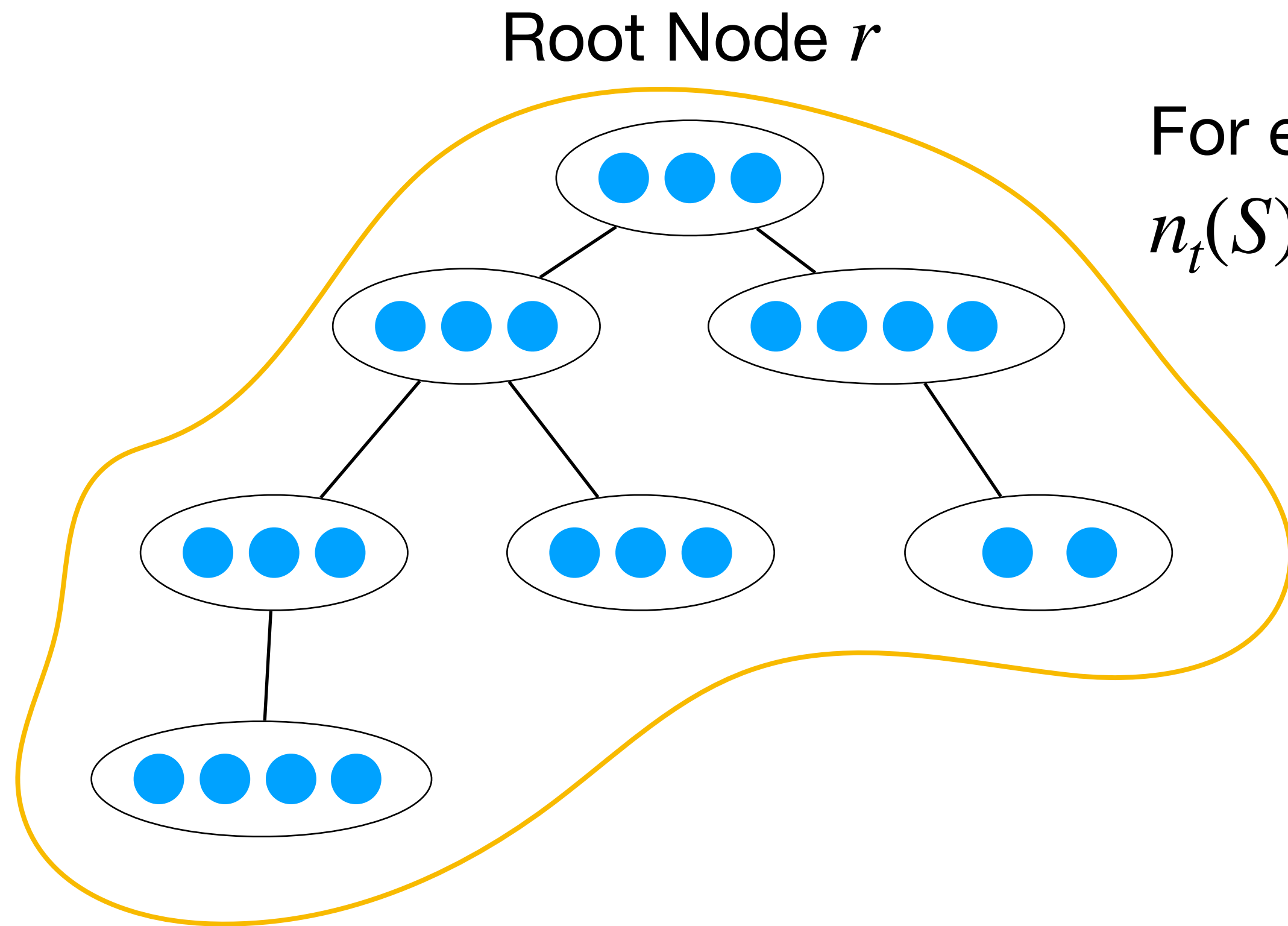
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Root Node  $r$



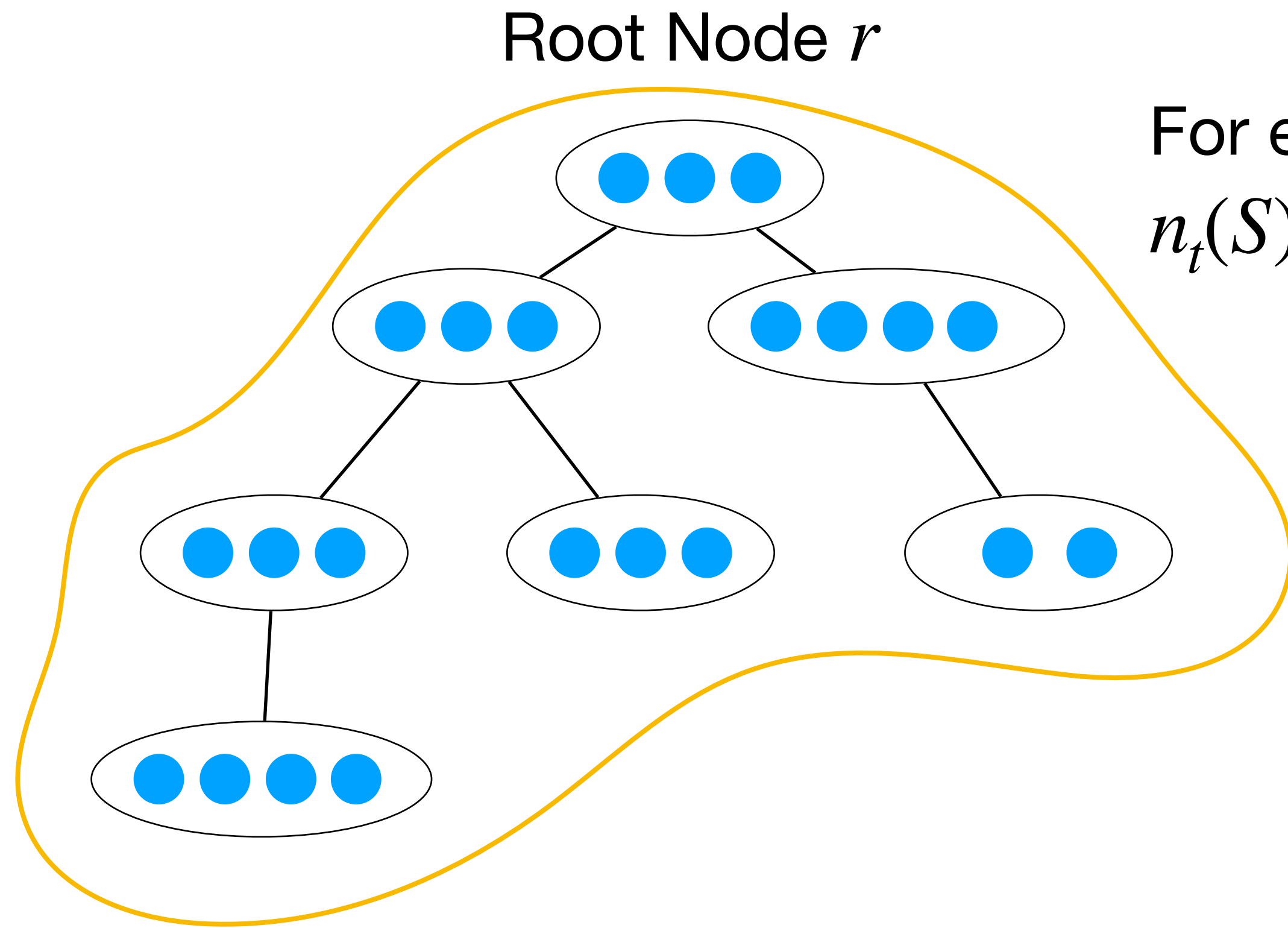
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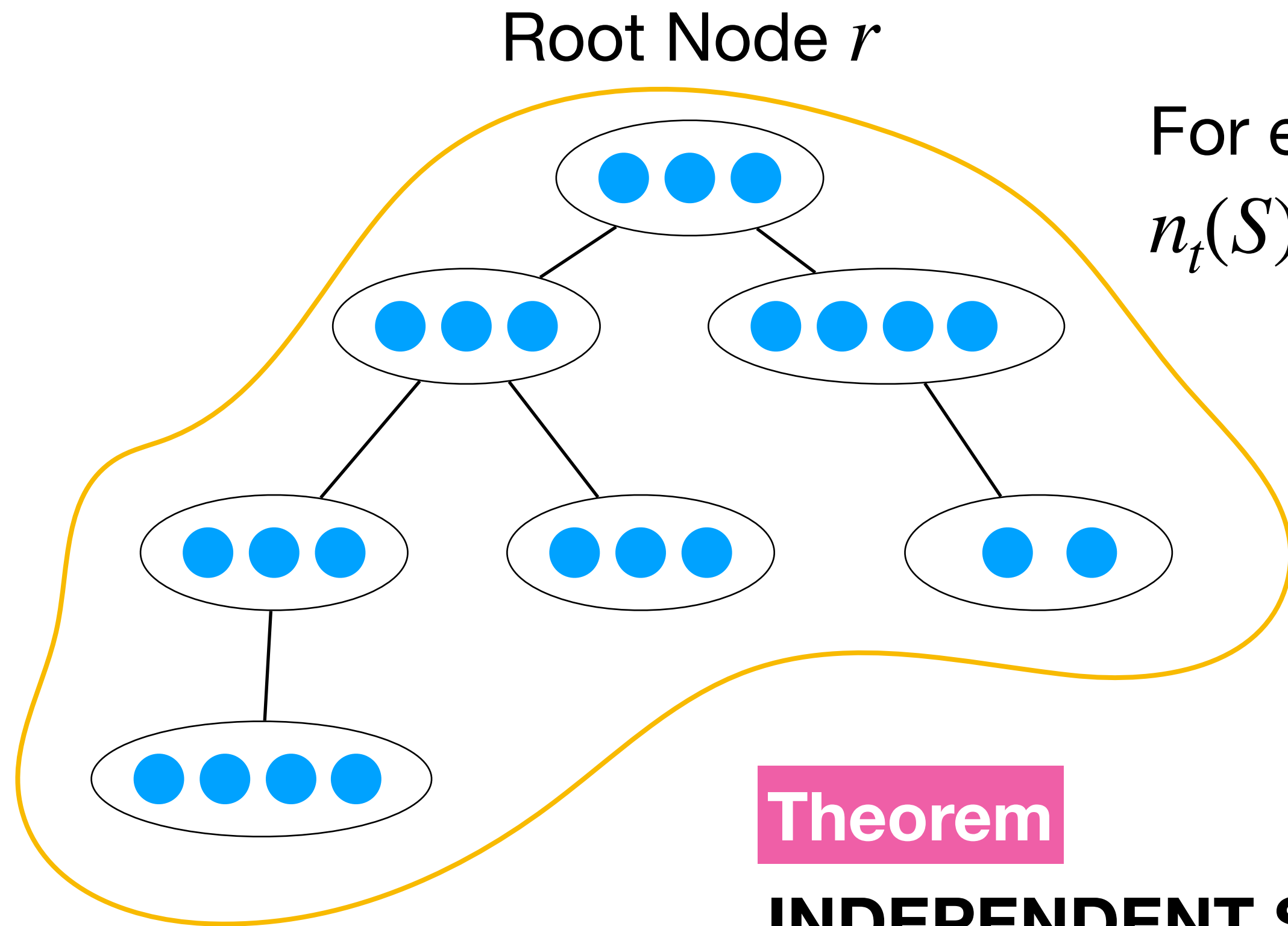


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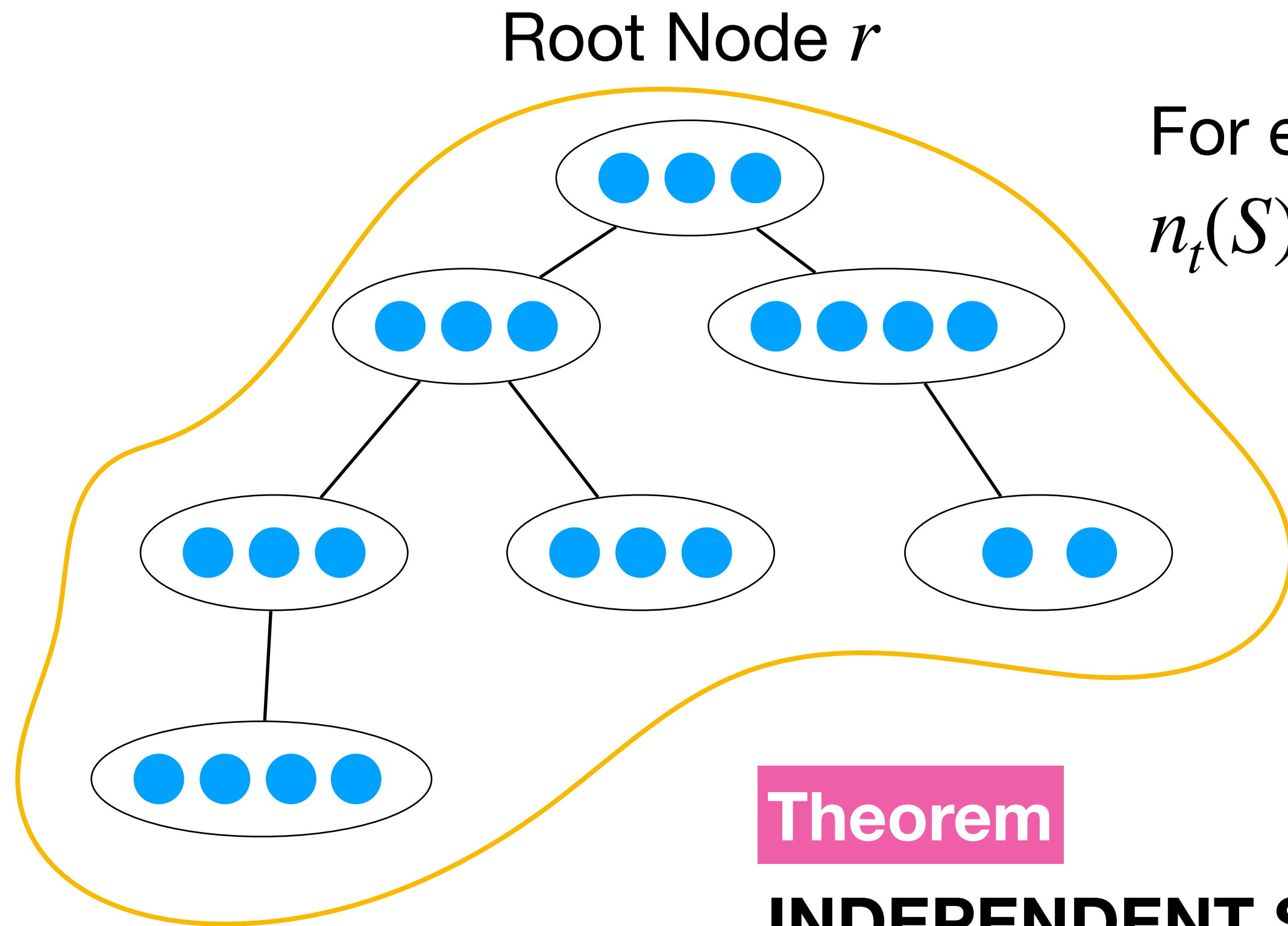
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**INDEPENDENT SET** is **FPT** parameterized by the treewidth of the input graph.\*

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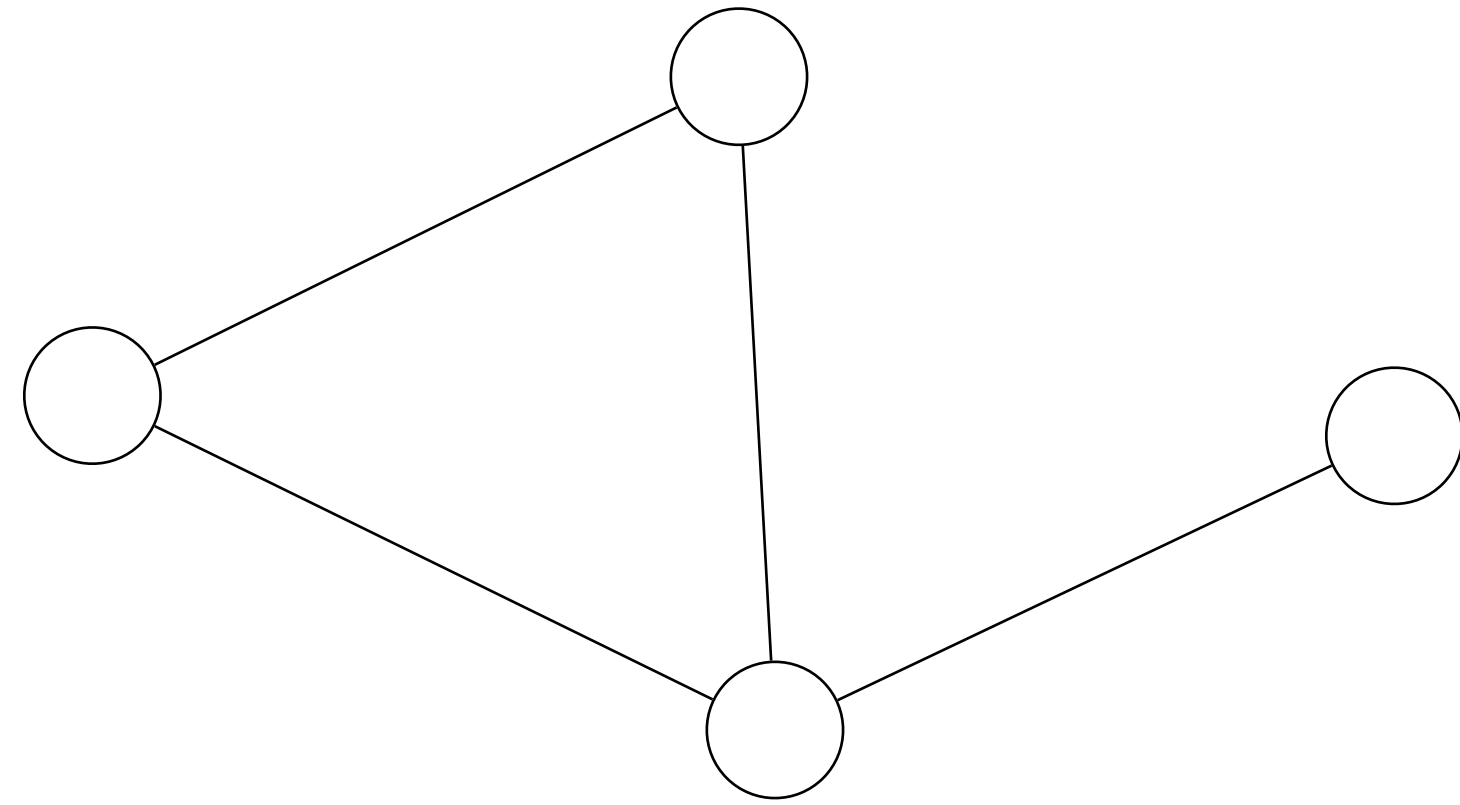
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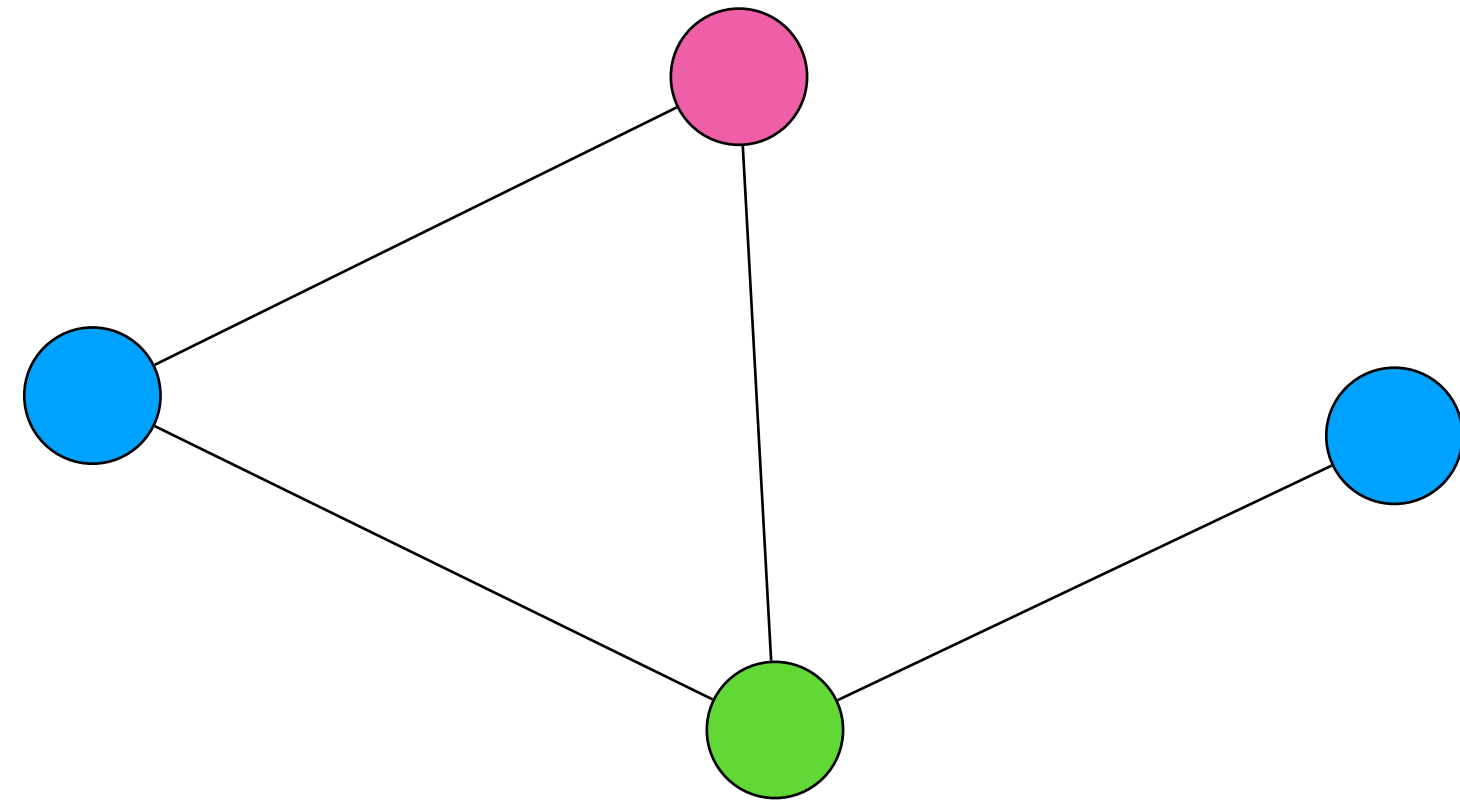
**INDEPENDENT SET** is **FPT** parameterized by the treewidth of the input graph.\*

\*If we can compute a nice tree decomposition of width  $g(k)$  in FPT time.

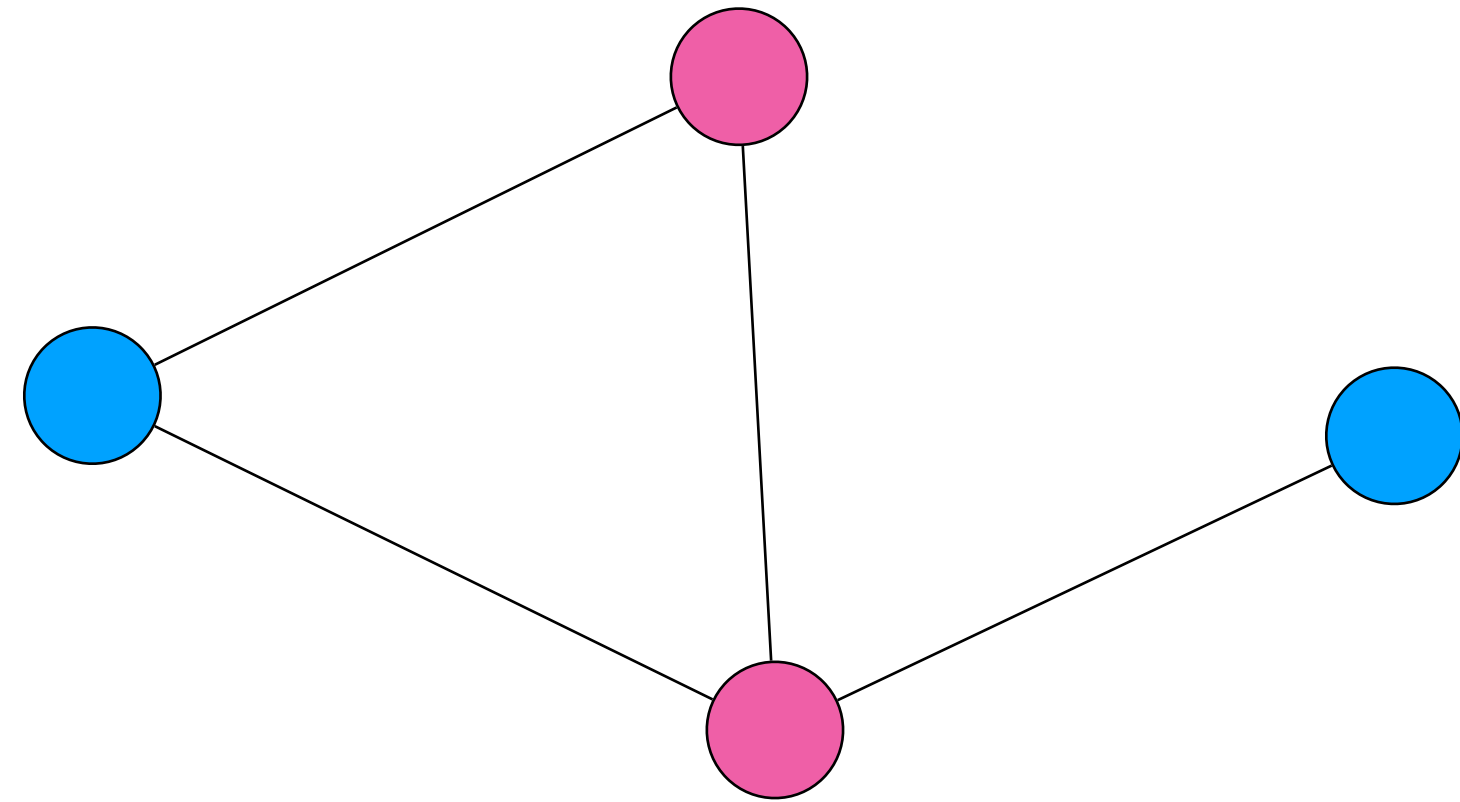
# 3-Colorability



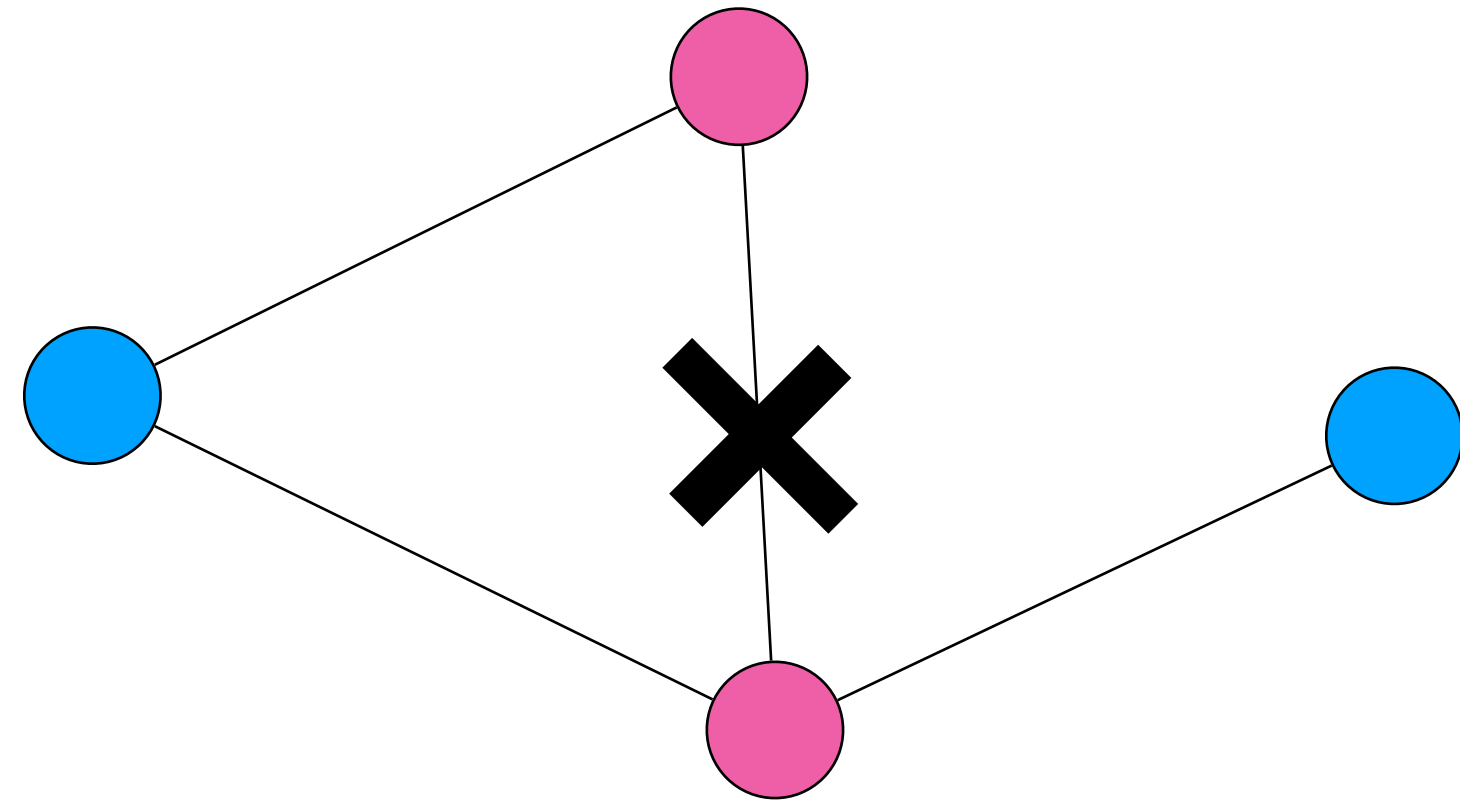
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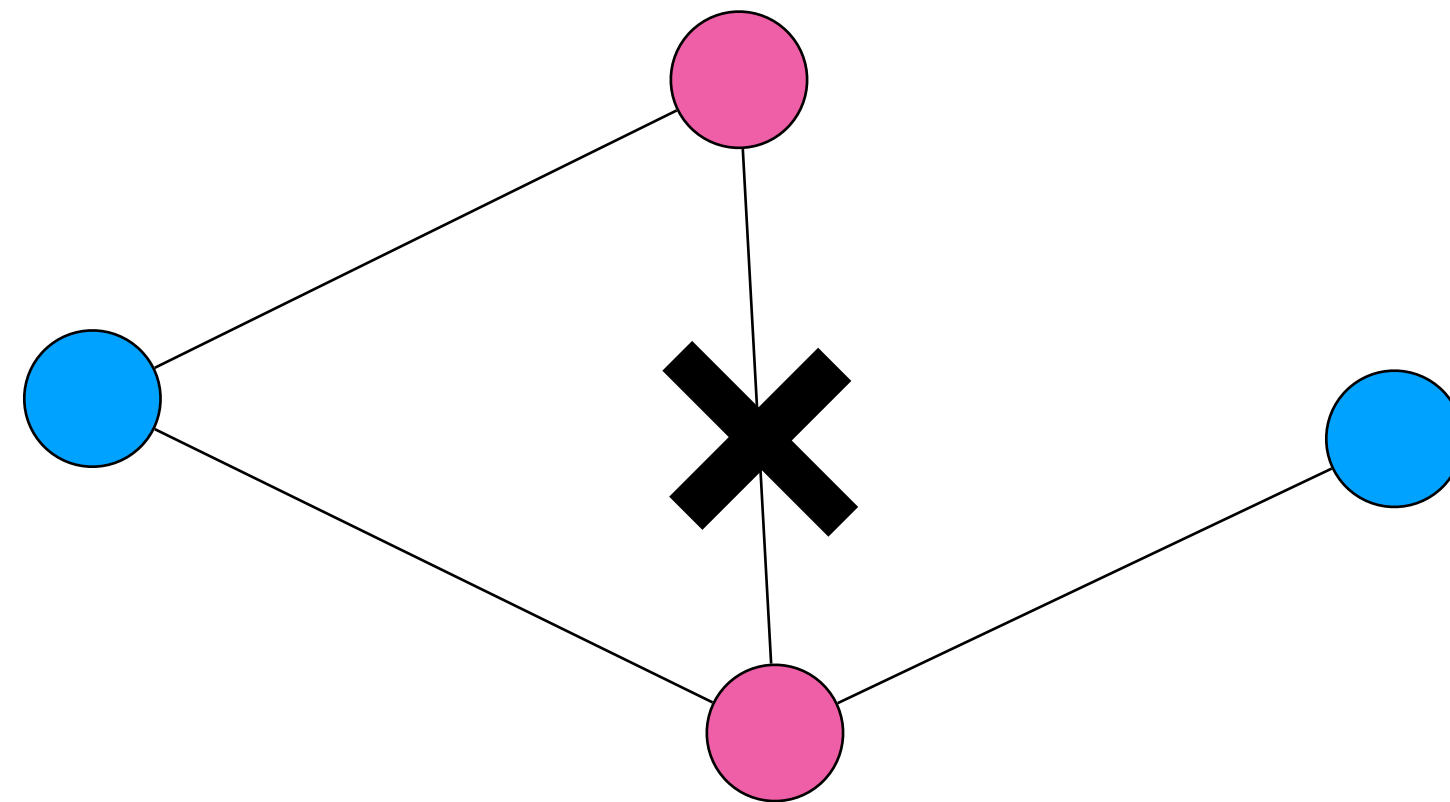
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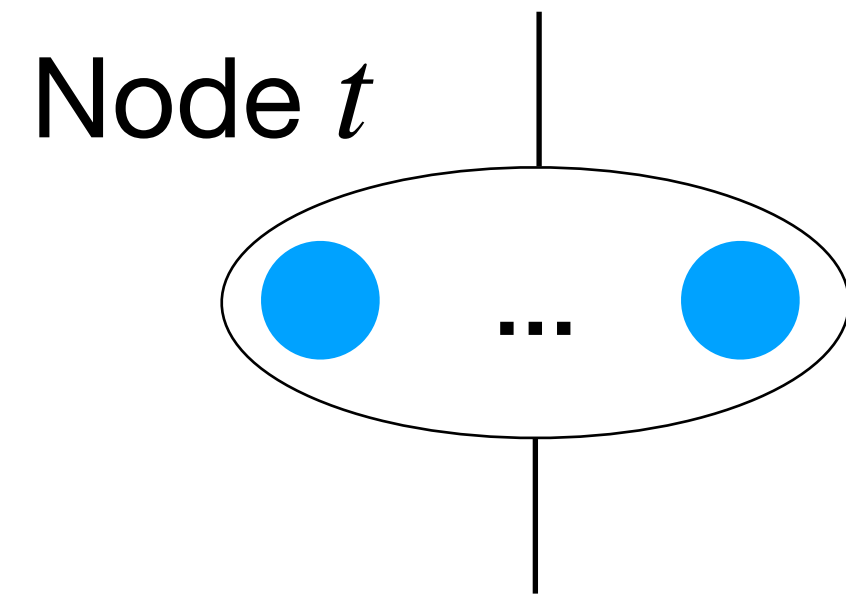


## 3-COLORABILITY

**Input:** A graph  $G$ .

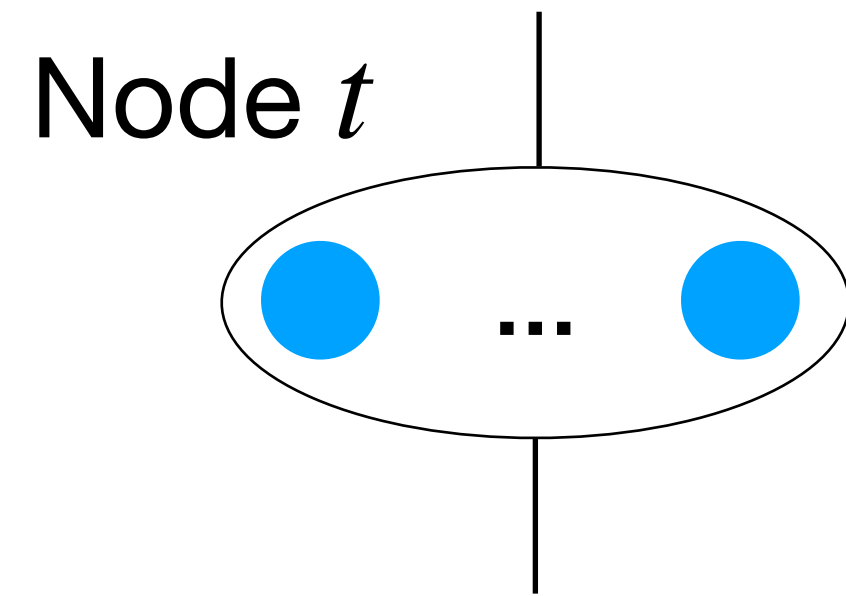
**Question:** Does  $G$  have a vertex-coloring with 3 colors?

# 3-Colorability on Tree Decompositions



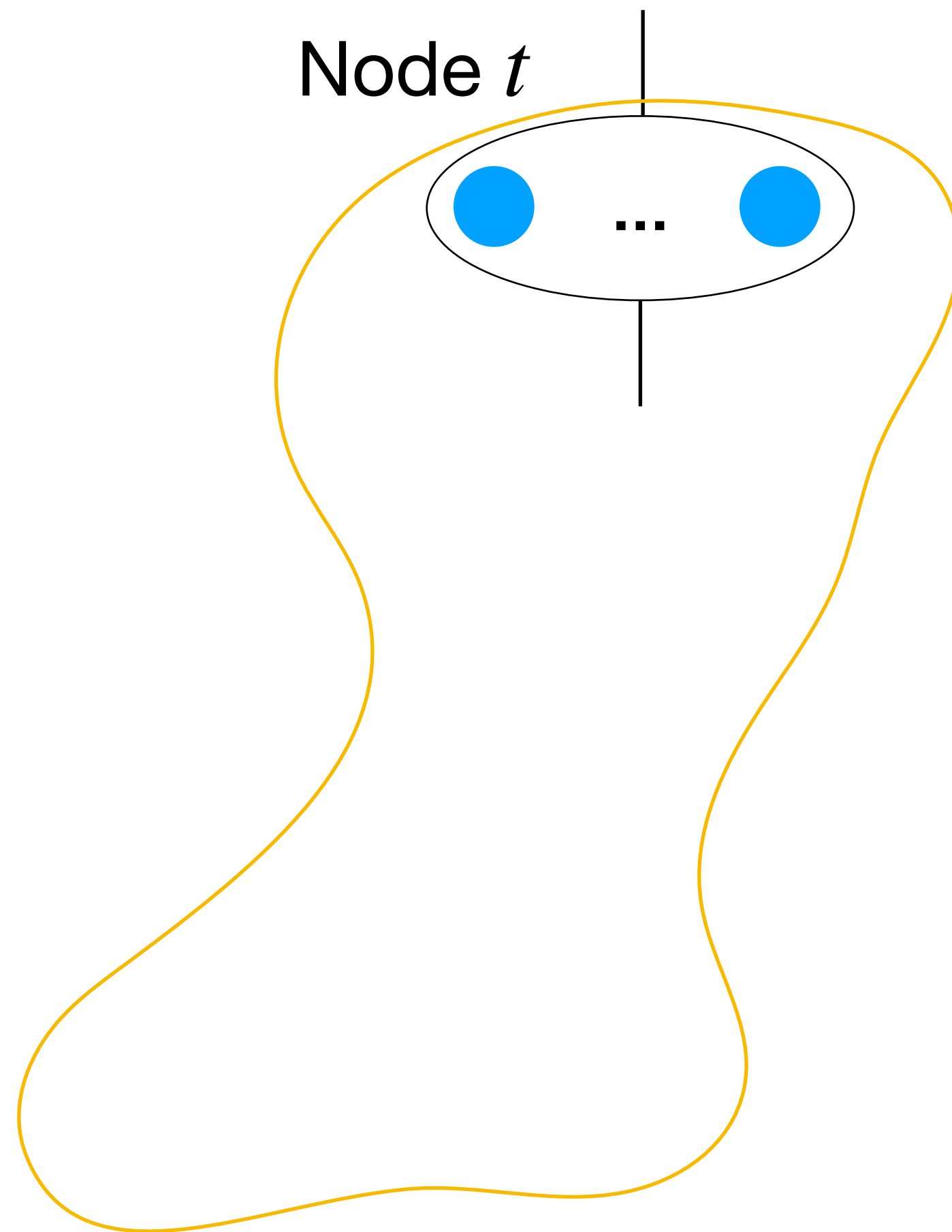


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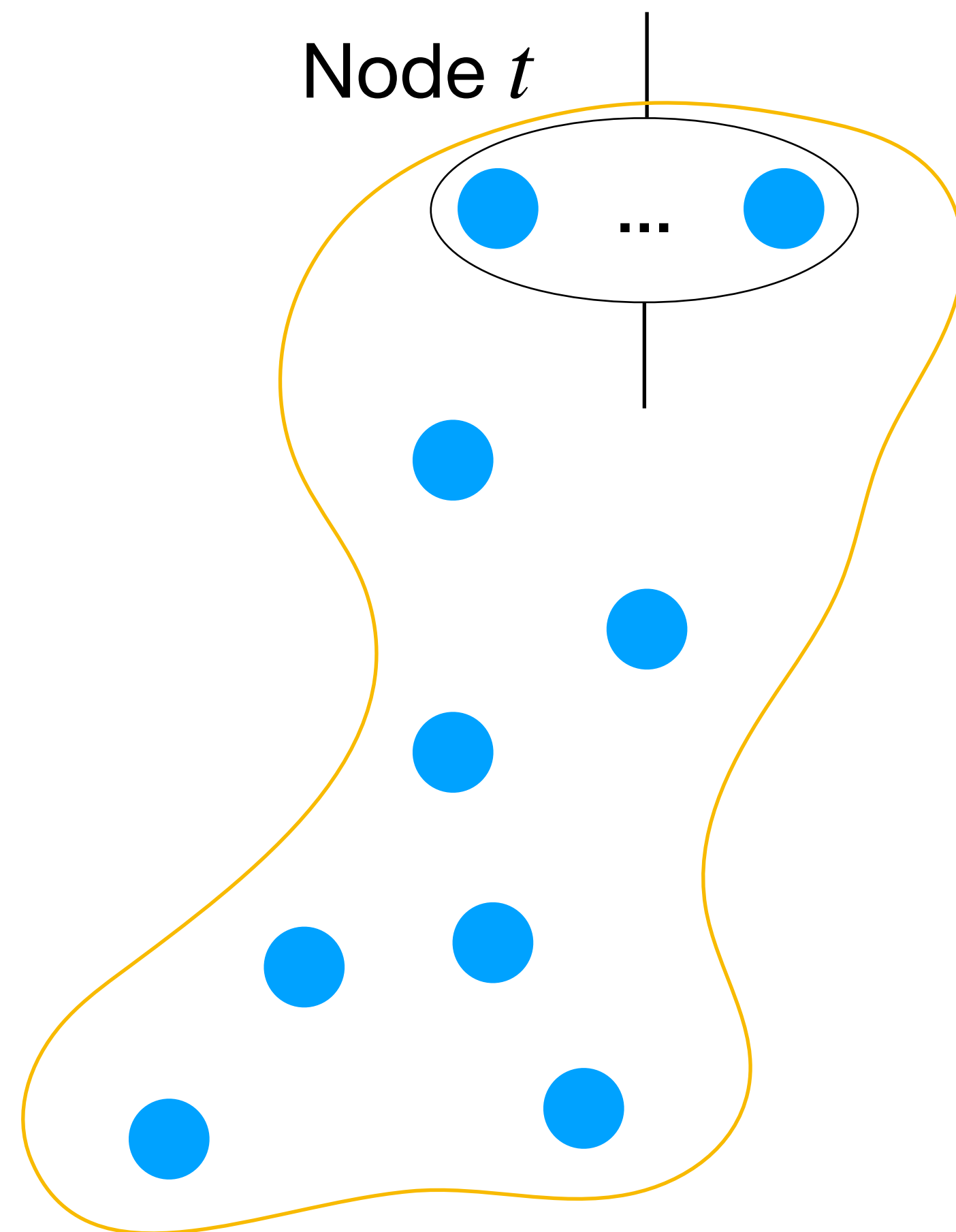
for  $\sigma : \chi(t) \rightarrow \{red, green, blue\}$

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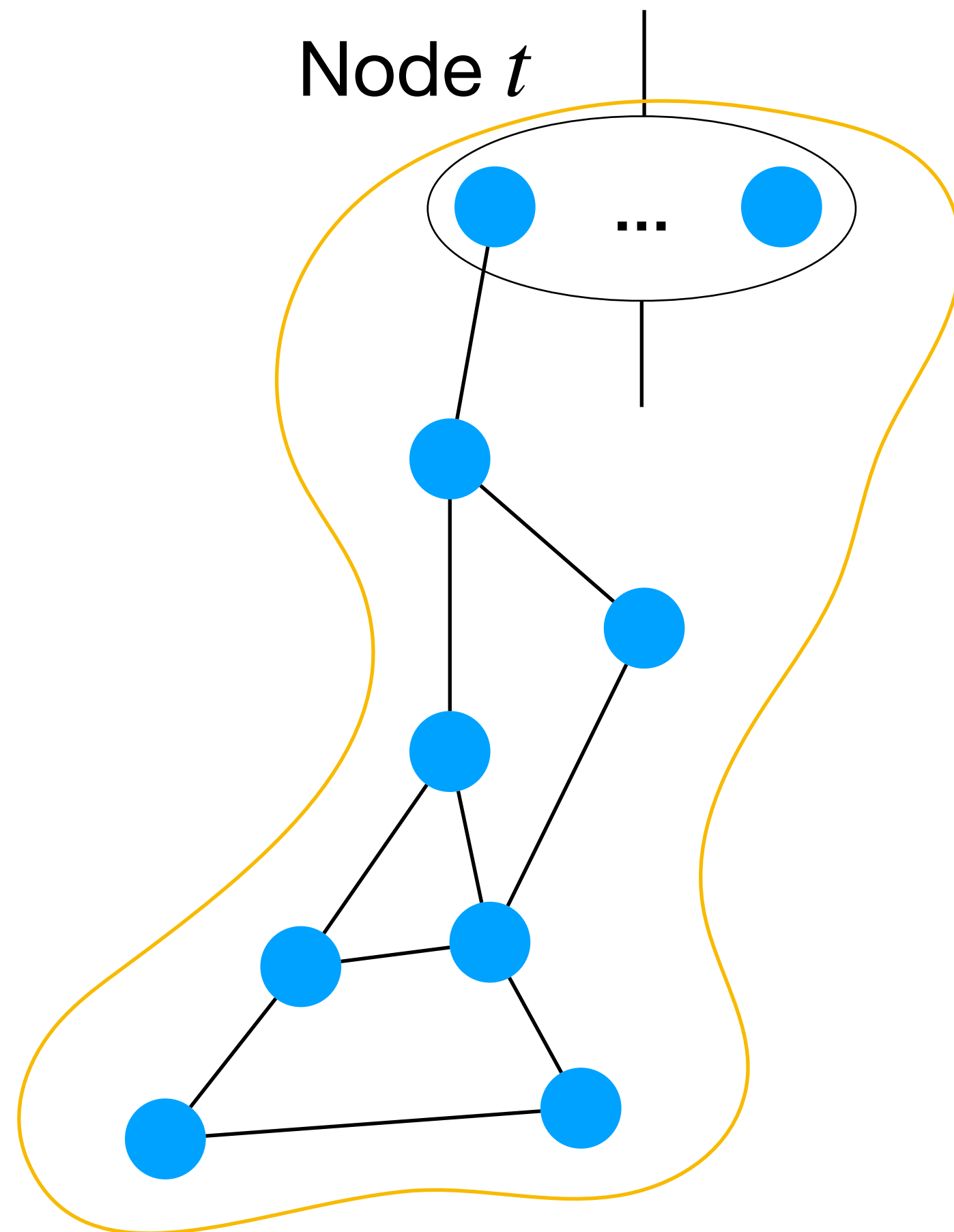
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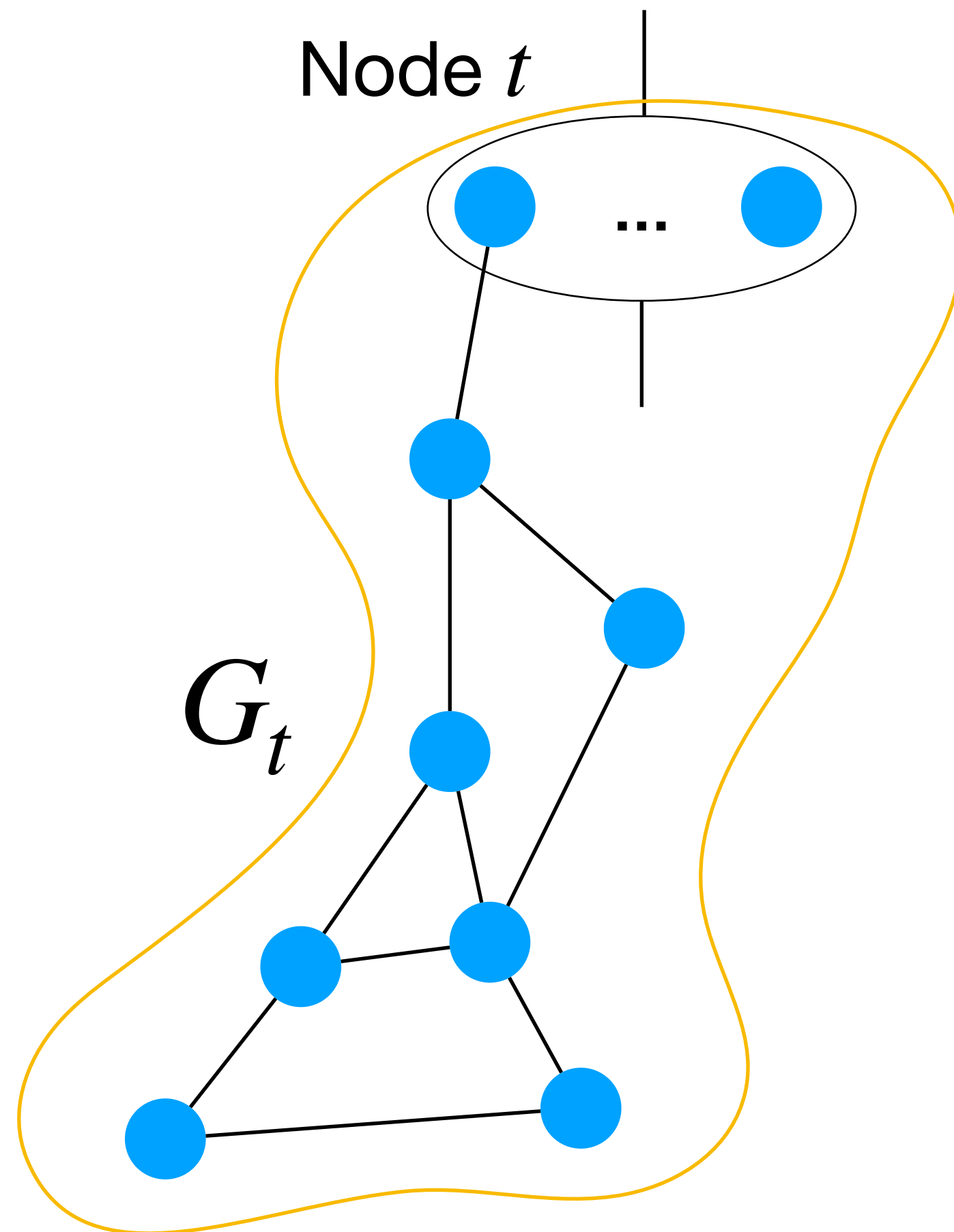
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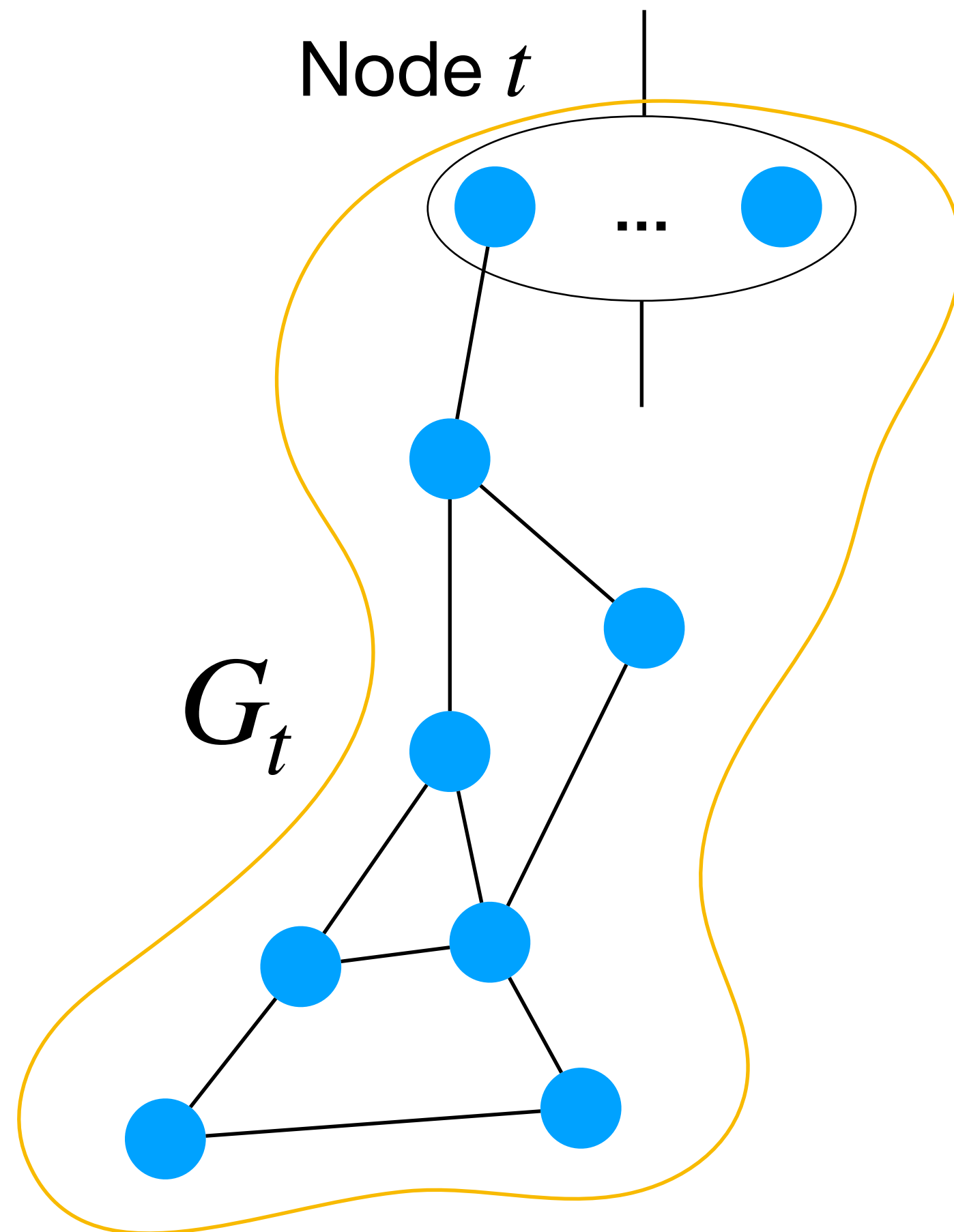
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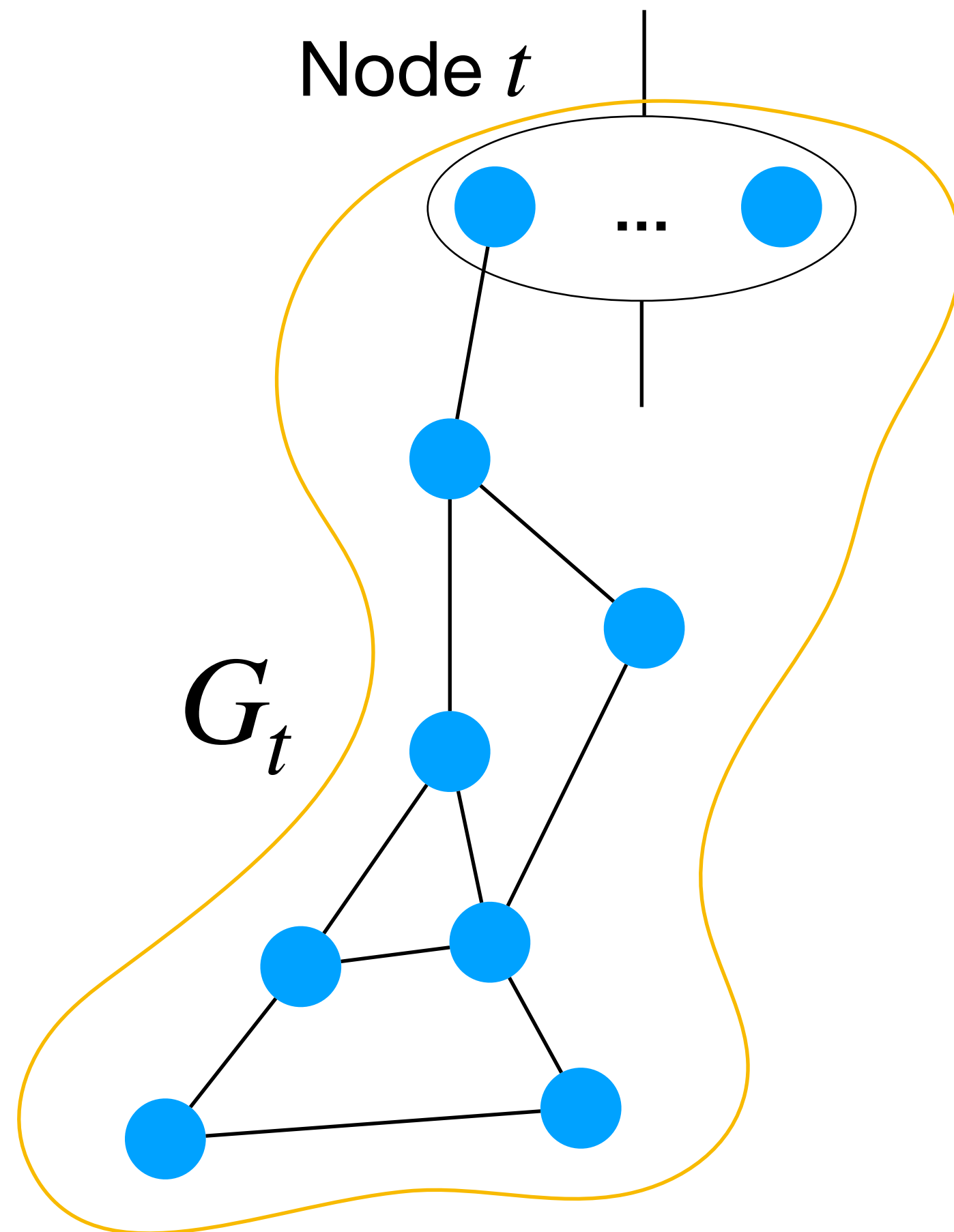
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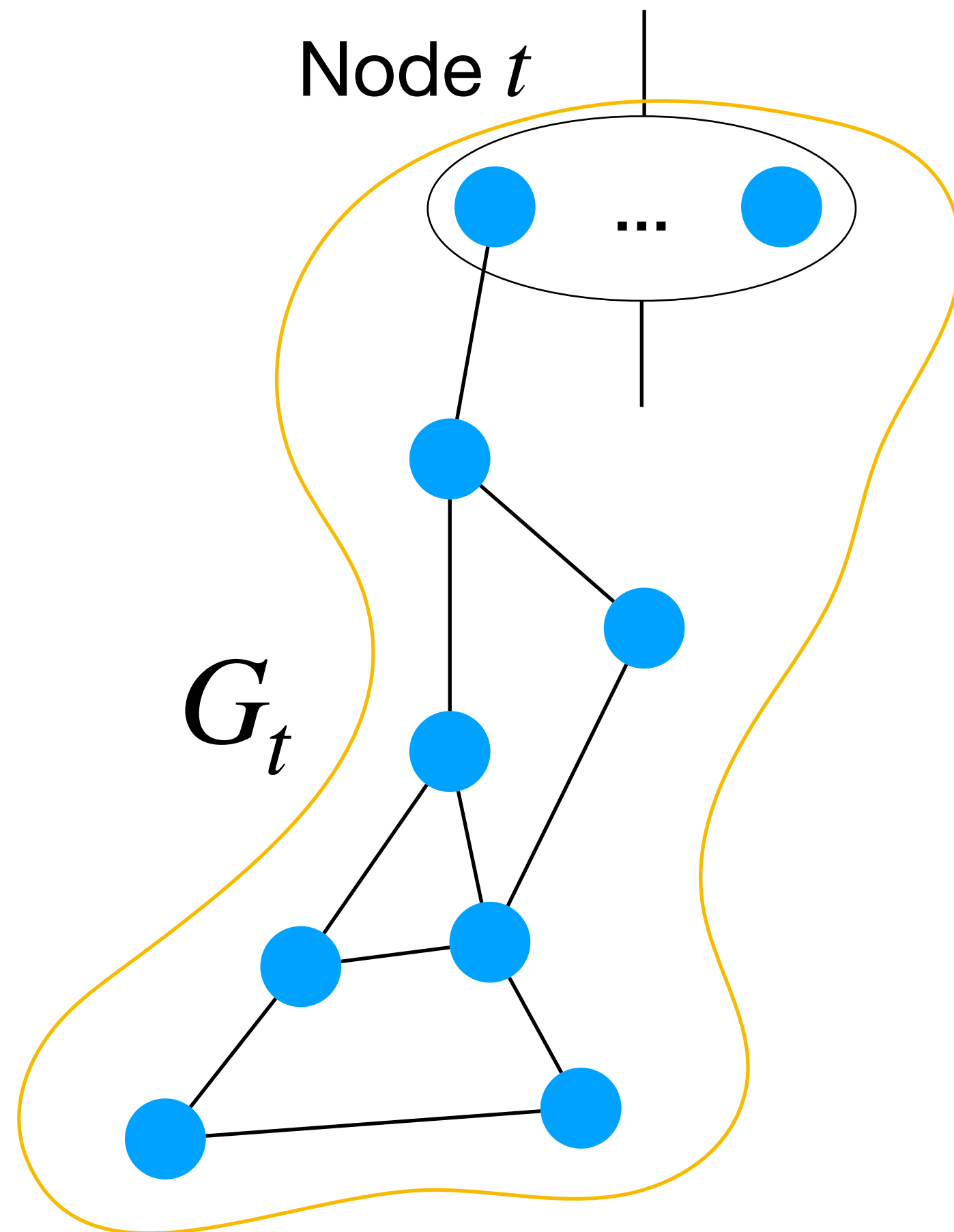
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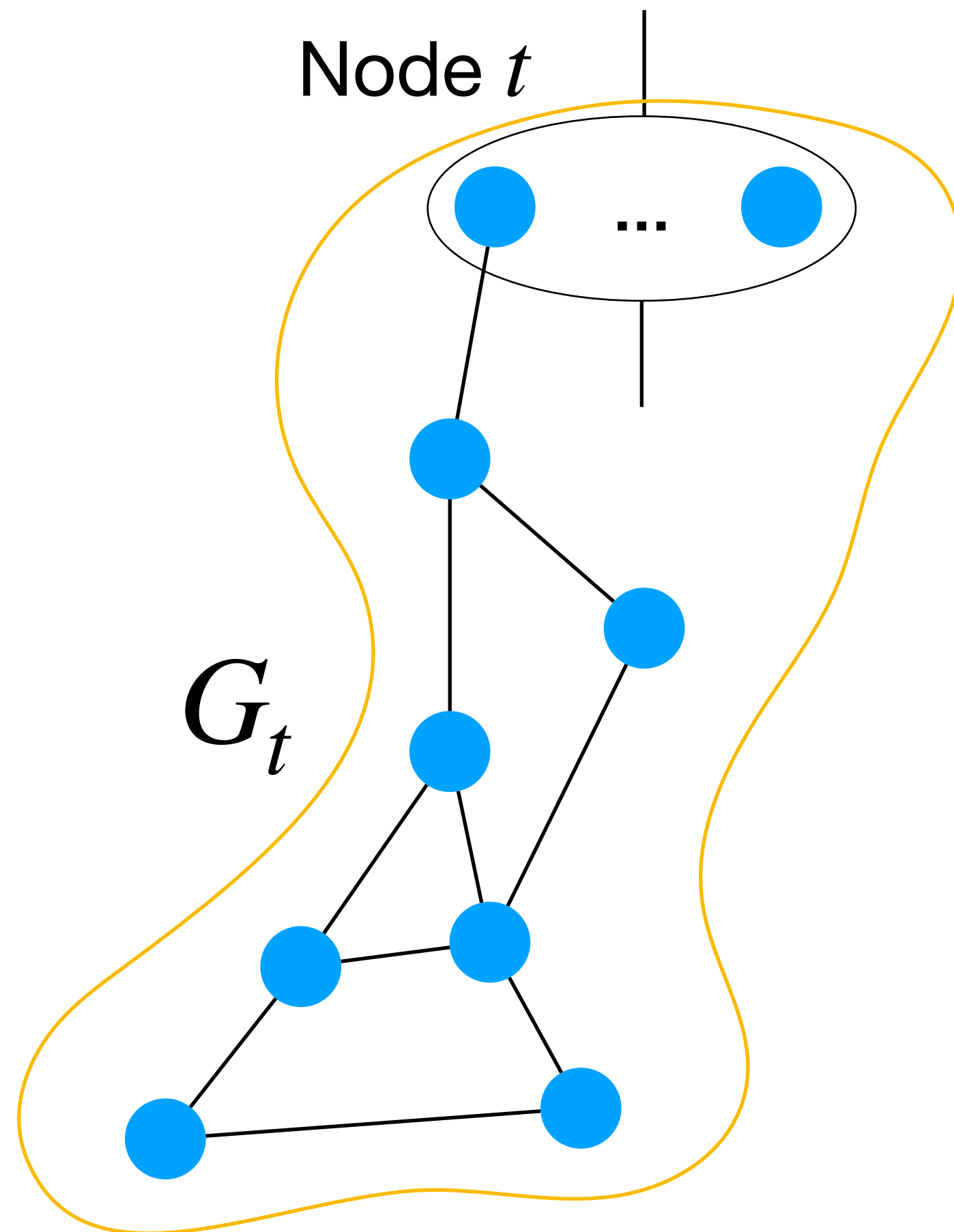


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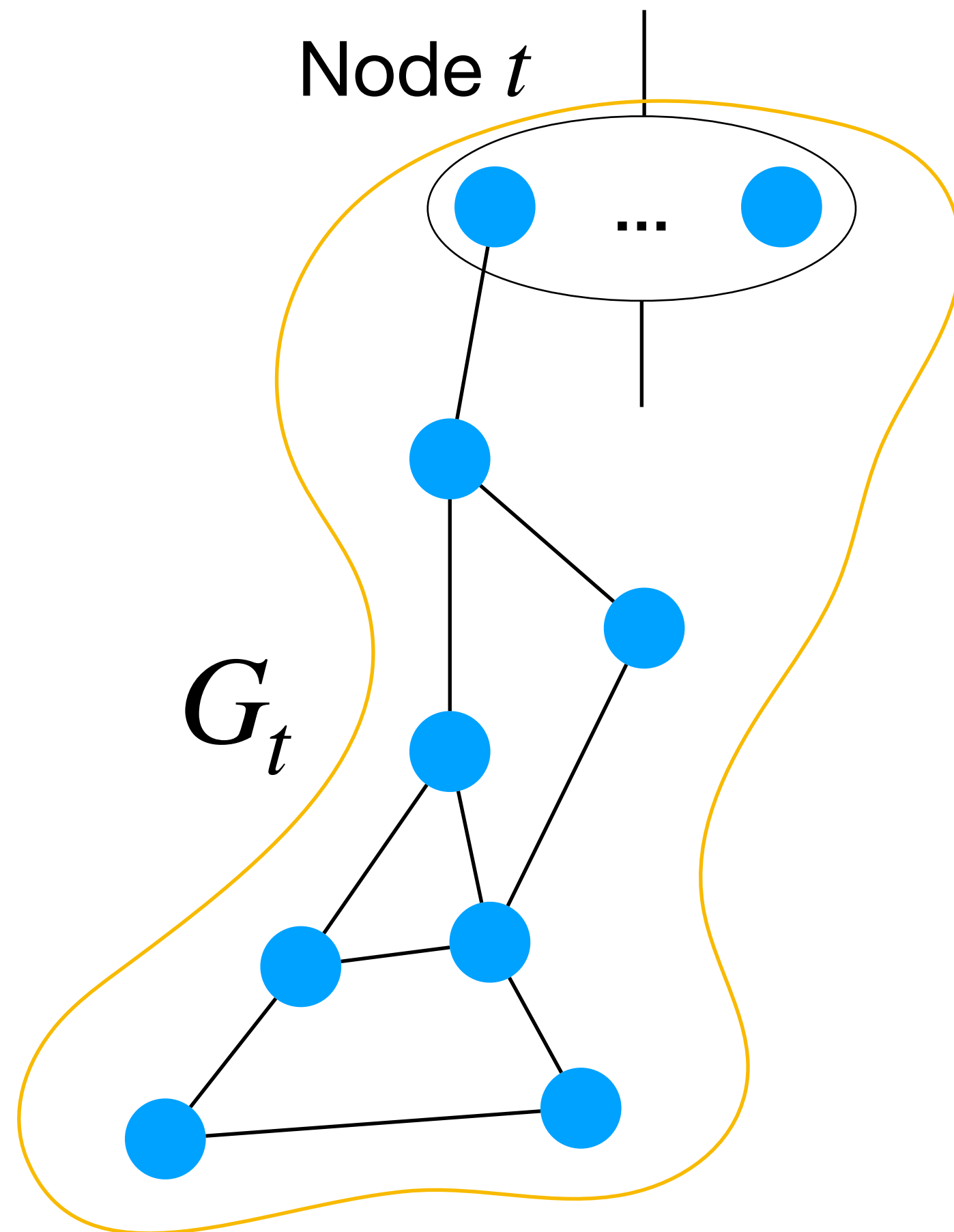


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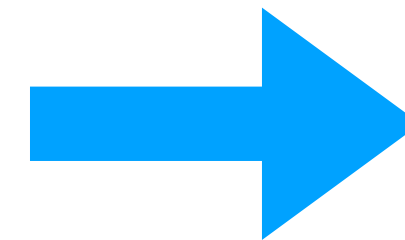
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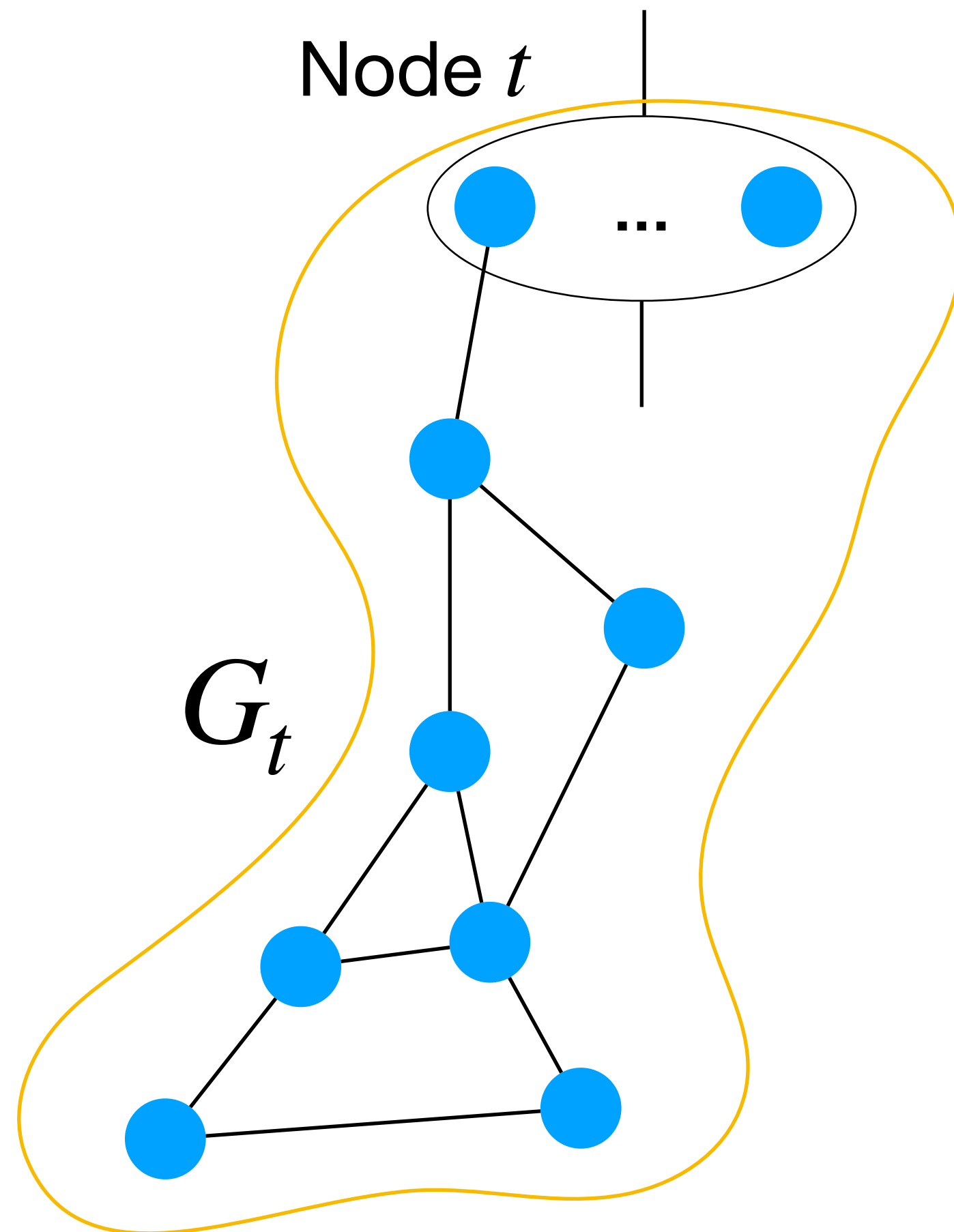
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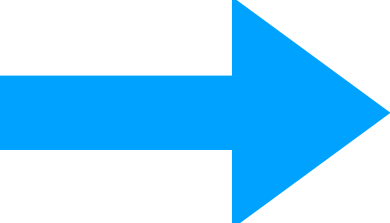


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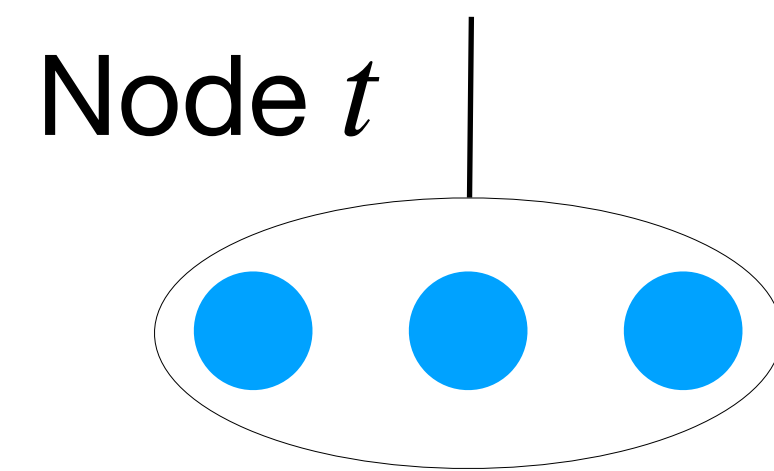


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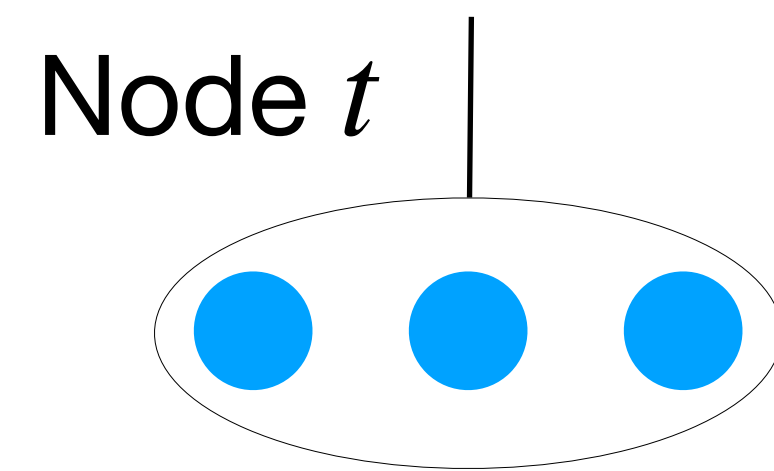
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**width  $k$**    $\leq 3^{k+1}$  colorings

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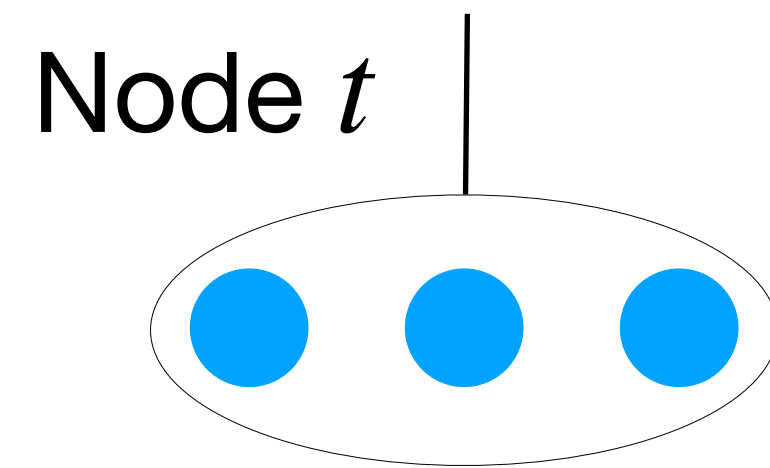


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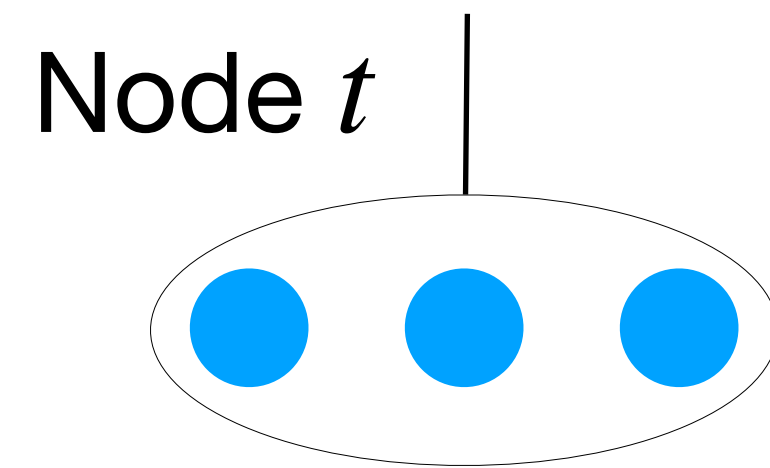
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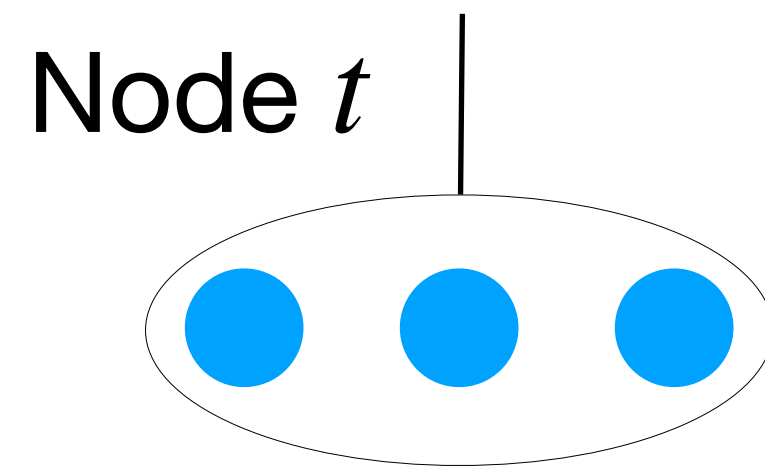


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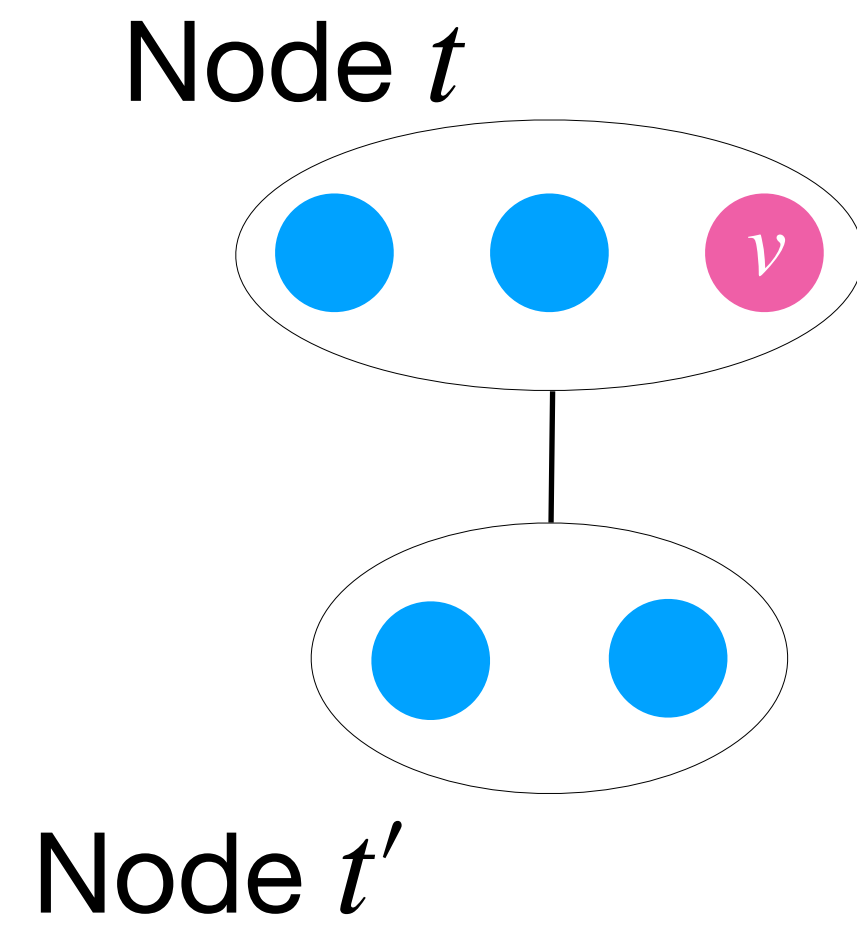
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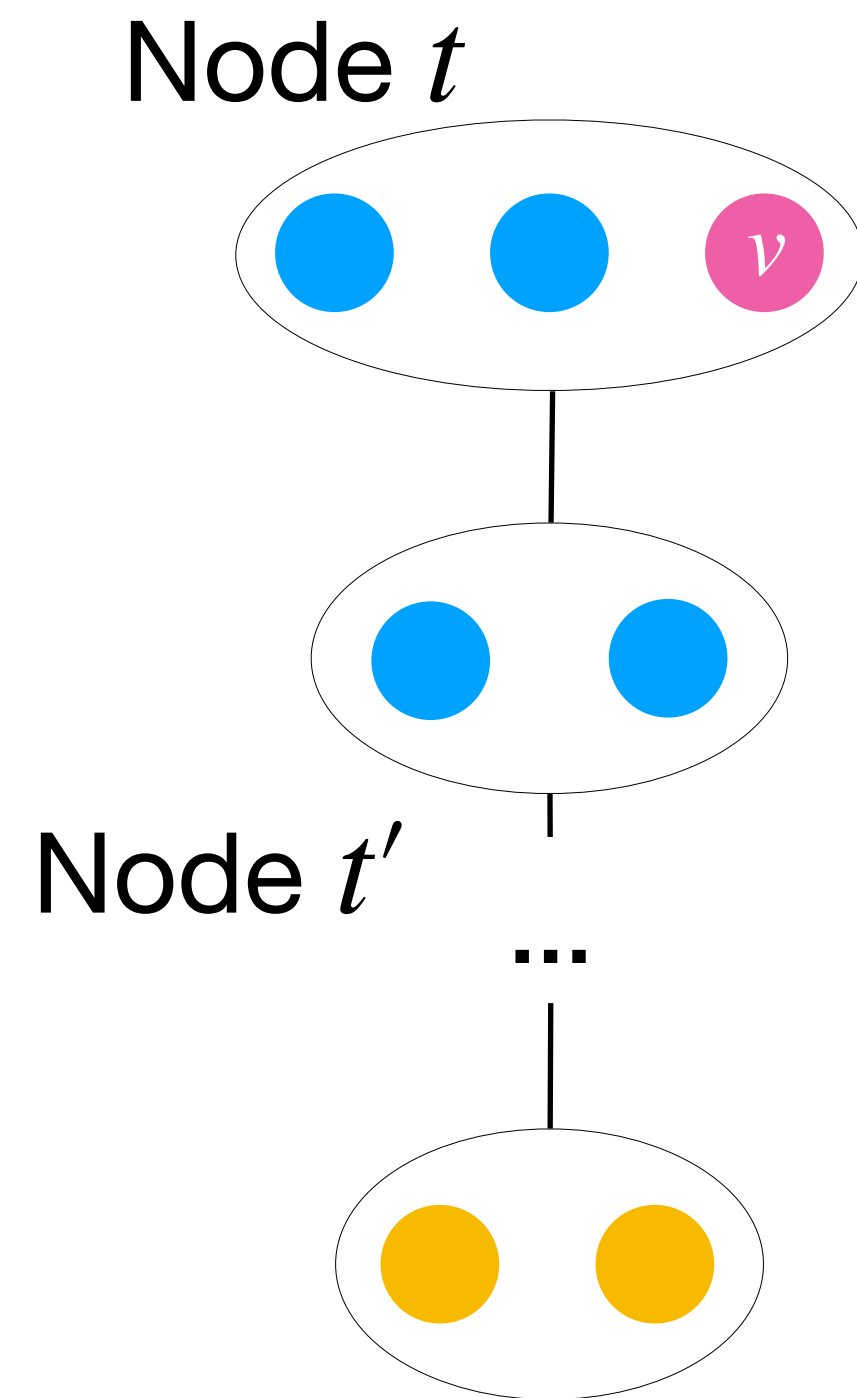
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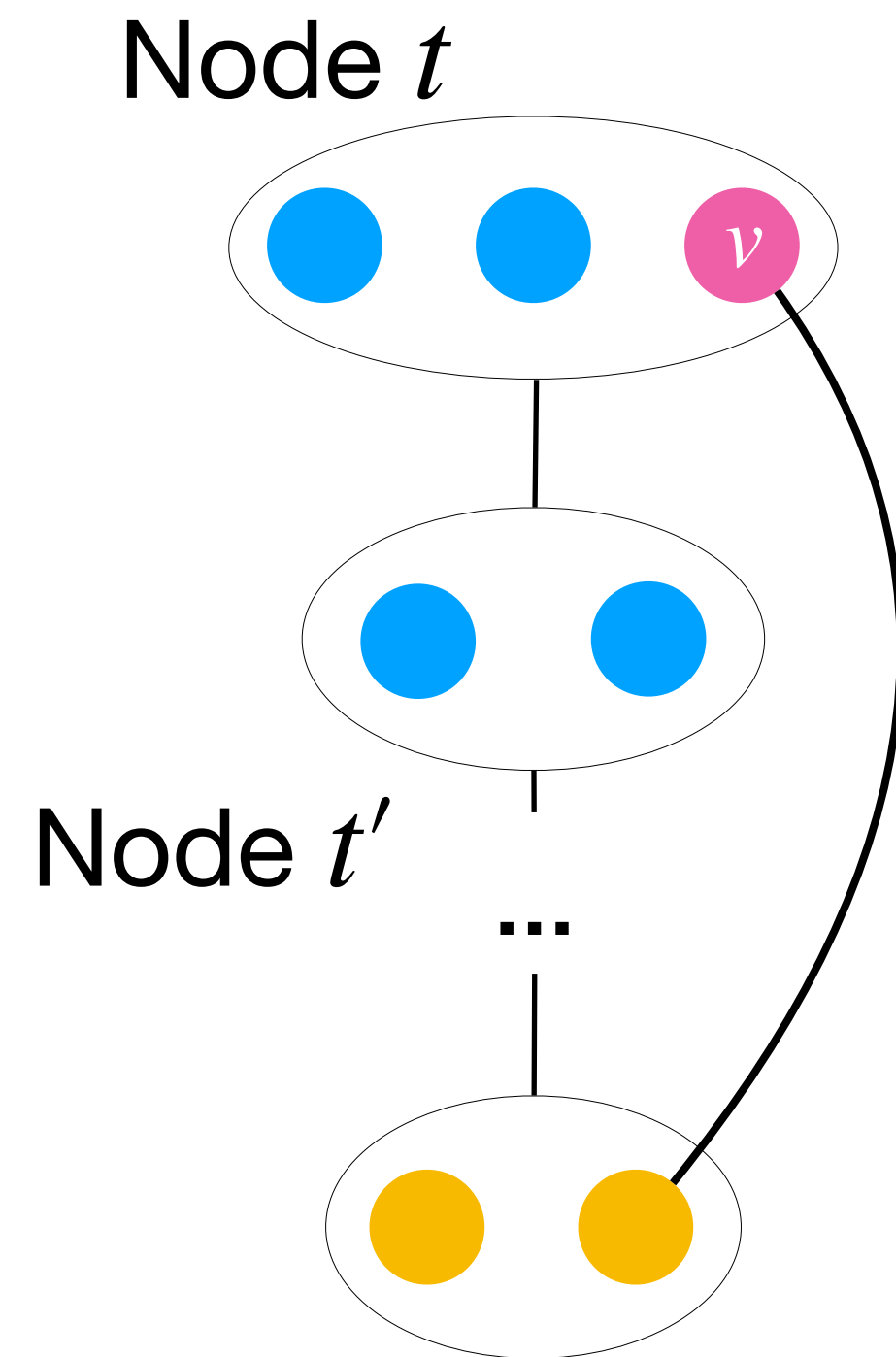
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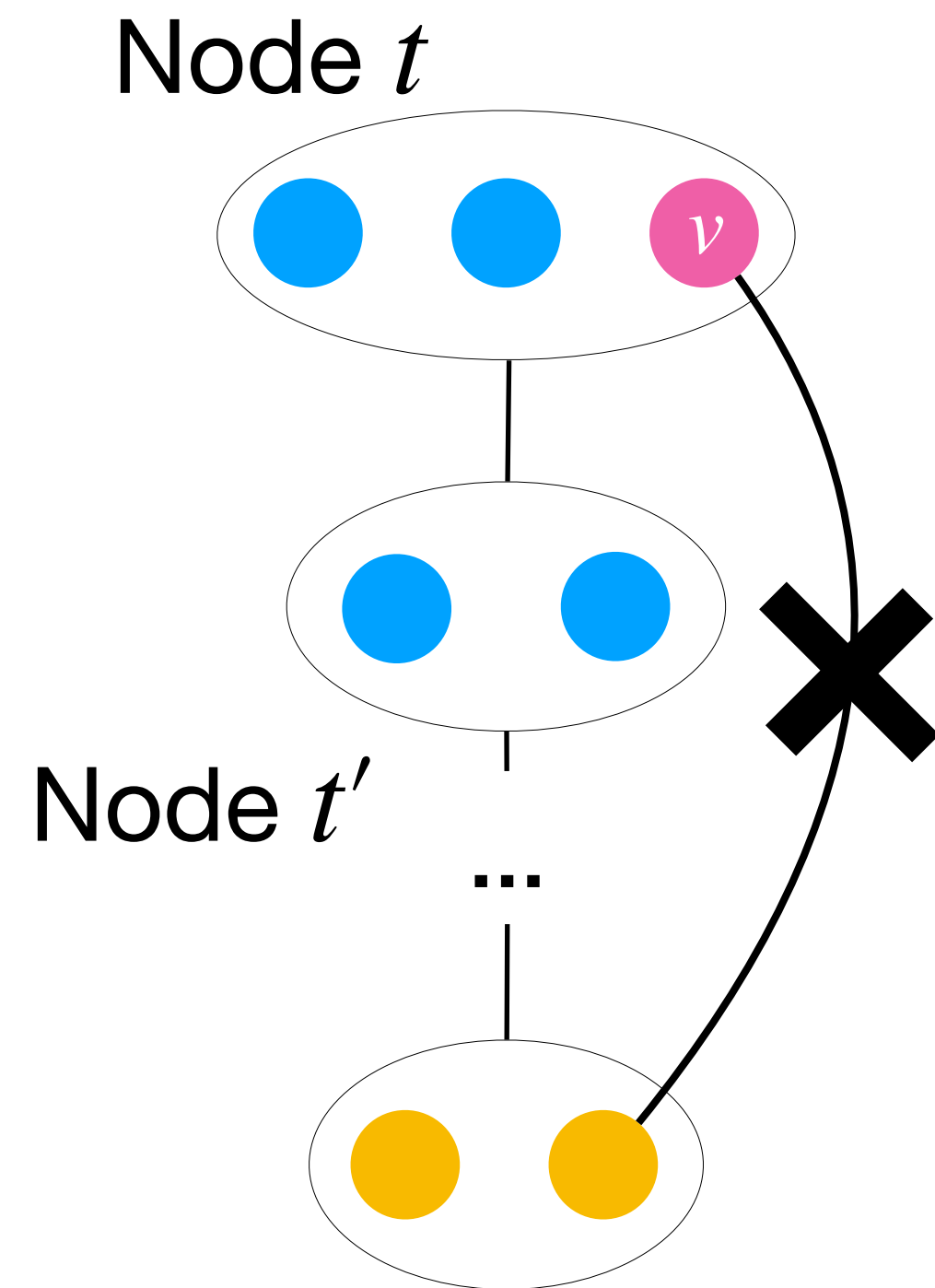
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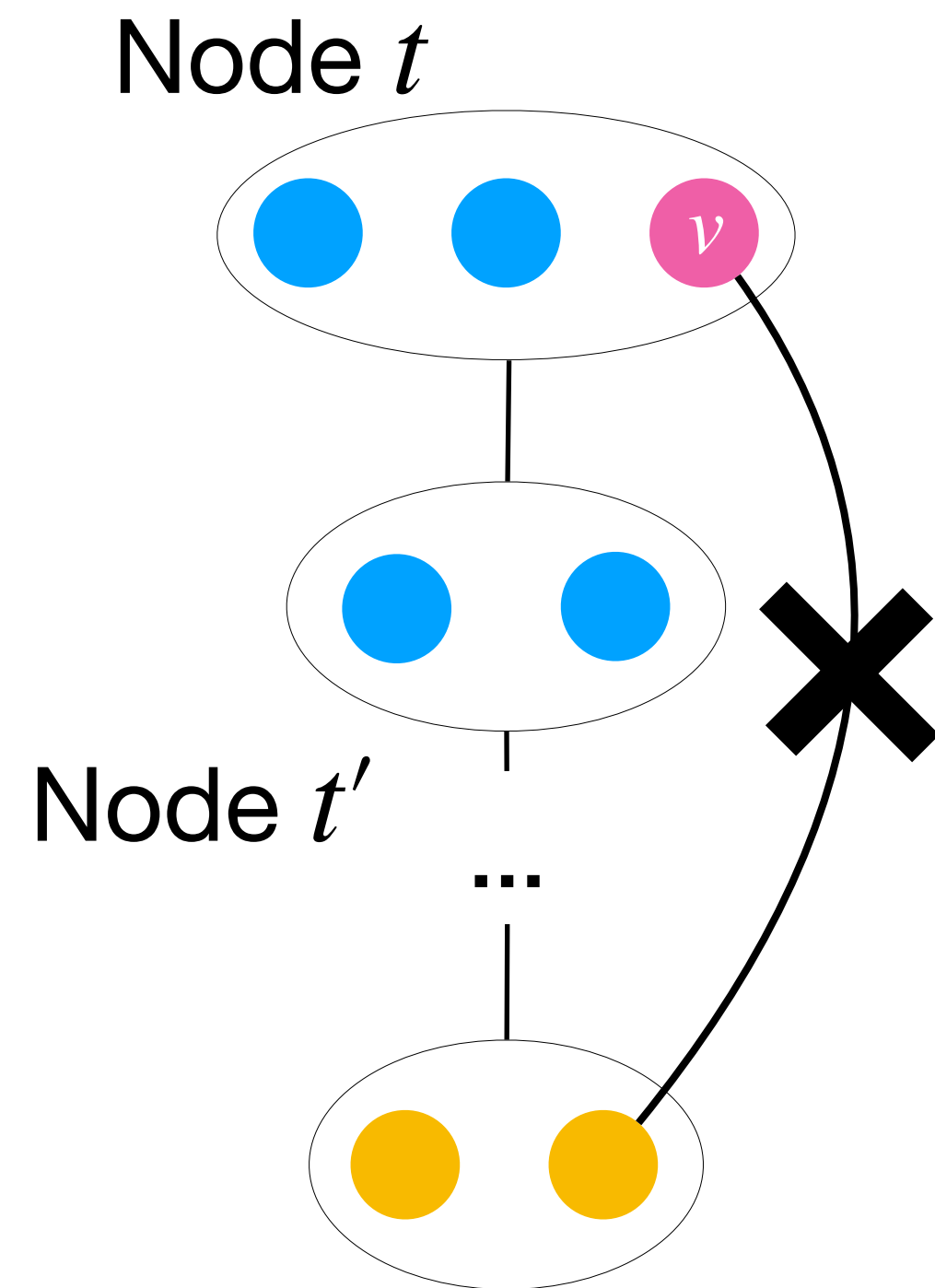
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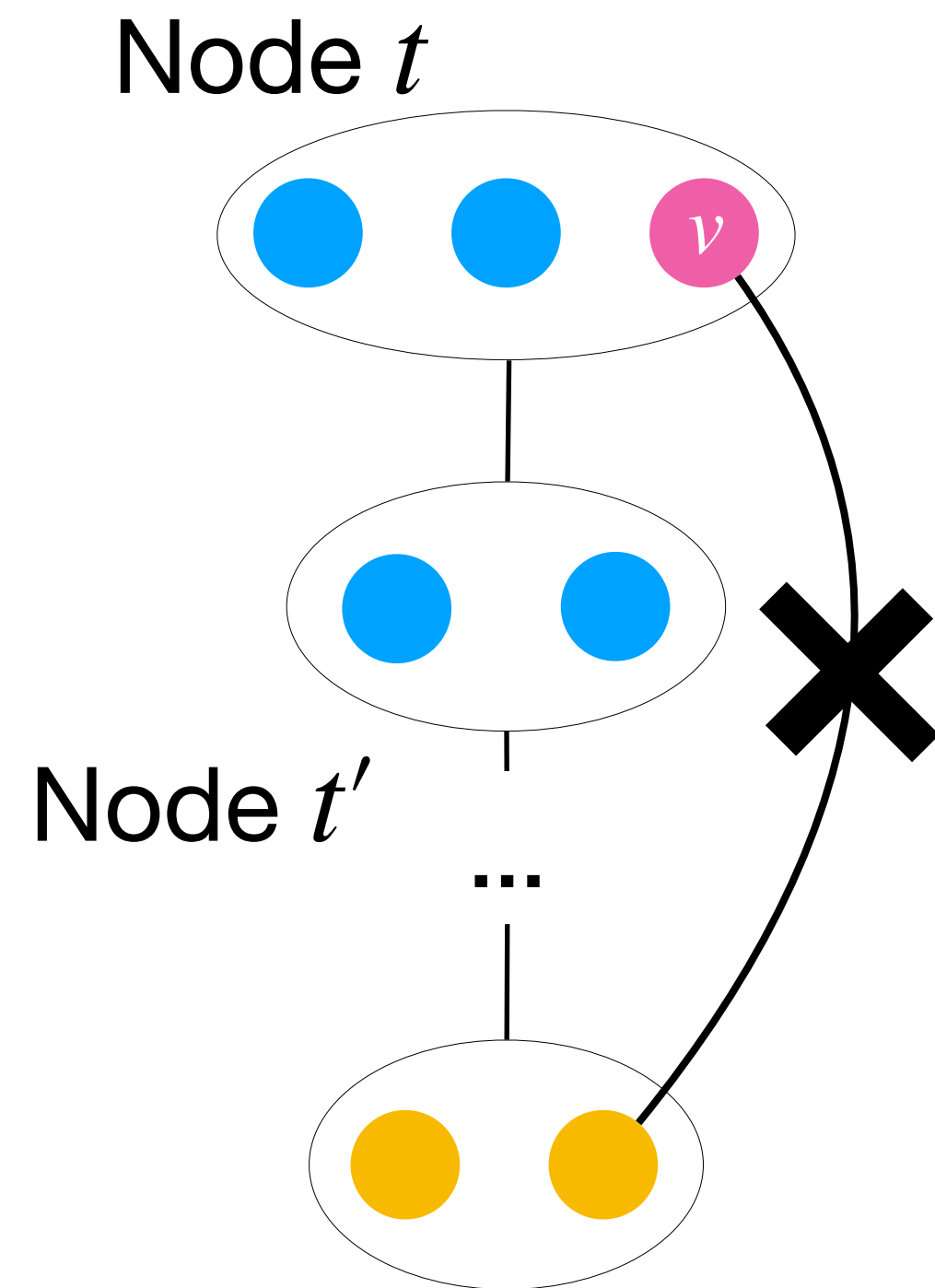


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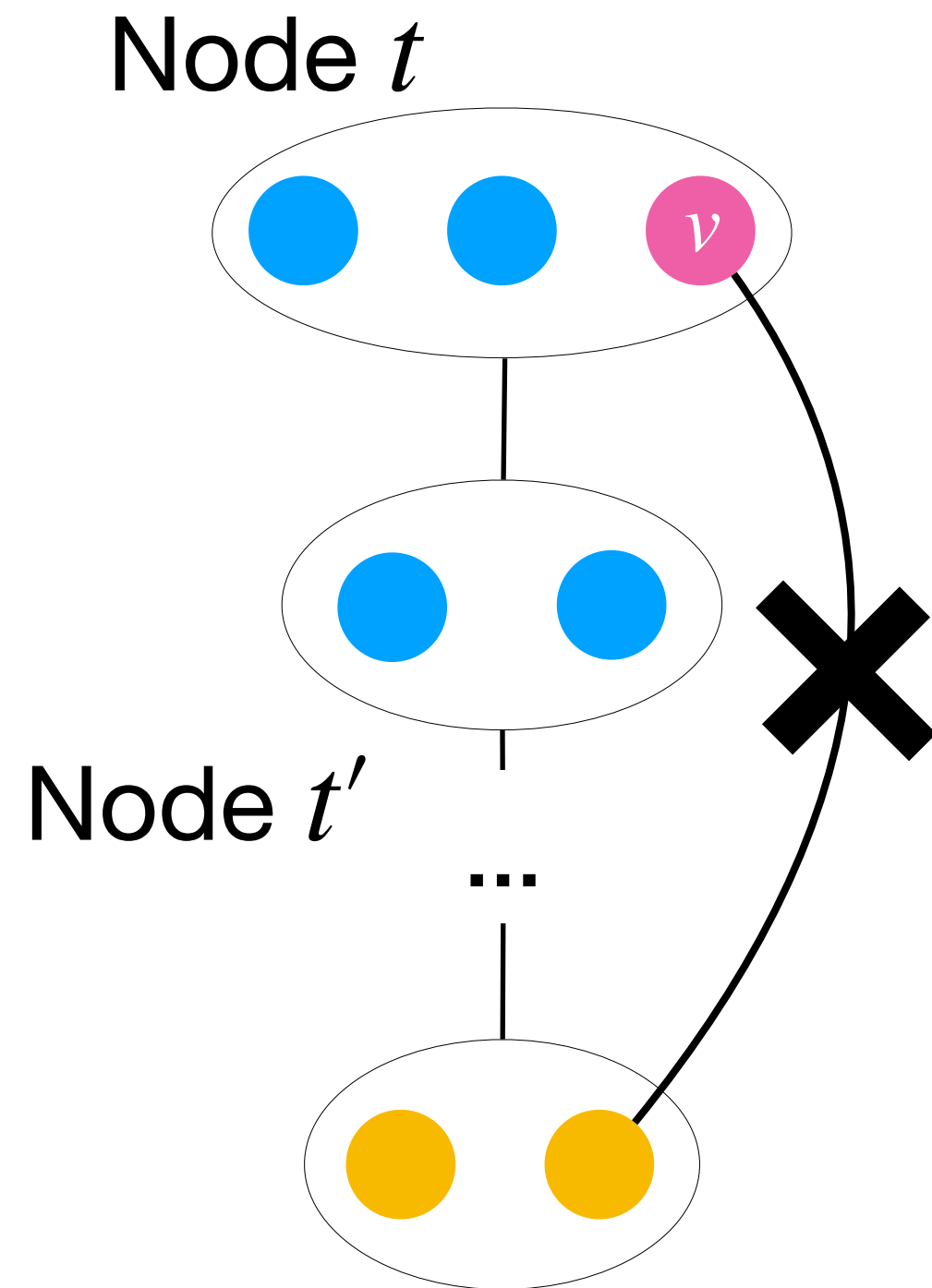
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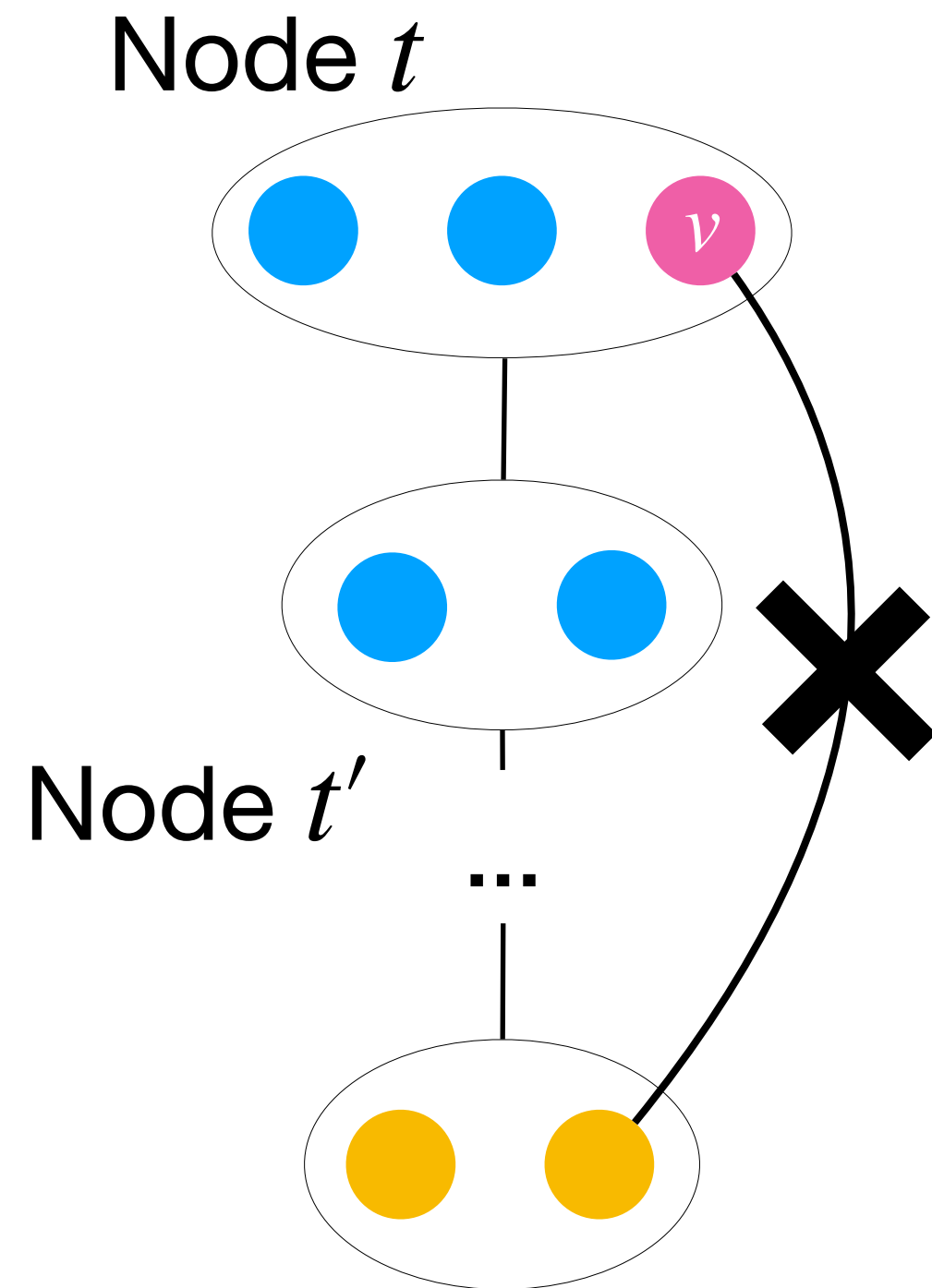


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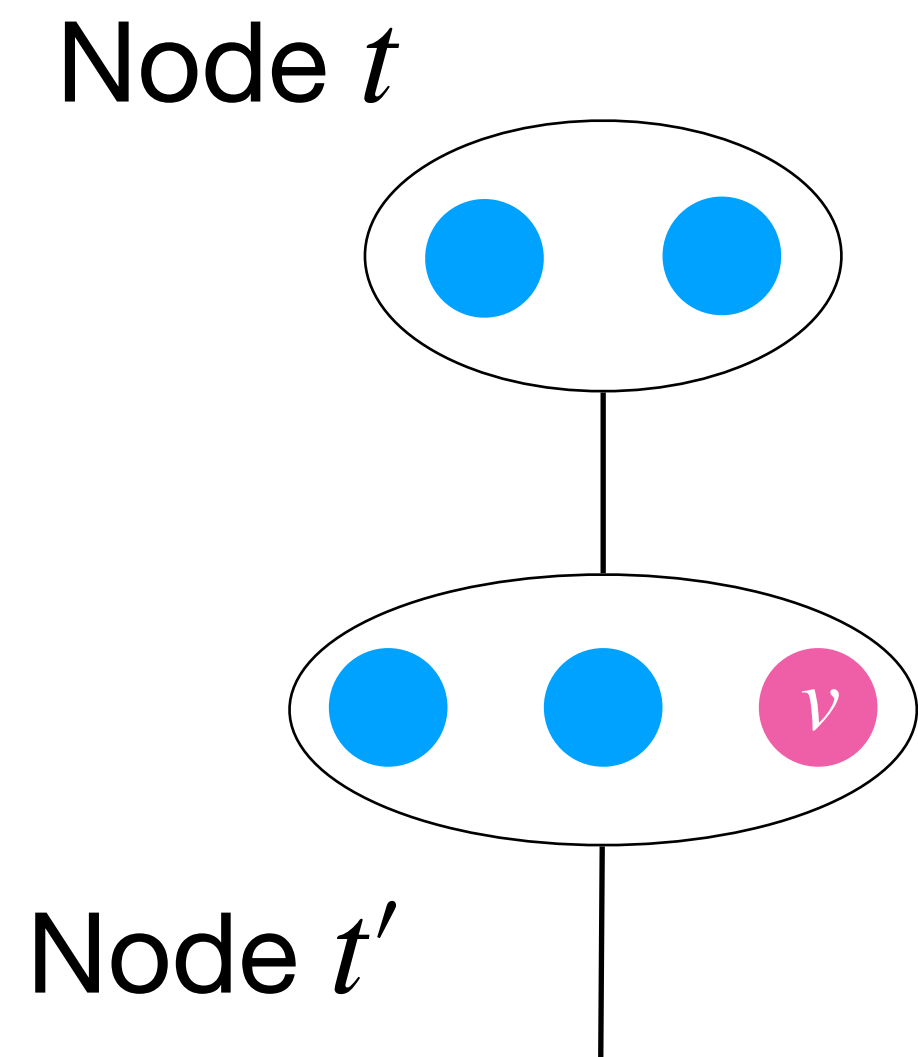
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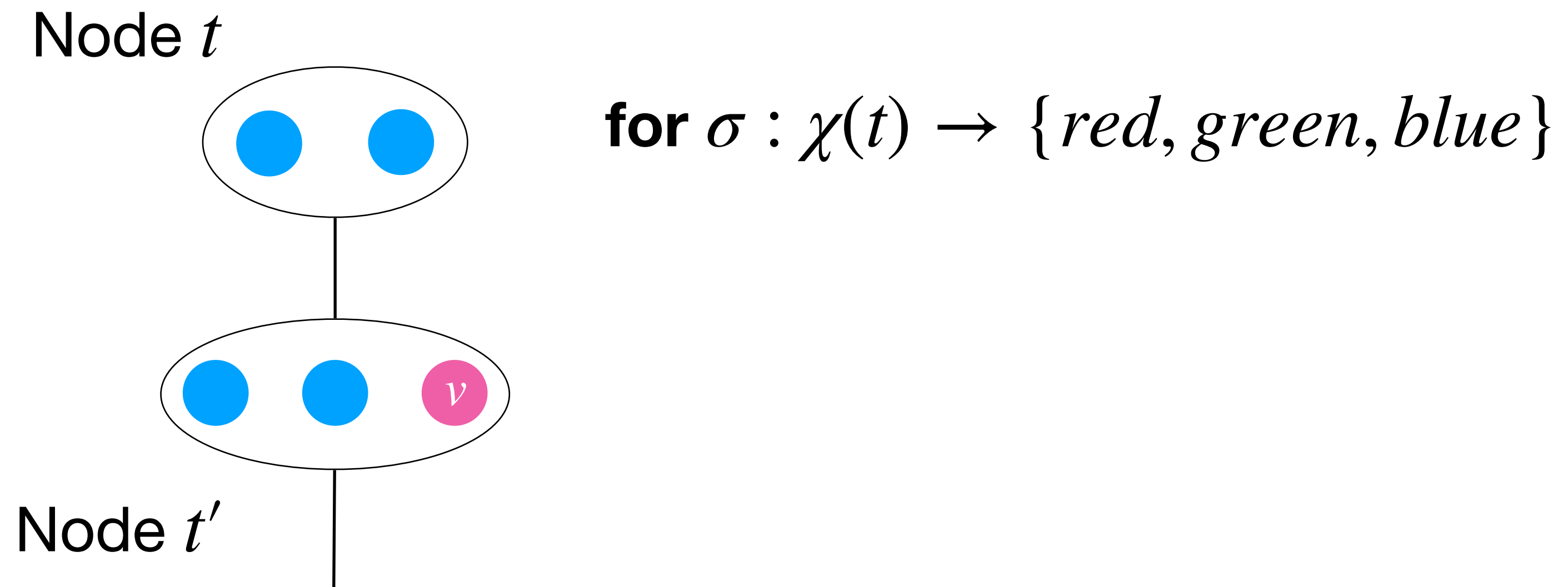
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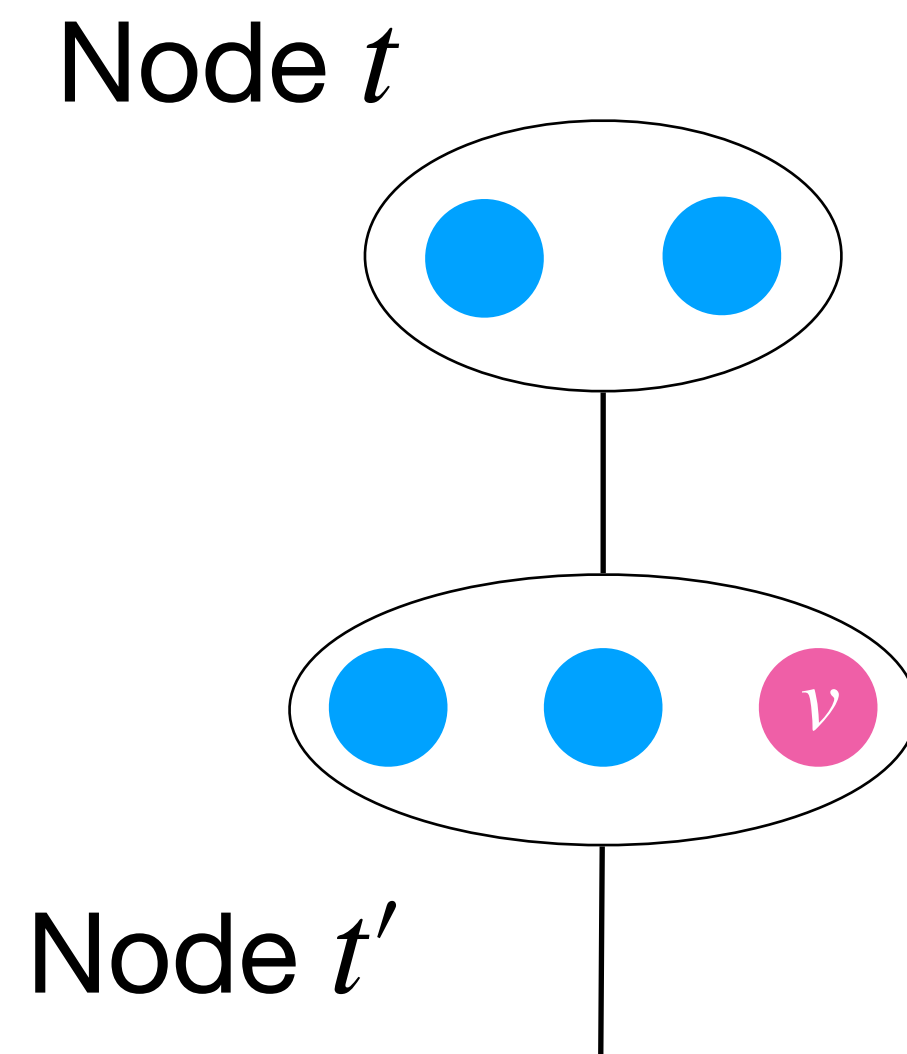
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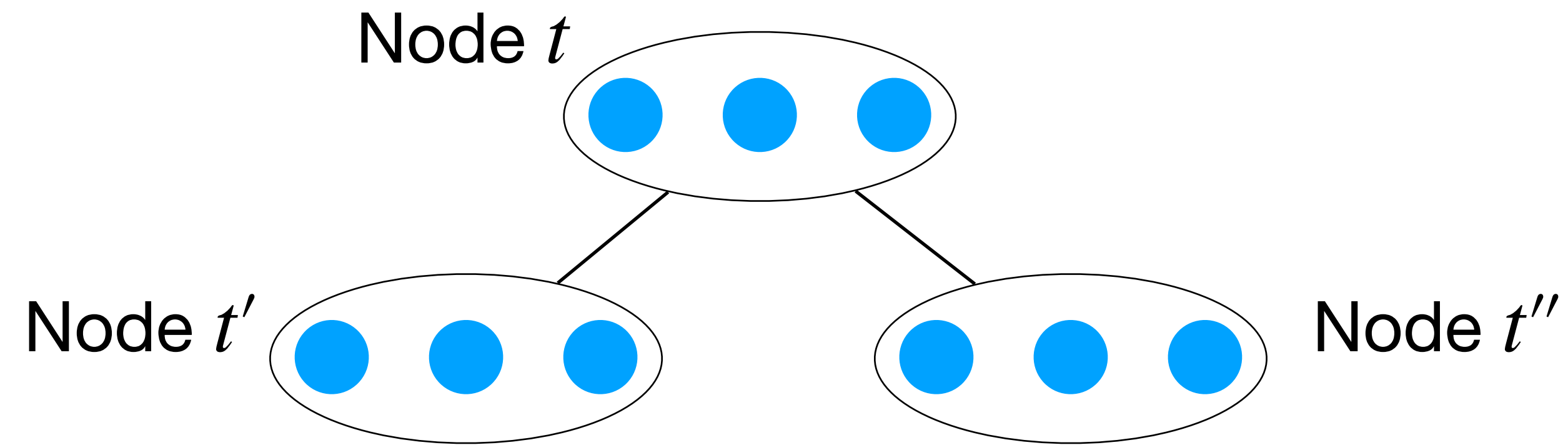
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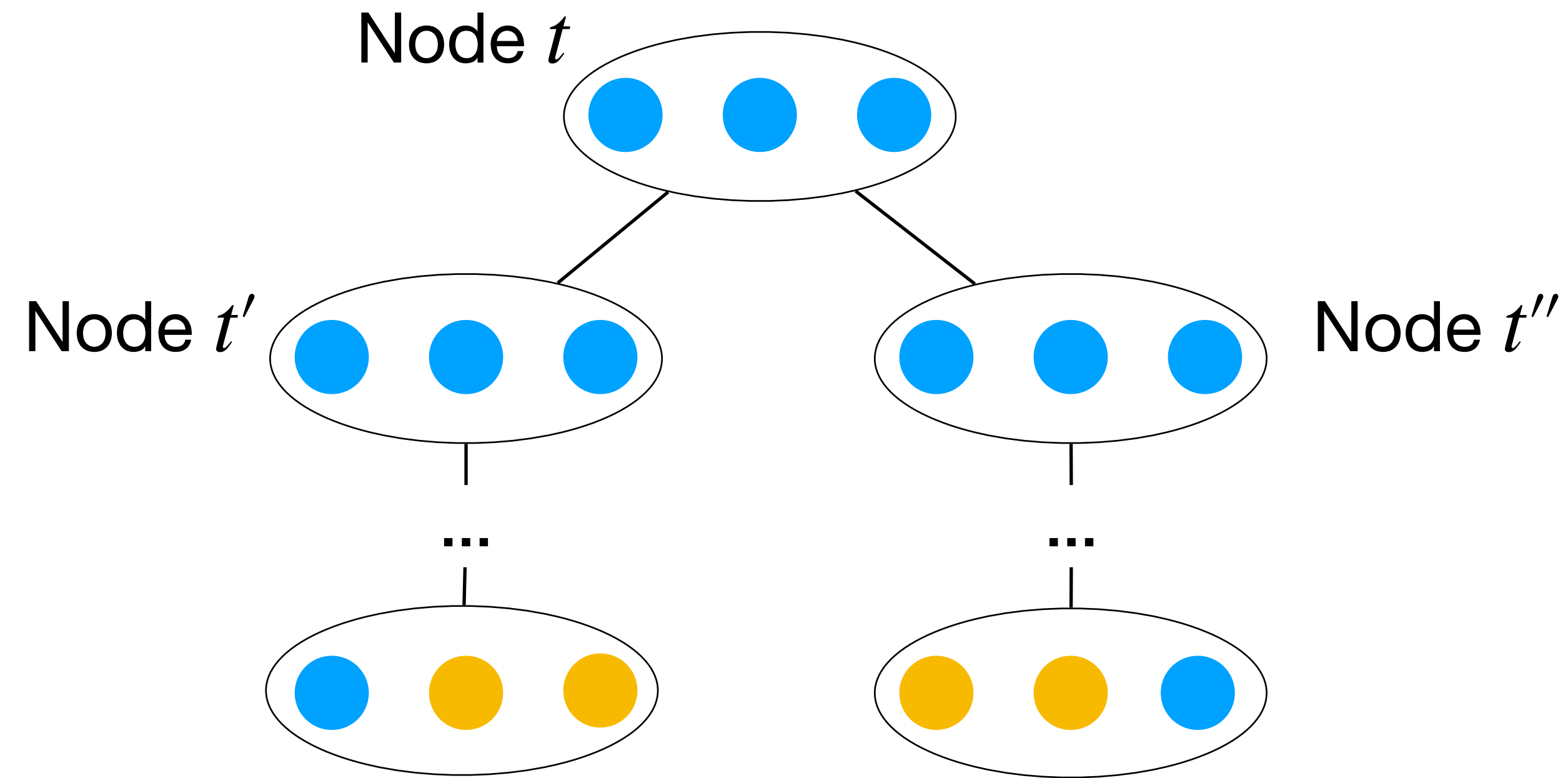
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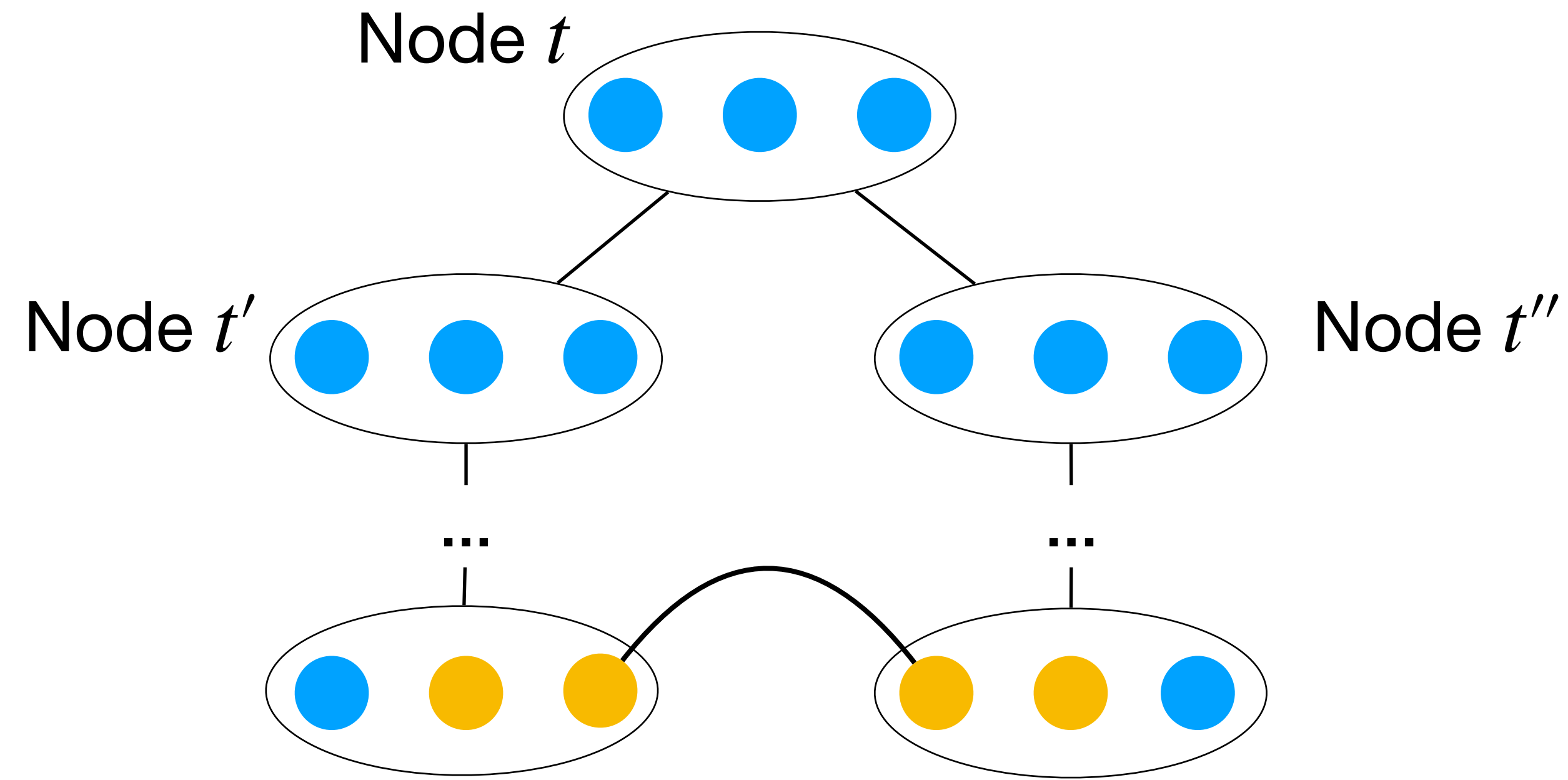
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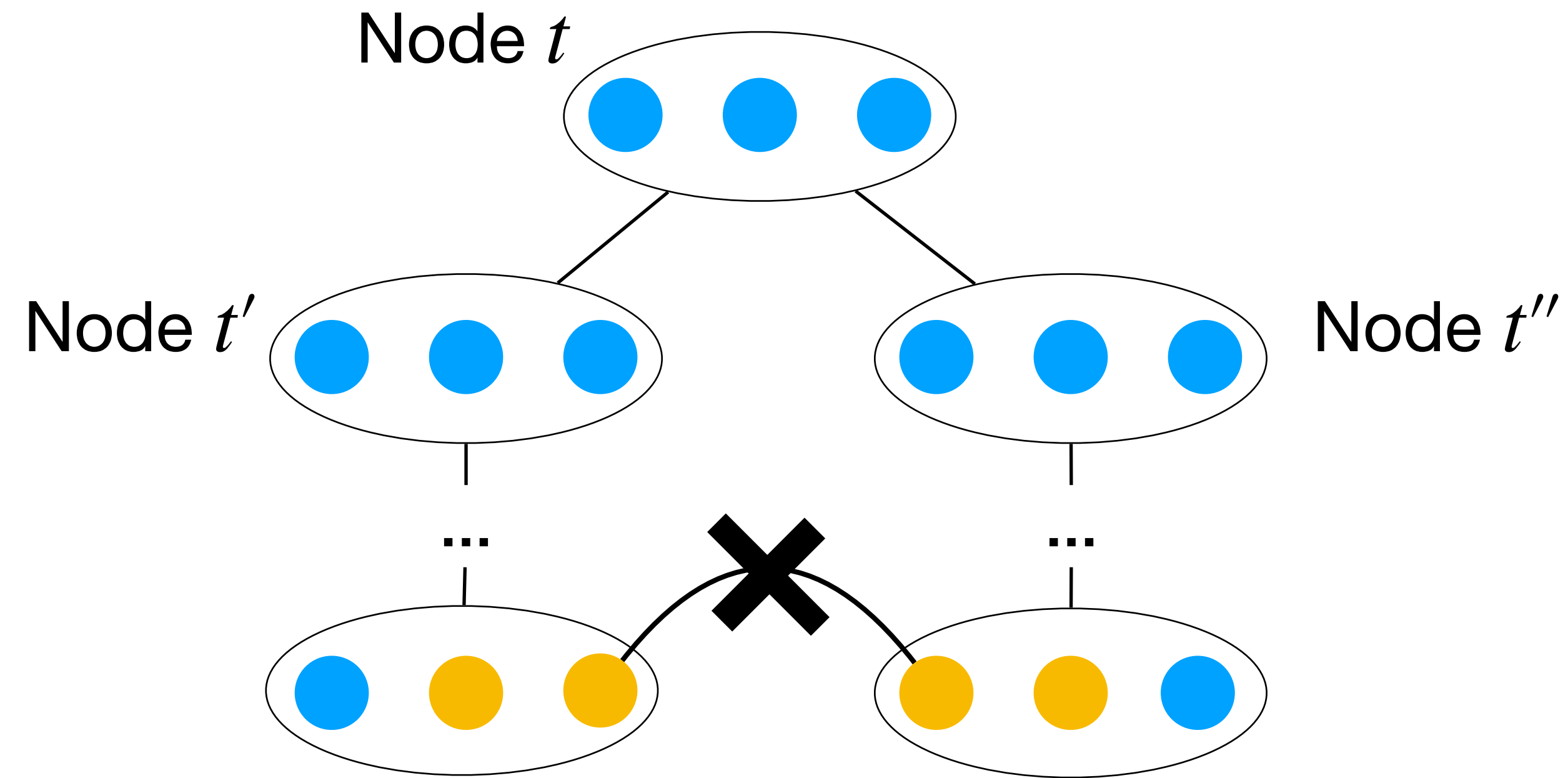
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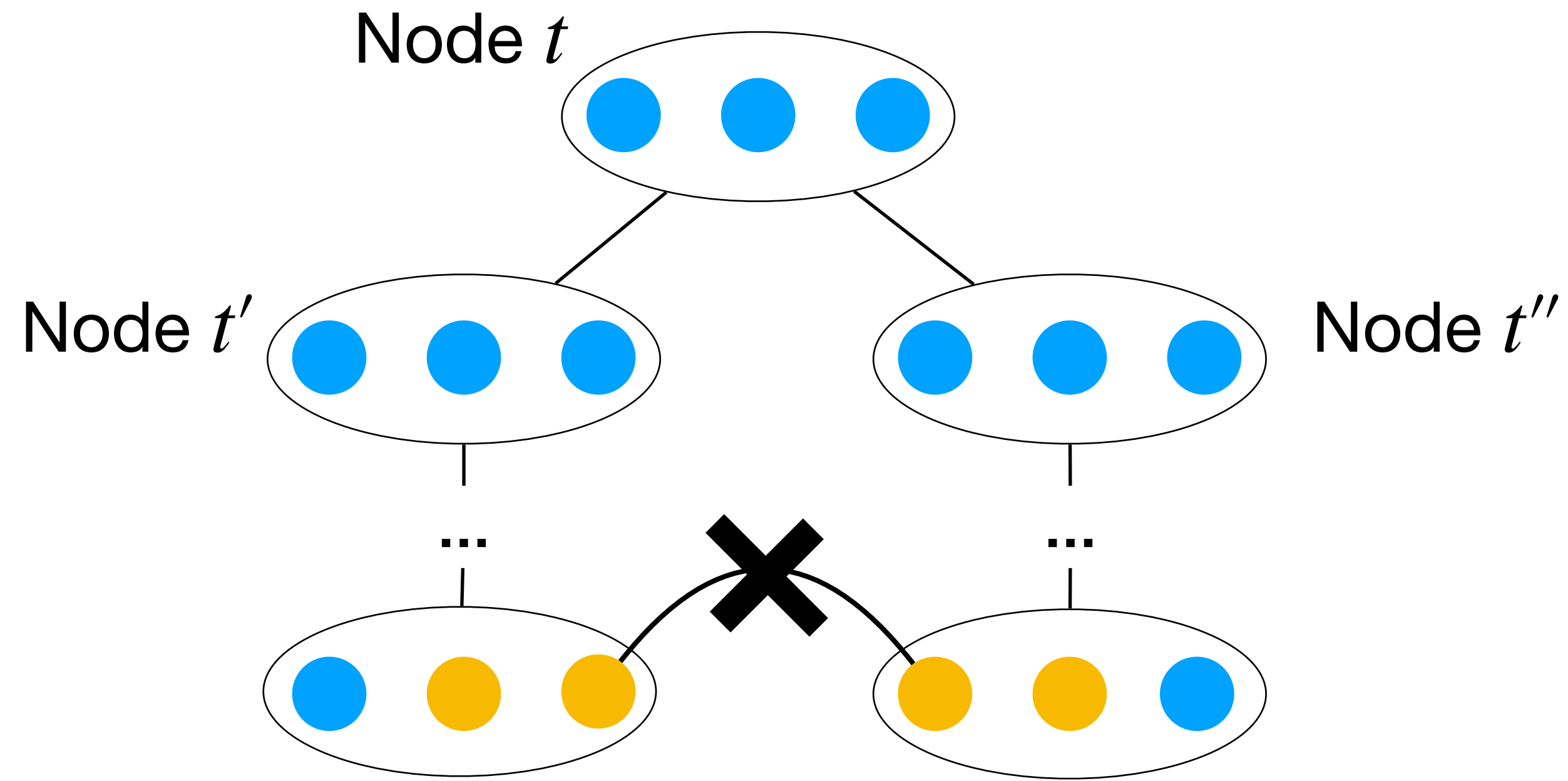
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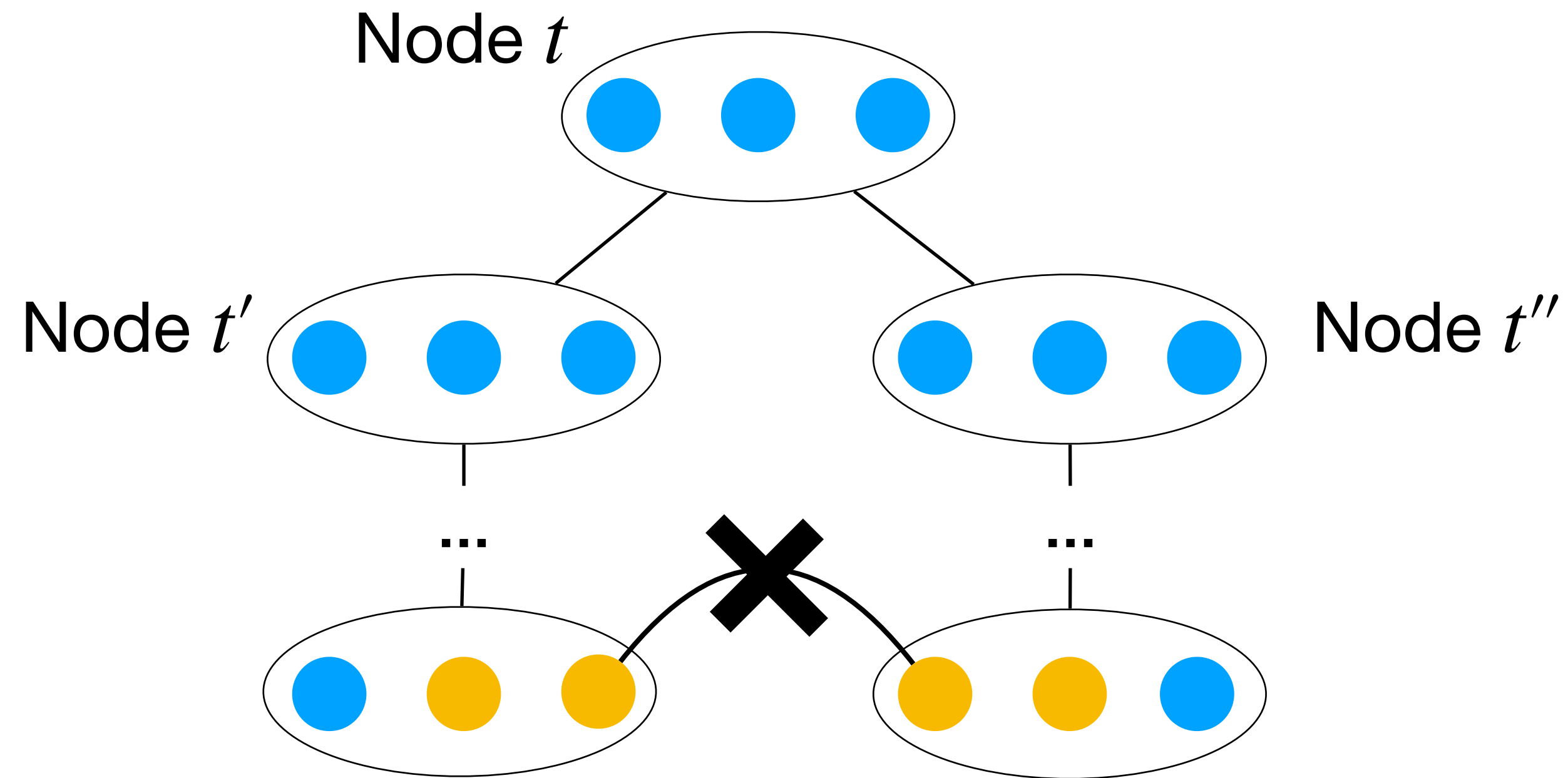
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**Con:** coming up with the right notion of “partial solution” ( $n_t(U)$ ) is difficult

# Computing Treewidth

# Treewidth is FPT

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**Input:** A graph  $G$  and an integer  $k$ .

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**Parameter:**  $k$

### Theorem (Bodlaender)

There is a function  $f$  and an algorithm  $A$  that computes a tree decomposition of a graph  $G$  in time  $f(k) |V(G)|$  or decides that its treewidth is greater than  $k$ .

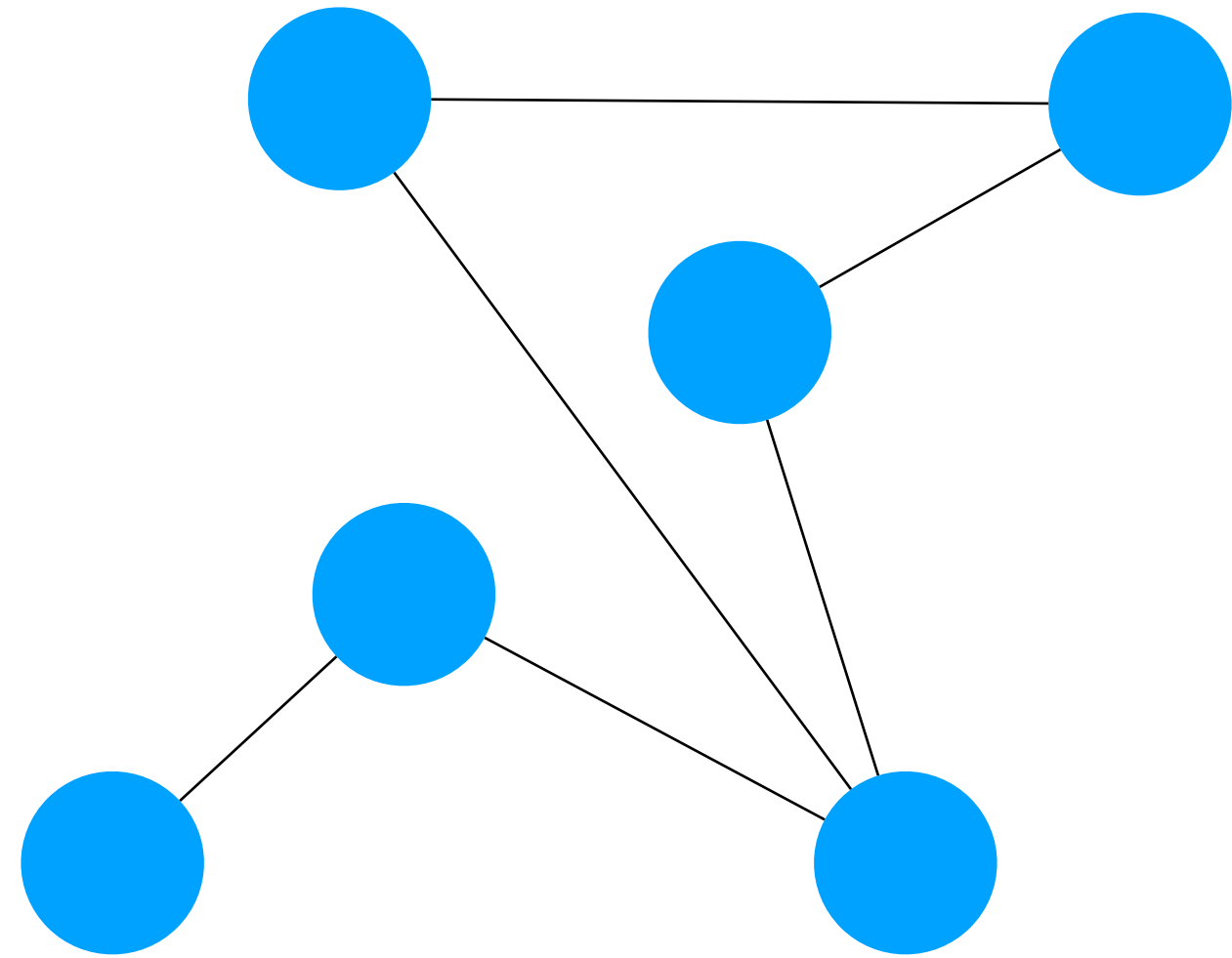
# FPT 2-Approximation

## Theorem (Korhonen 2021)

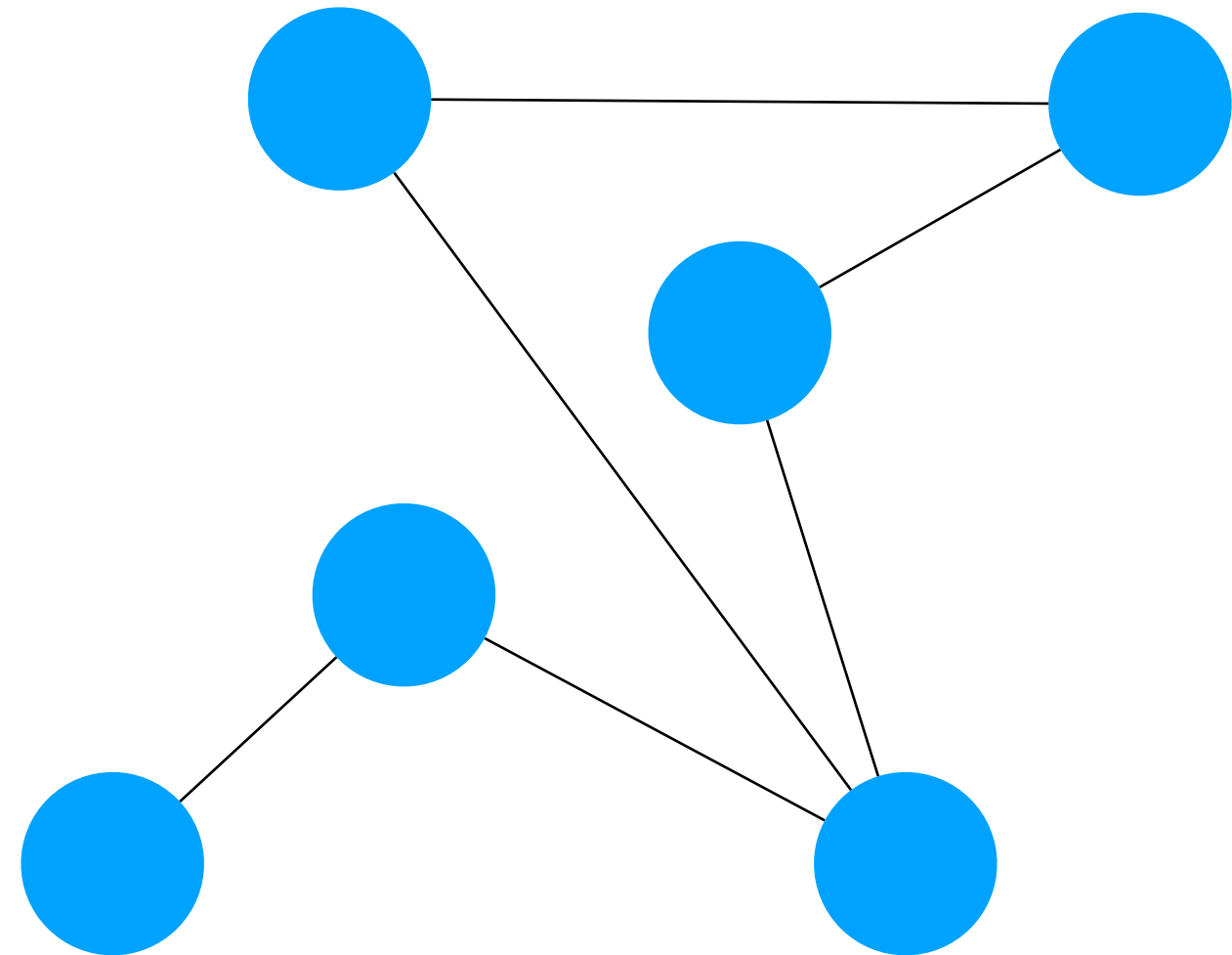
There is an algorithm that, given an  $n$ -vertex graph  $G$  and an integer  $k$ , in time  $2^{O(k)}n$  either outputs a tree decomposition of width at most  $2k + 1$  or determines that the treewidth of  $G$  is larger than  $k$ .

# Heuristics

# Elimination Orderings



# Elimination Orderings

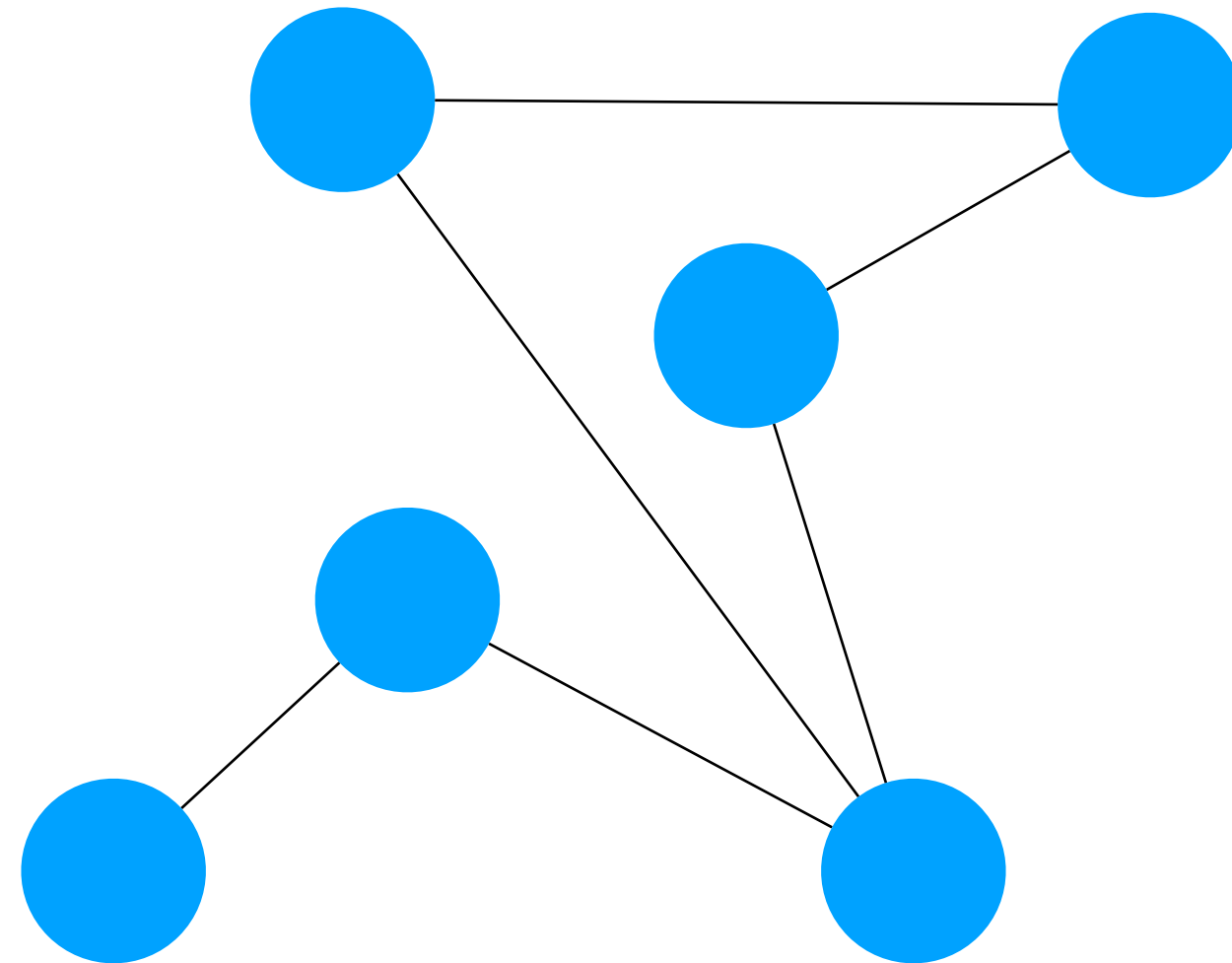


## Definition

Let  $G = (V, E)$  be a graph. An **elimination ordering** of  $G$  is simply an ordering  $\sigma = (v_1, \dots, v_n)$  of  $V$ .



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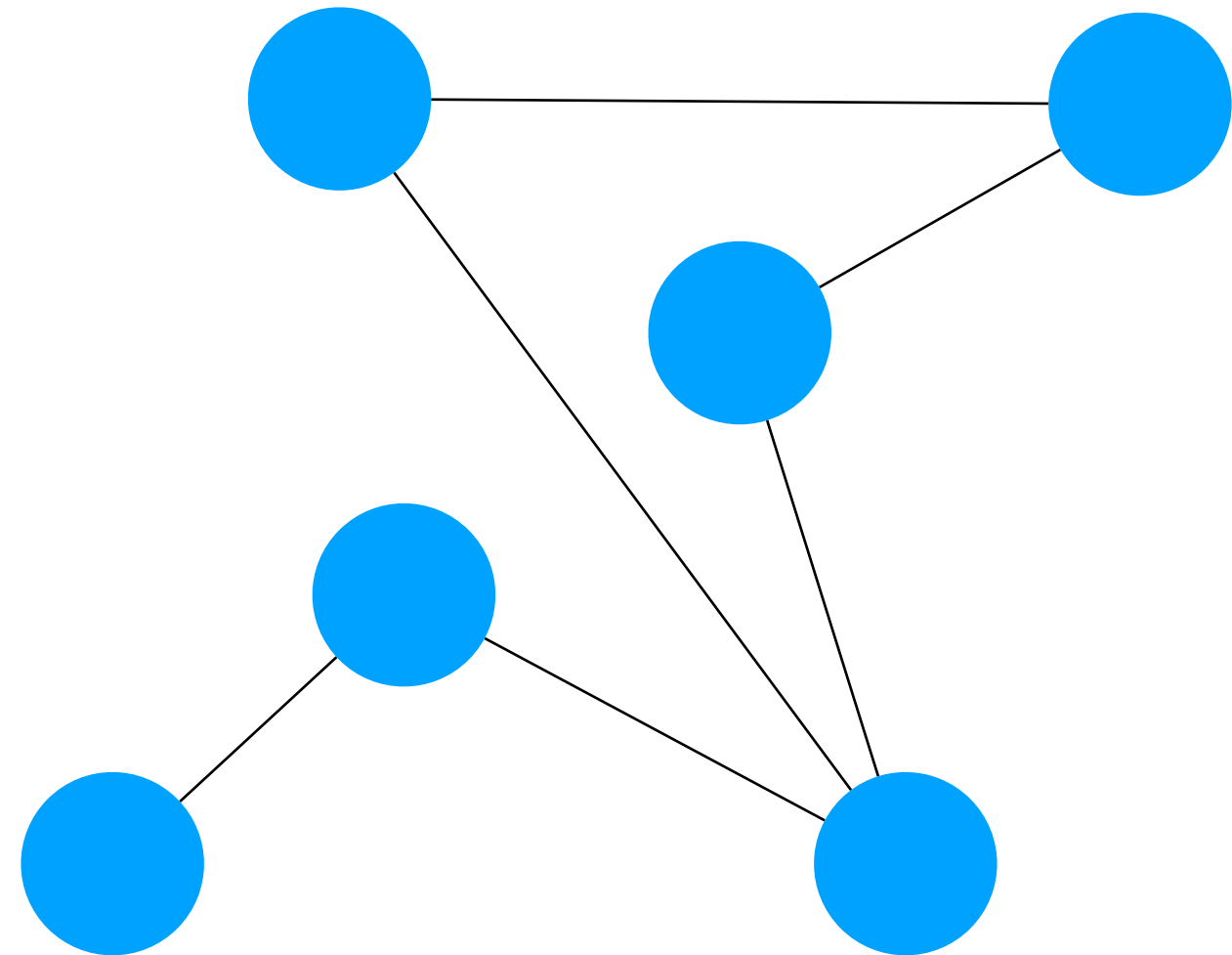


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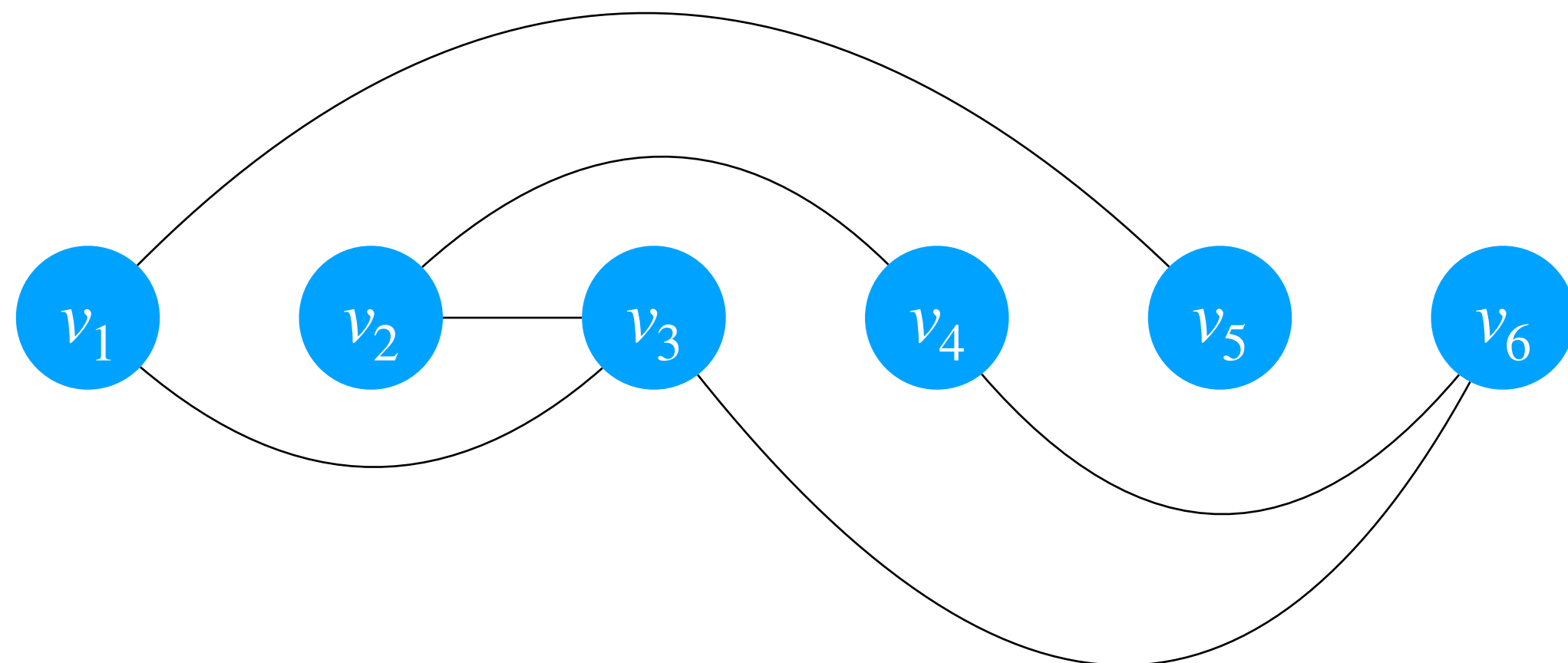


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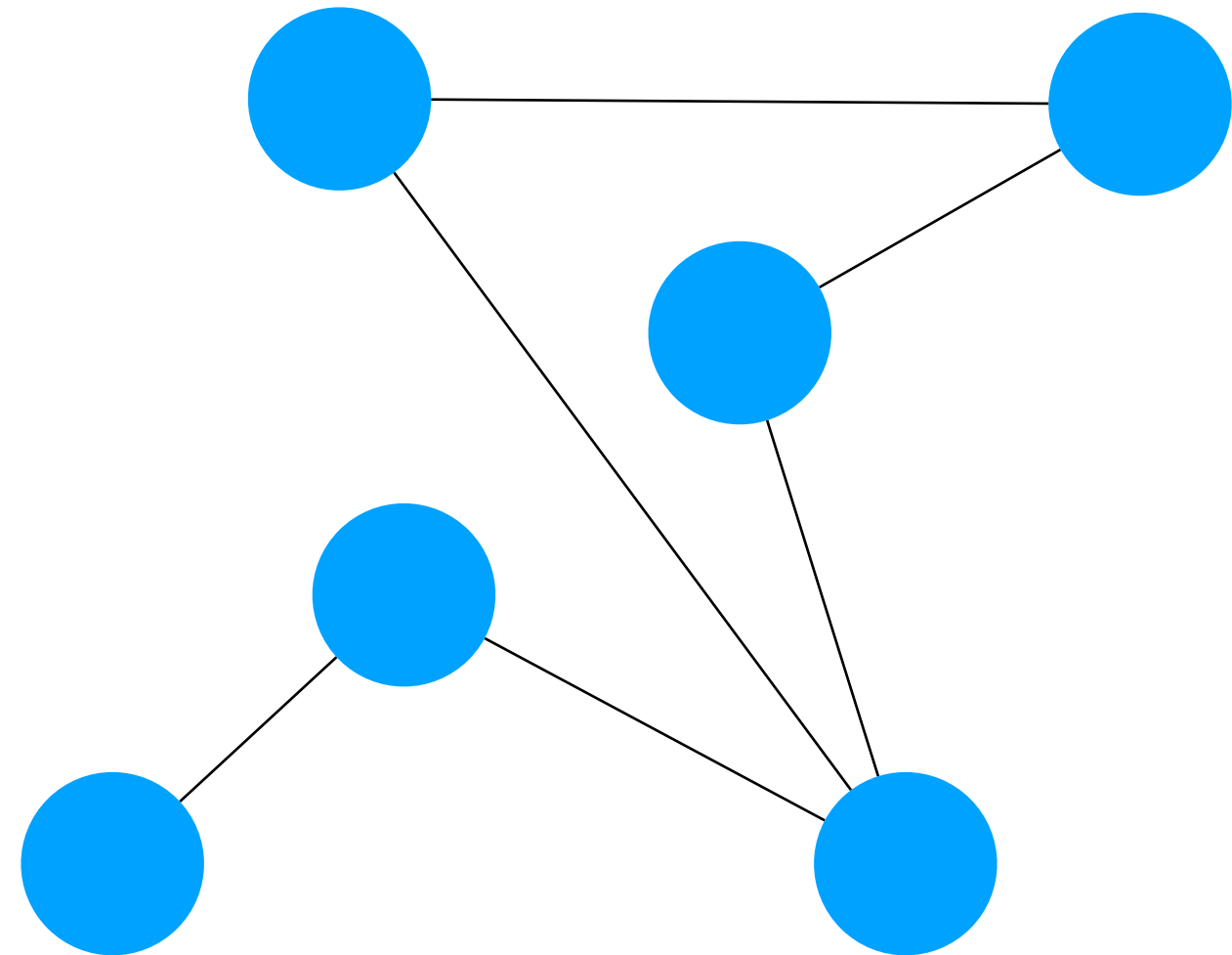


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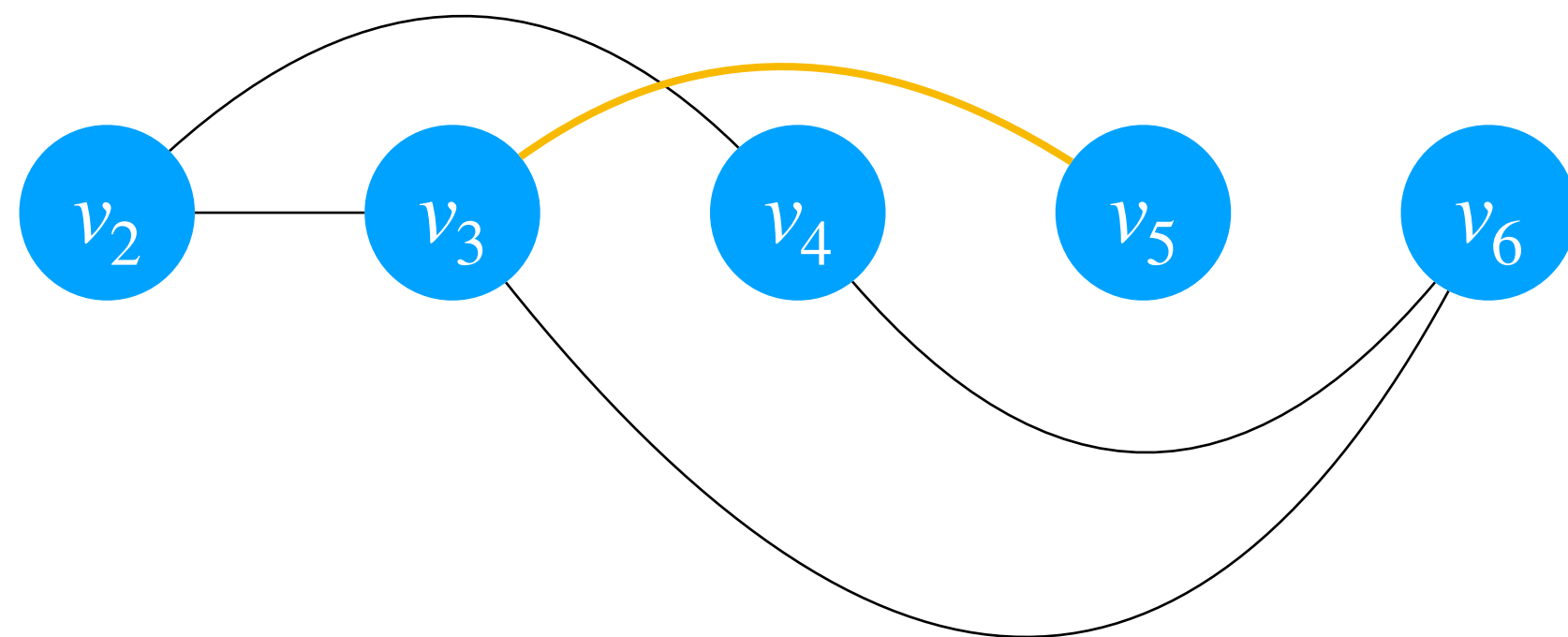


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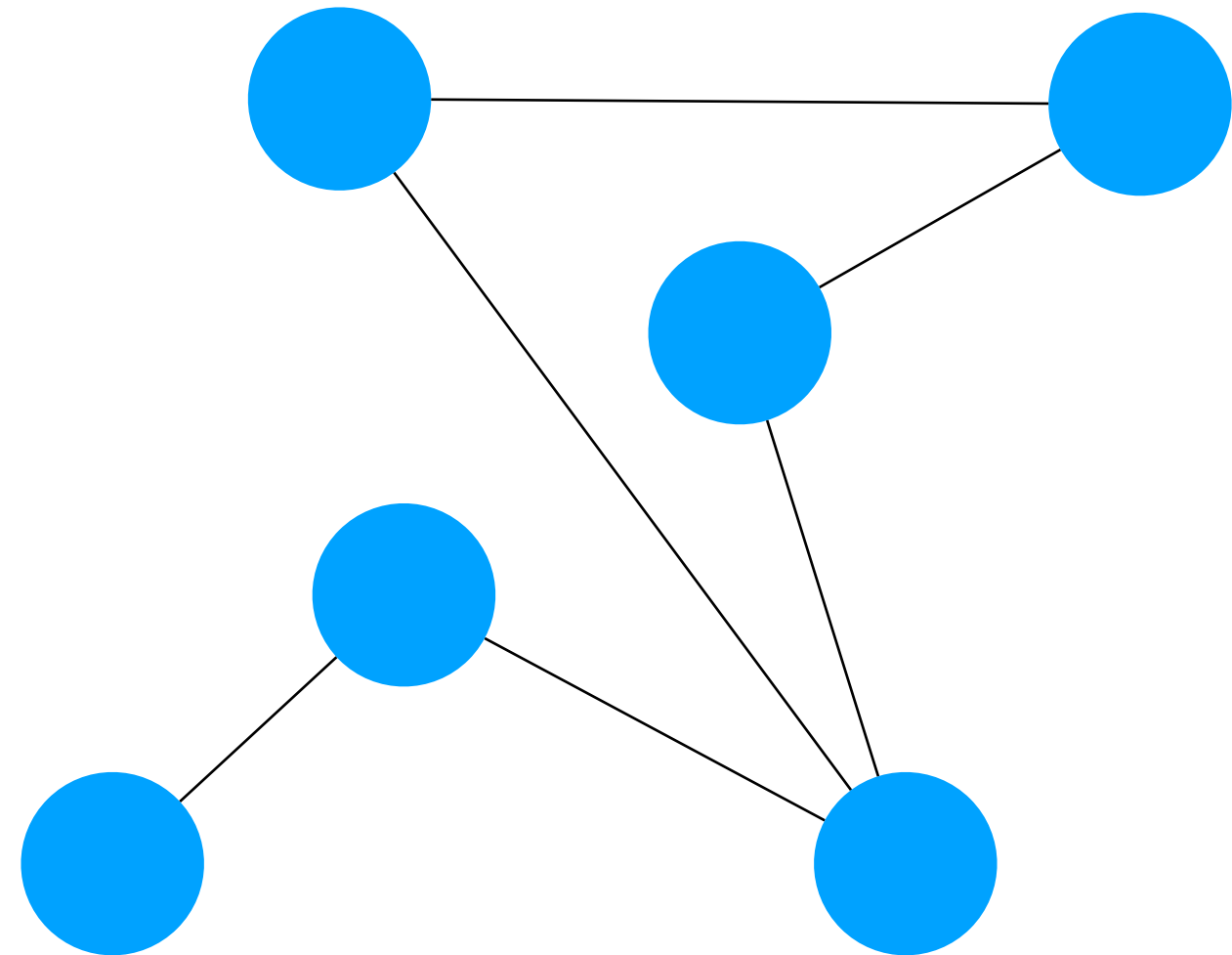


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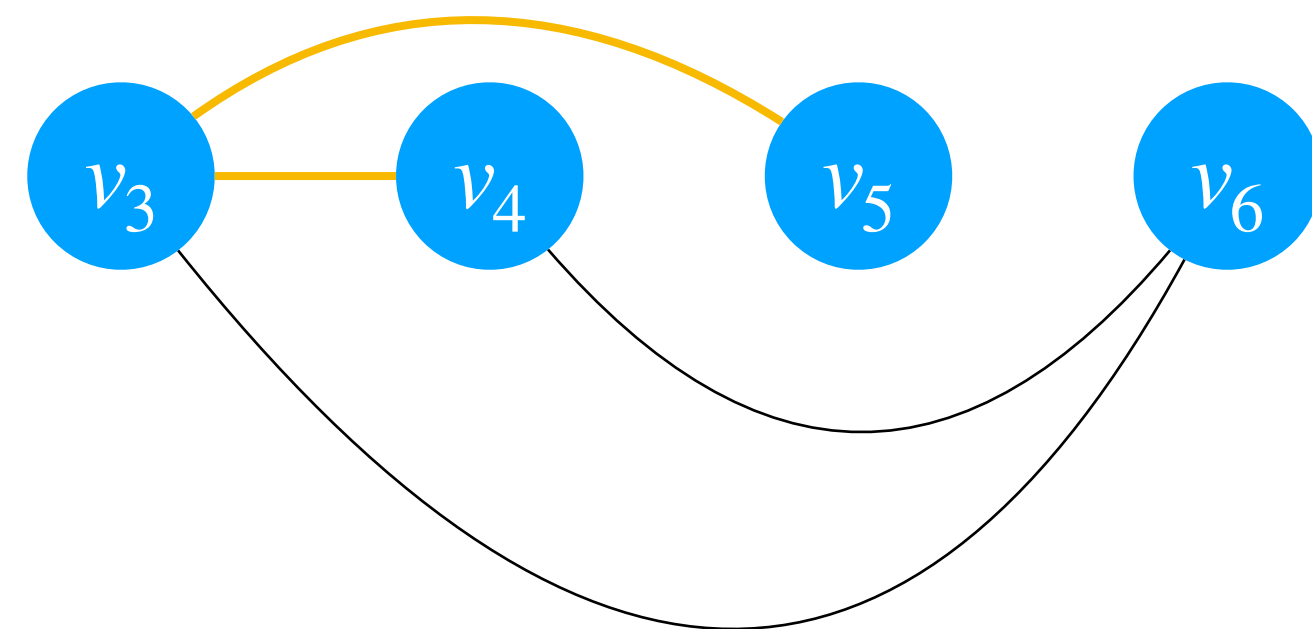


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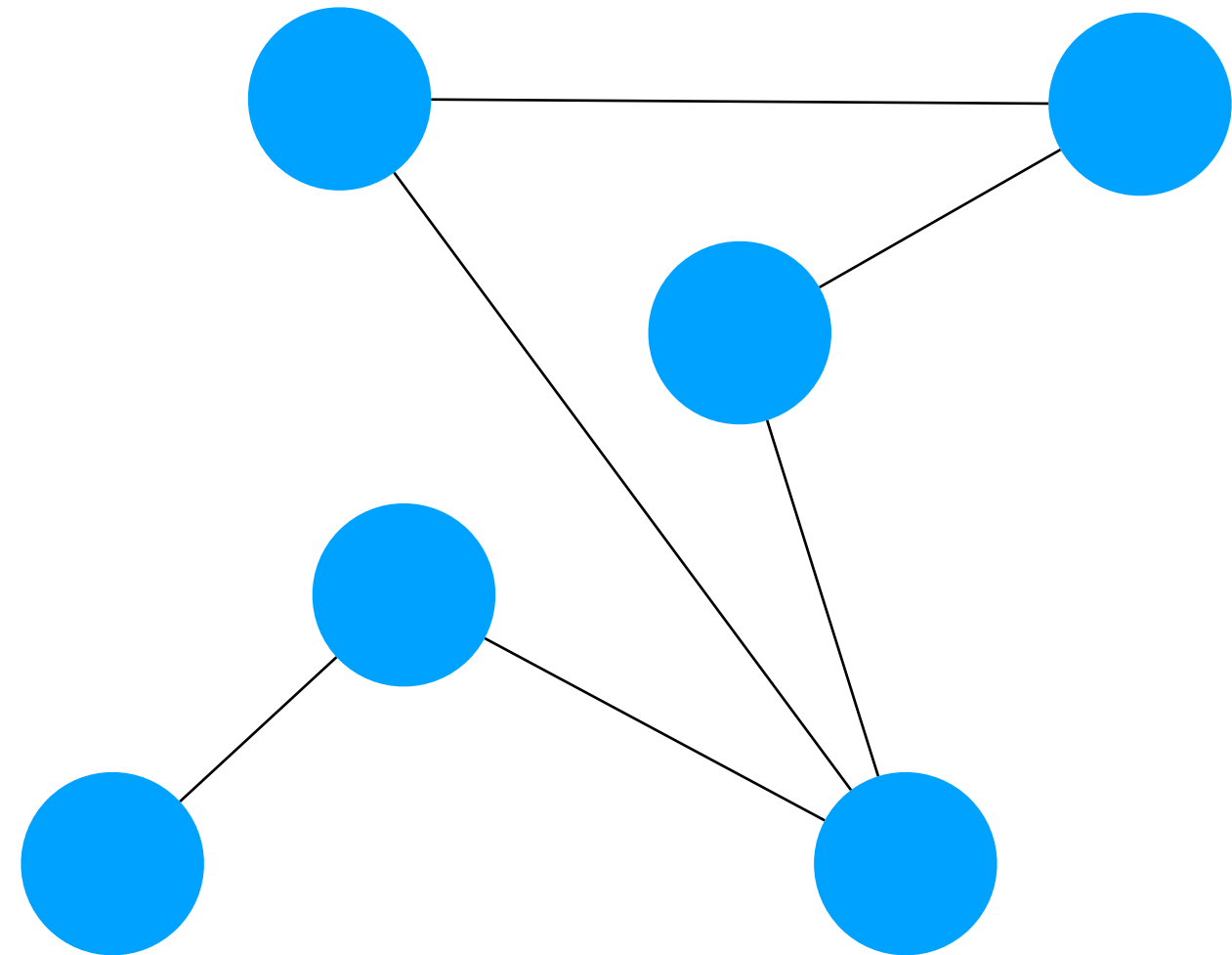


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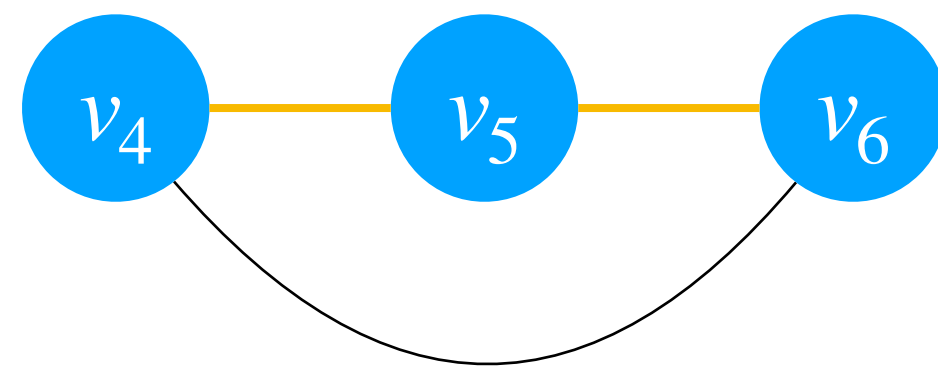


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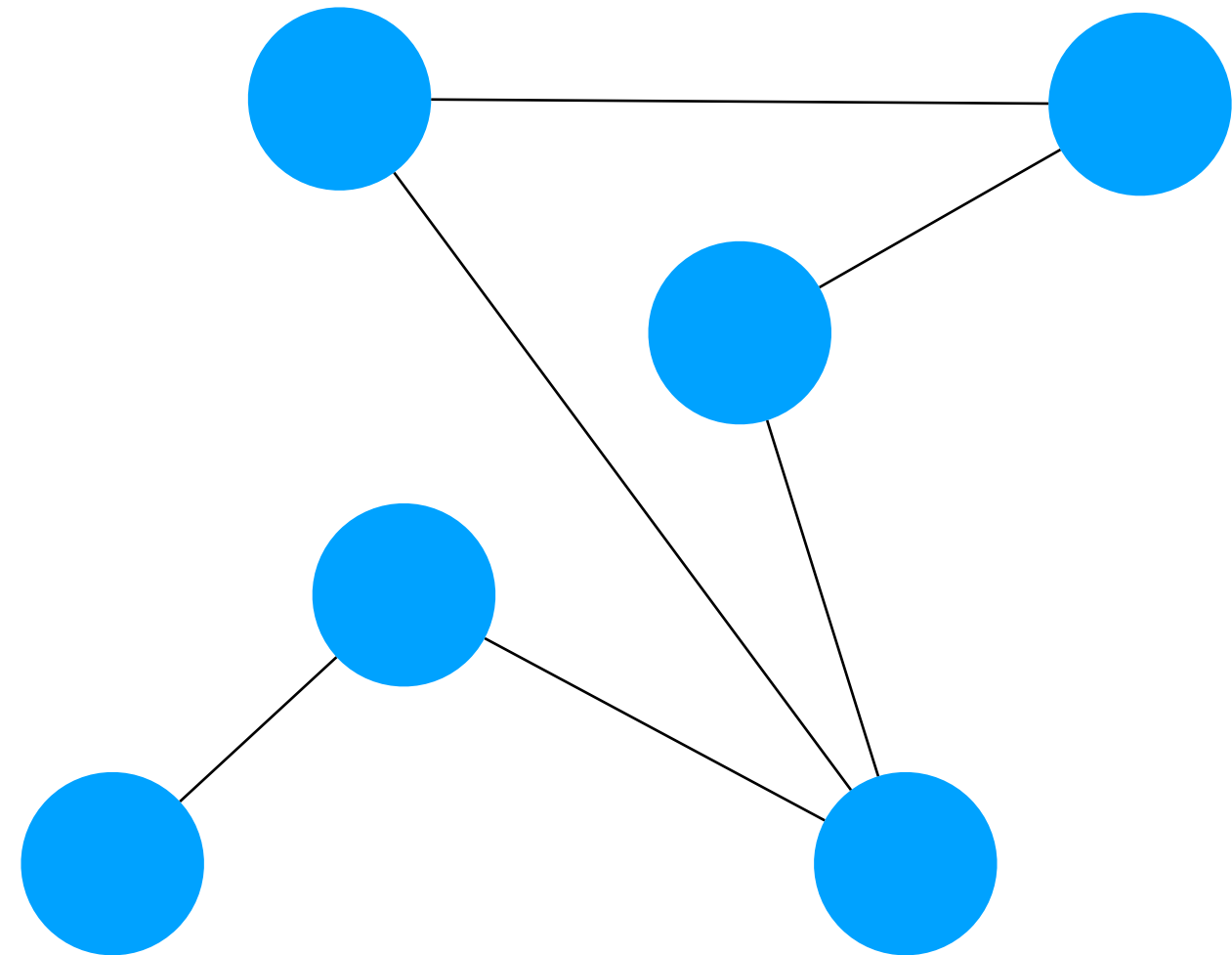


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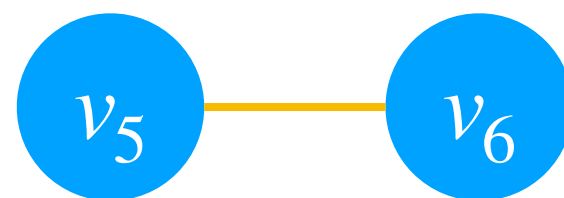


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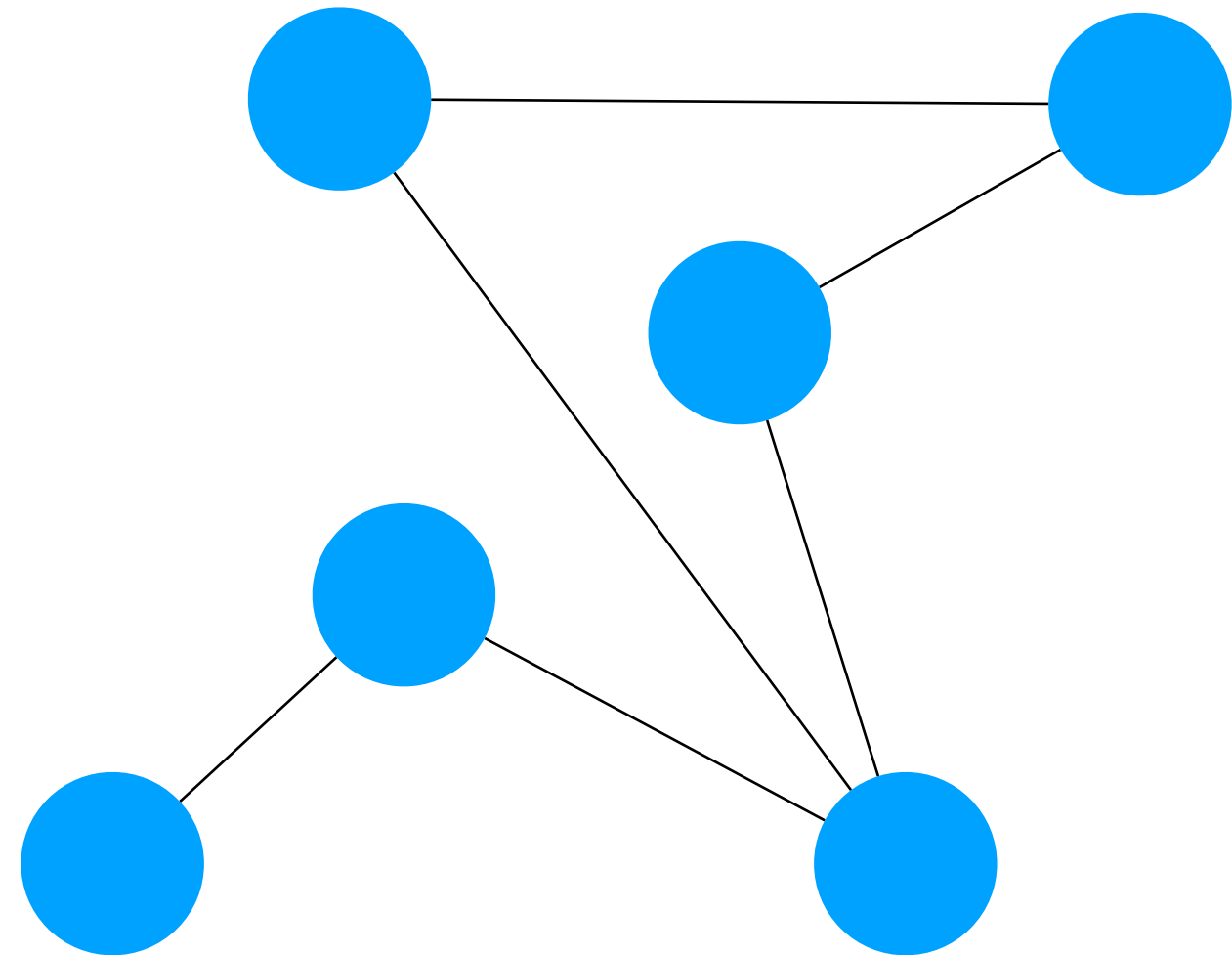


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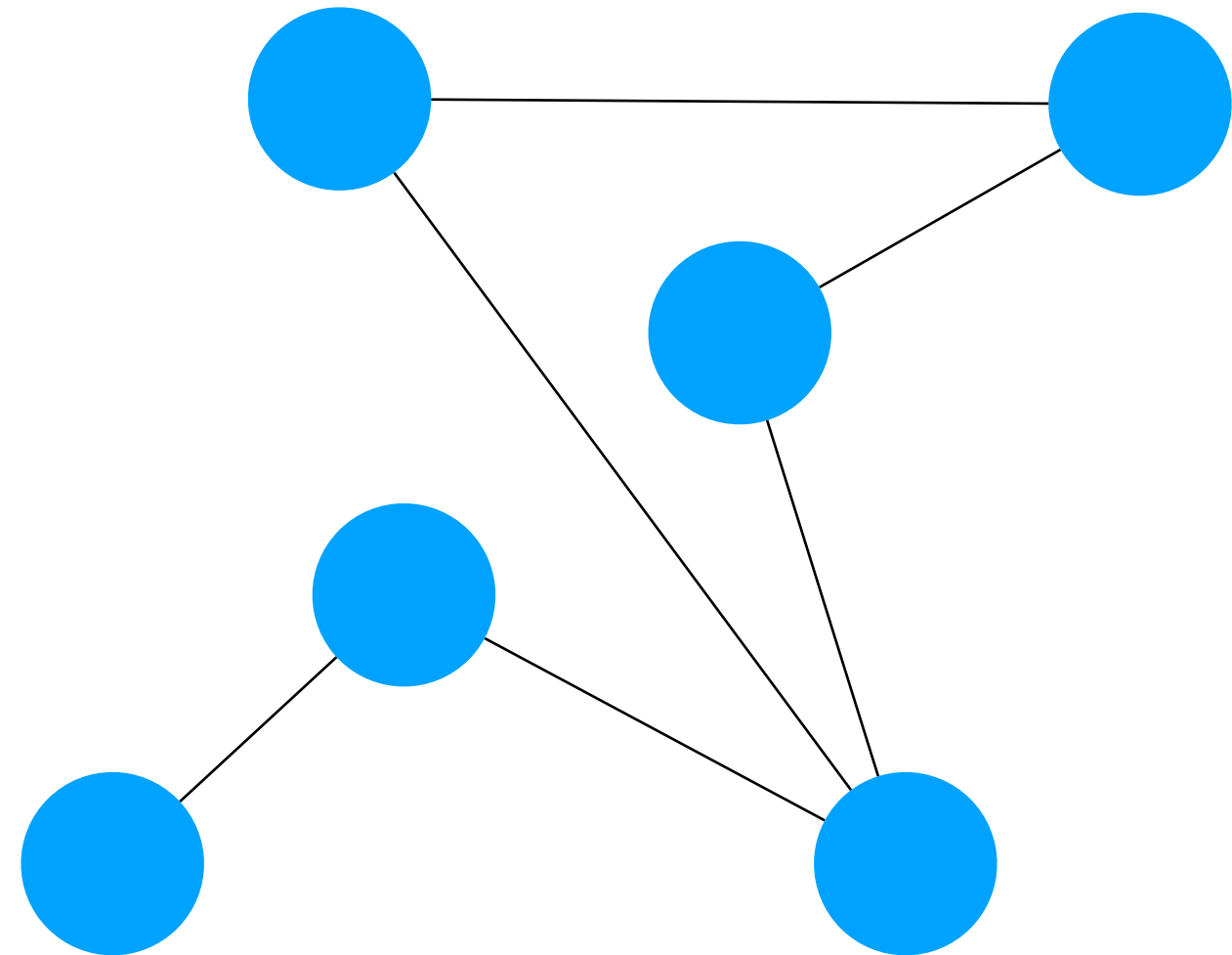


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$v_6$

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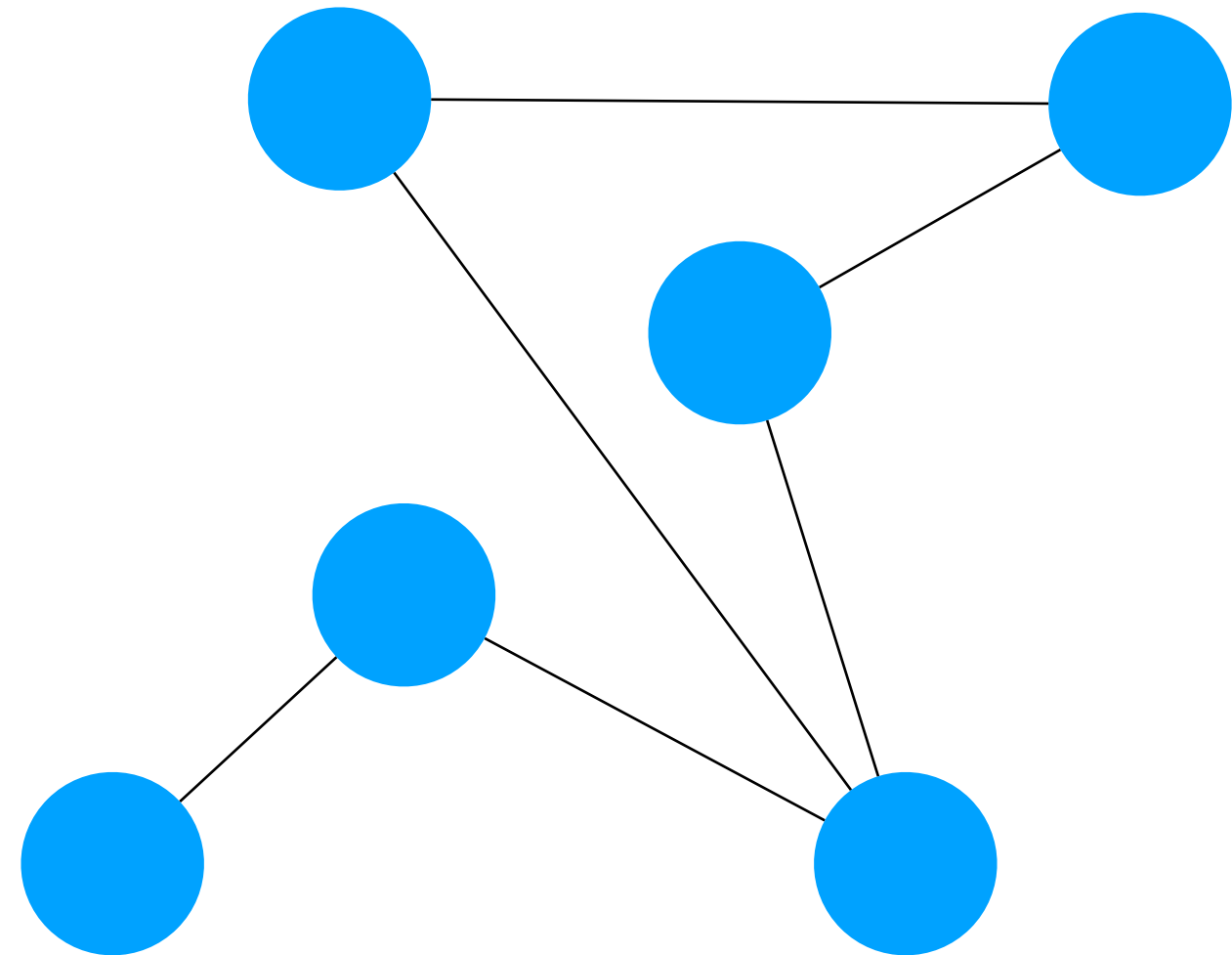


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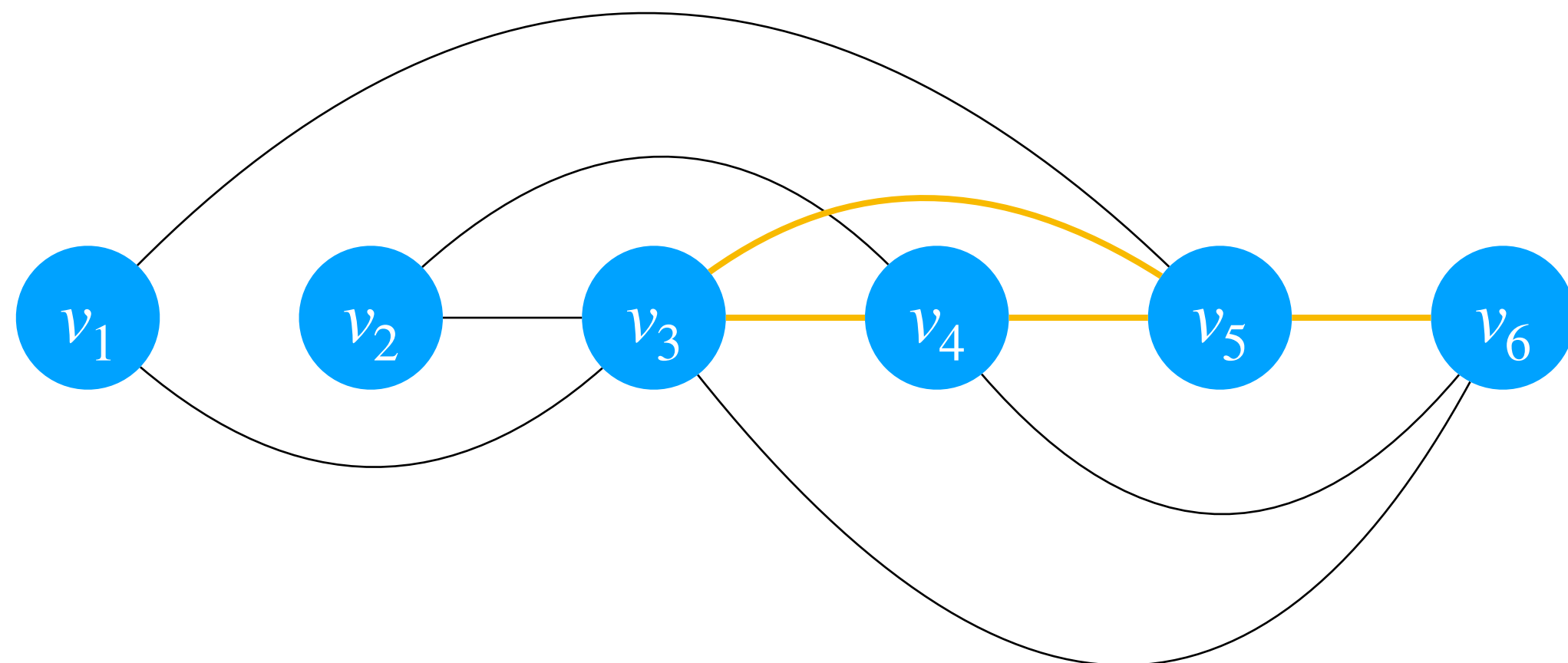


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The **width** of  $\sigma$  is the maximum degree of a vertex upon elimination.

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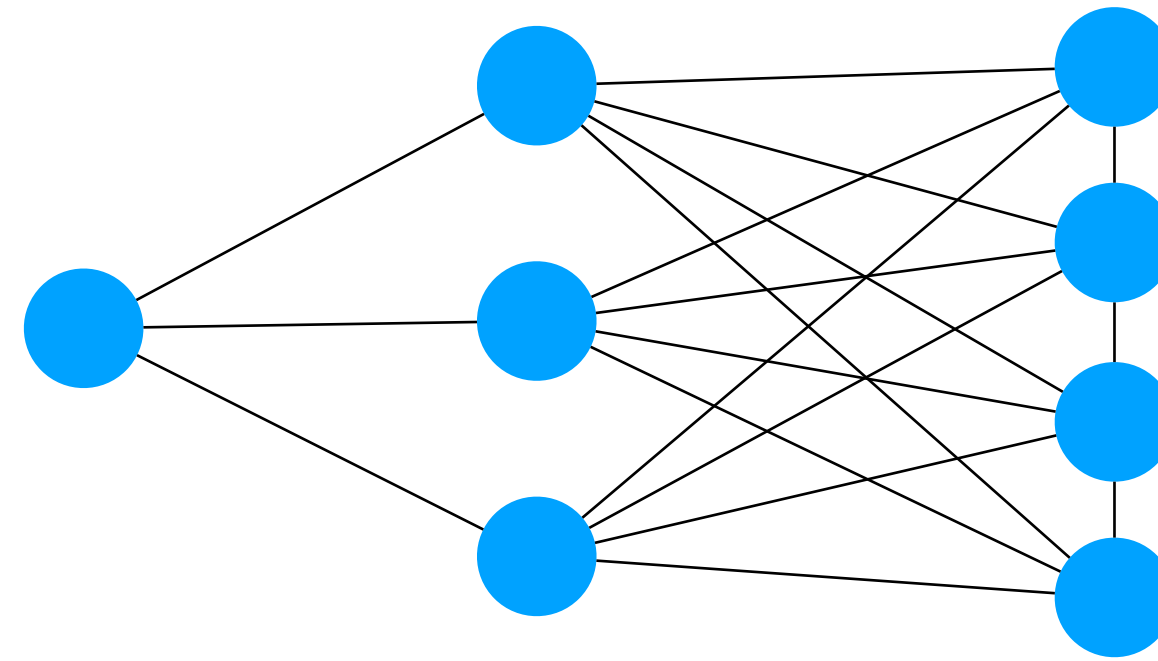
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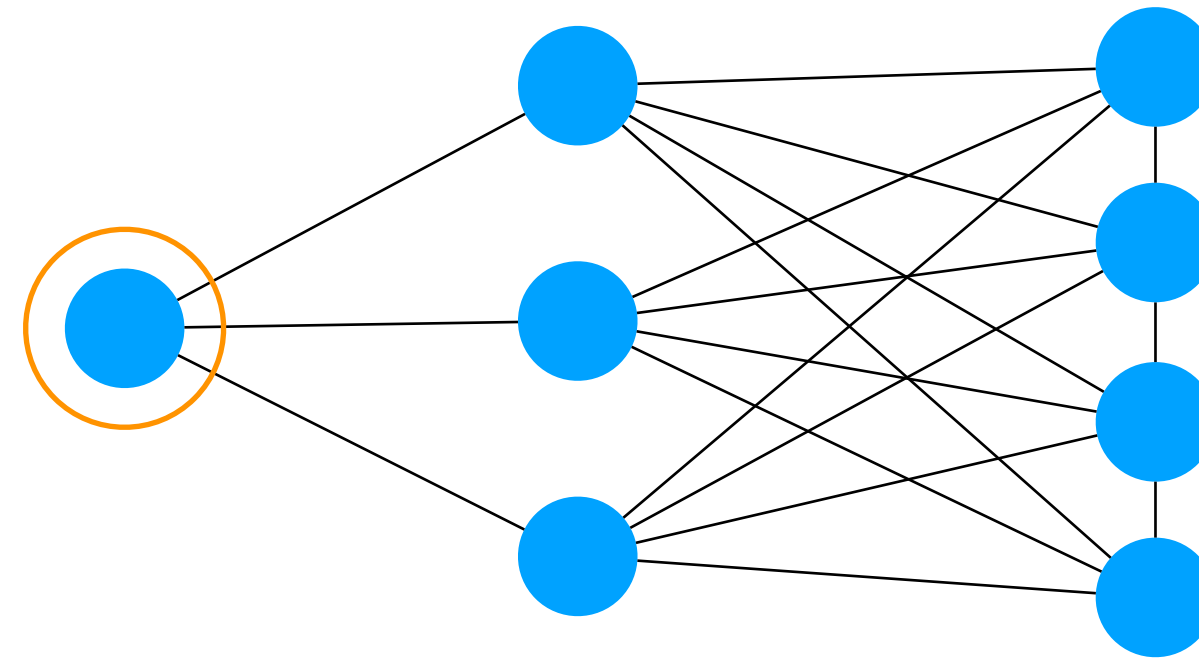
**return**  $\pi$

- 1. Min Degree/Size** minimum degree
- 2. Min Fill** fewest fill-in edges

# Min Degree and Min Fill

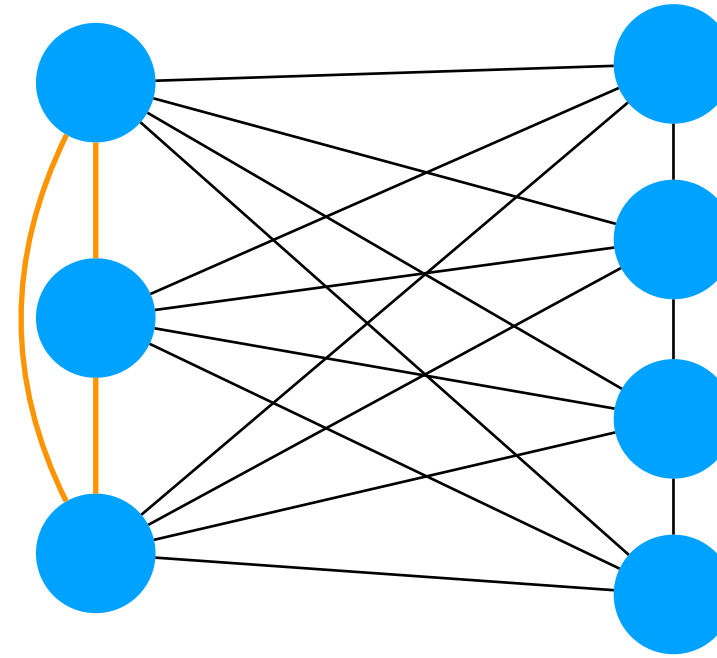


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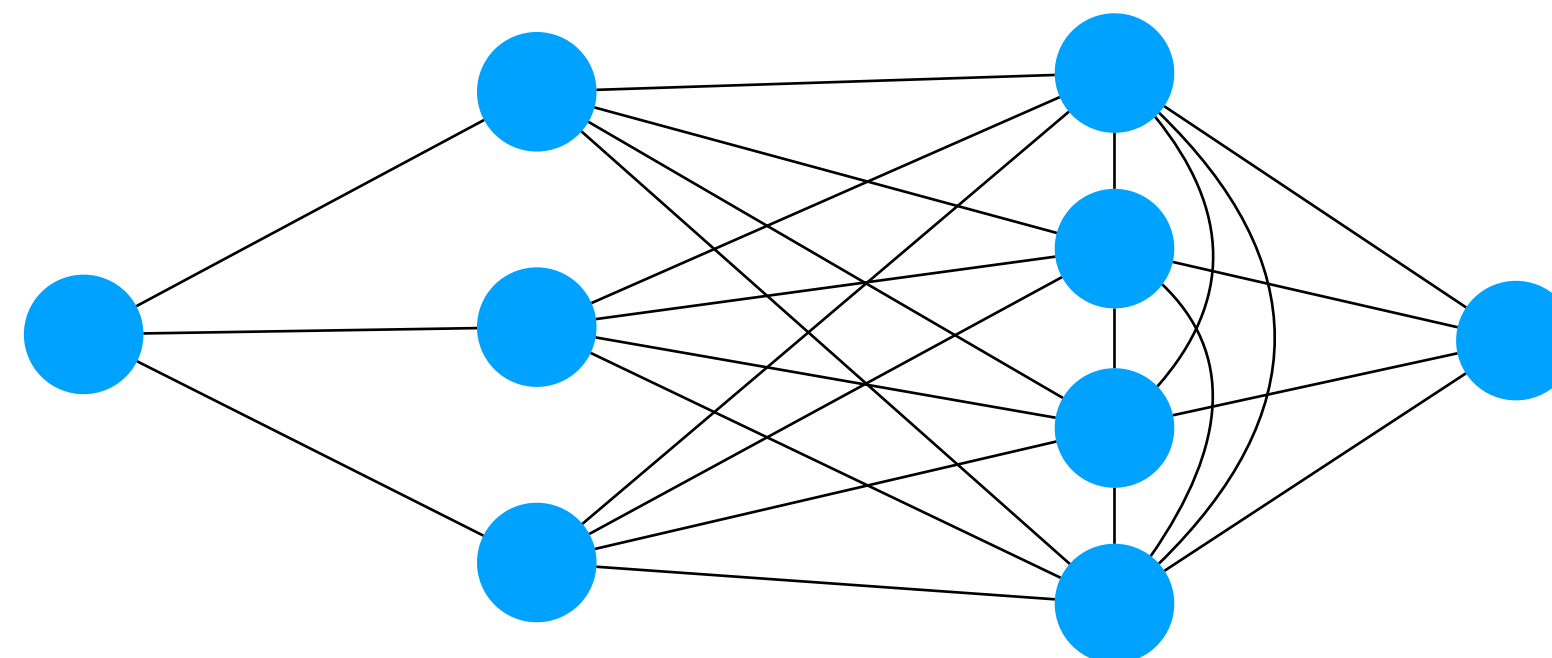
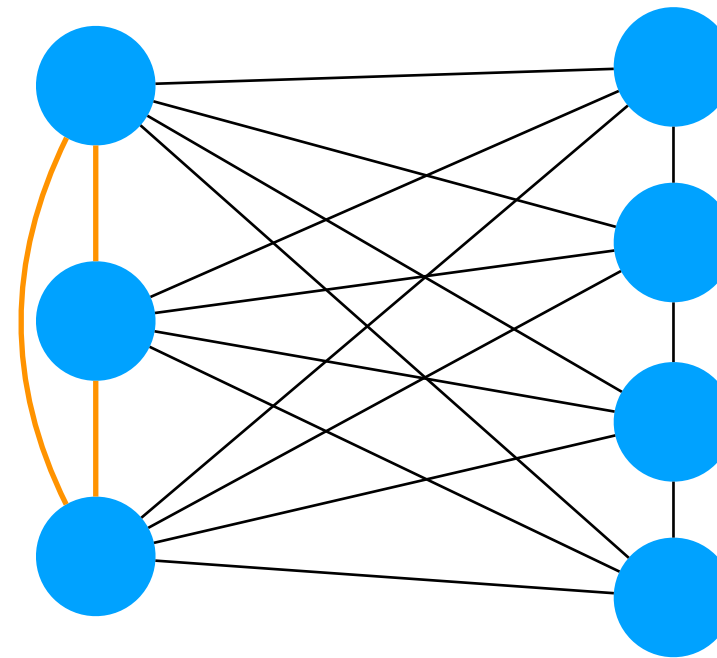




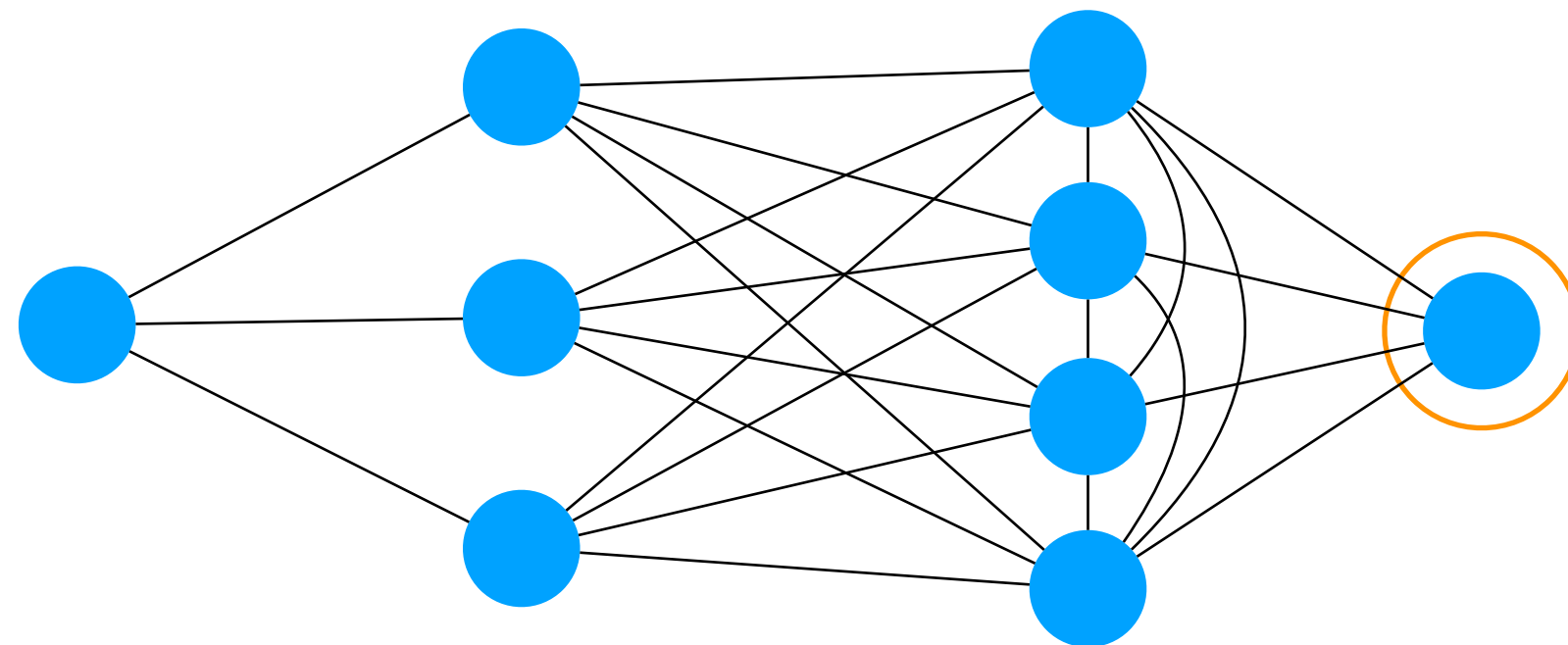
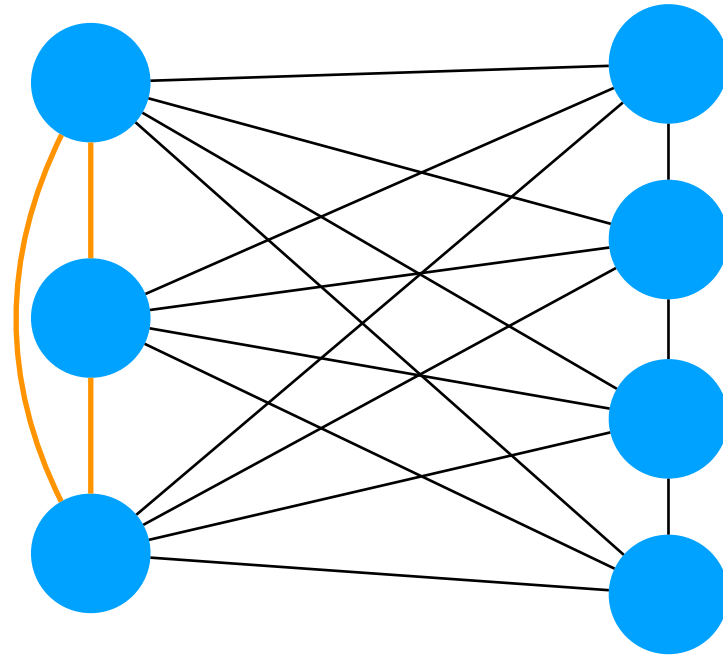
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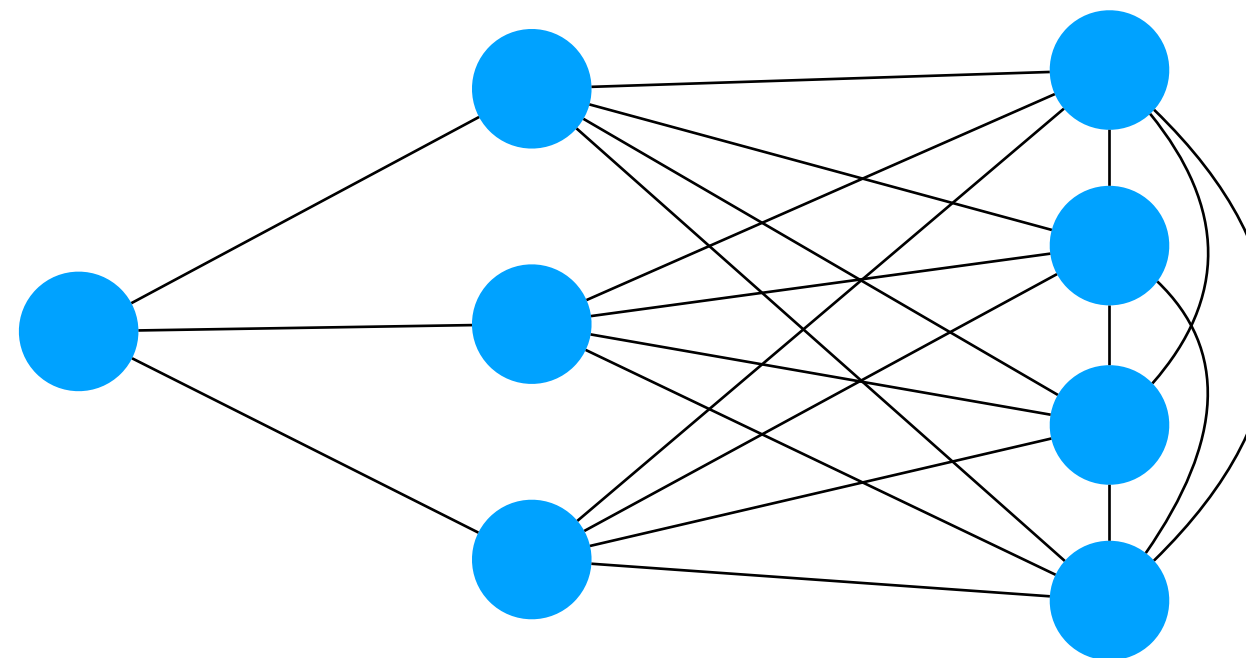
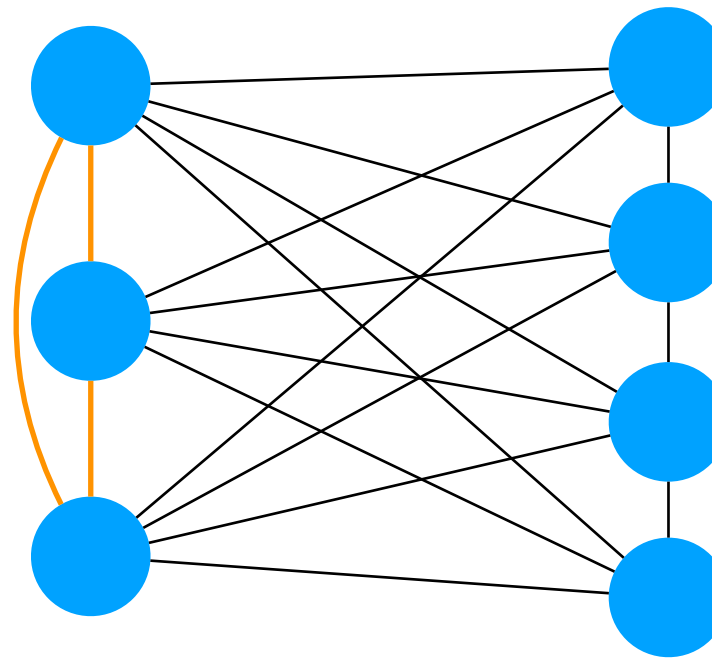
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$\pi := v, \pi$

**return**  $\pi$

# Heuristics in Practice

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- **Min-Fill** generates better (in terms of width) orderings.
- Can generate good elimination orderings for large graphs.

# Heuristics in Practice

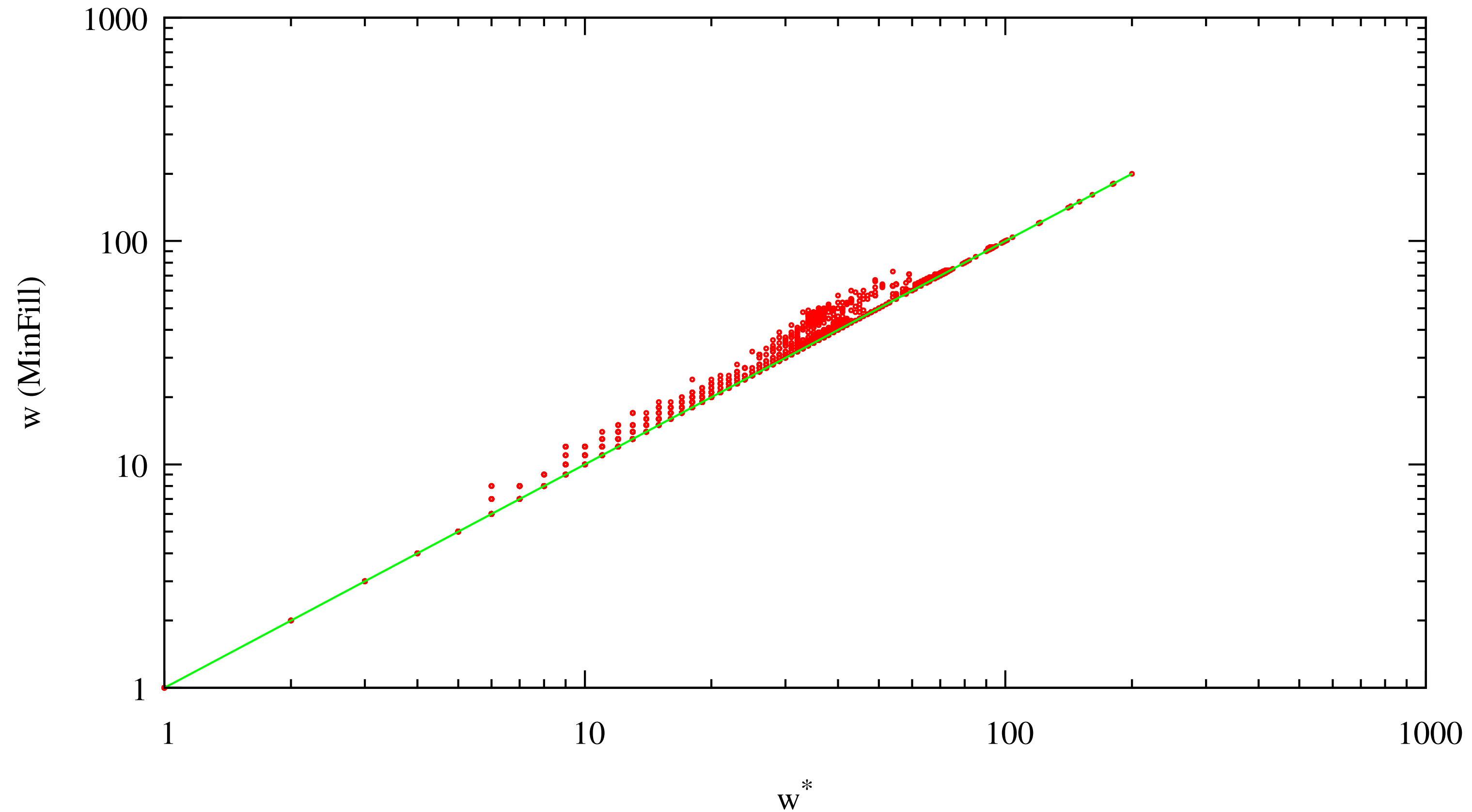


Fig. 2. The width obtained with MinFill vs the treewidth.

**Jegou et al. ICTAI 2018**