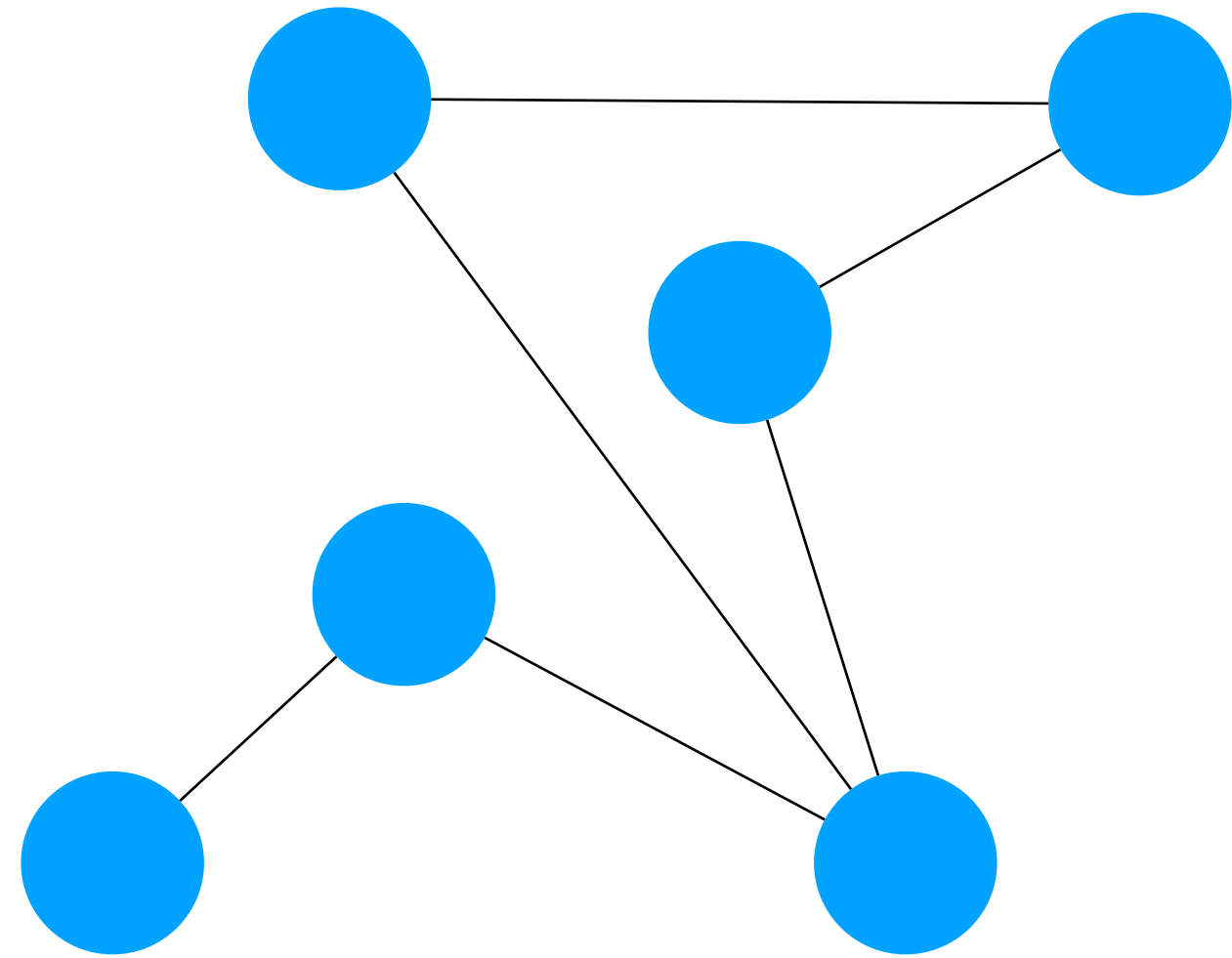


# Algorithmics Part 4

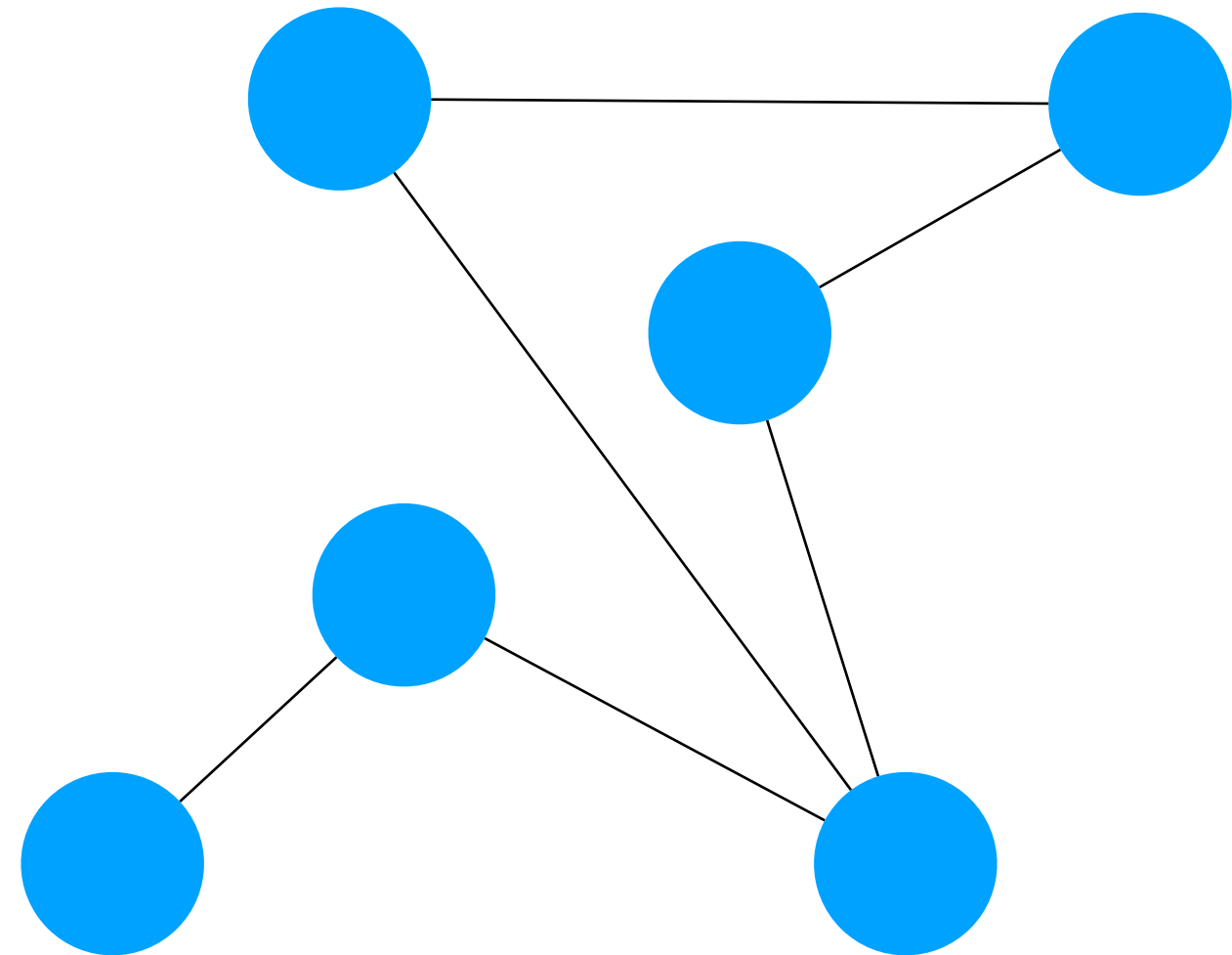
# Structural Decompositions and Algorithms

Friedrich Slivovsky

# Elimination Orderings



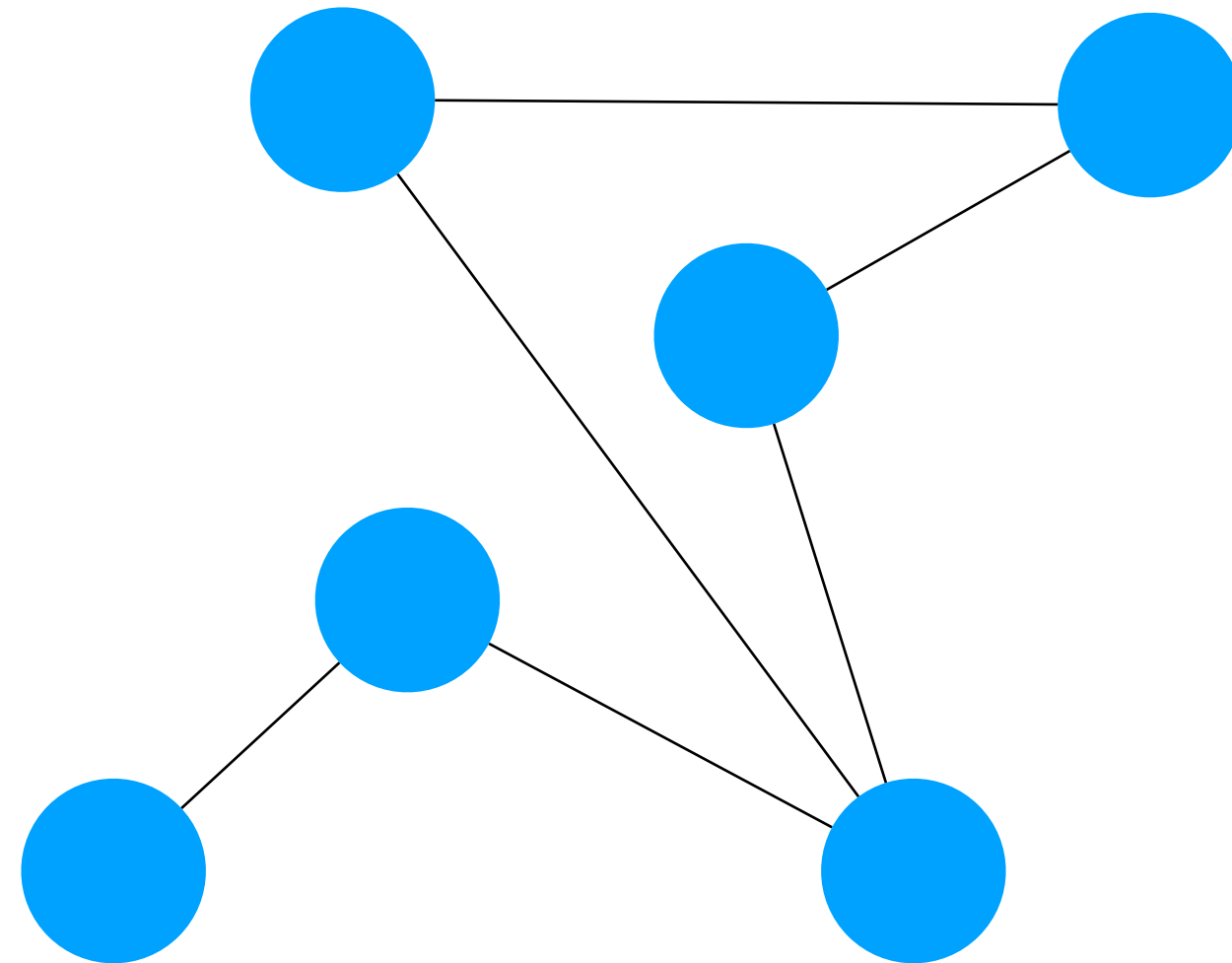
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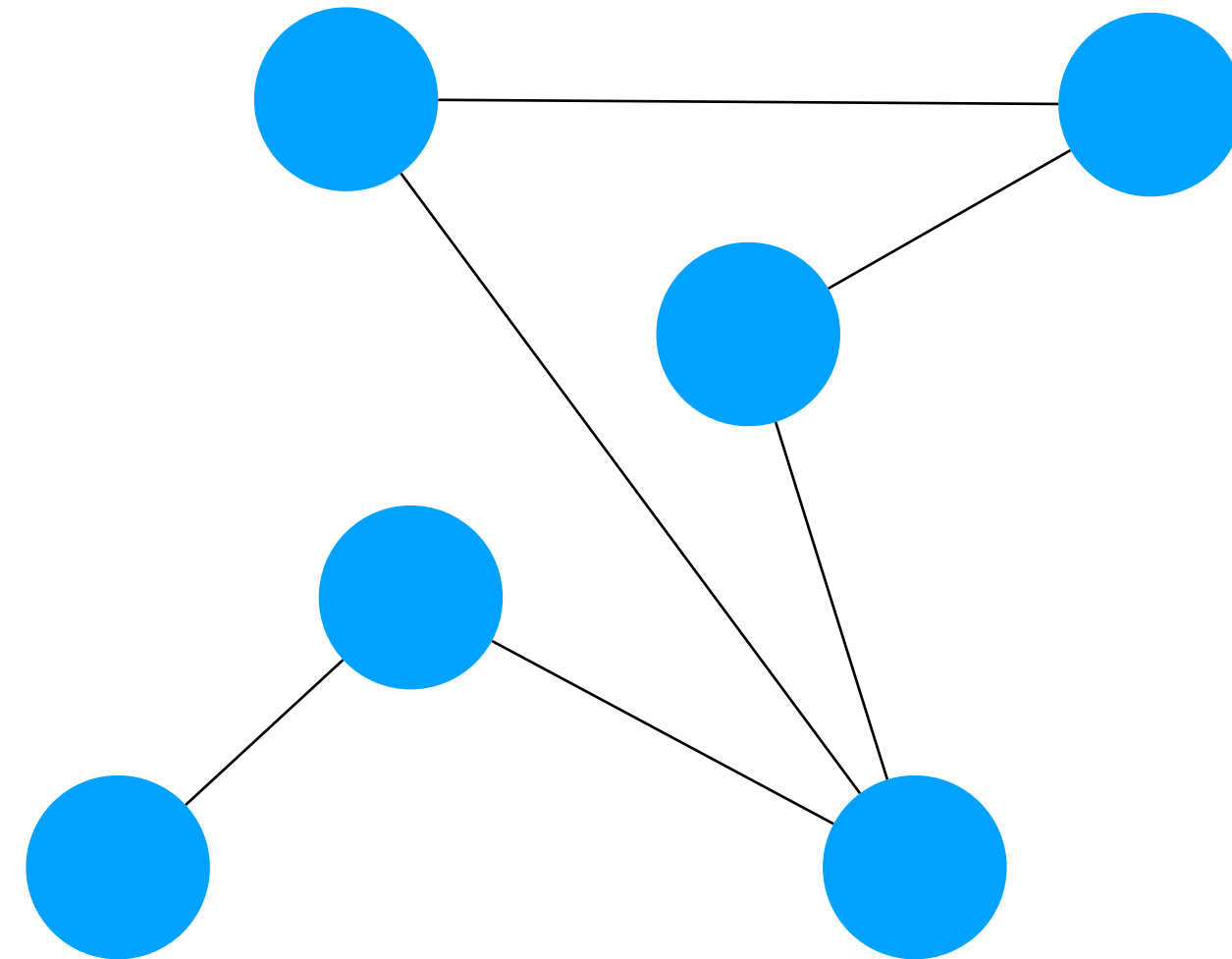


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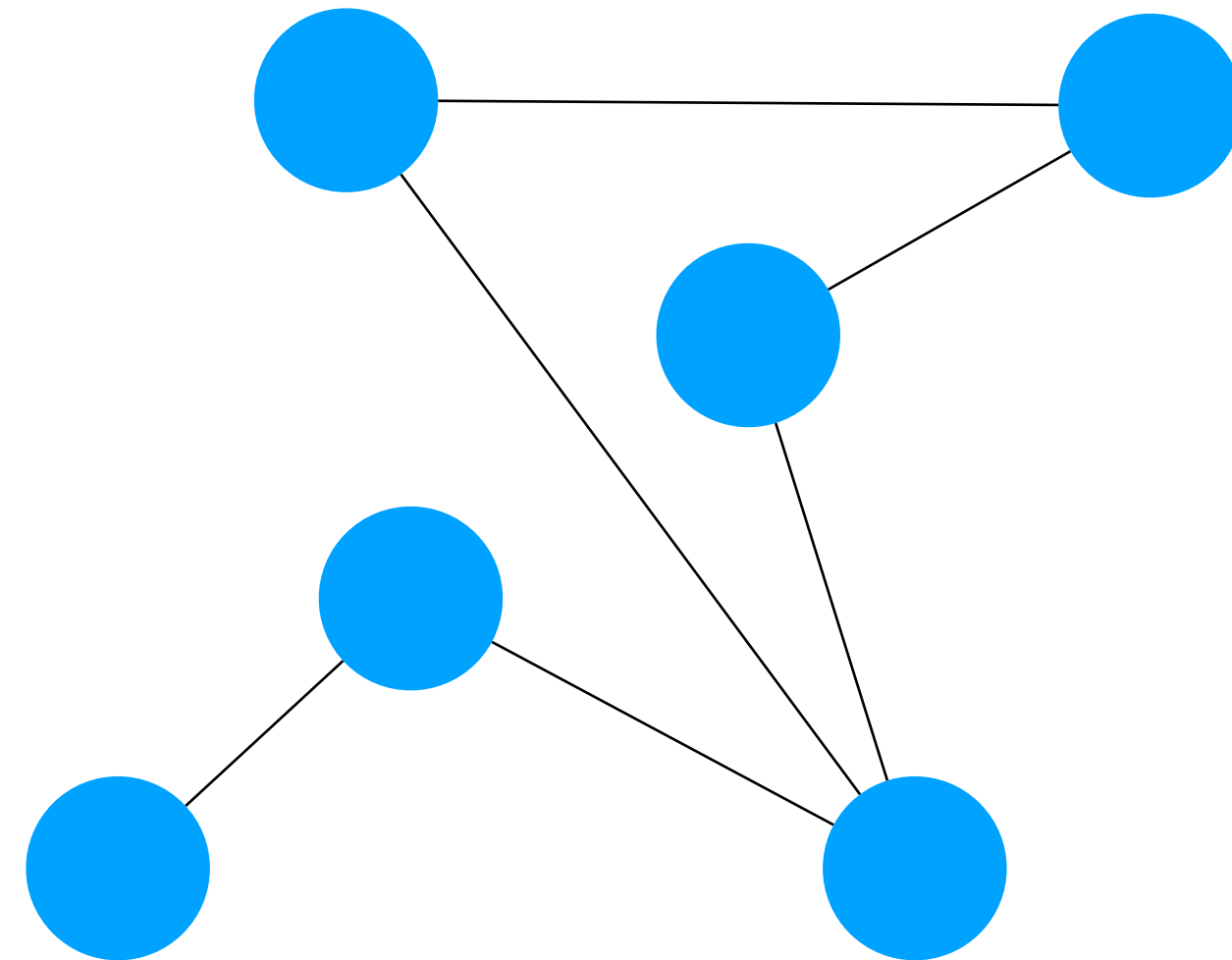
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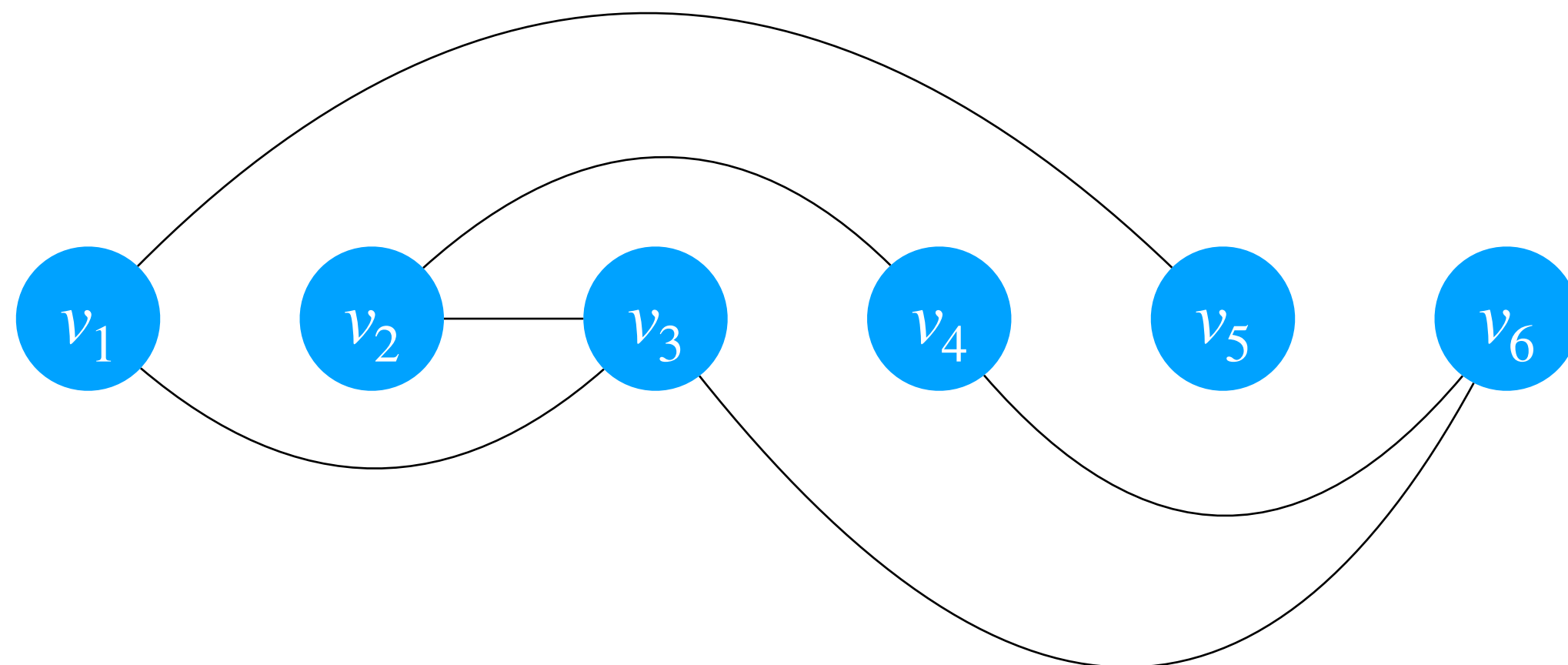
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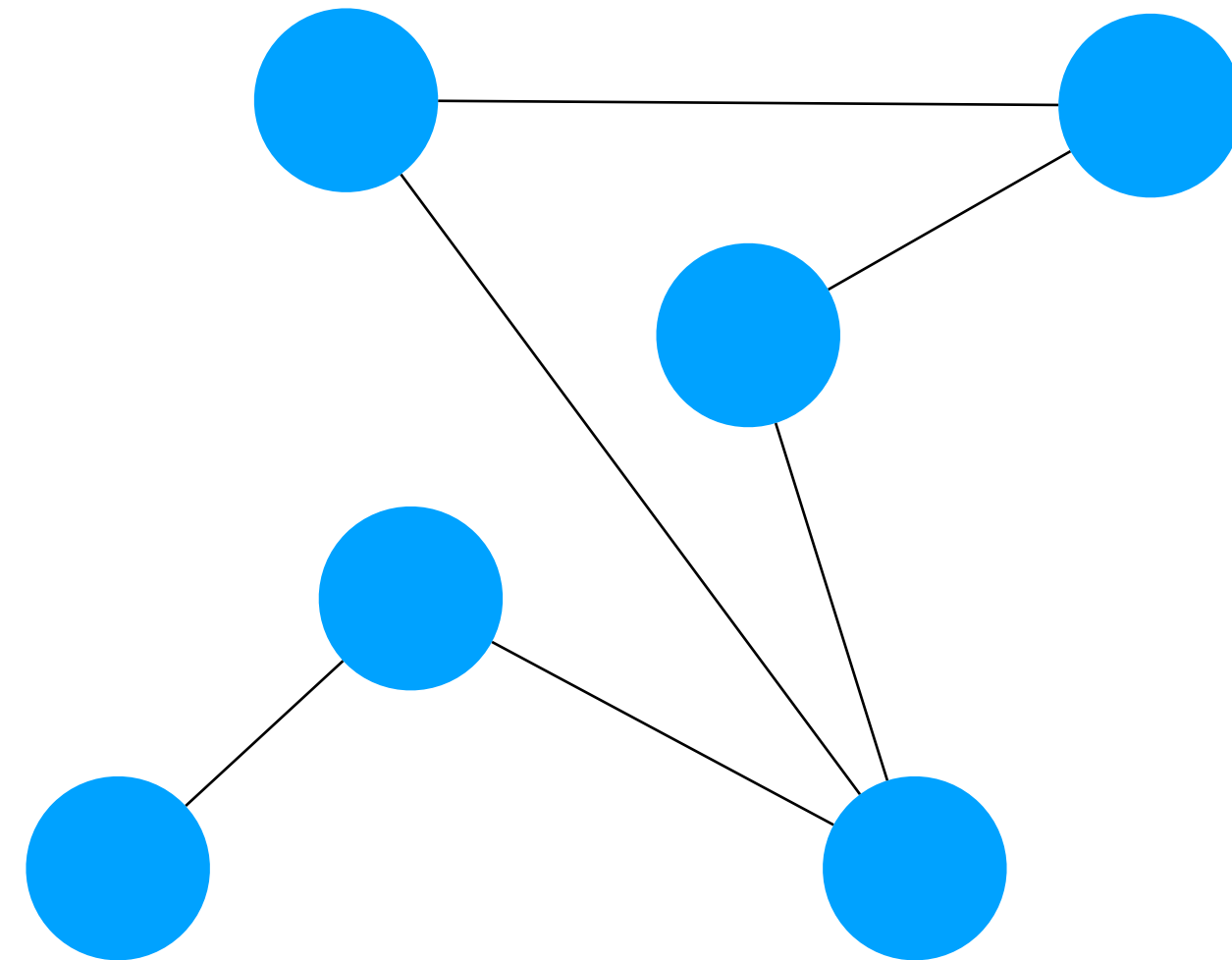
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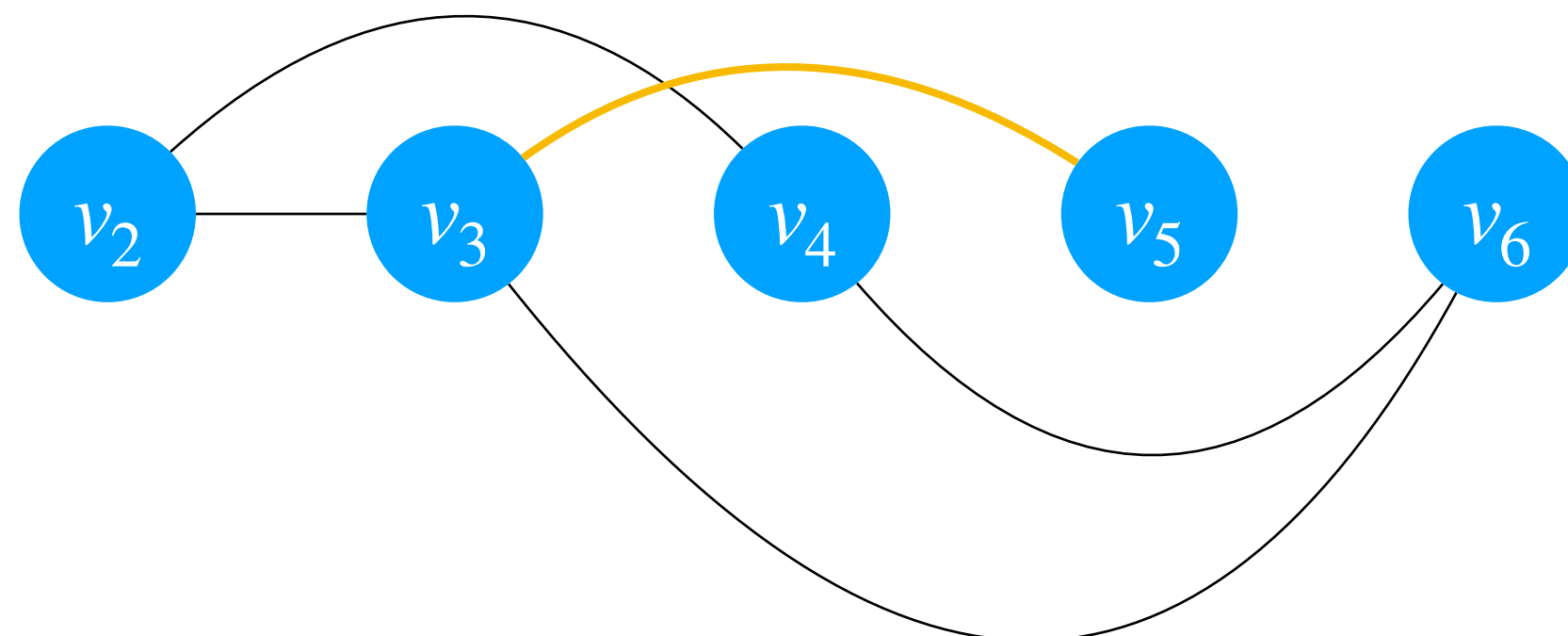
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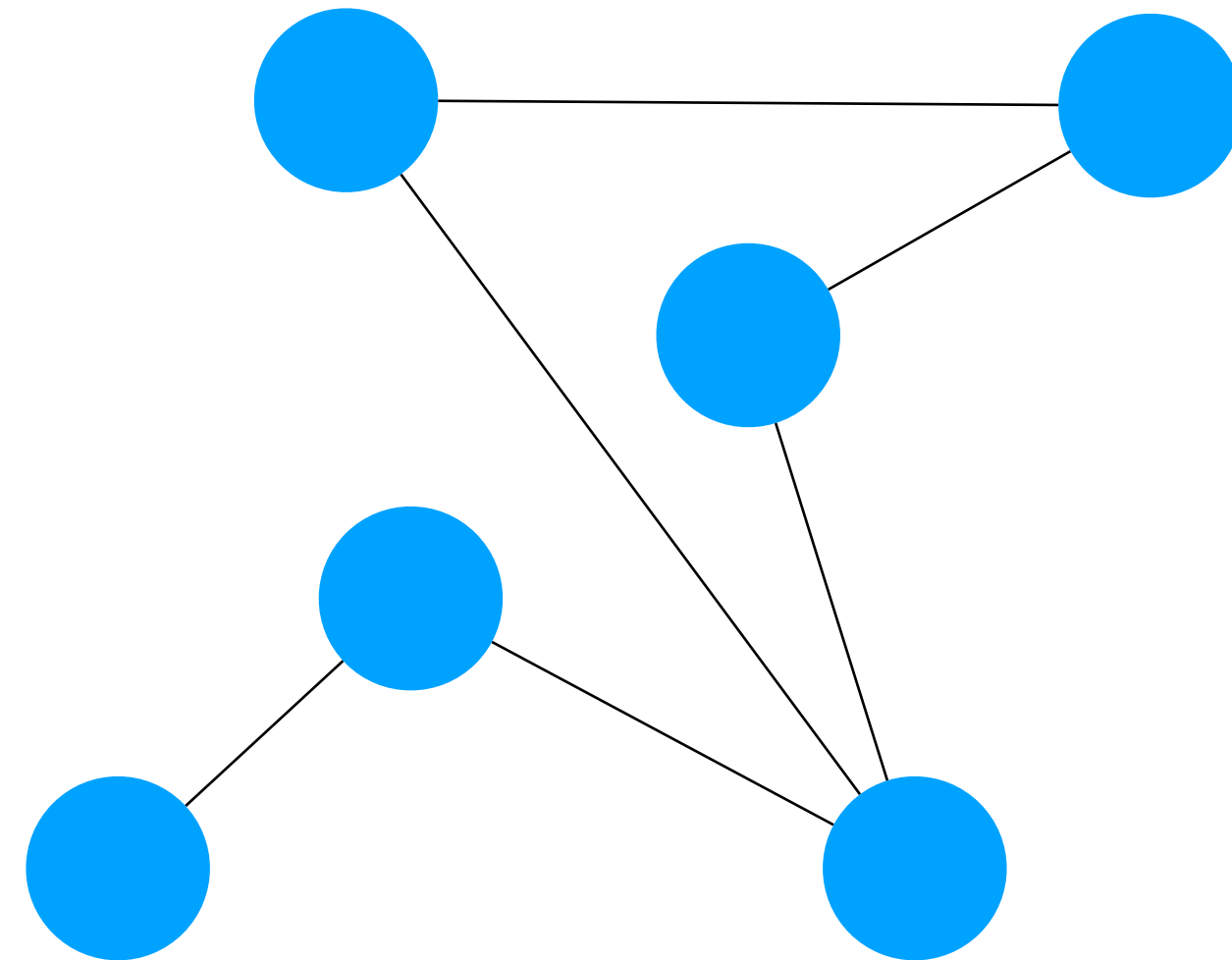
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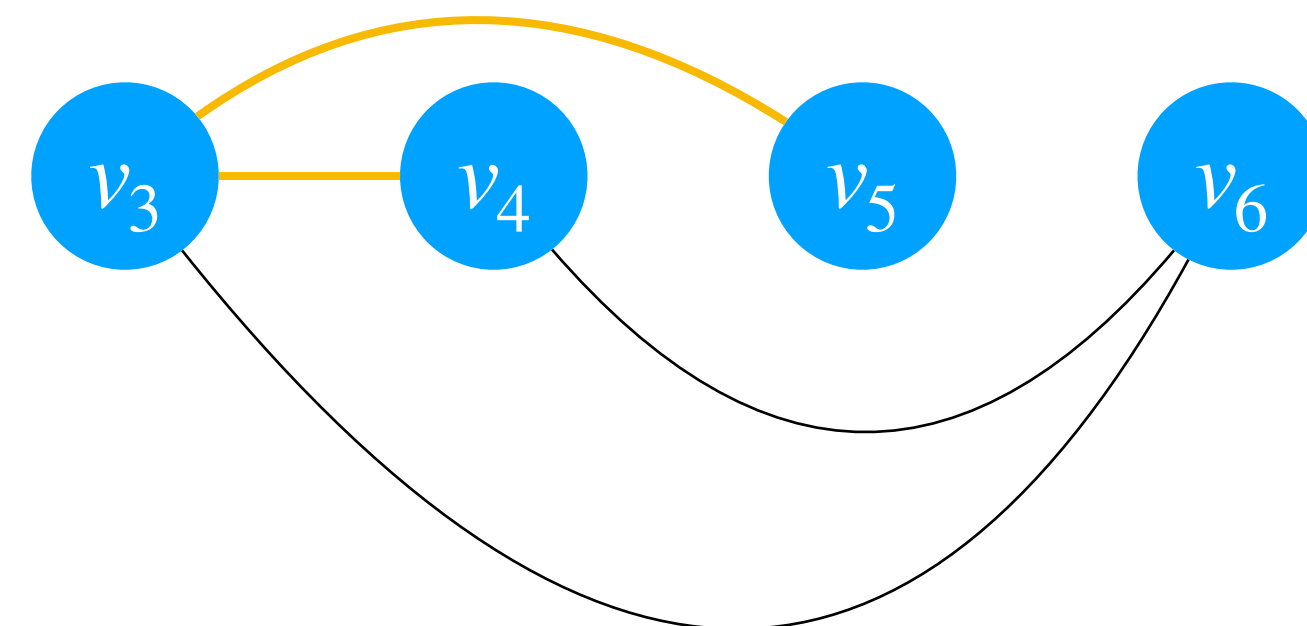
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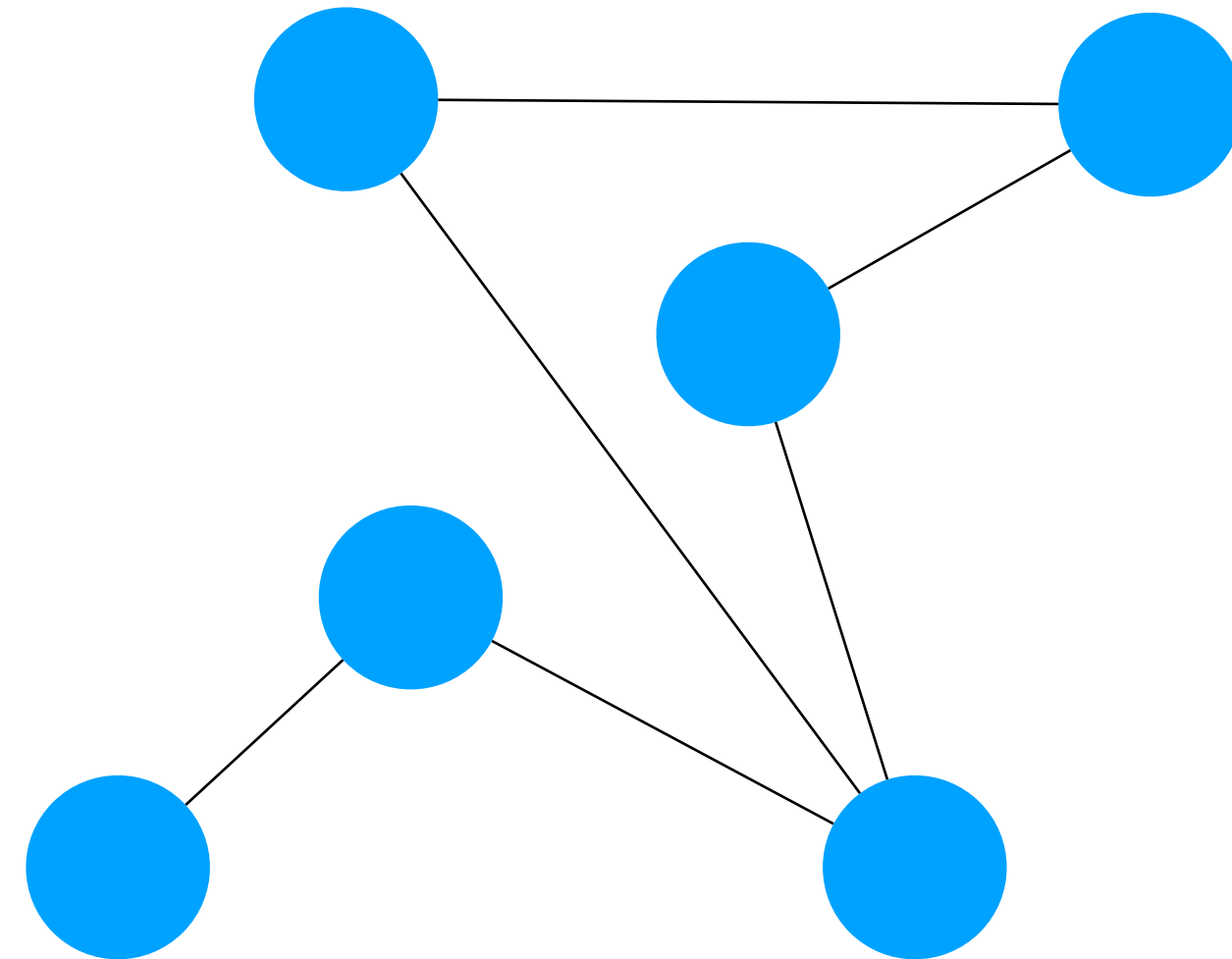
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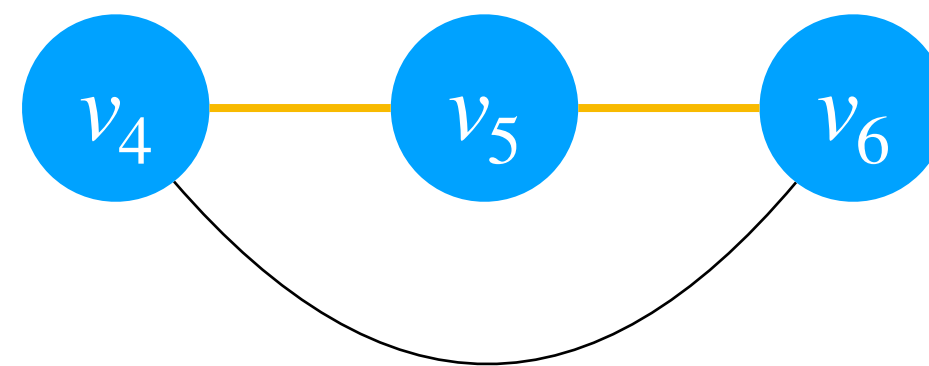
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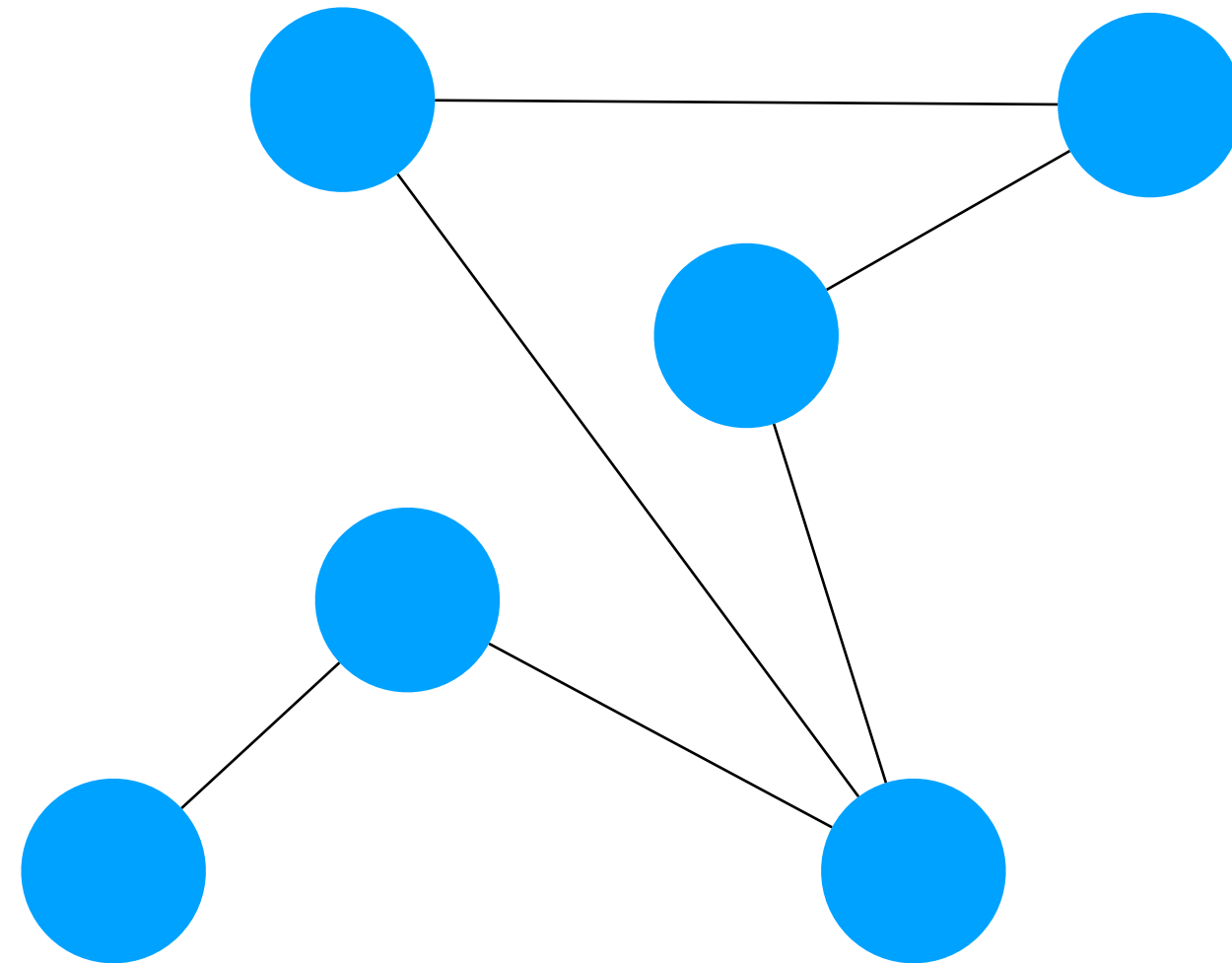
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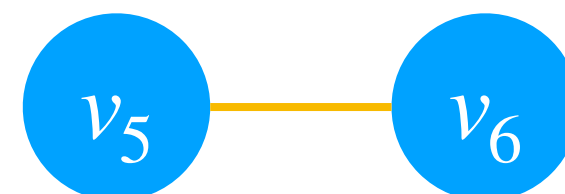
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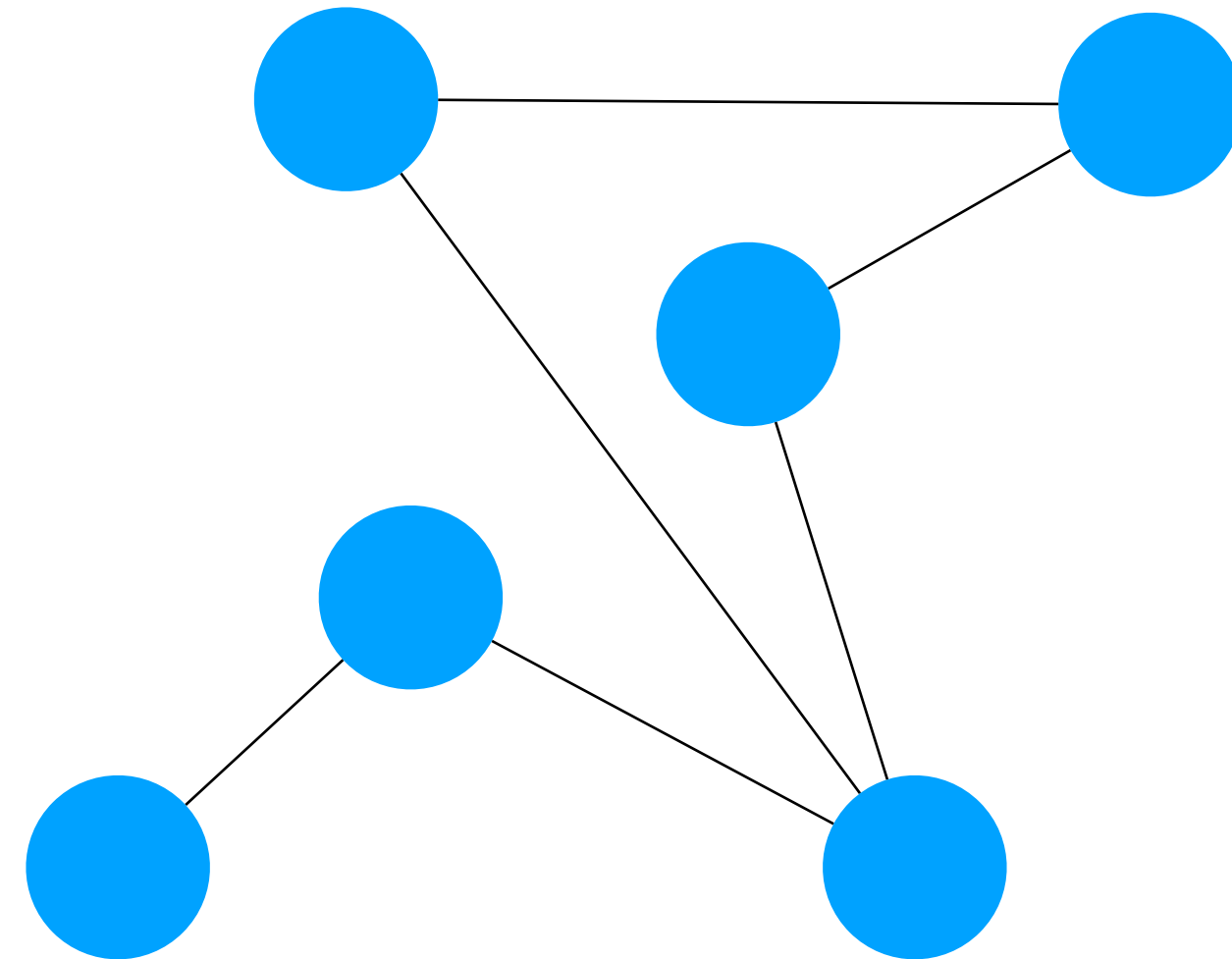
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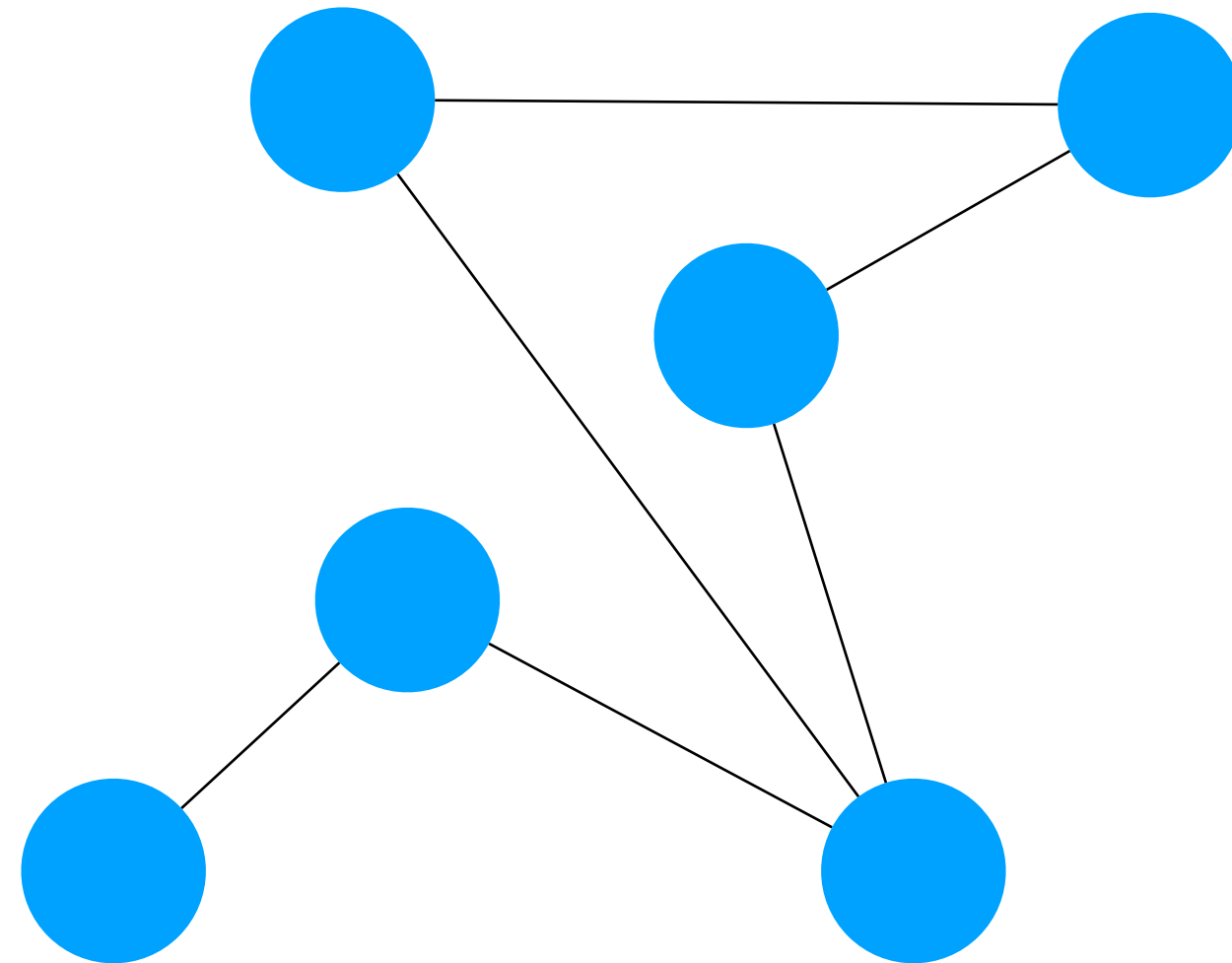
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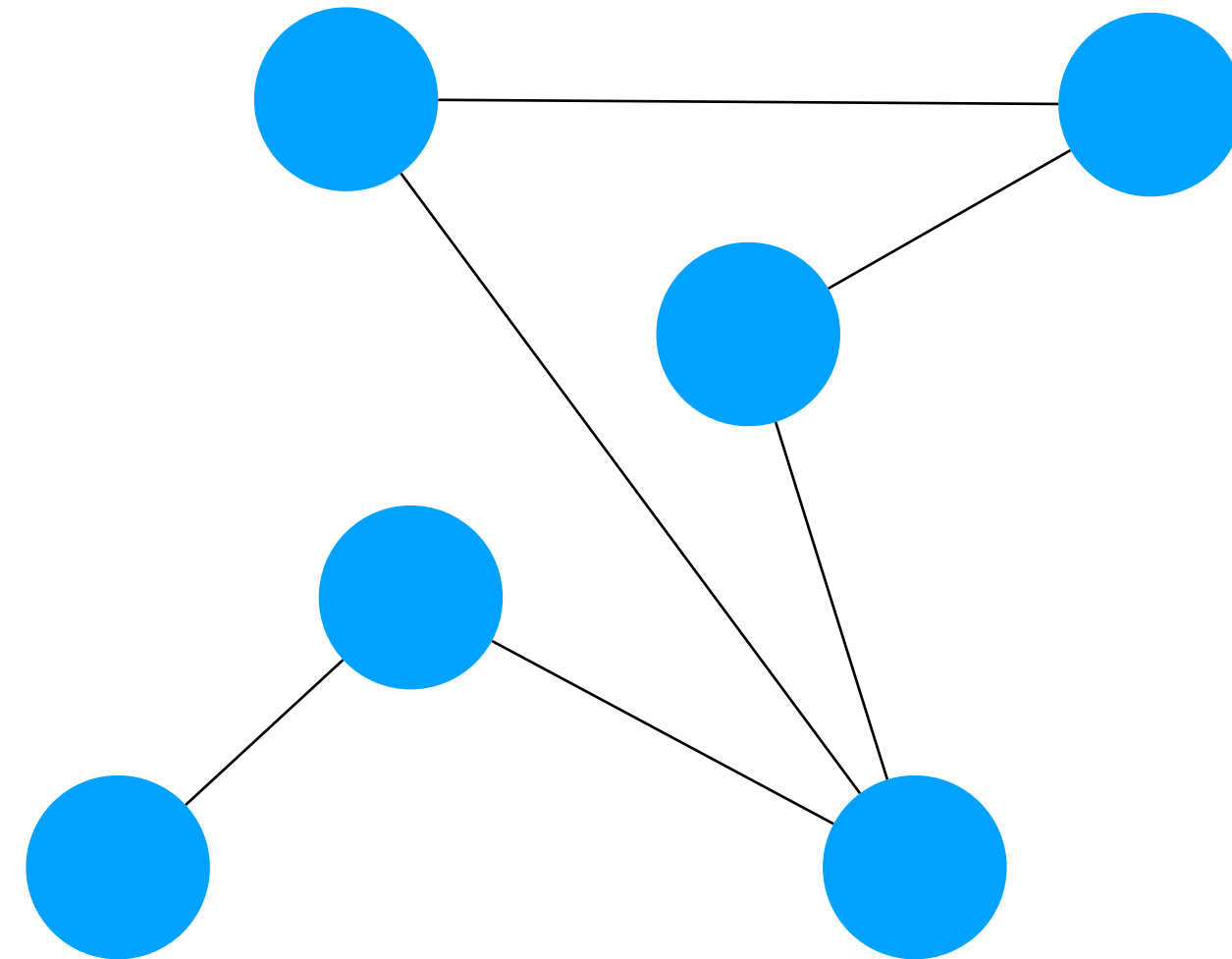


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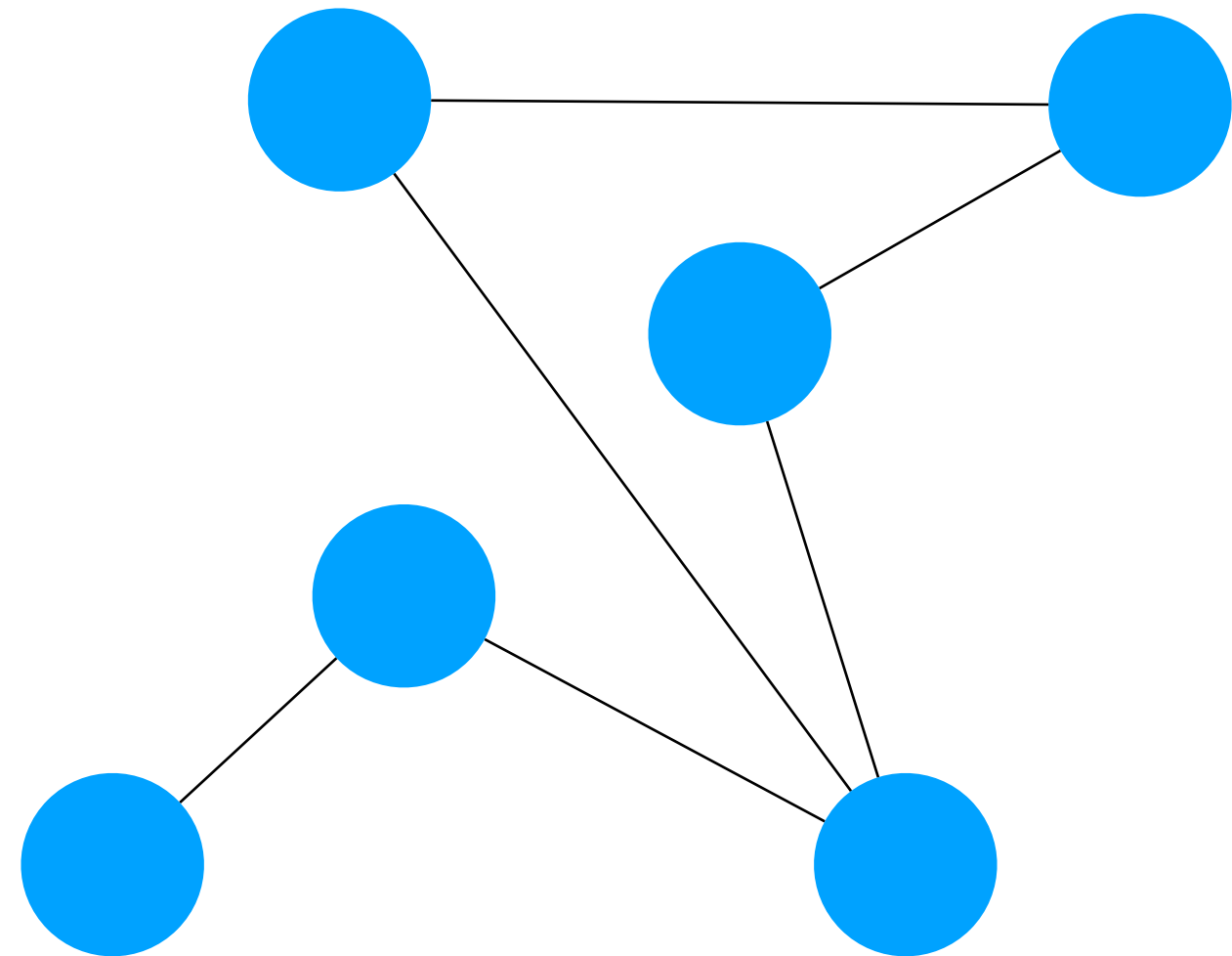
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fill-in edges

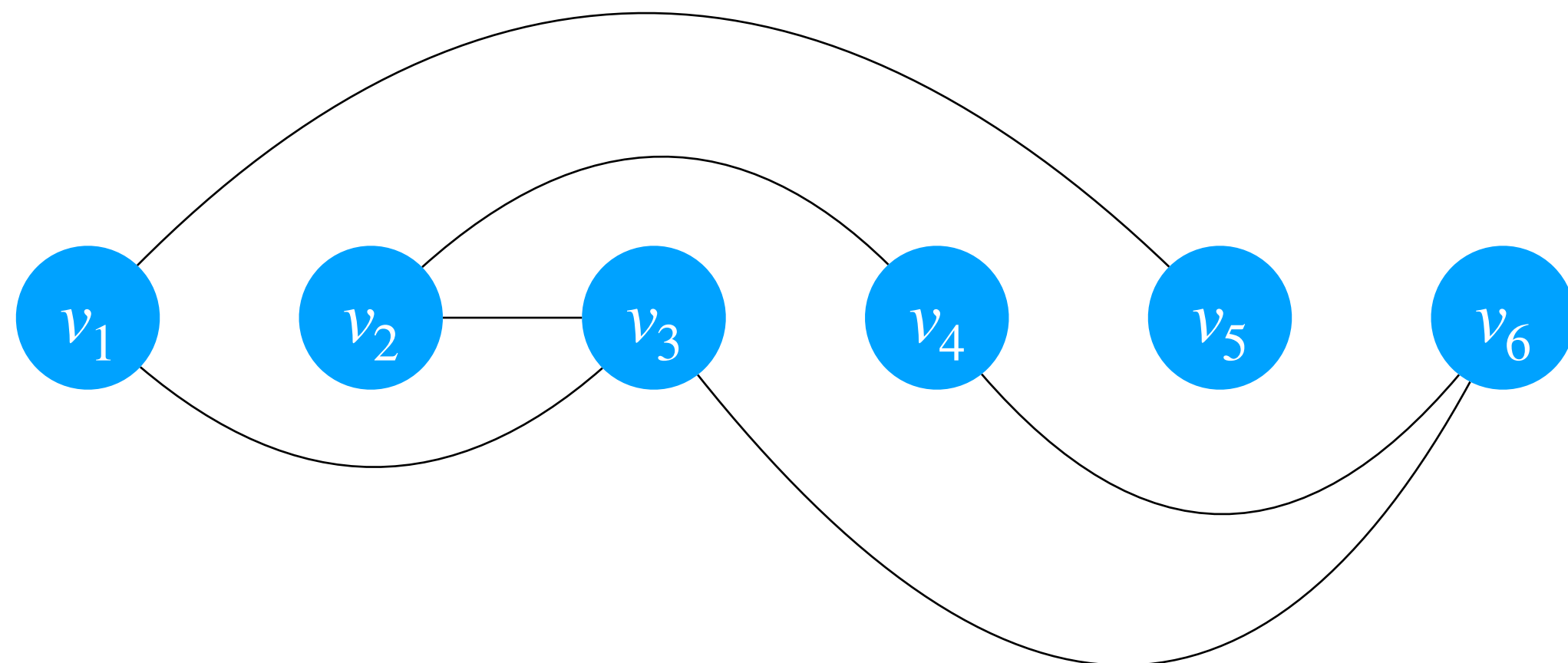
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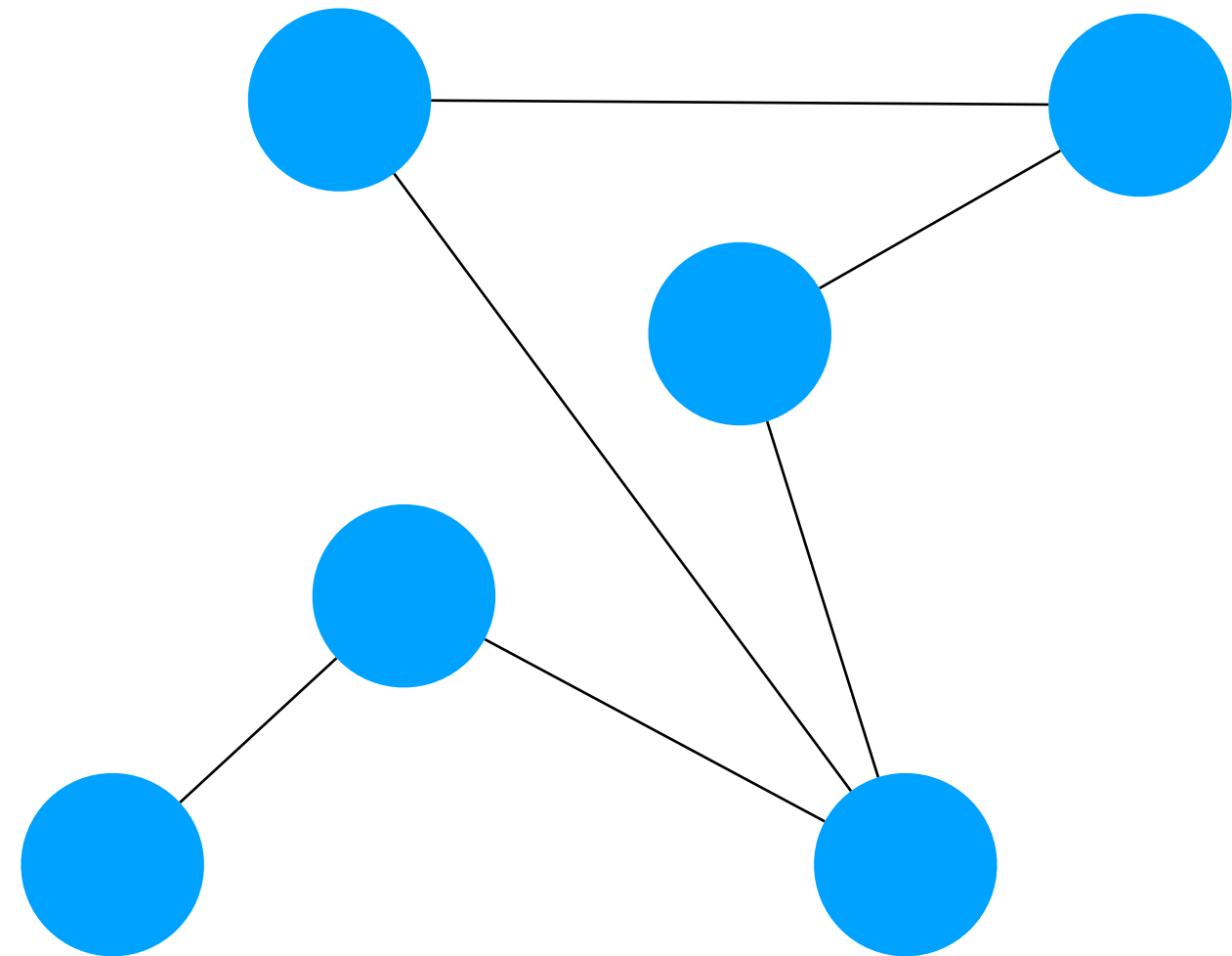
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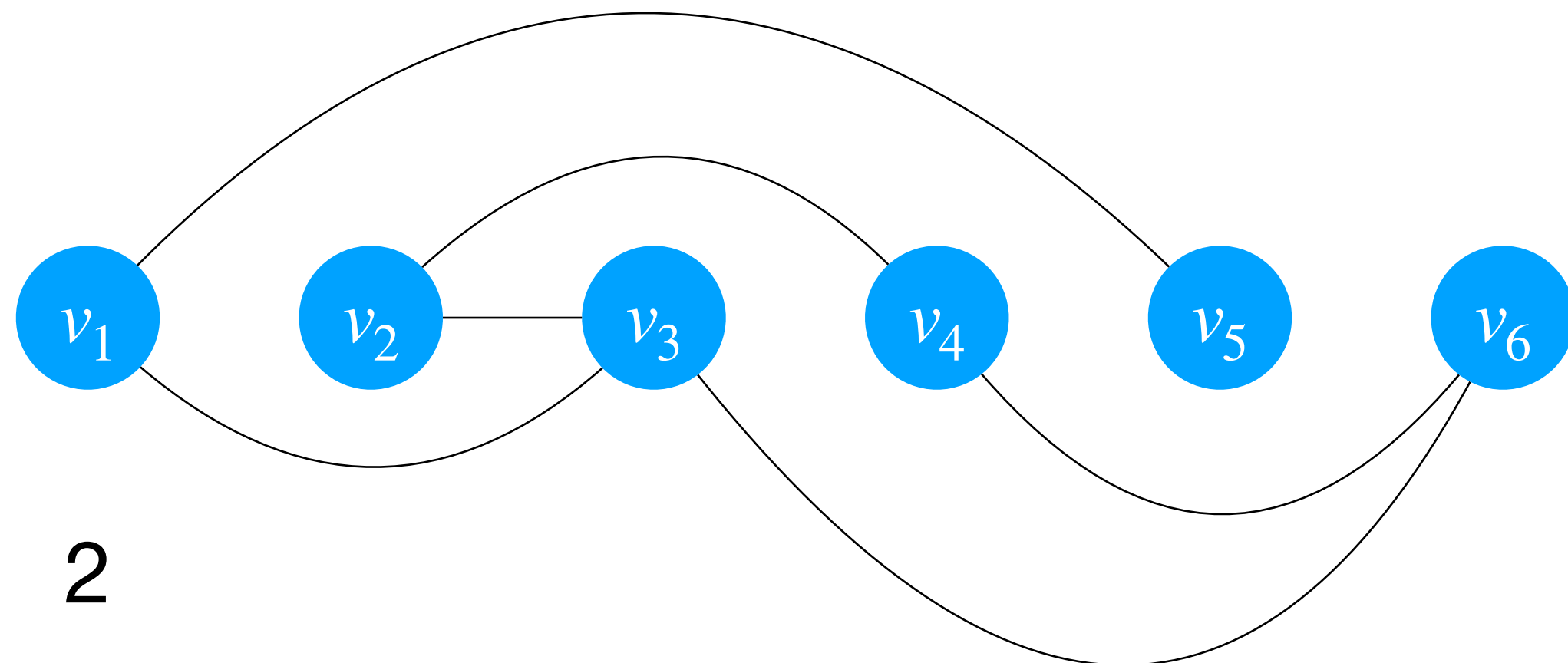
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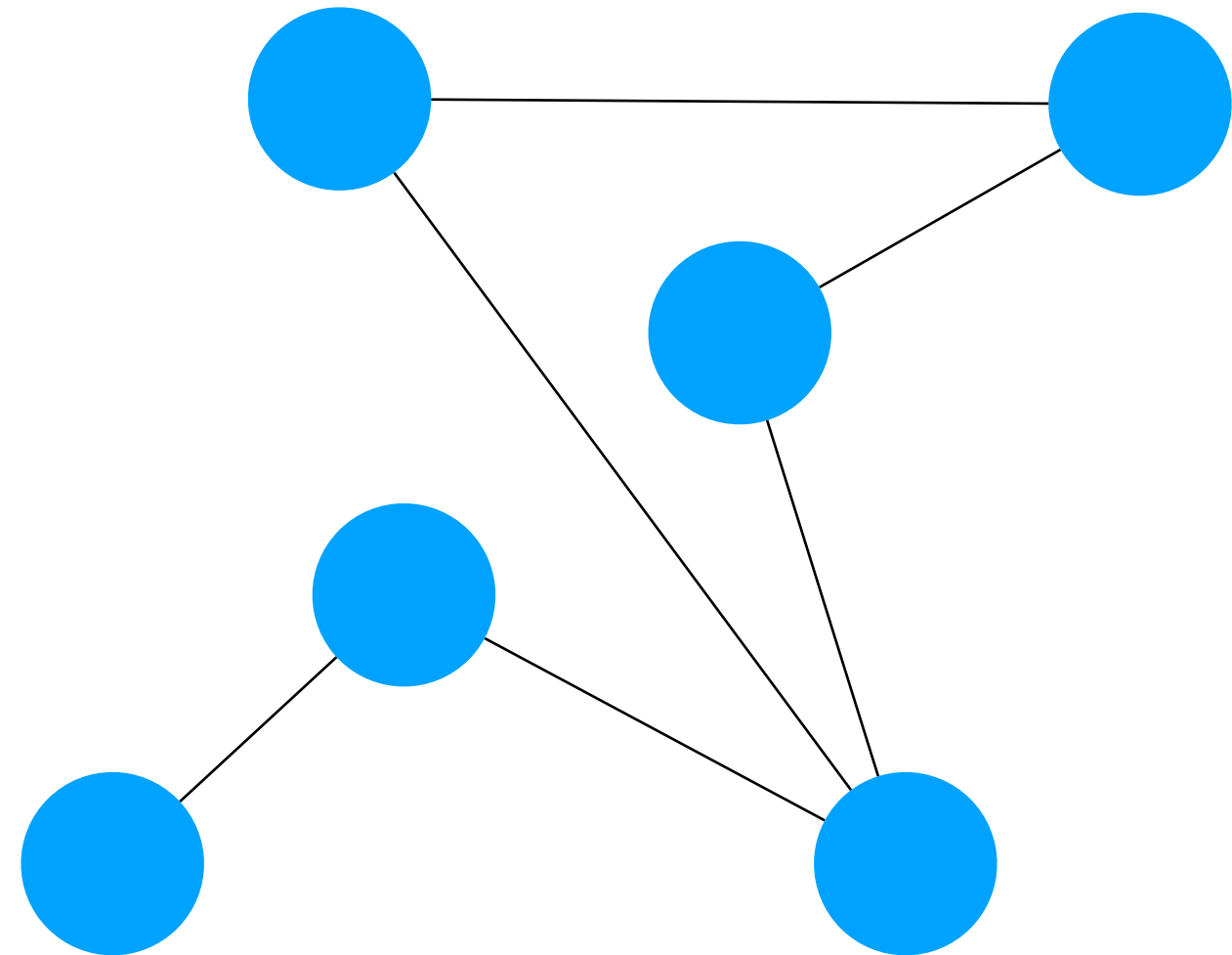
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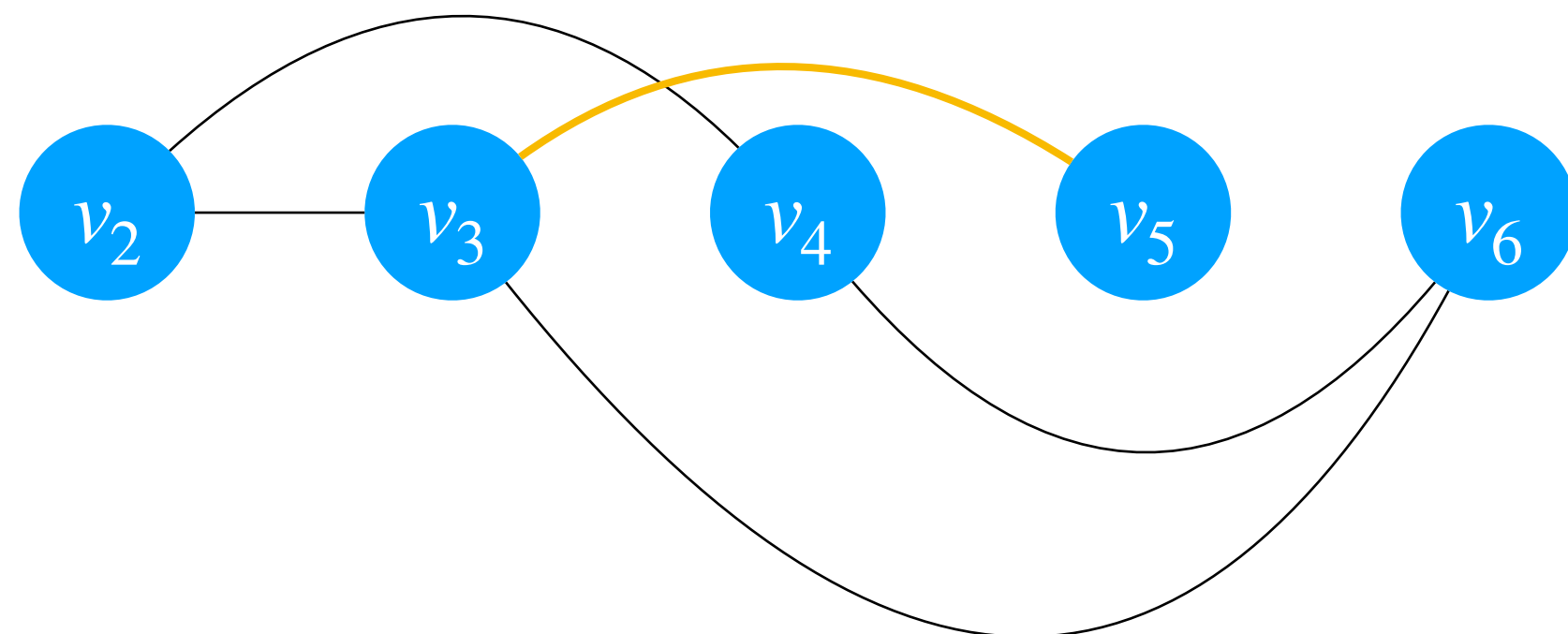
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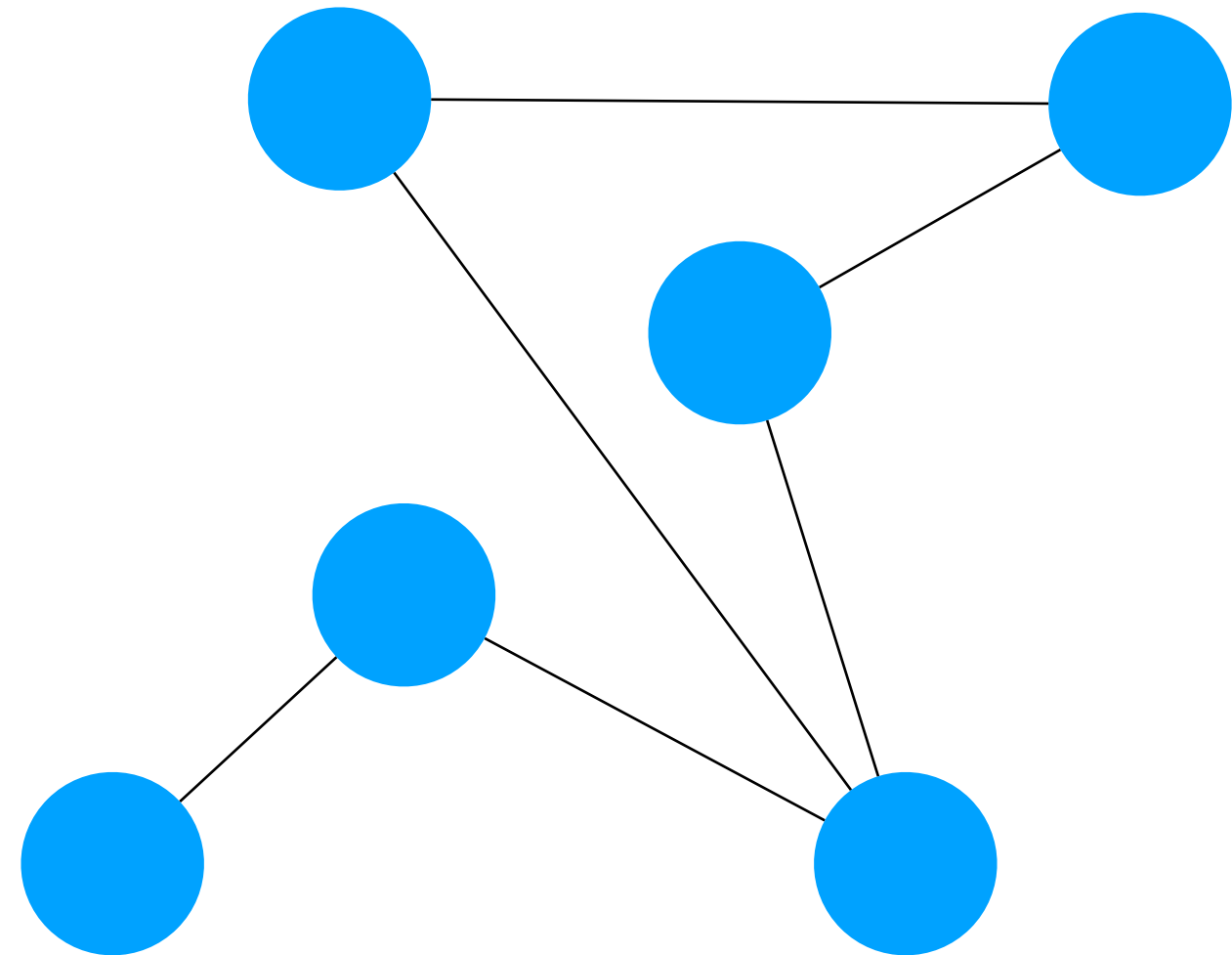
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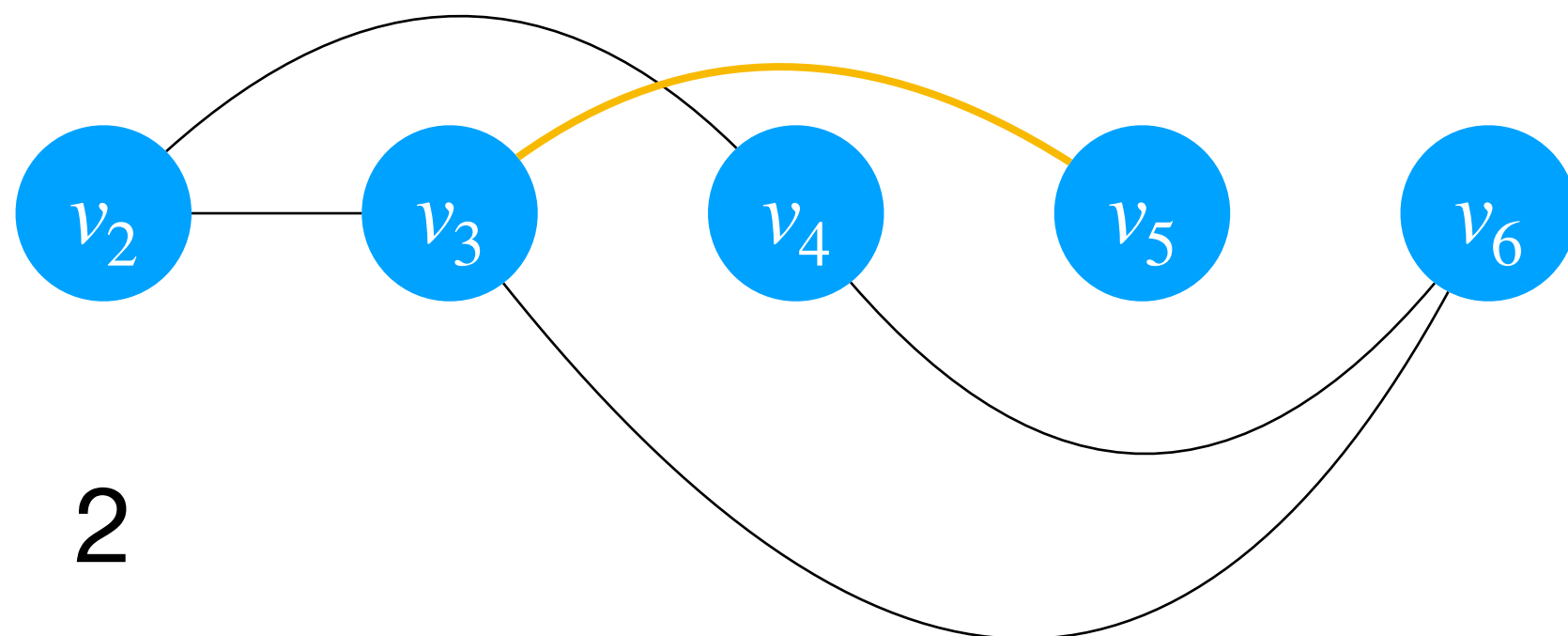
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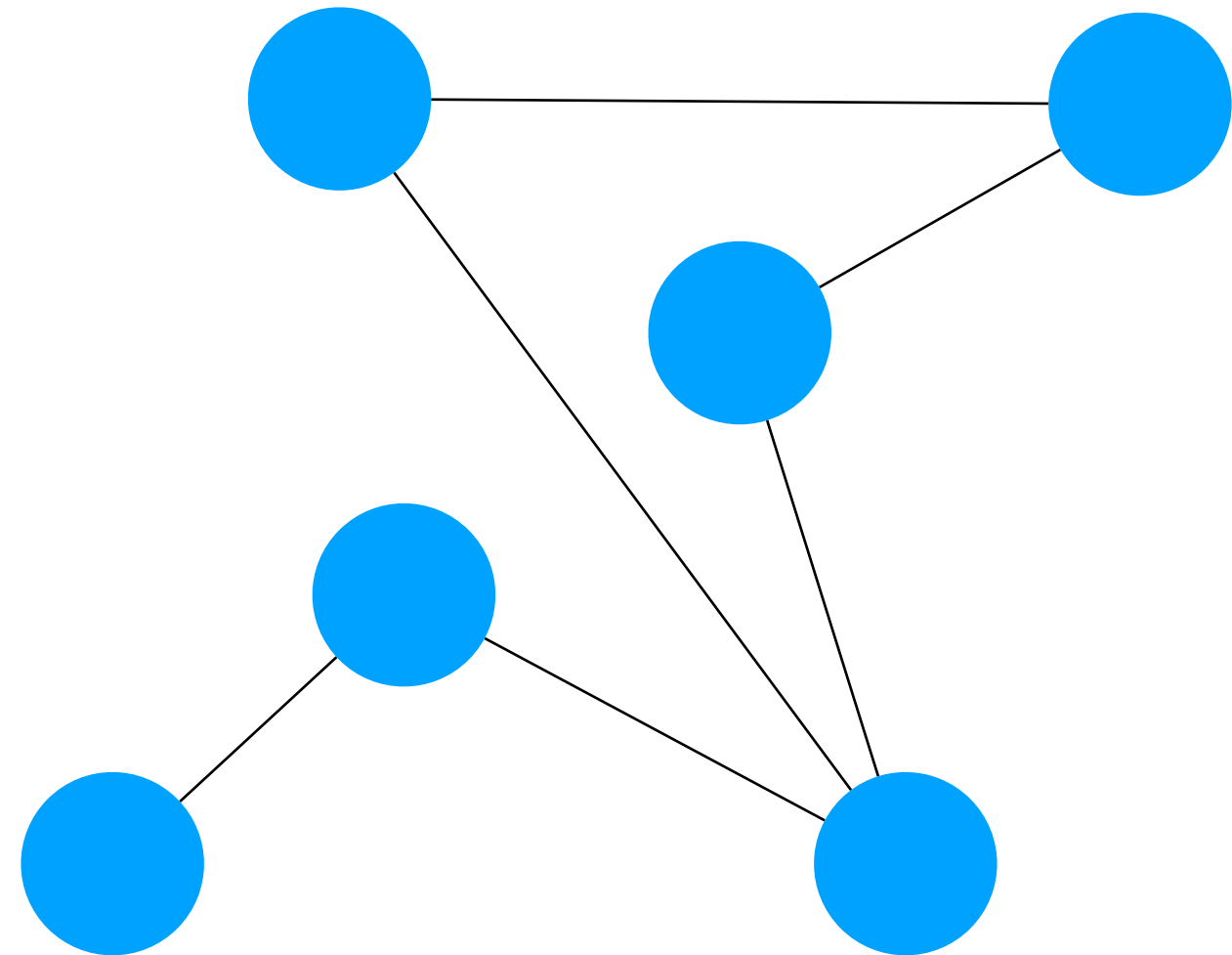


2

2

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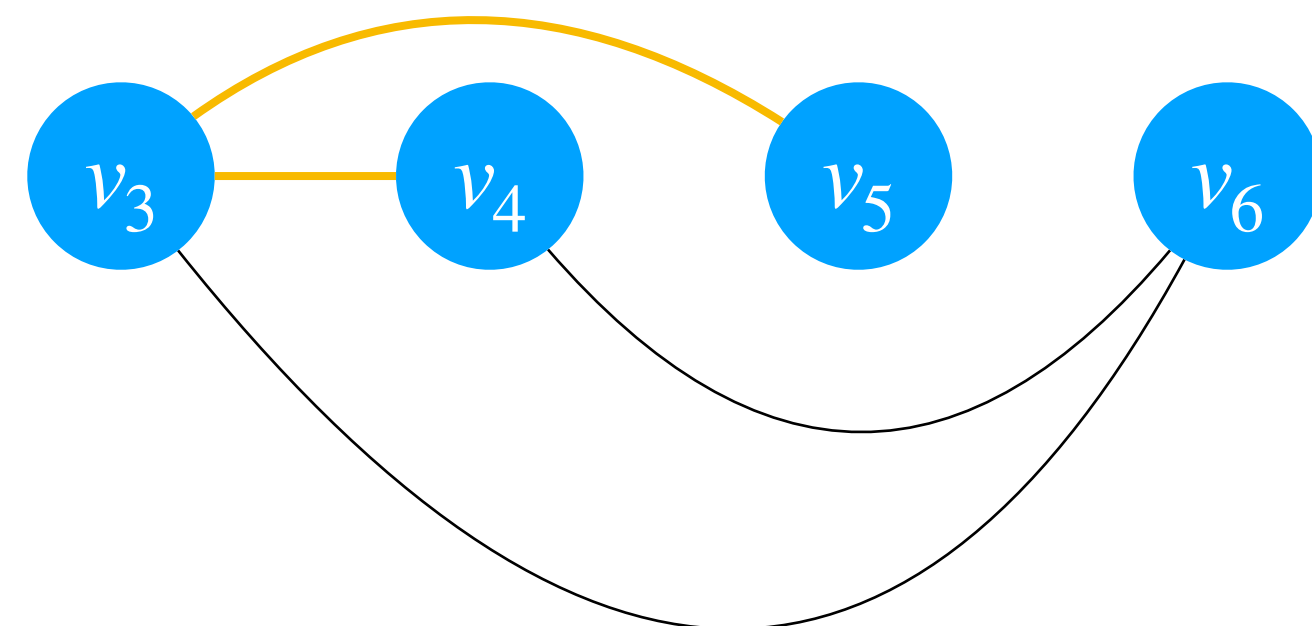
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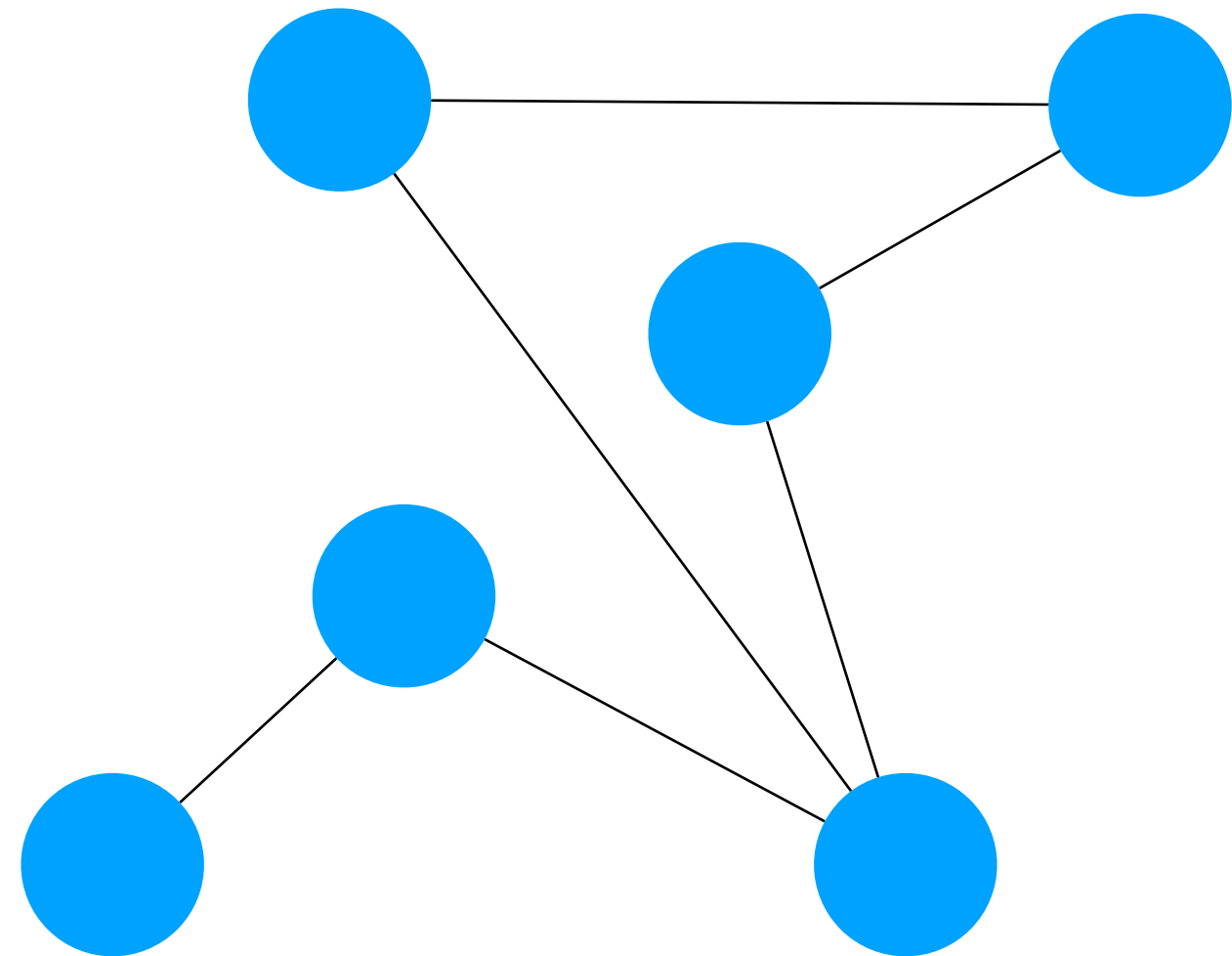


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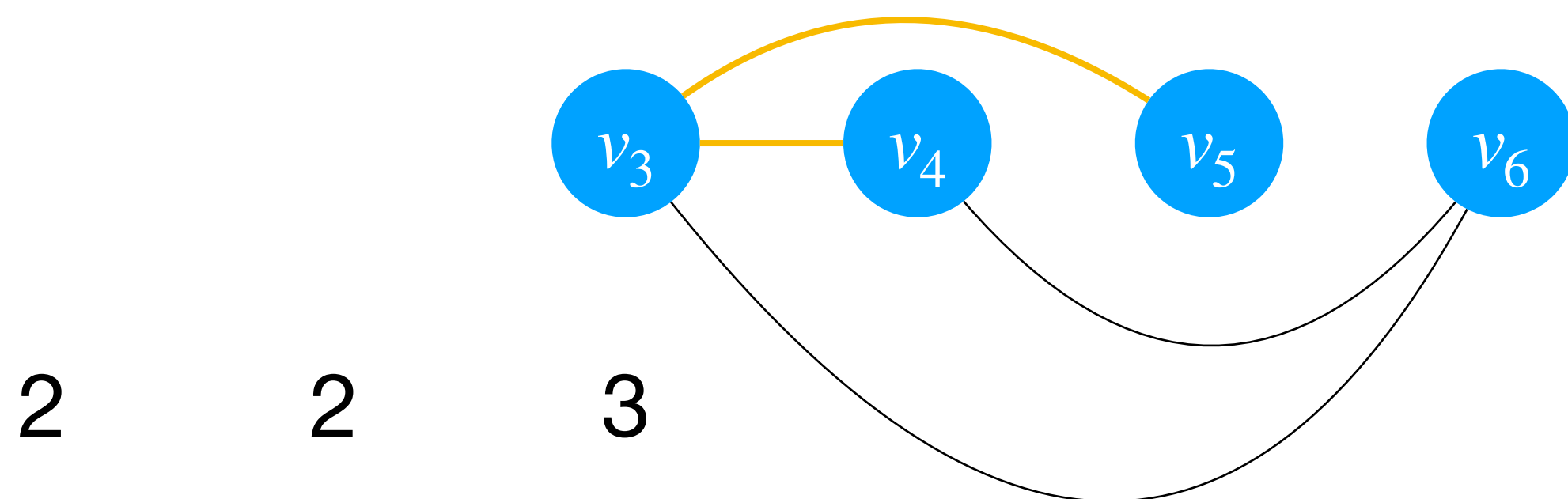
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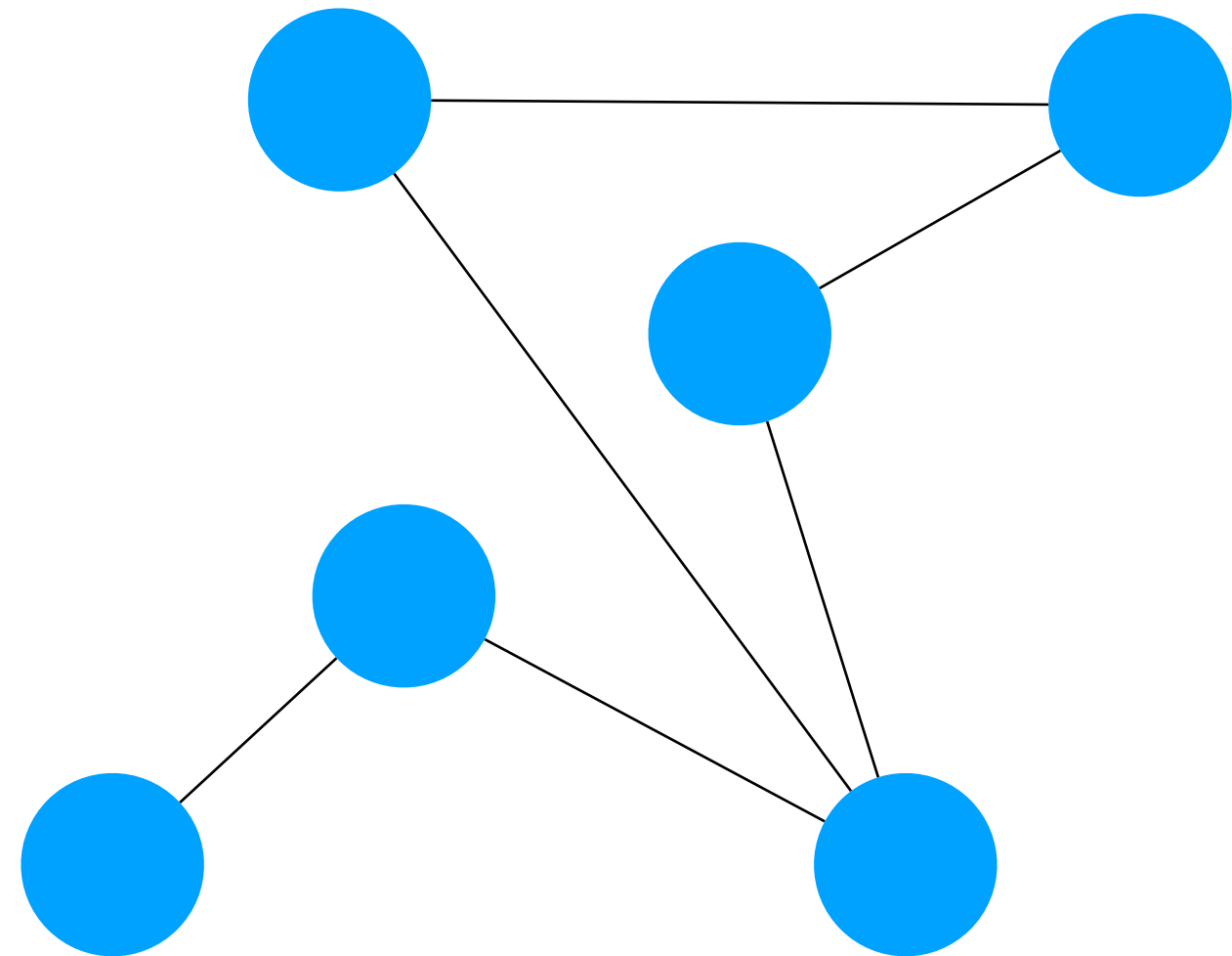
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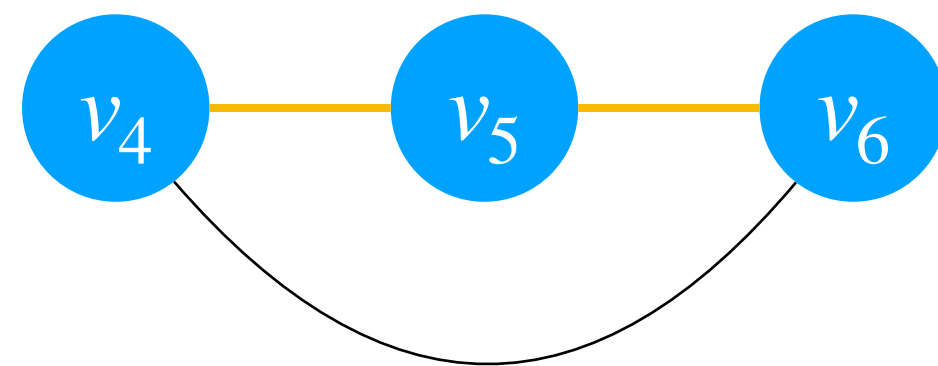
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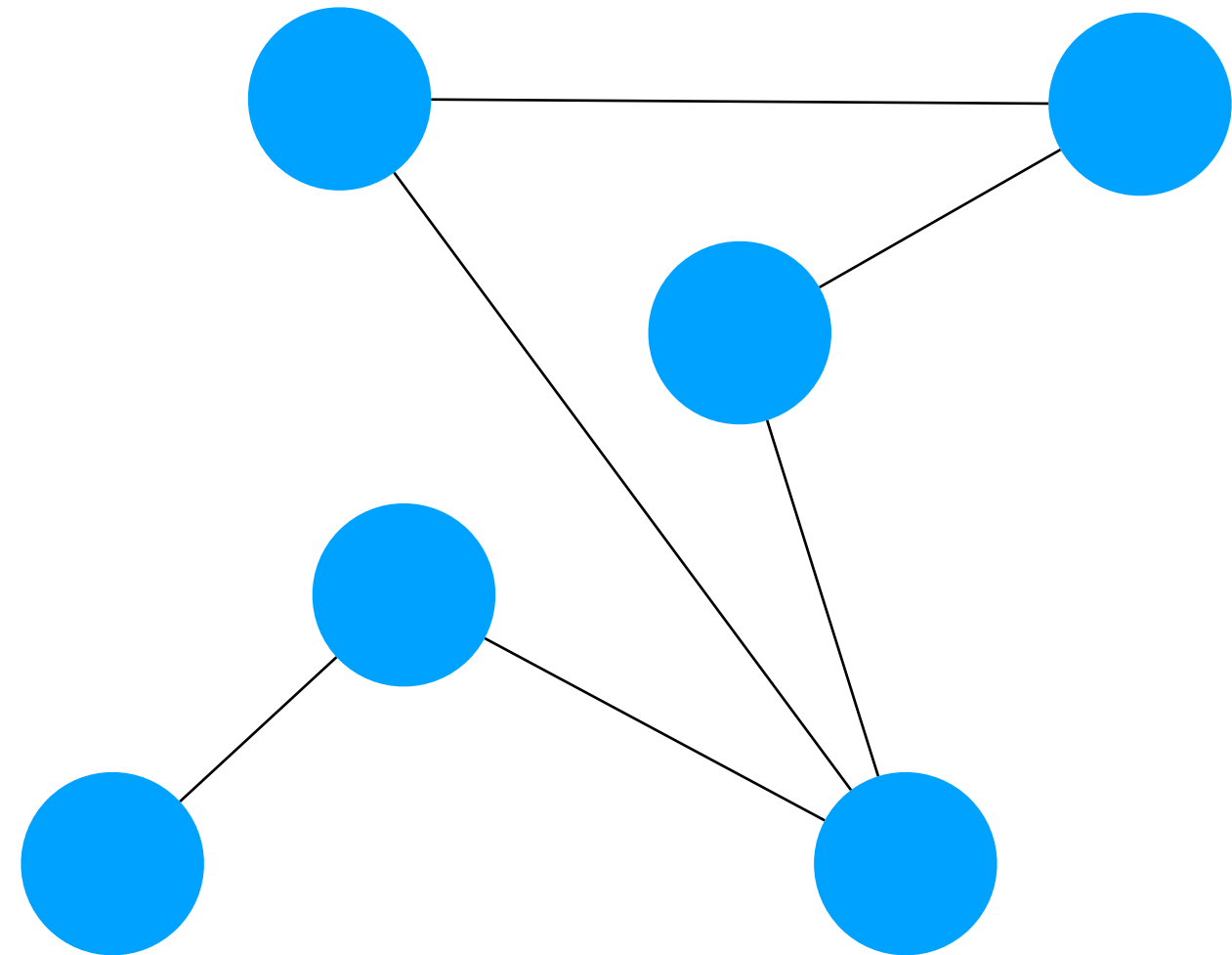
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3



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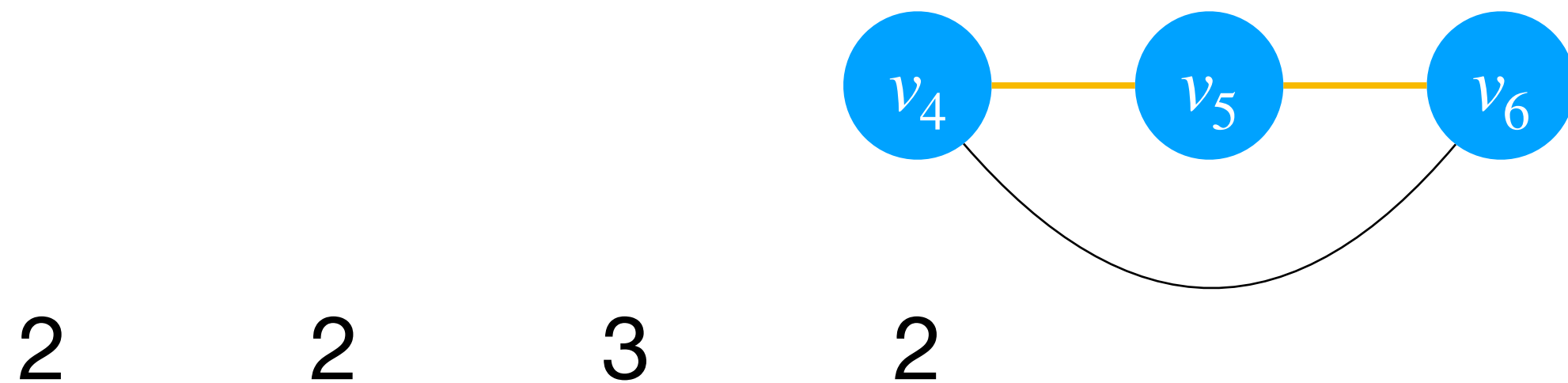
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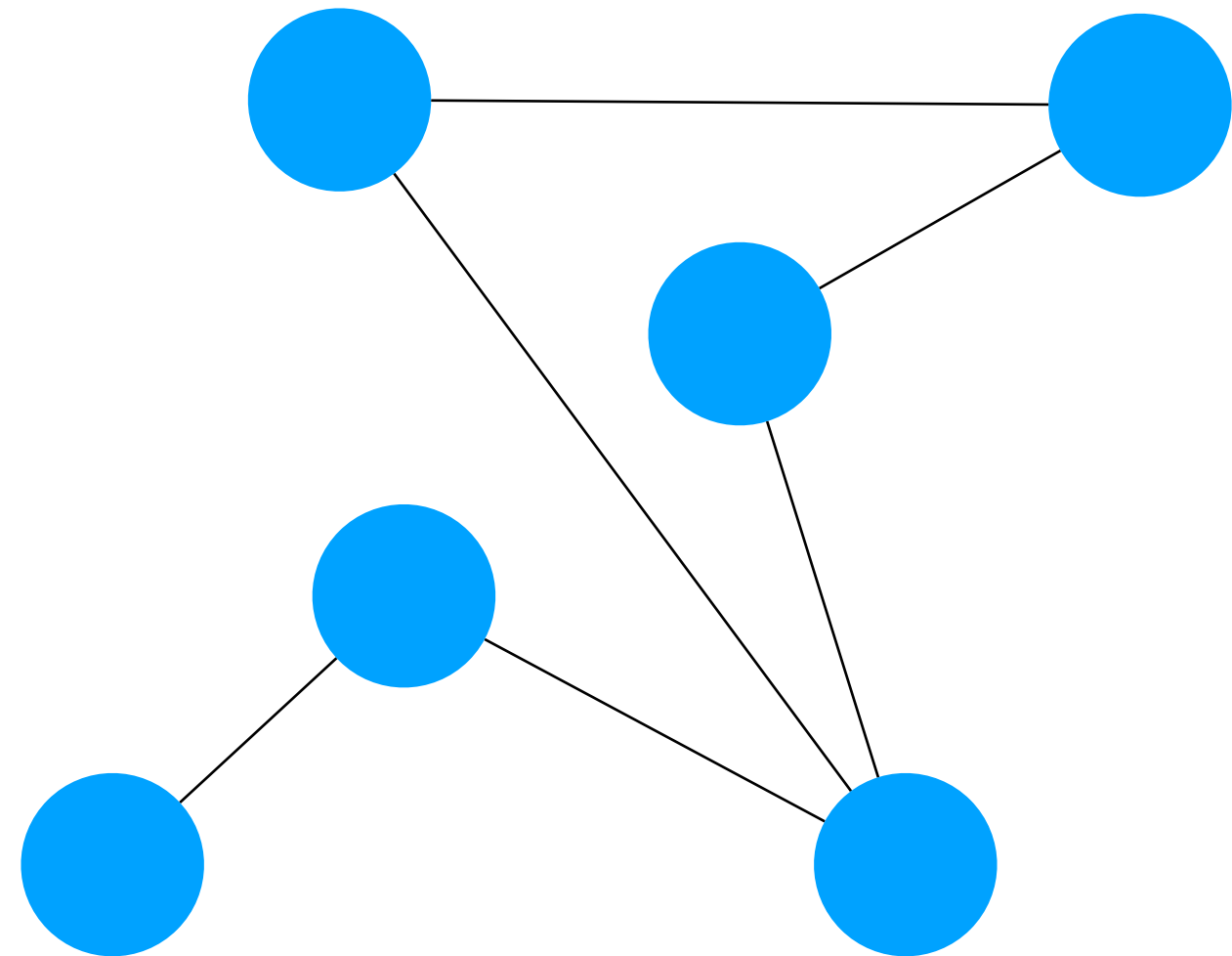
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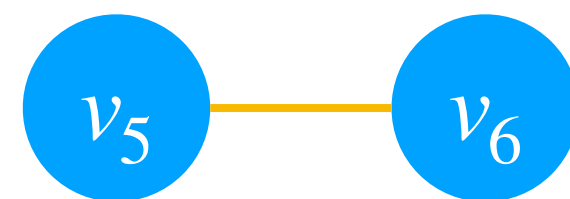
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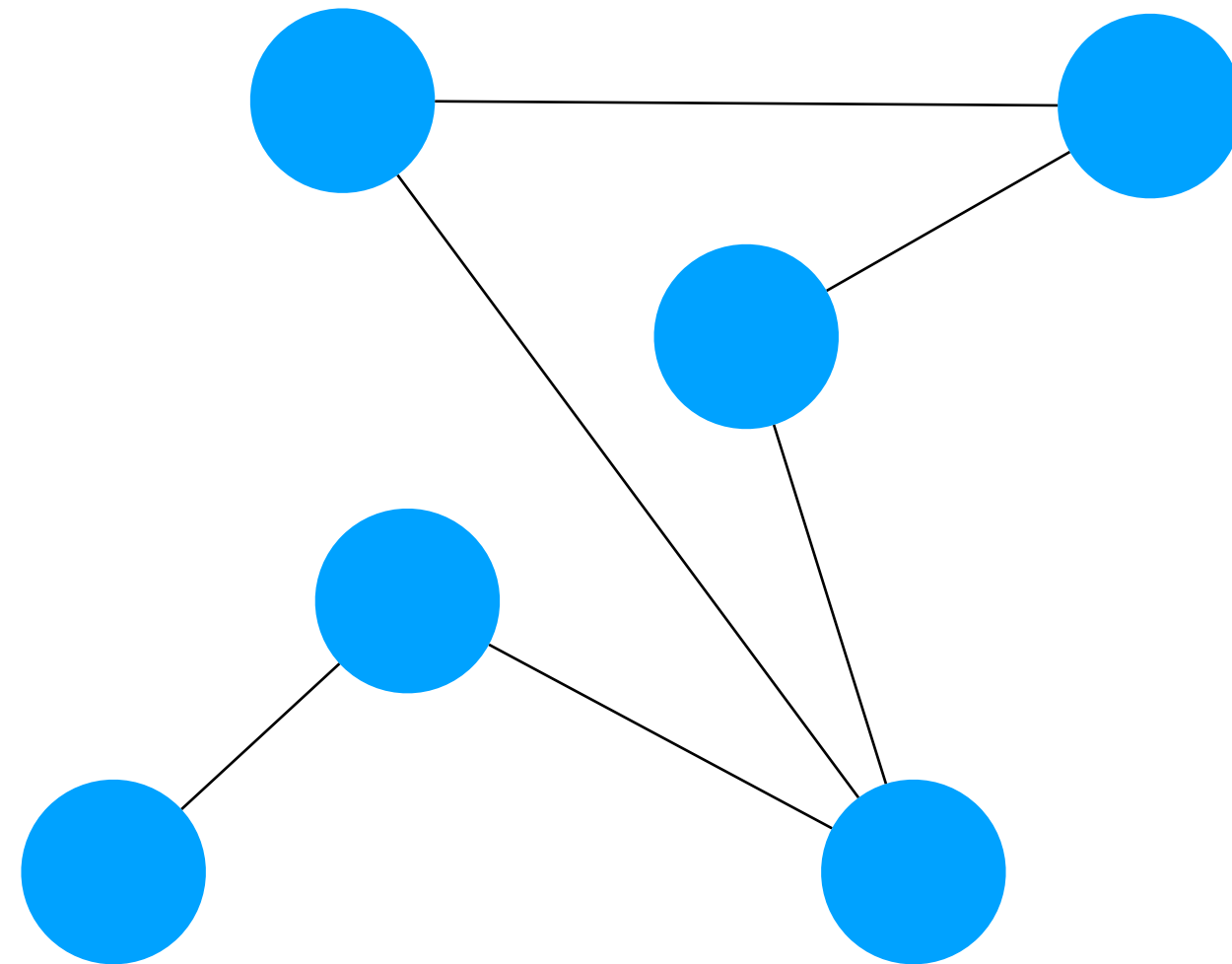
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2      2      3      2

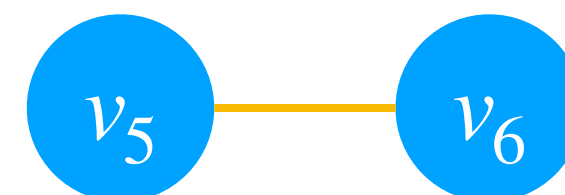
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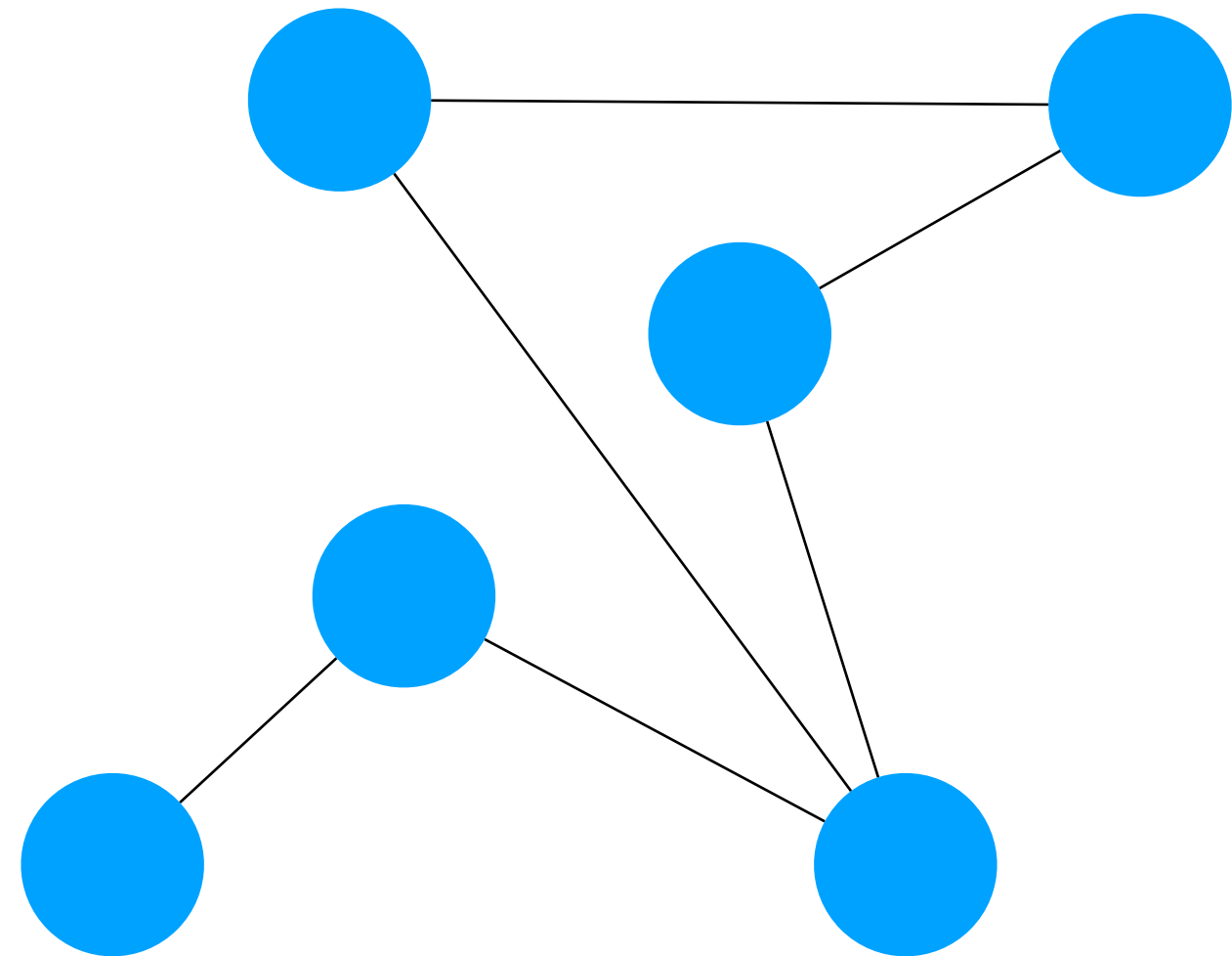
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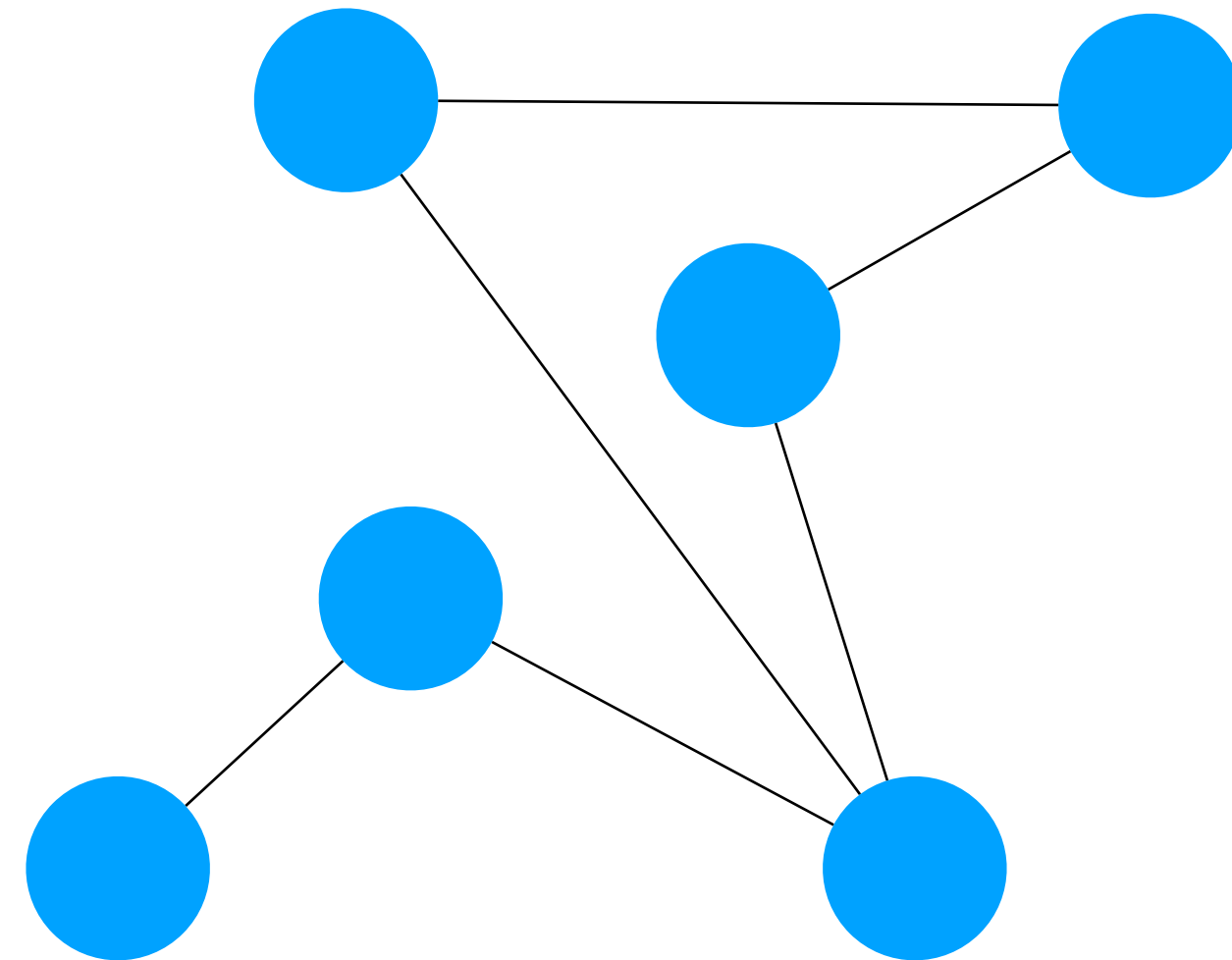


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2    2    3    2    1



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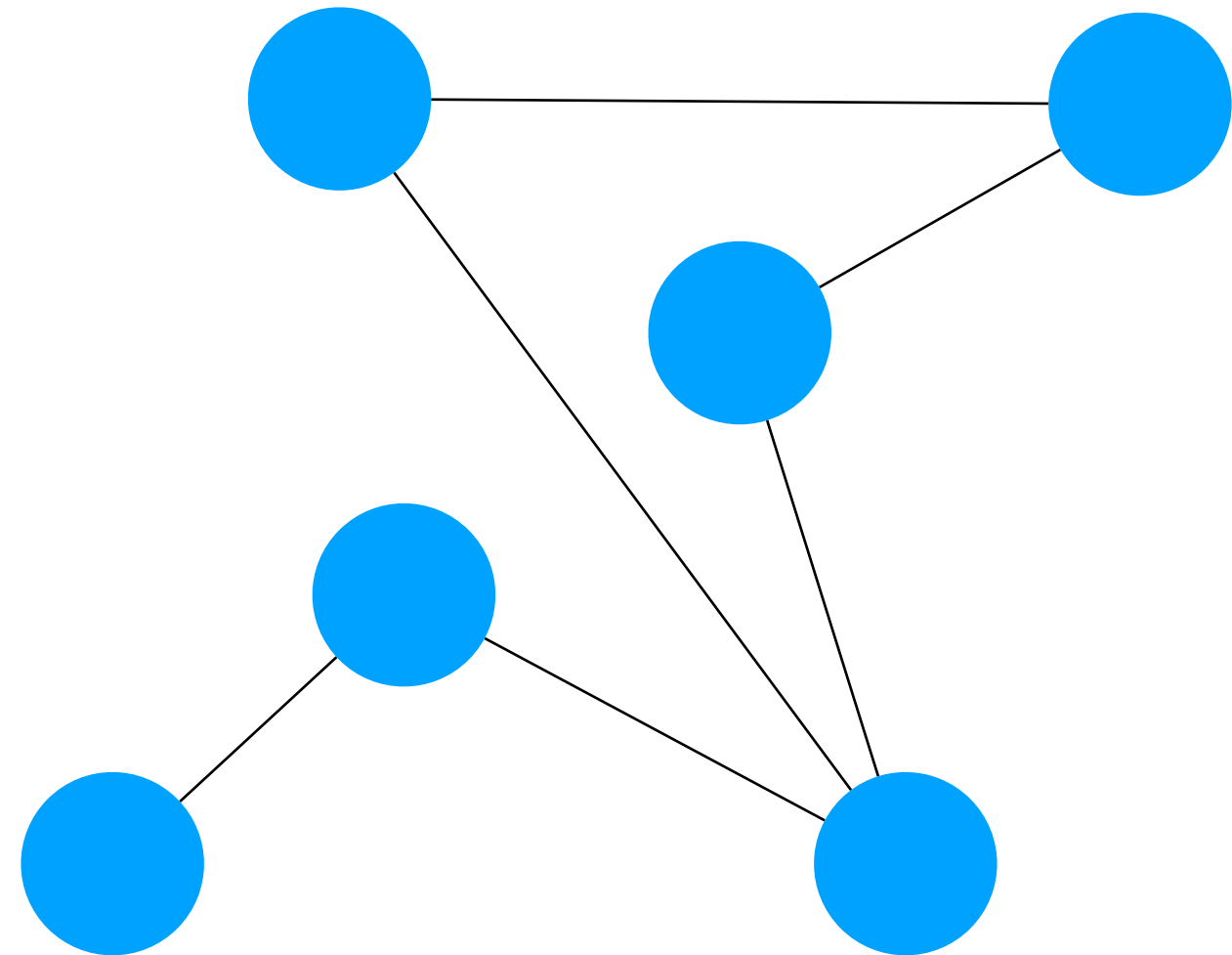
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2    2    3    2    1    0

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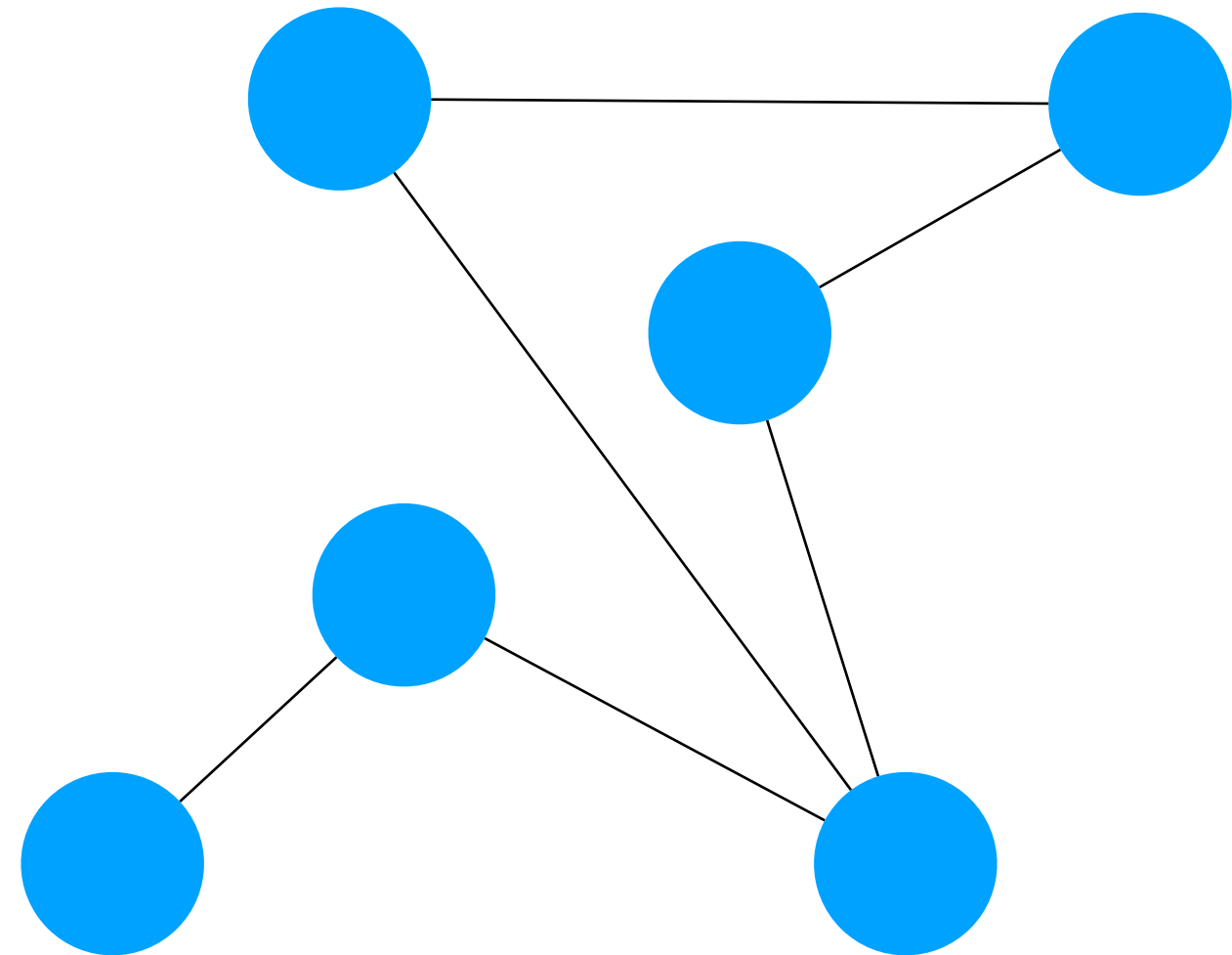
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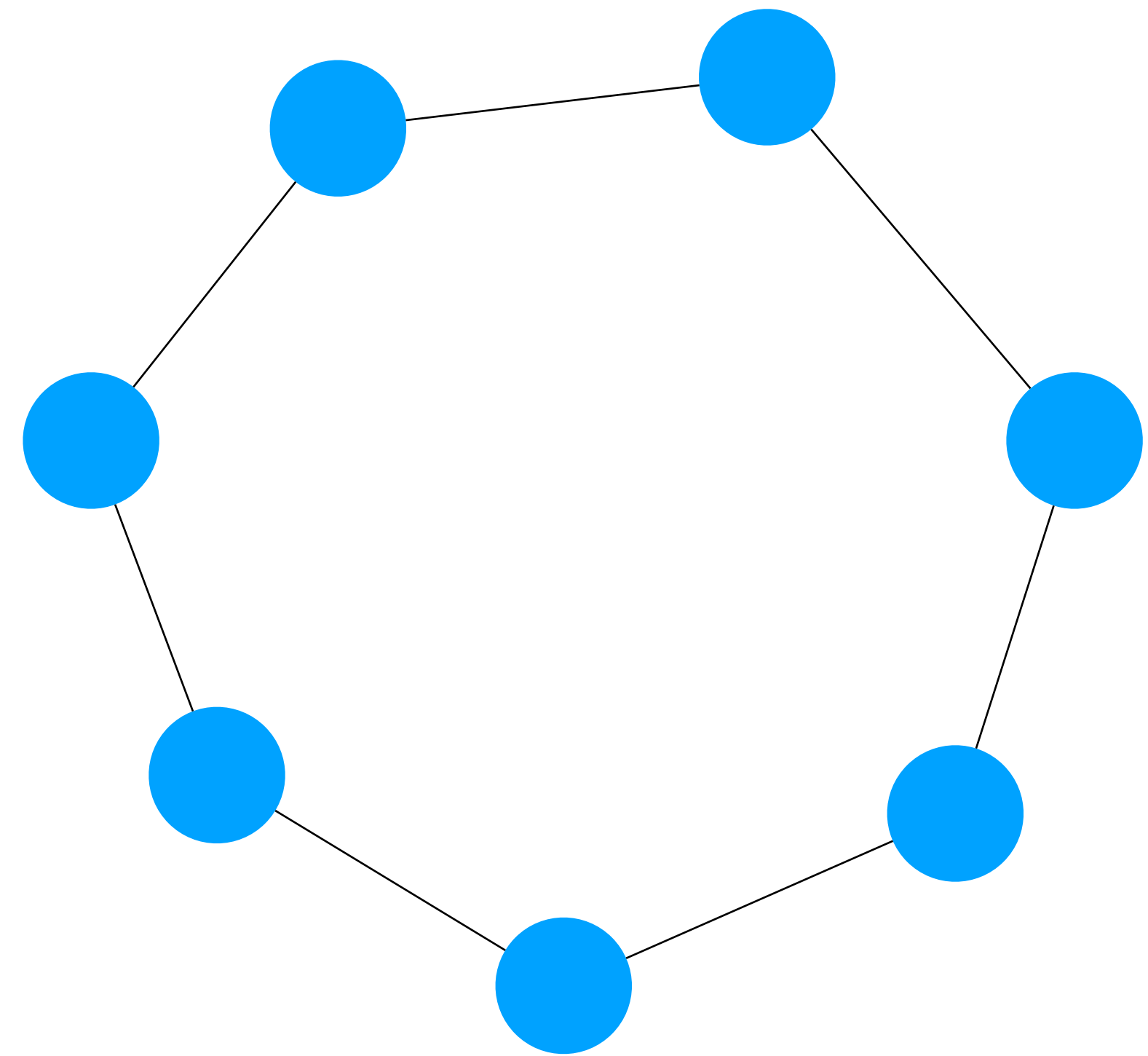
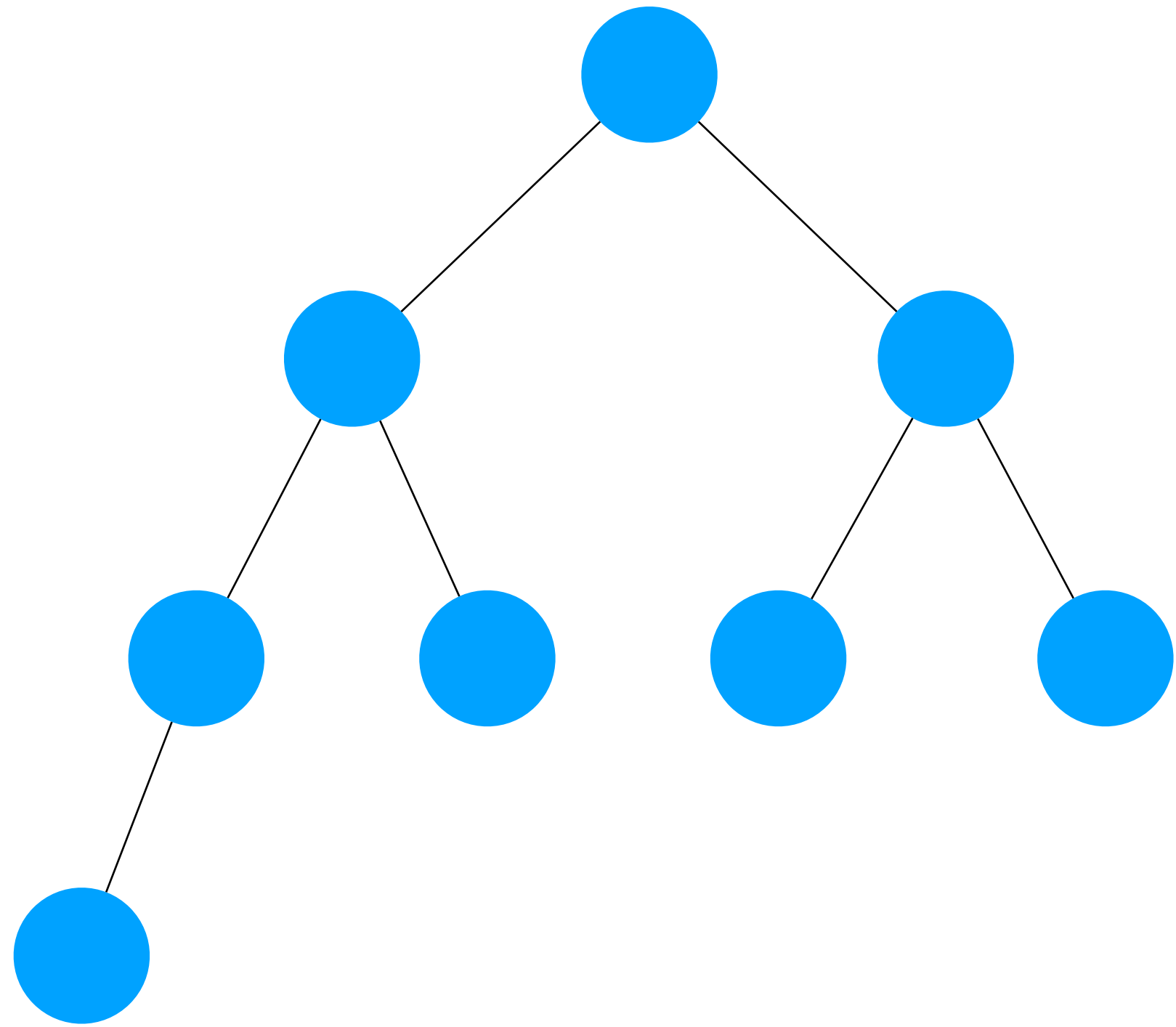
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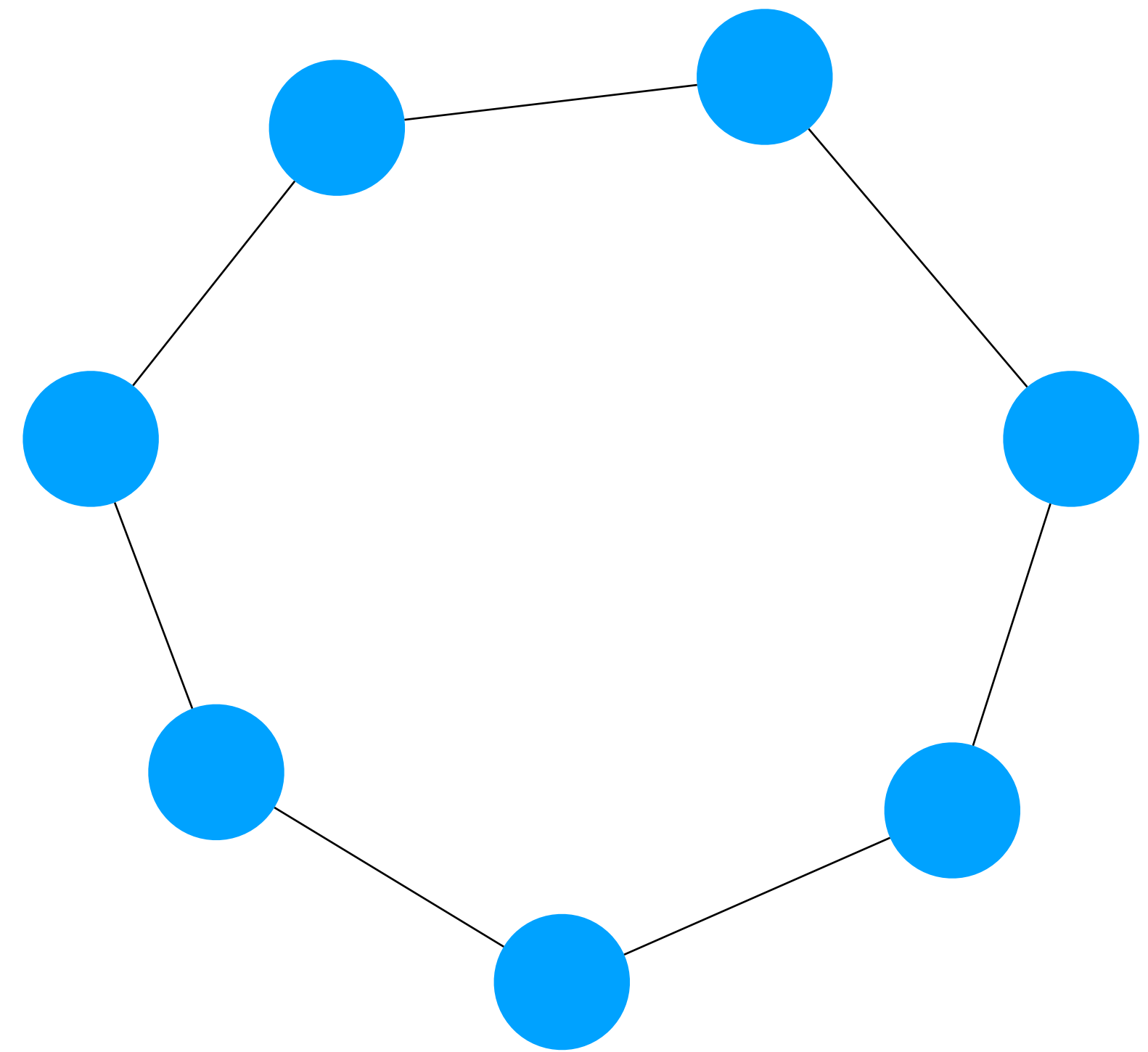
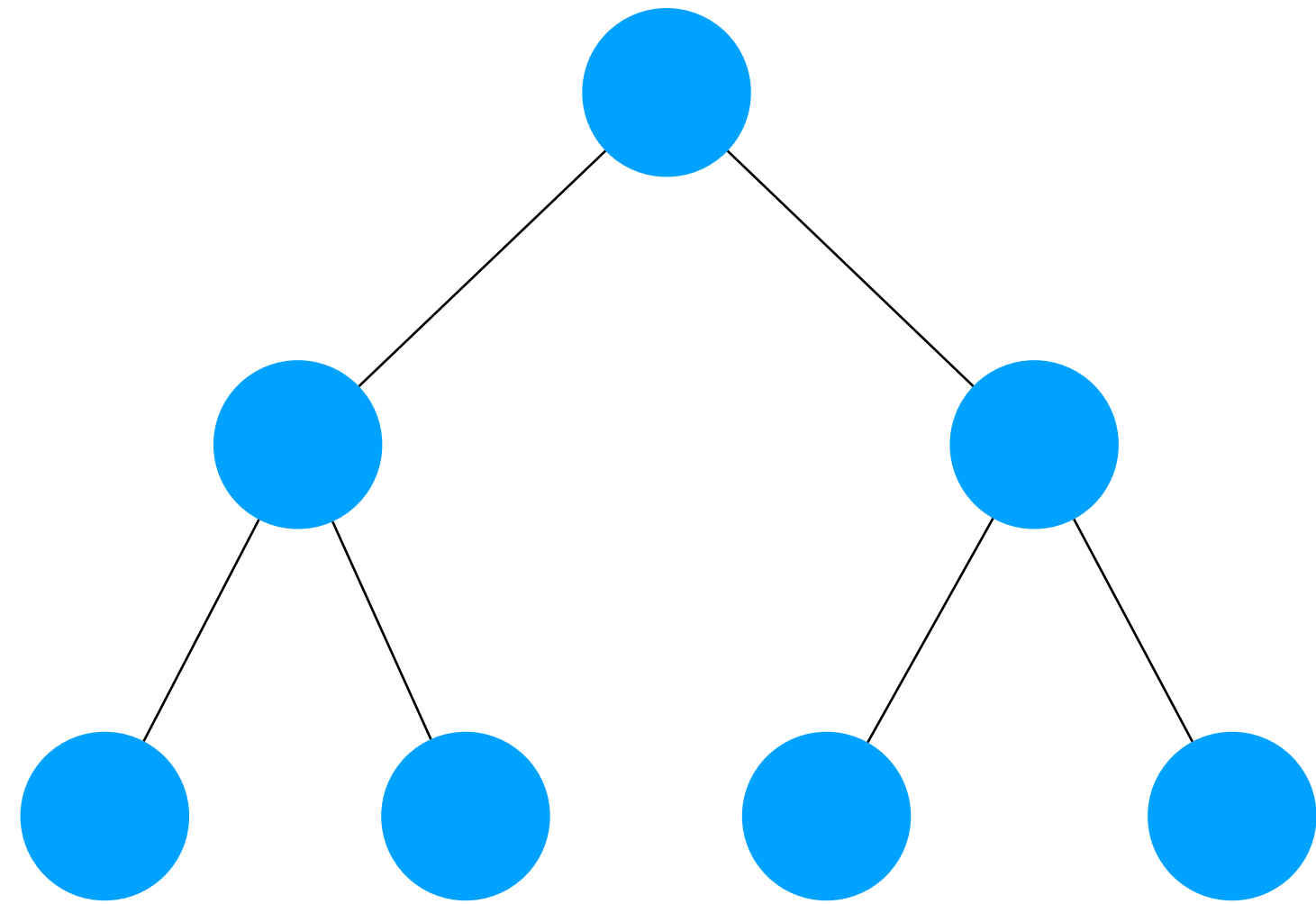
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2    2    3    2    1    0

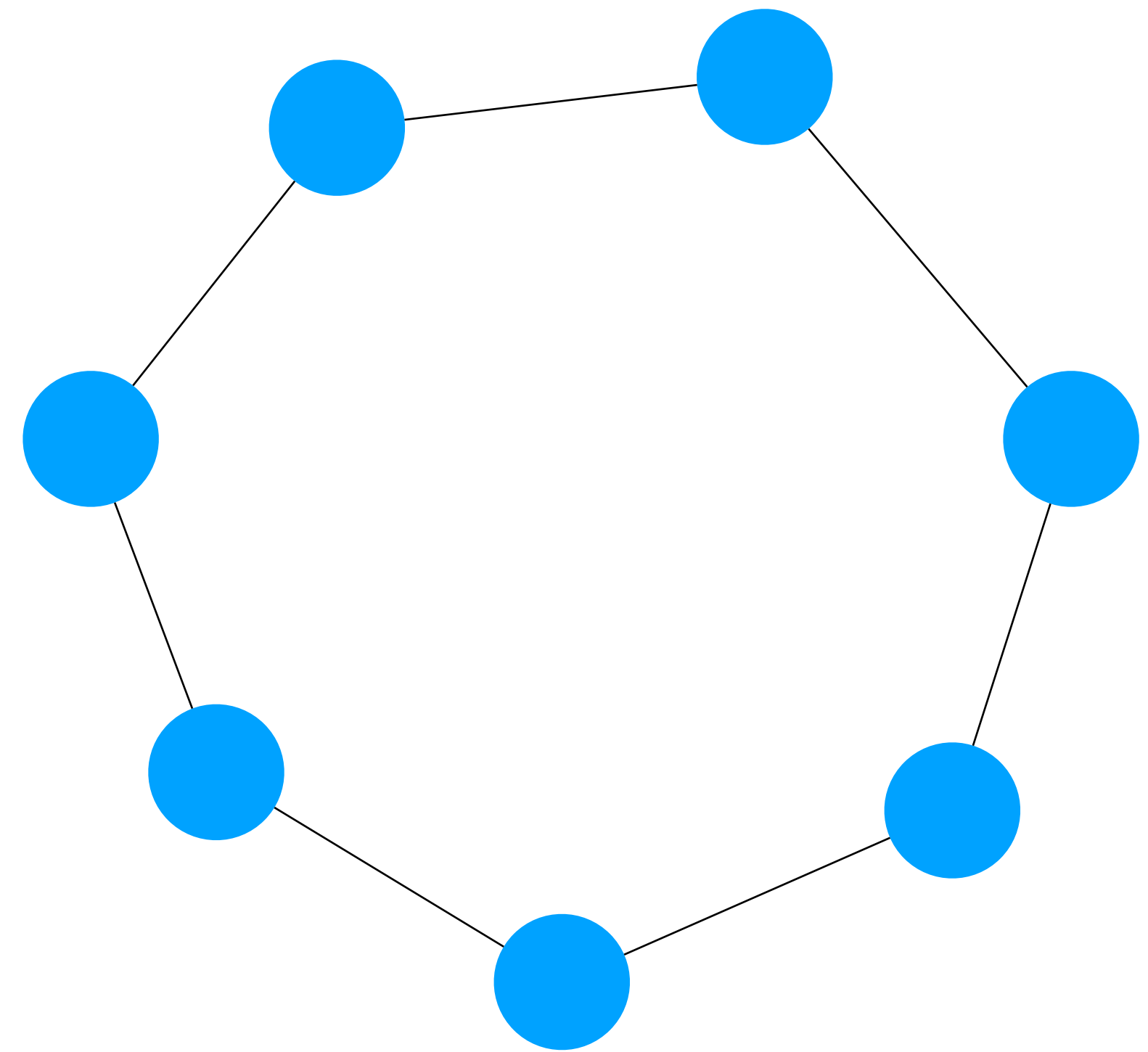
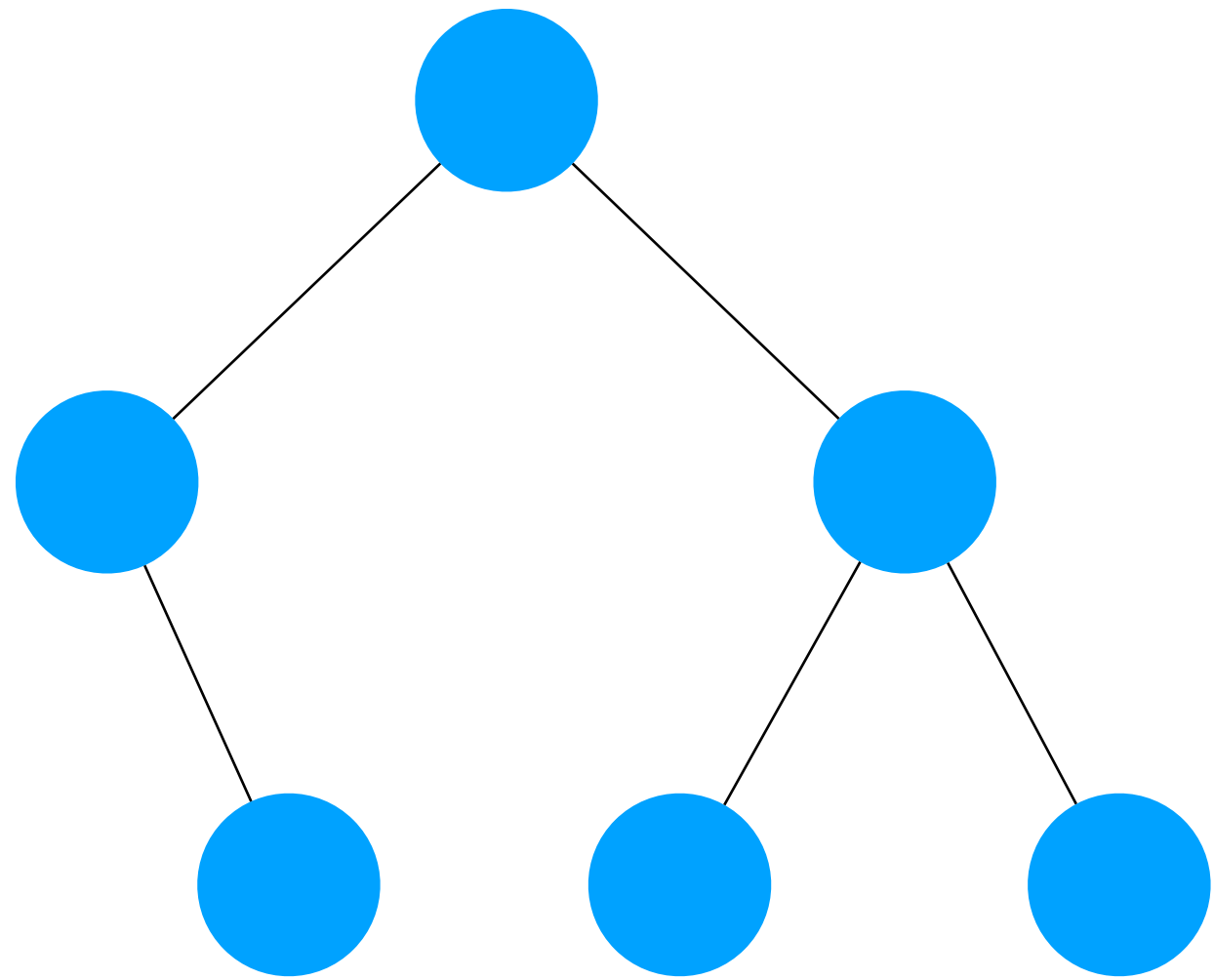
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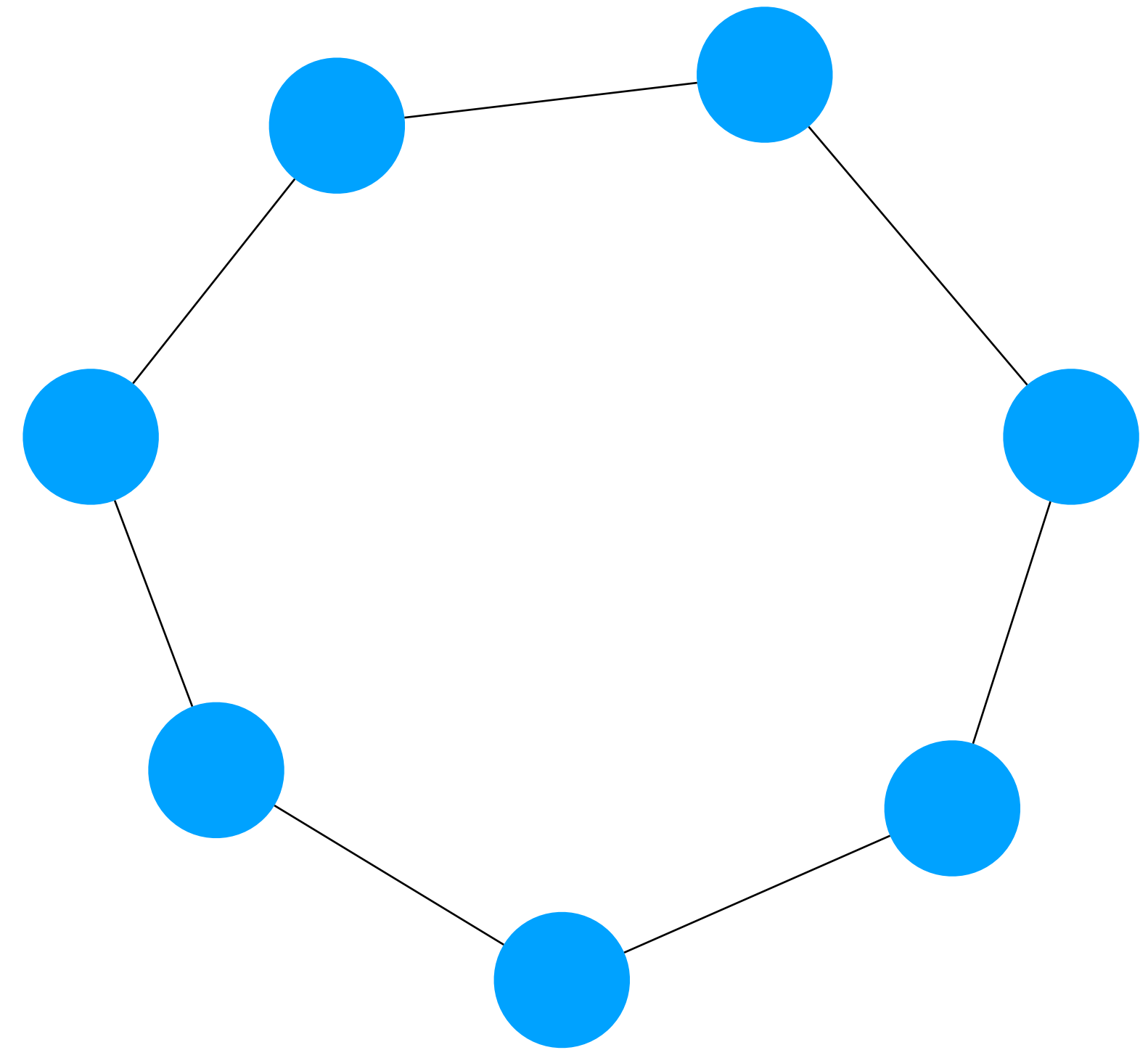
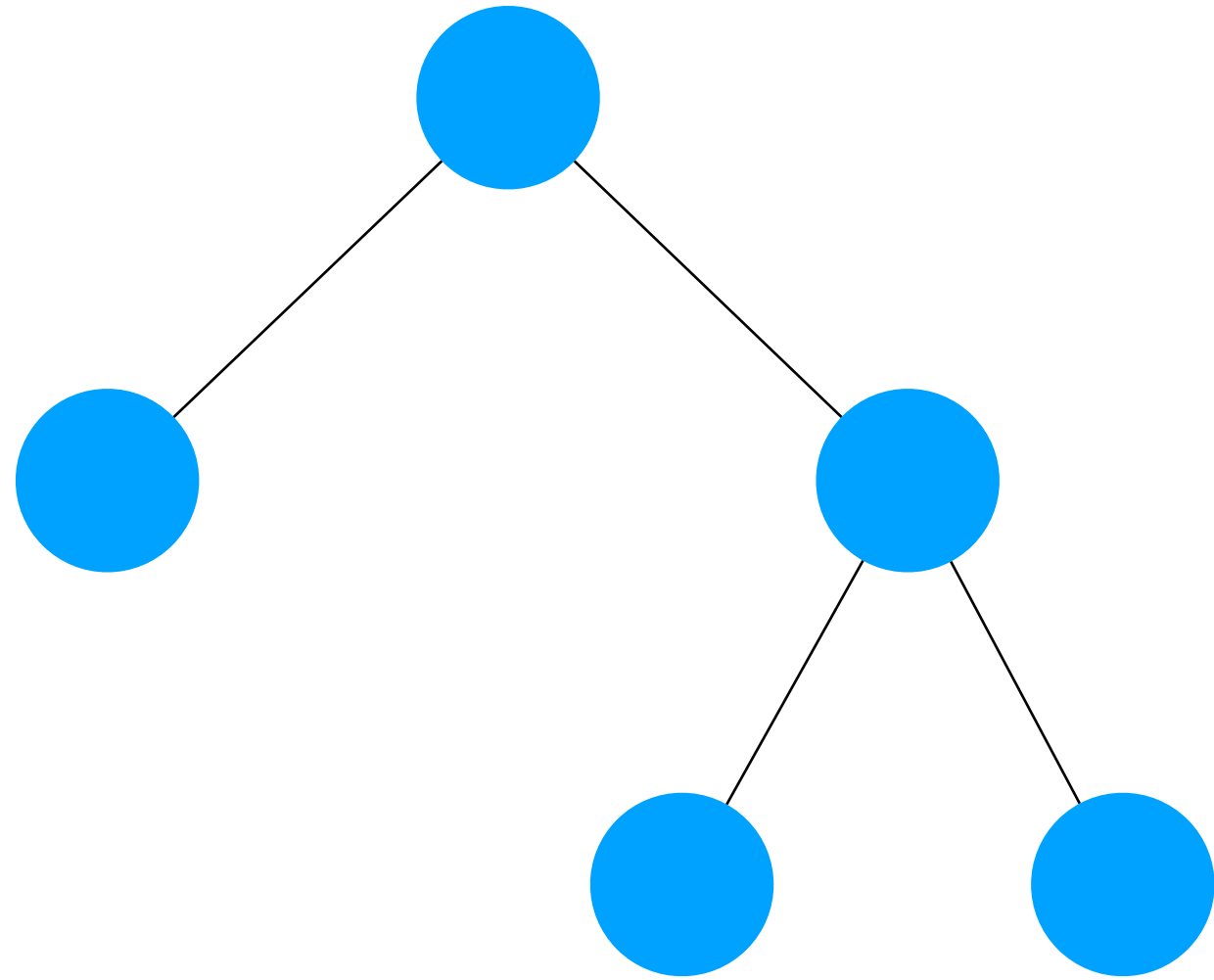
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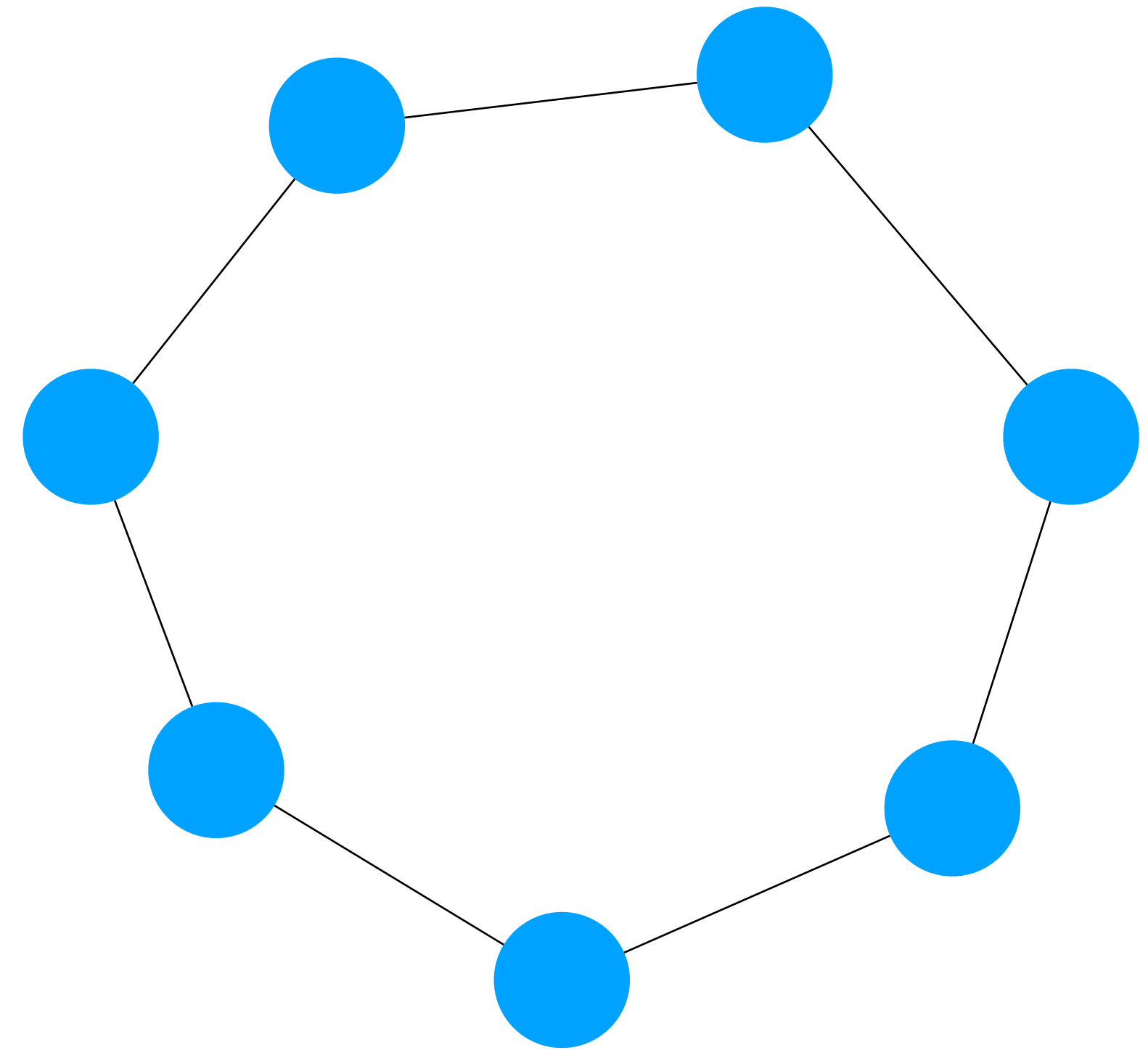
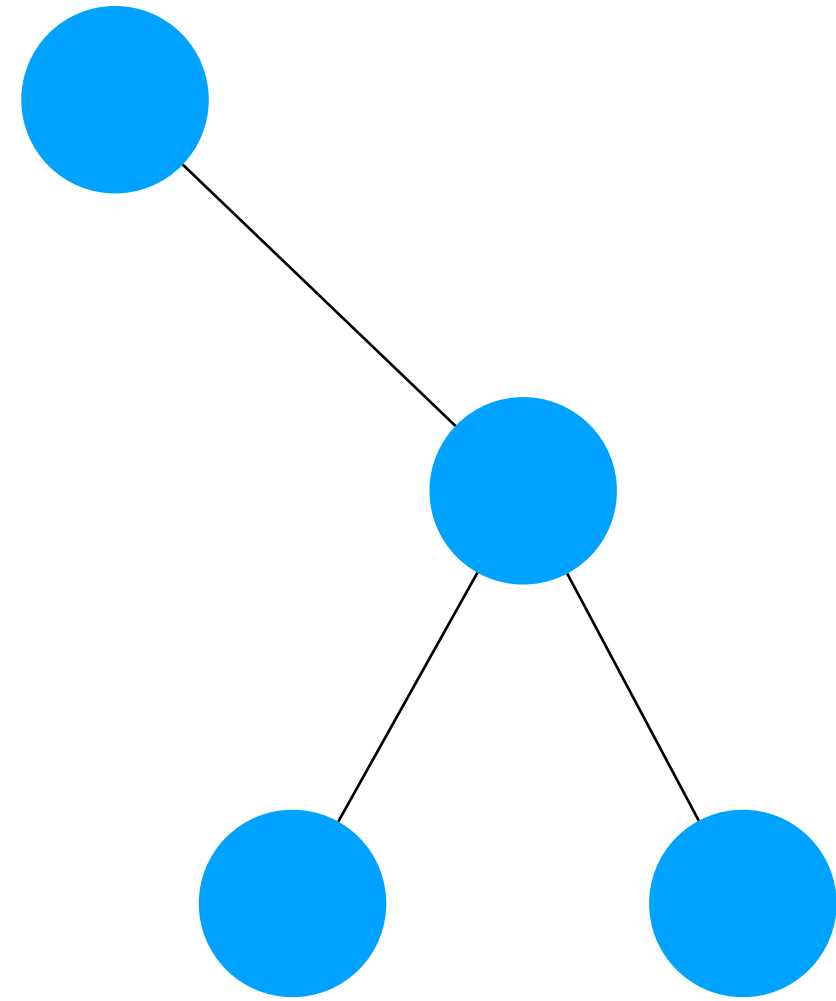
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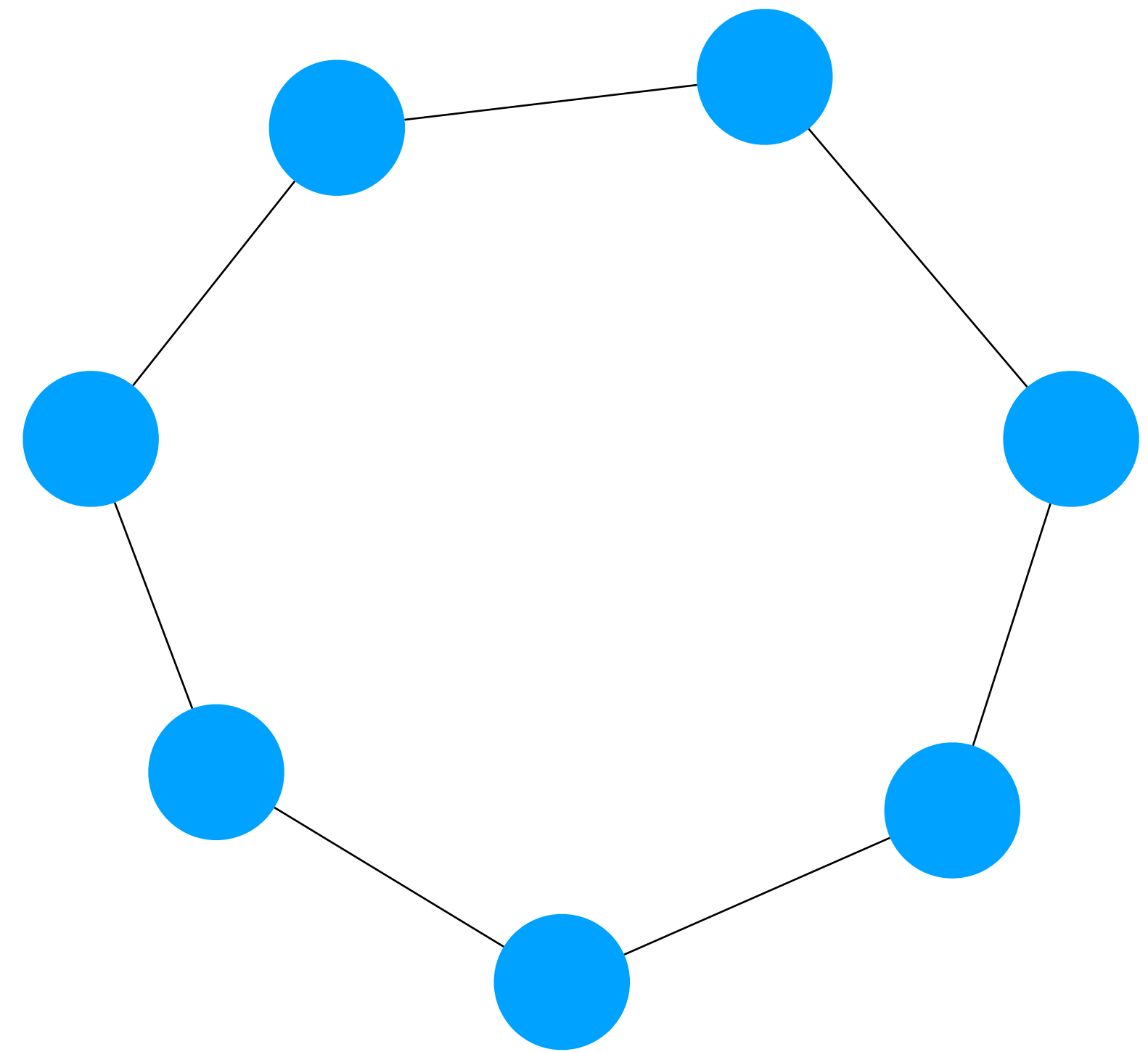
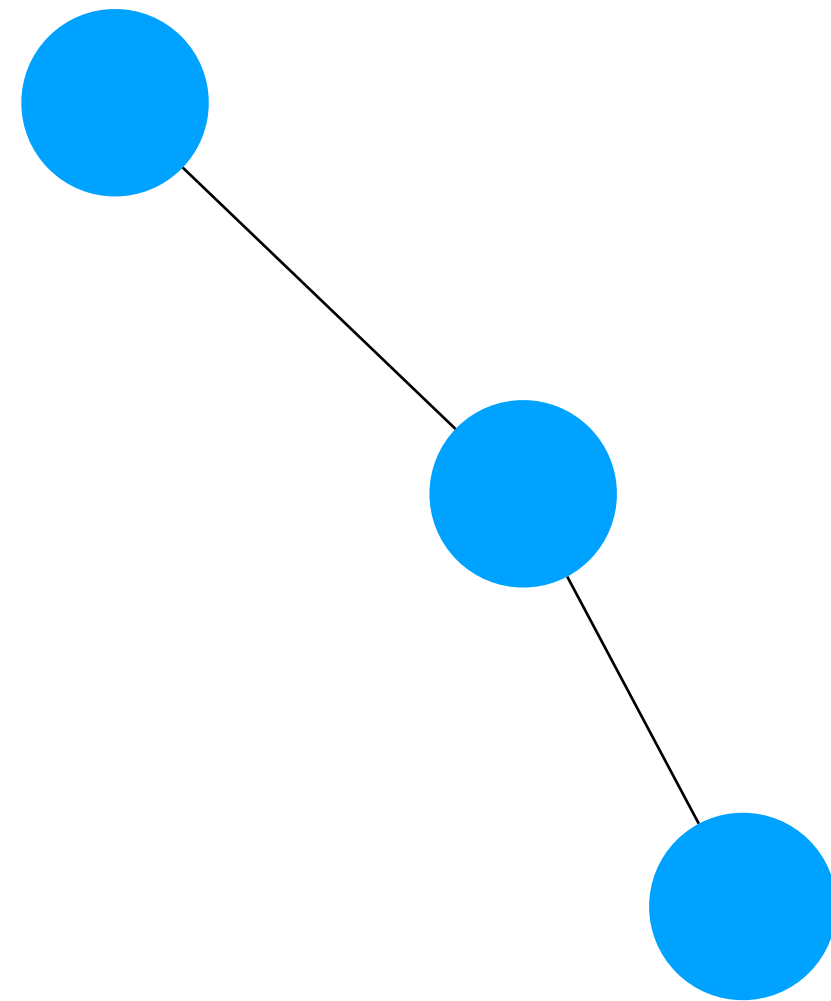


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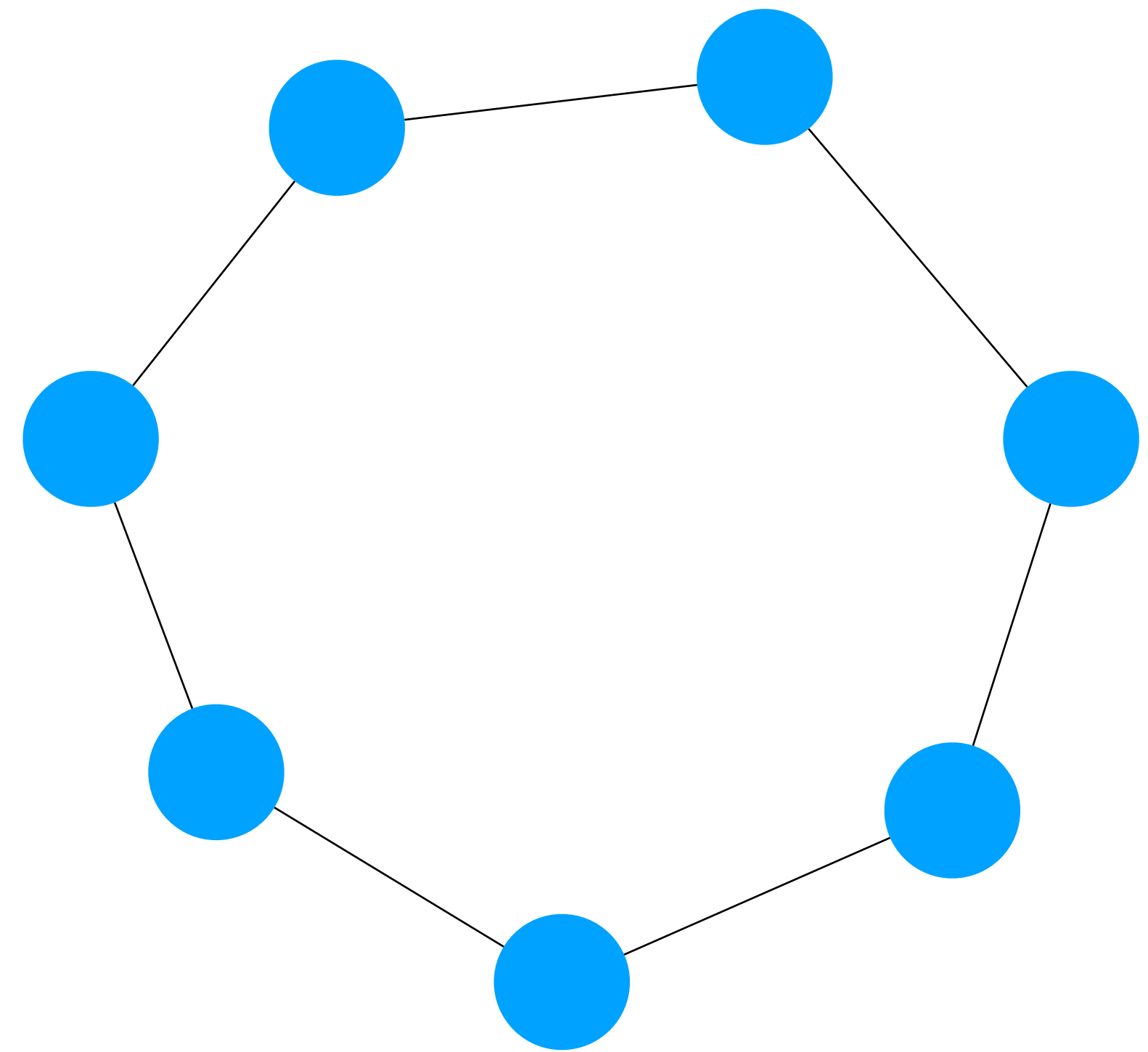
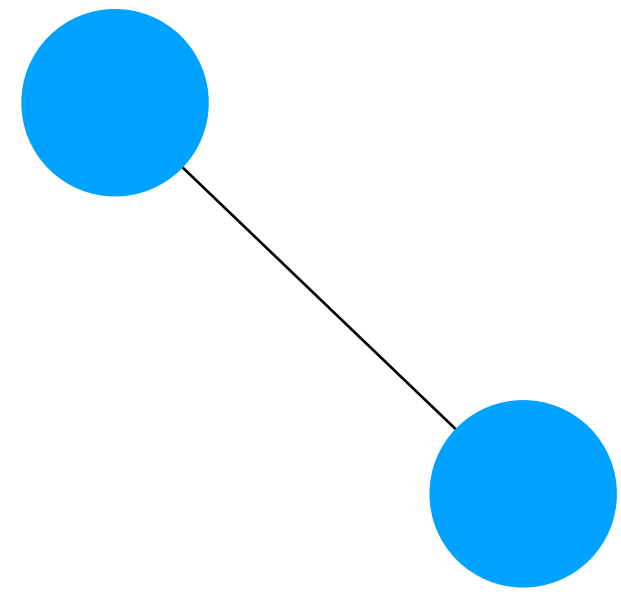




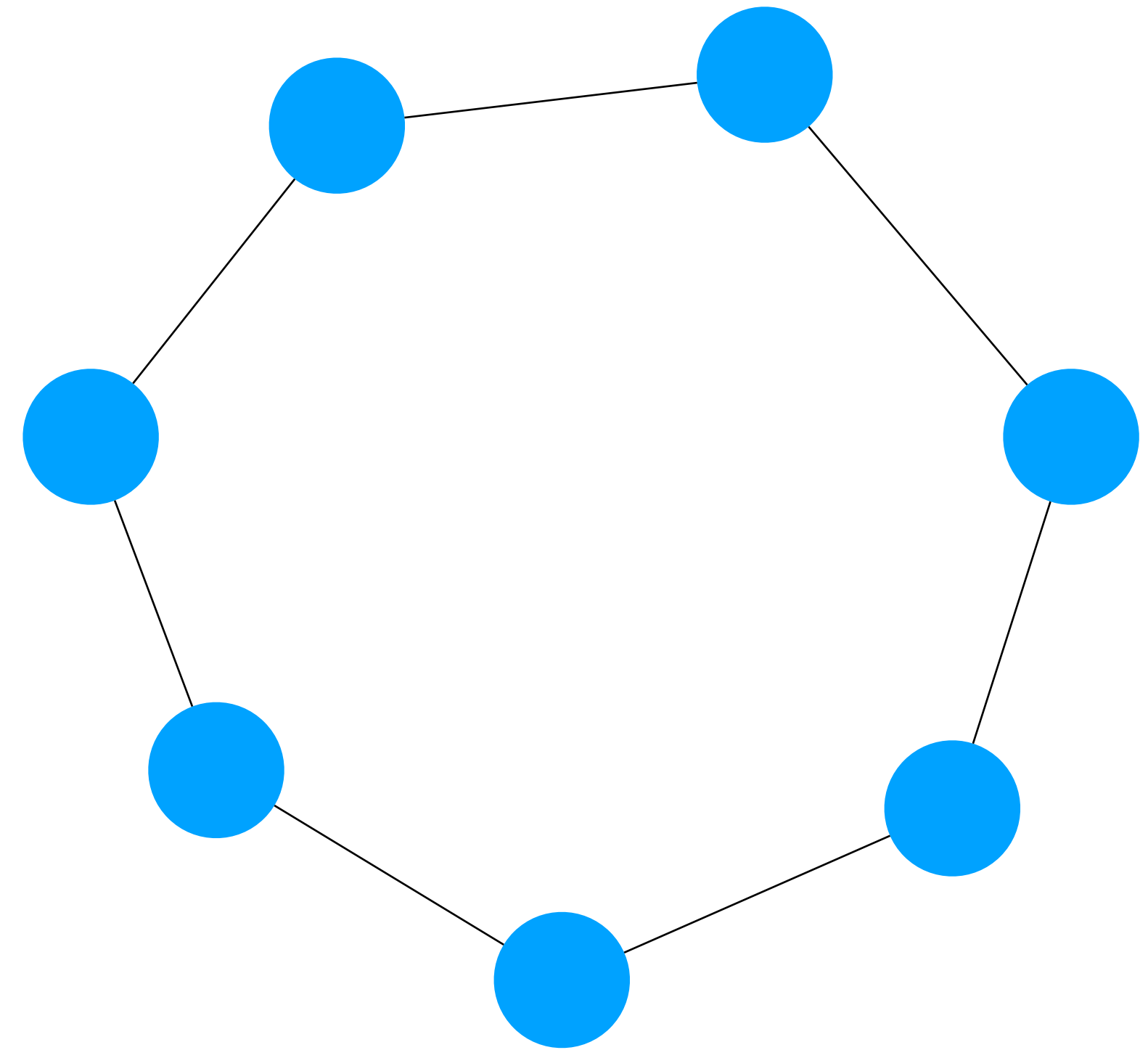
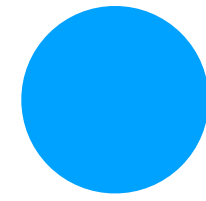
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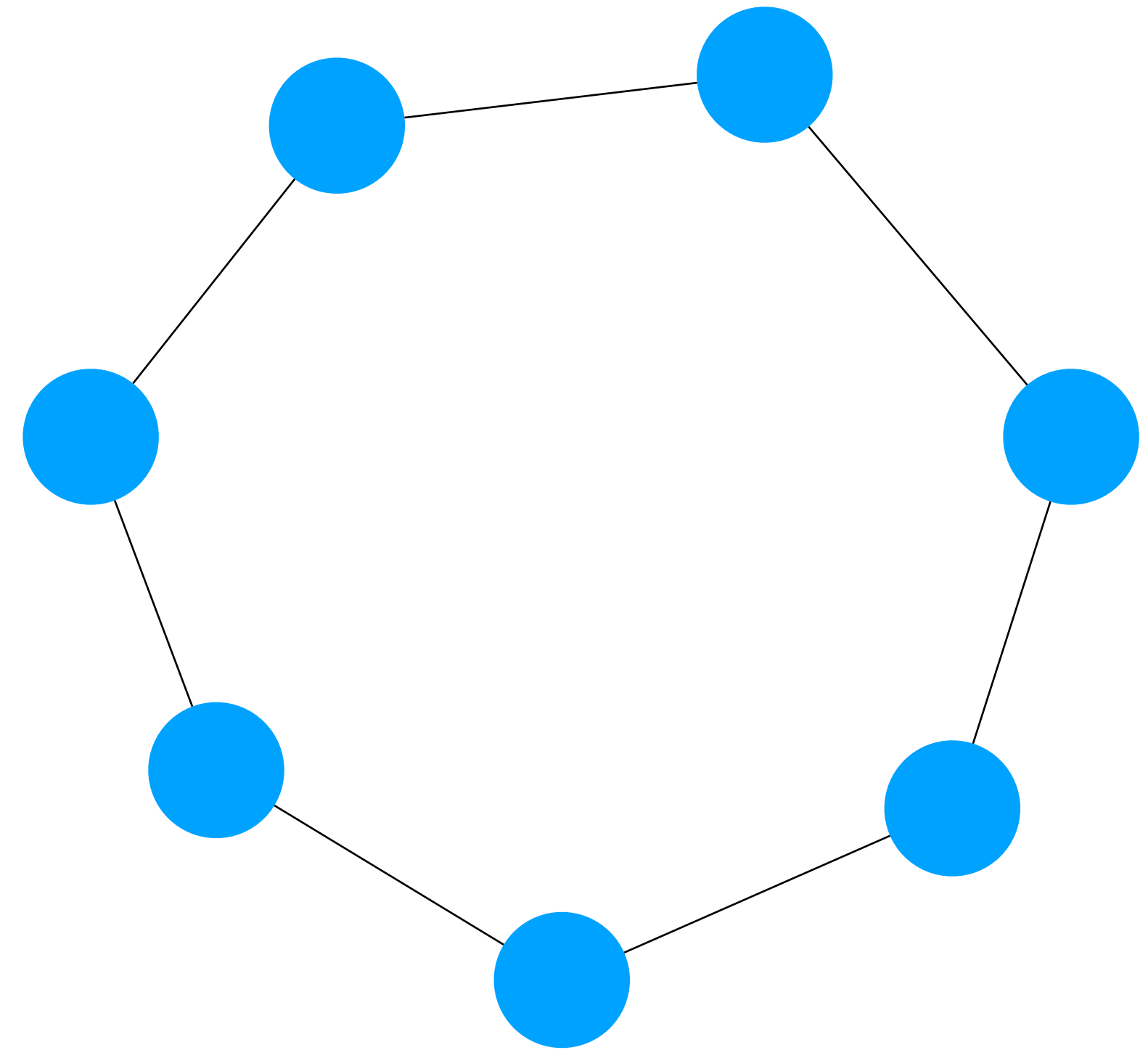
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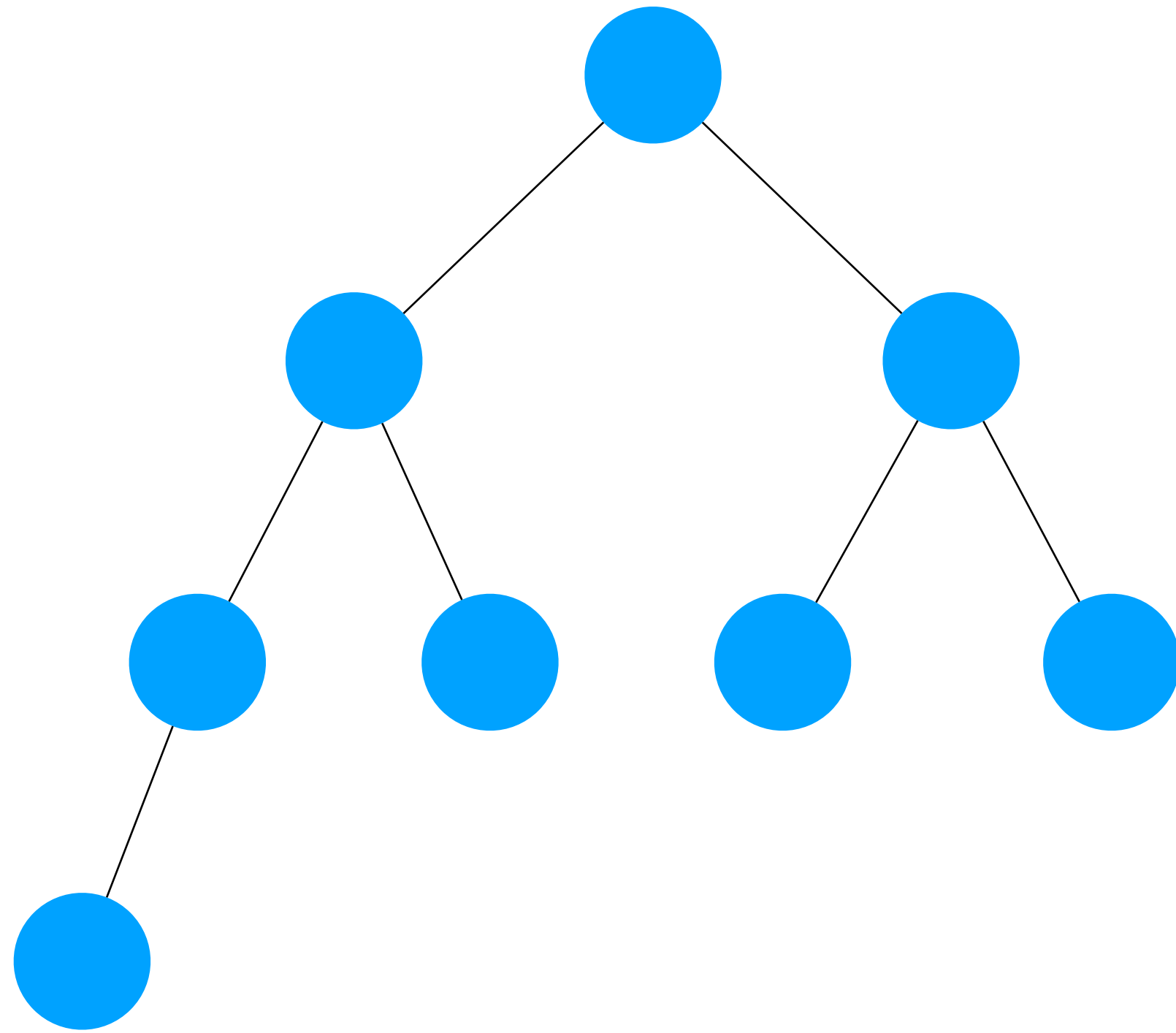
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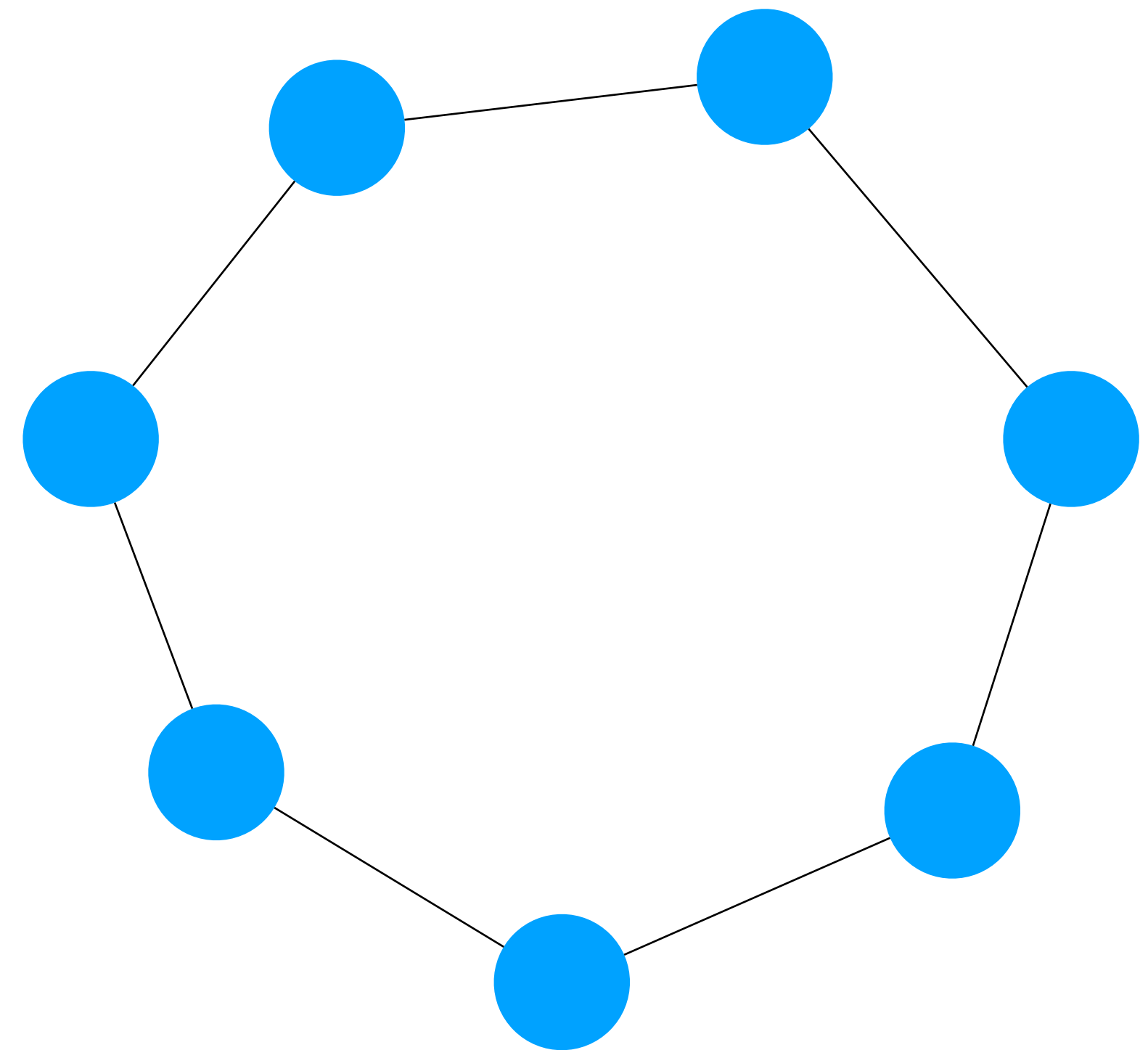
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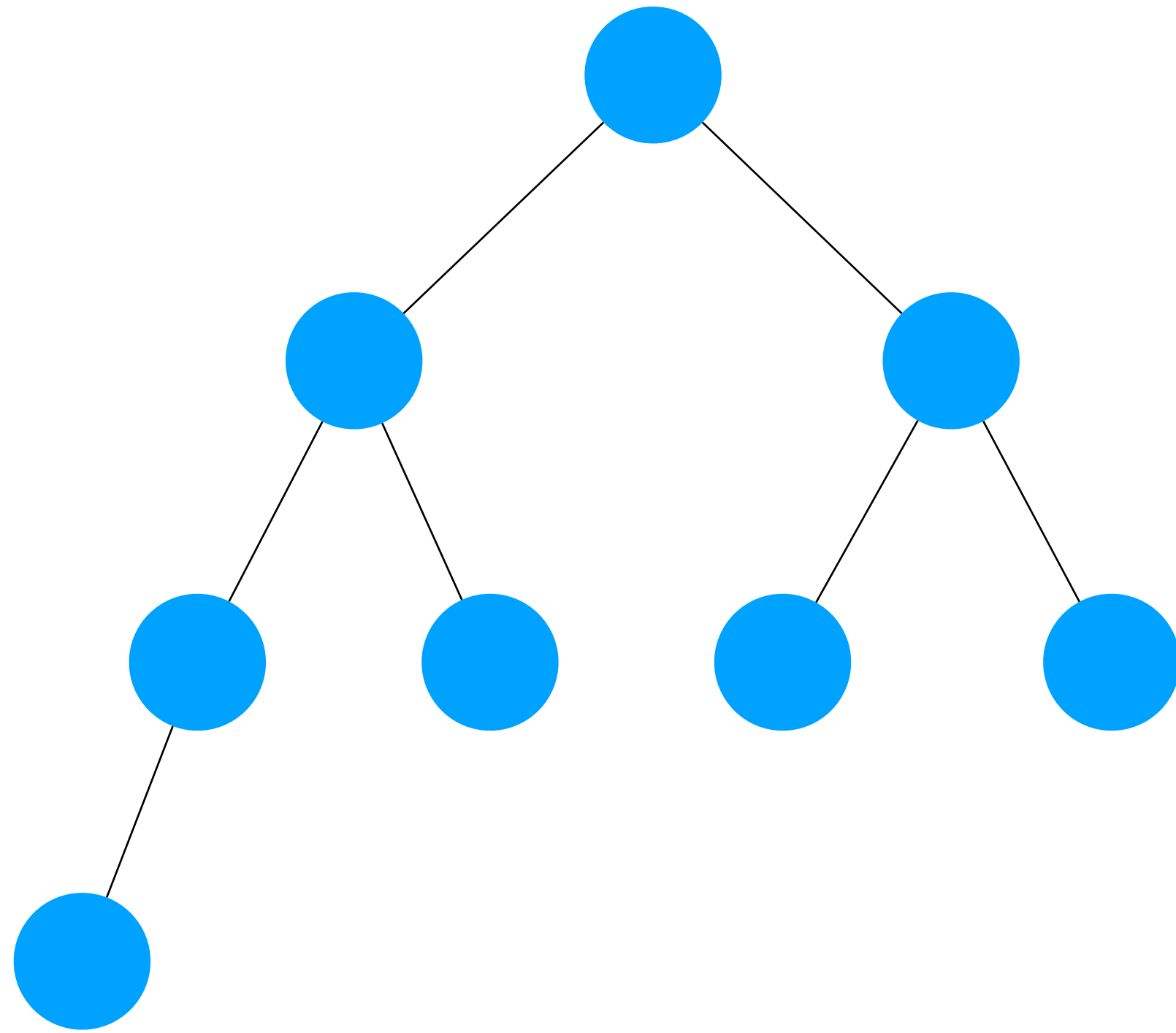
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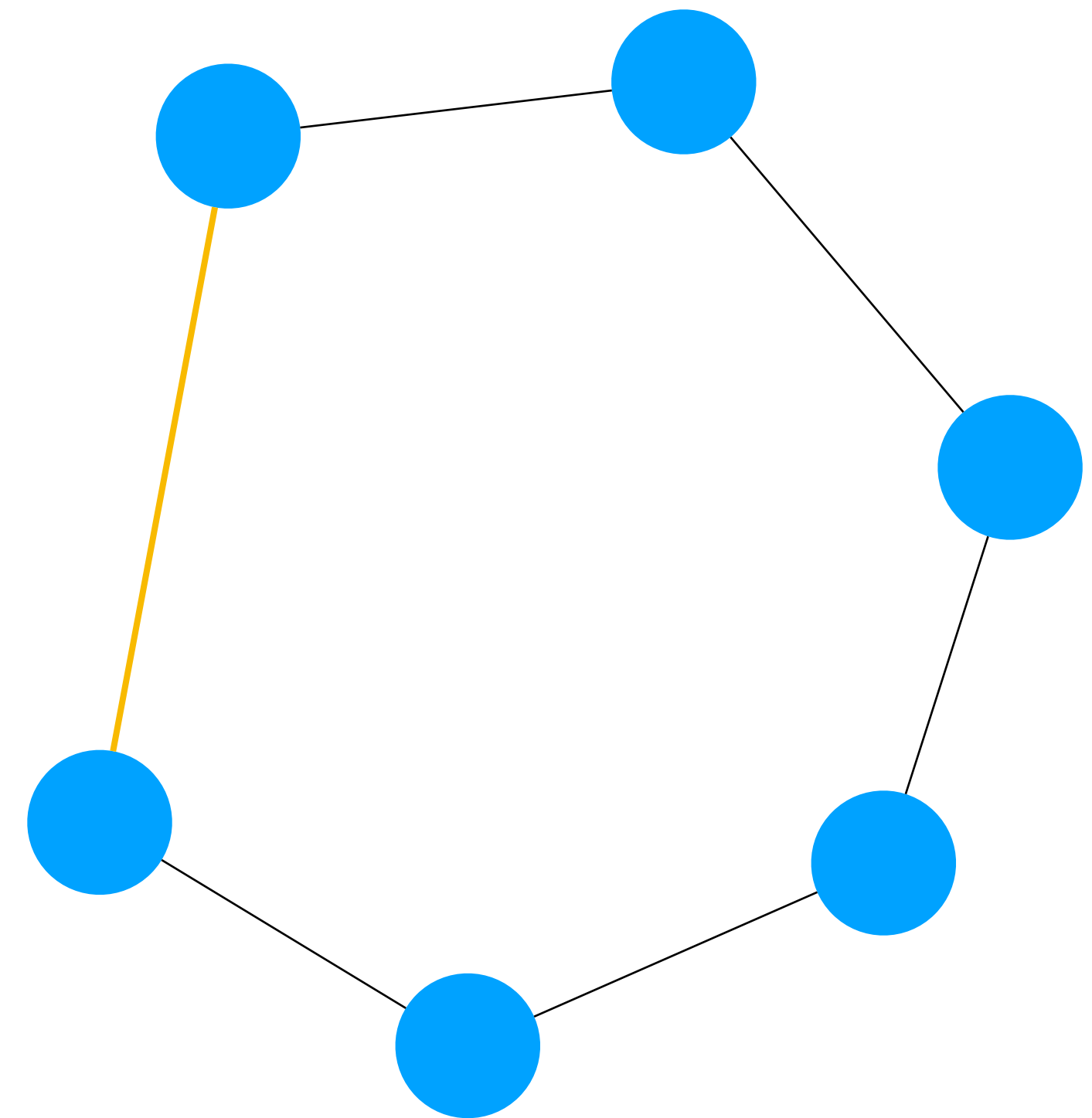
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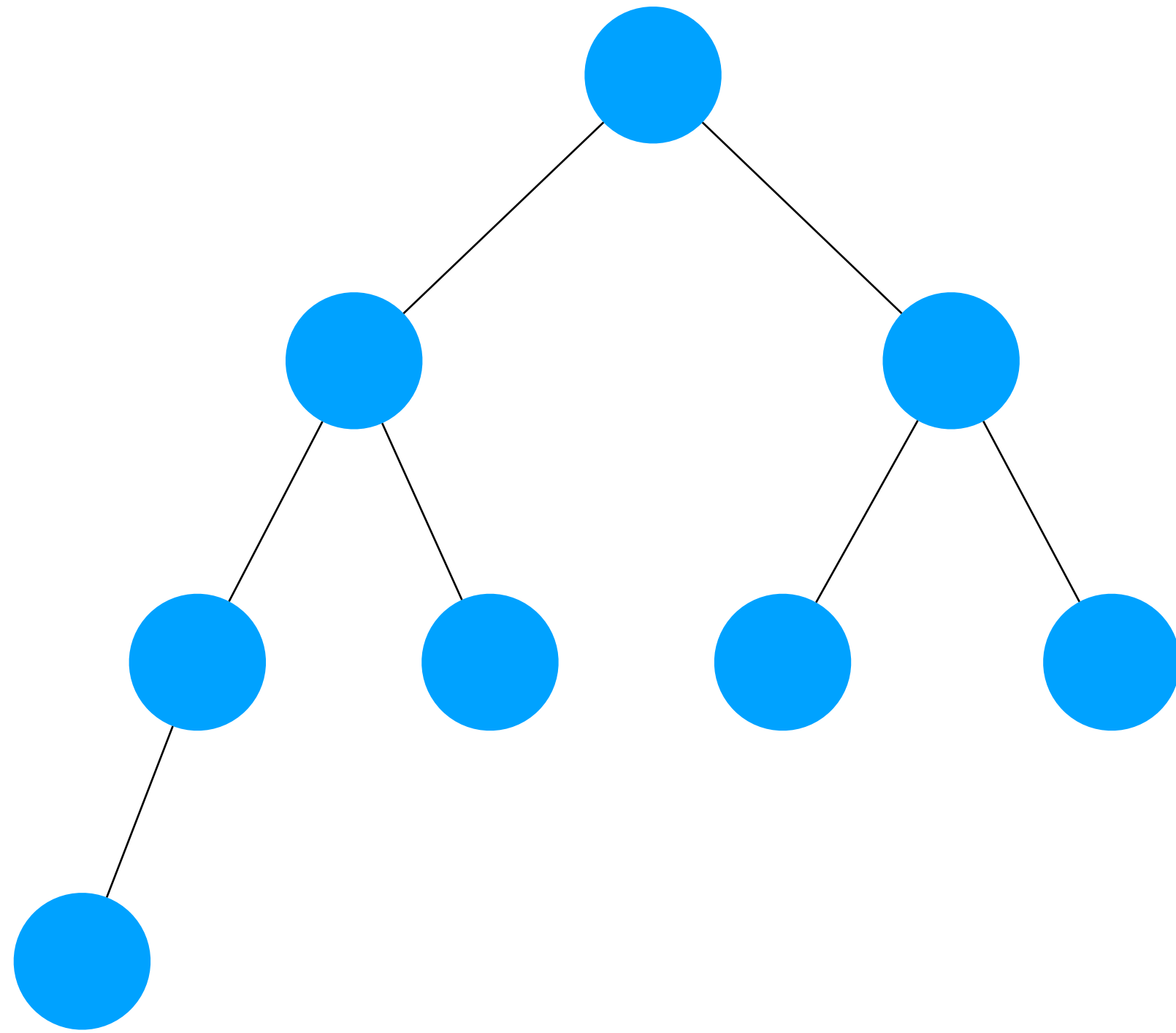
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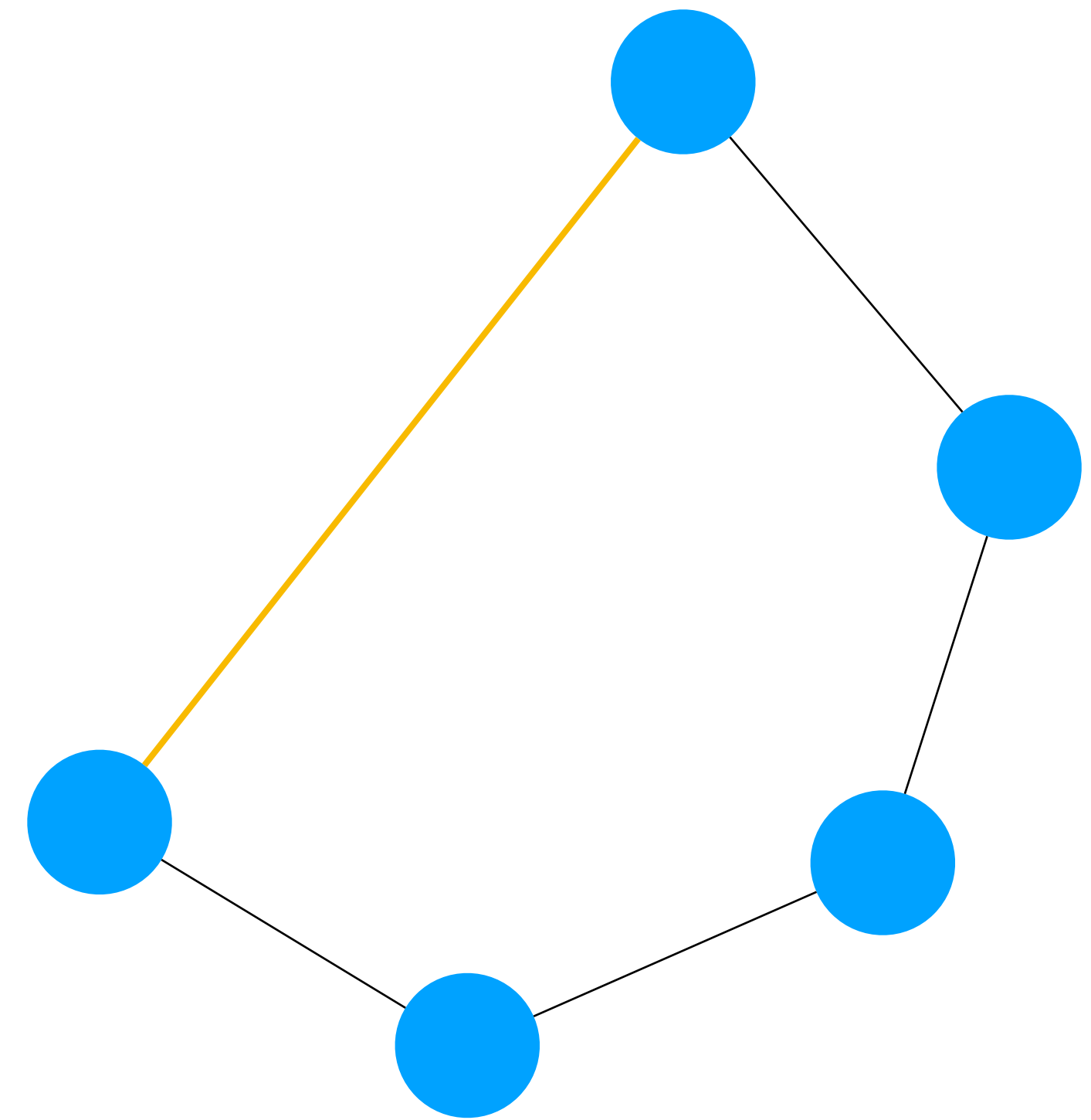
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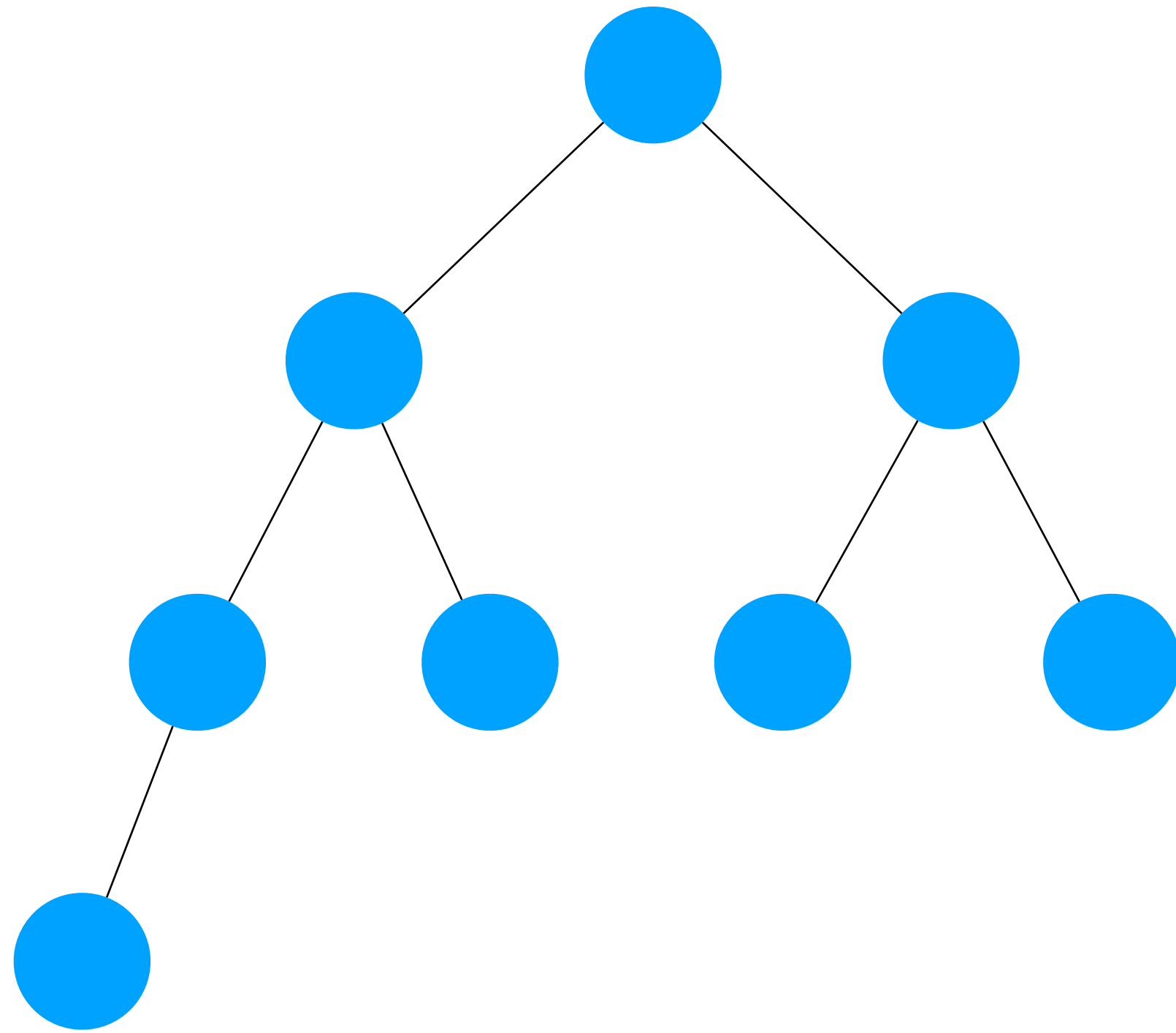
# Examples



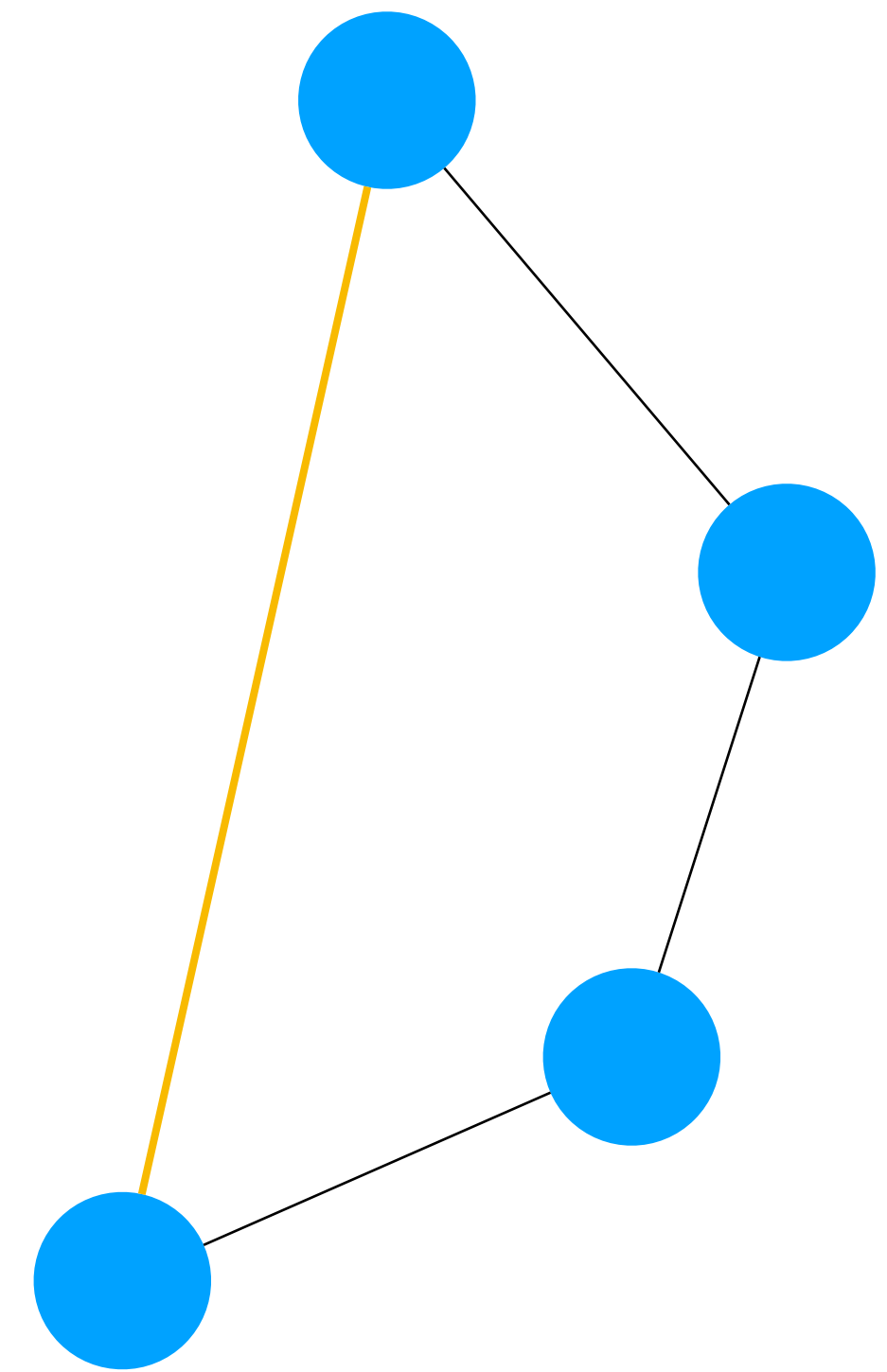
**width 1**



# Examples

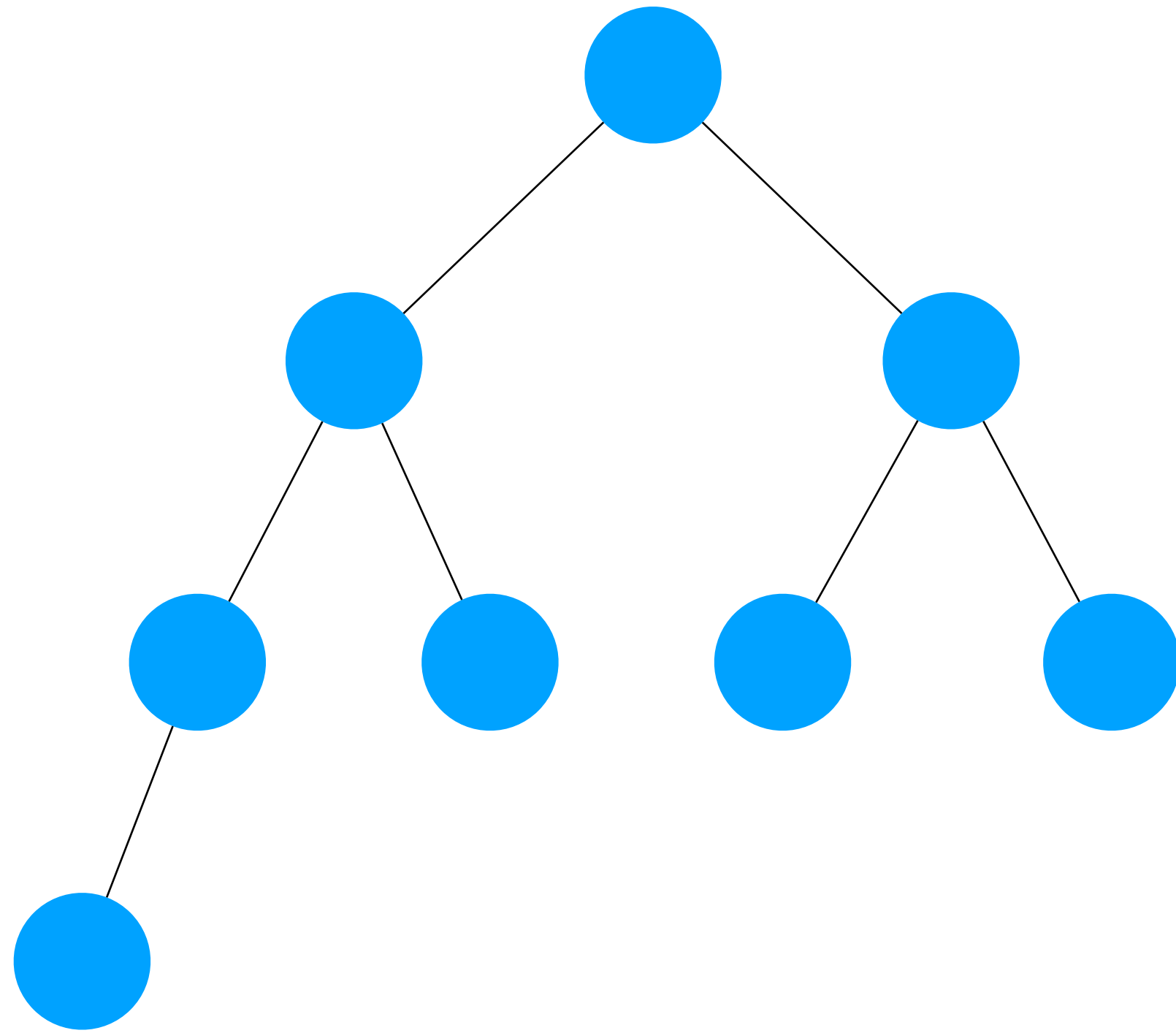


**width 1**

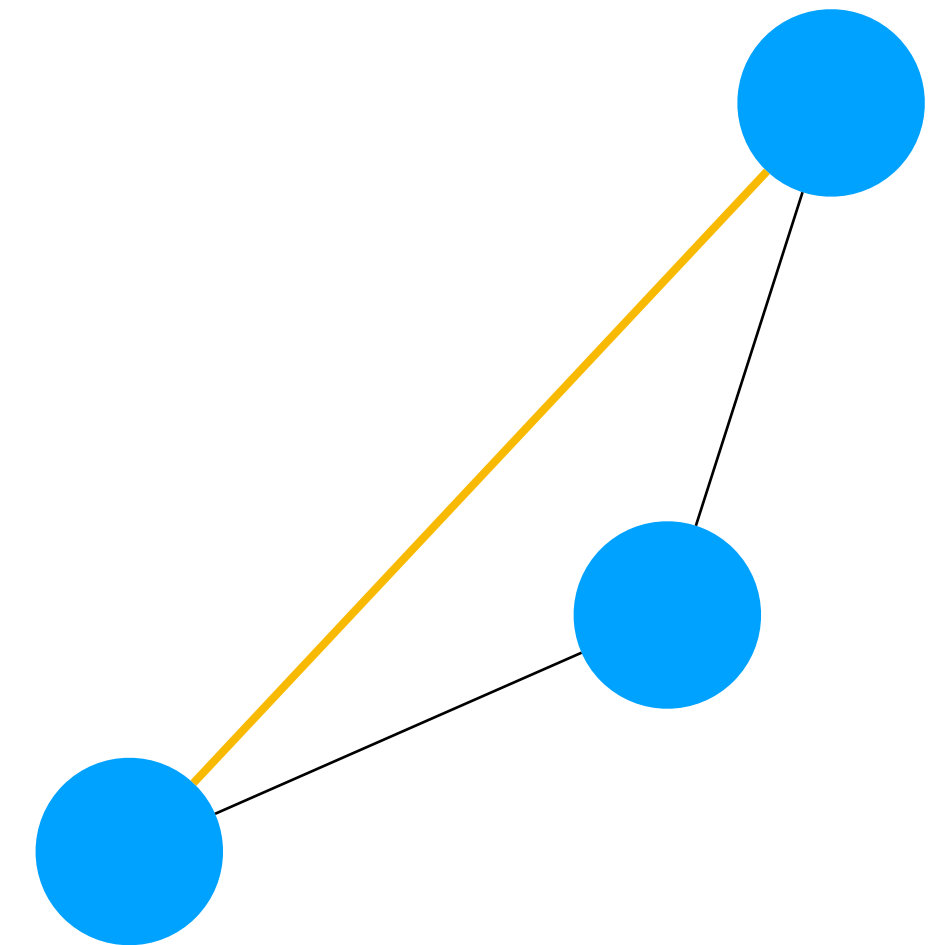




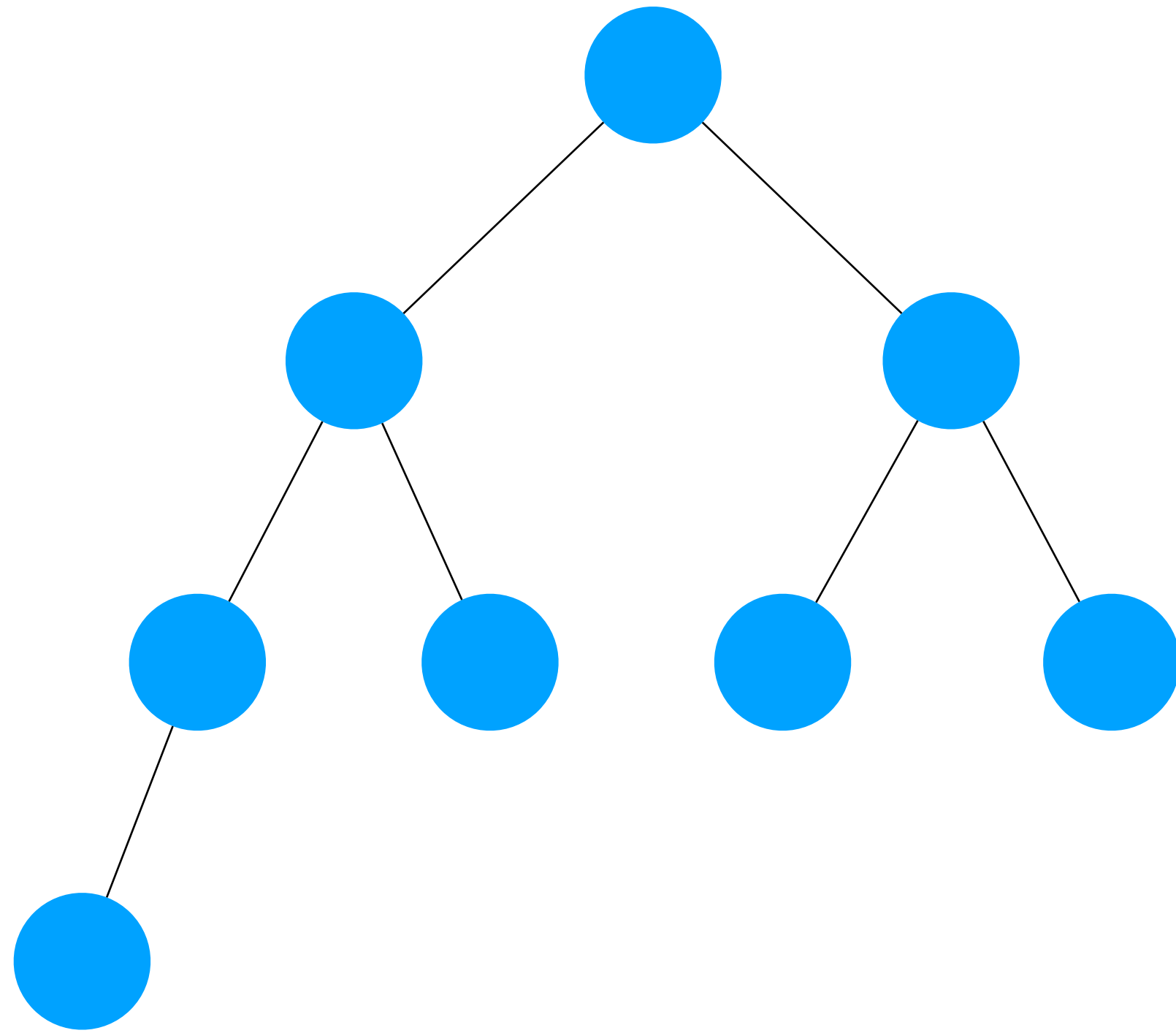
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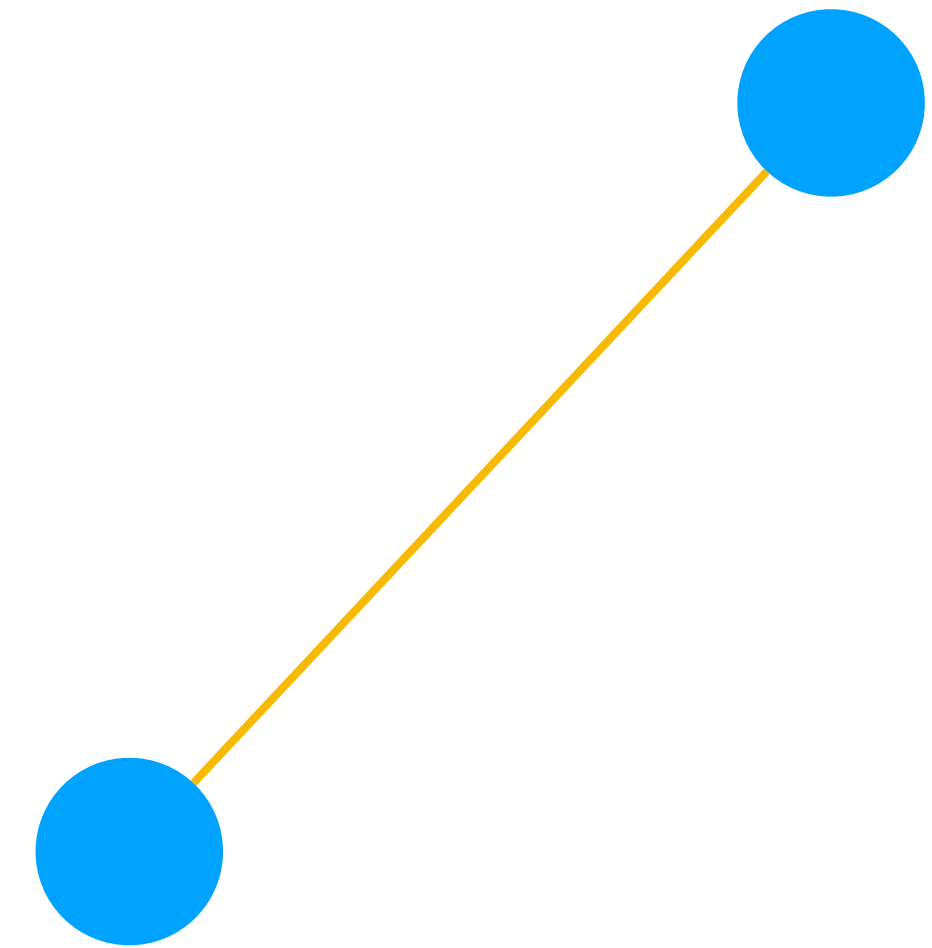
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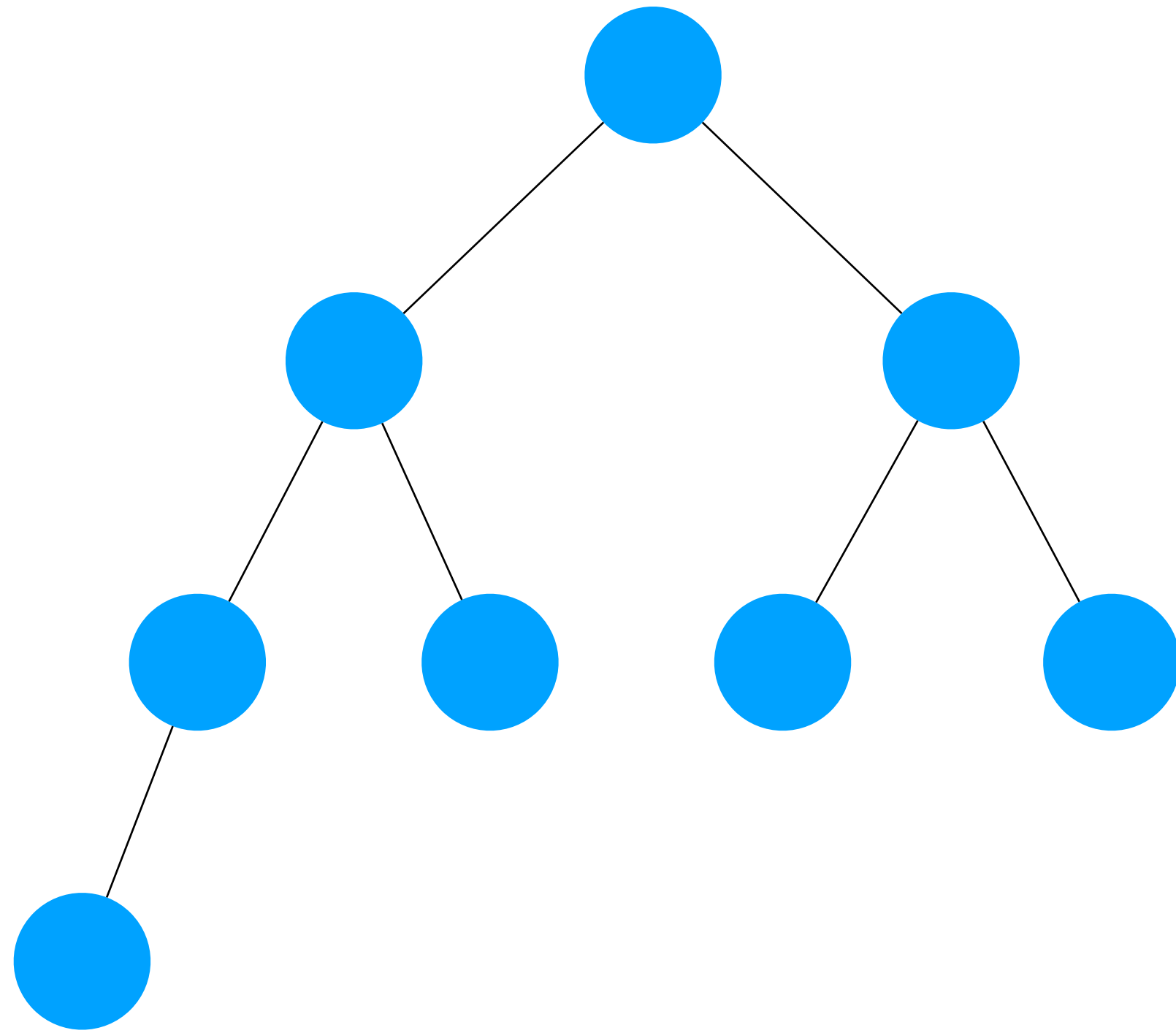
# Examples



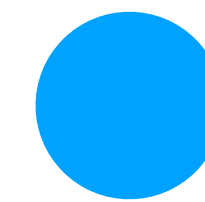
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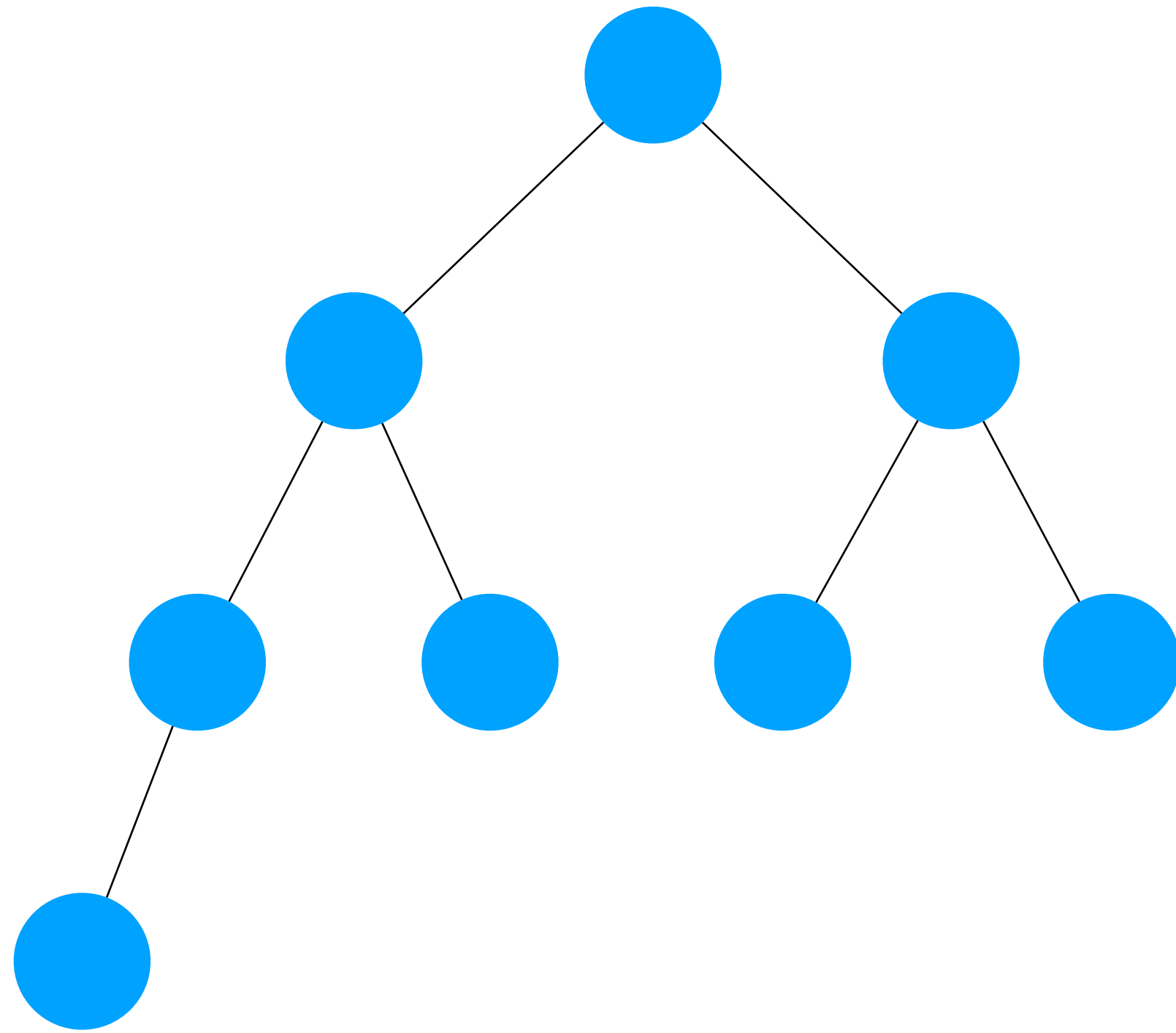
# Examples



**width 1**

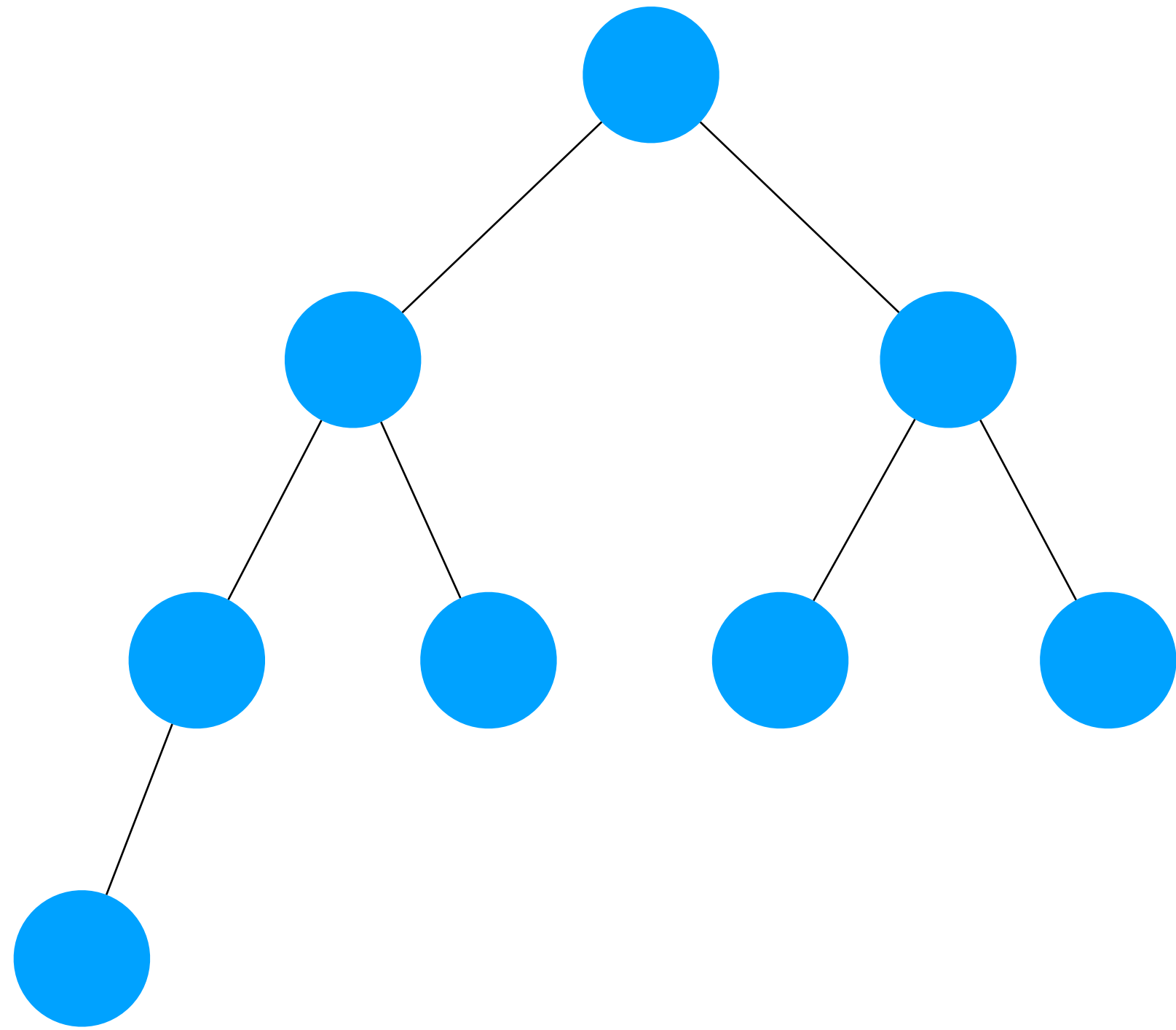


# Examples

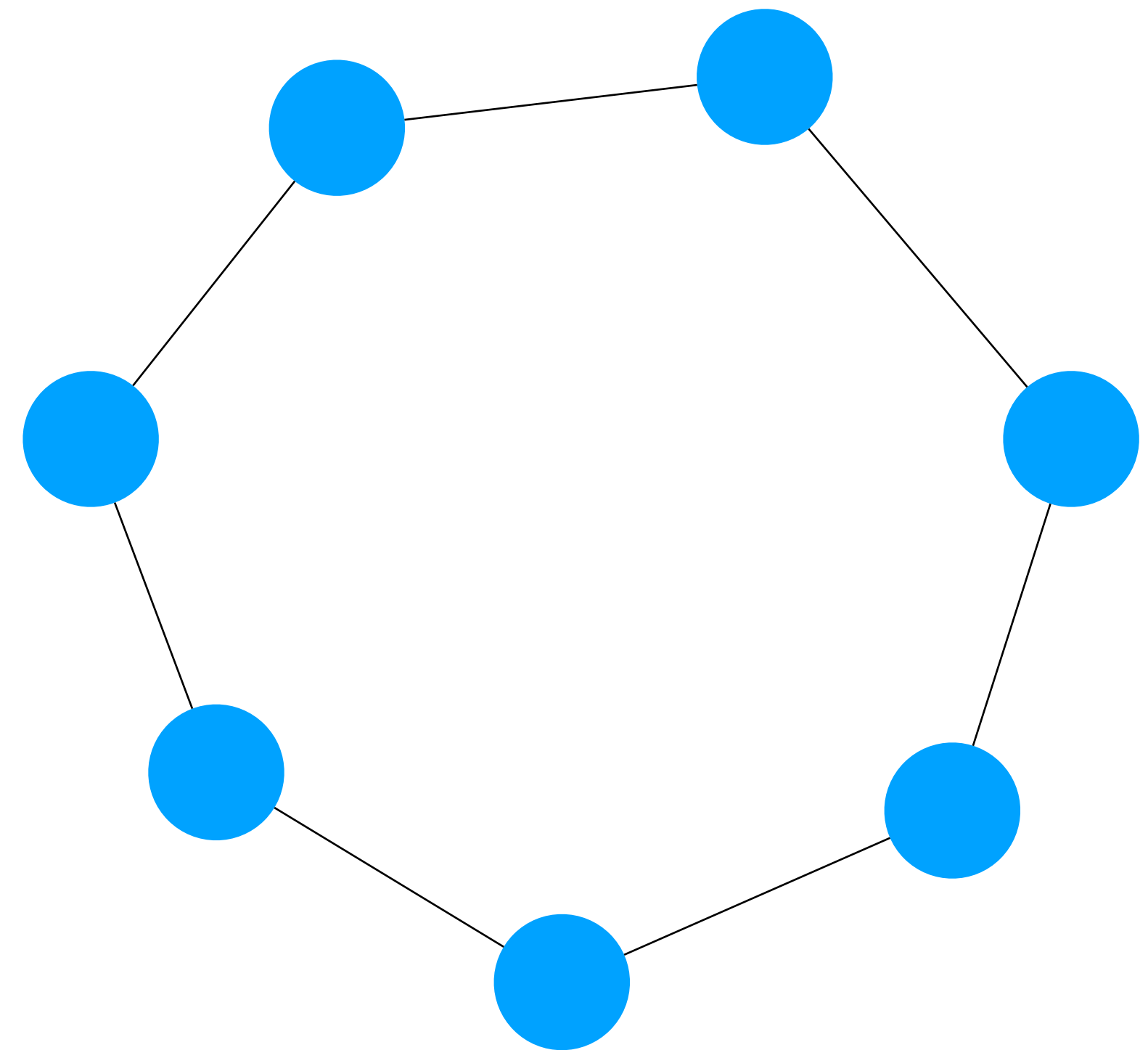


**width 1**

# Examples



**width 1**



**width 2**

# Treewidth

# Treewidth

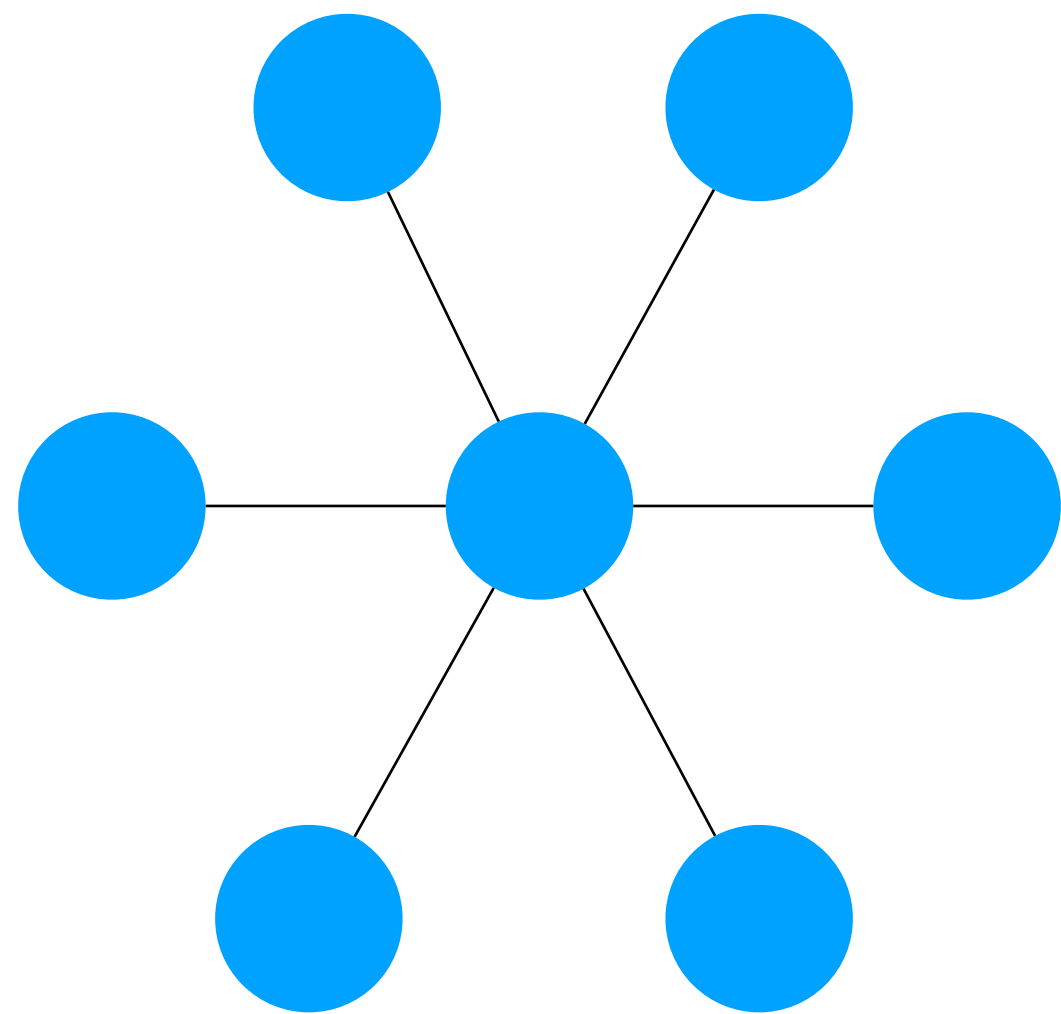
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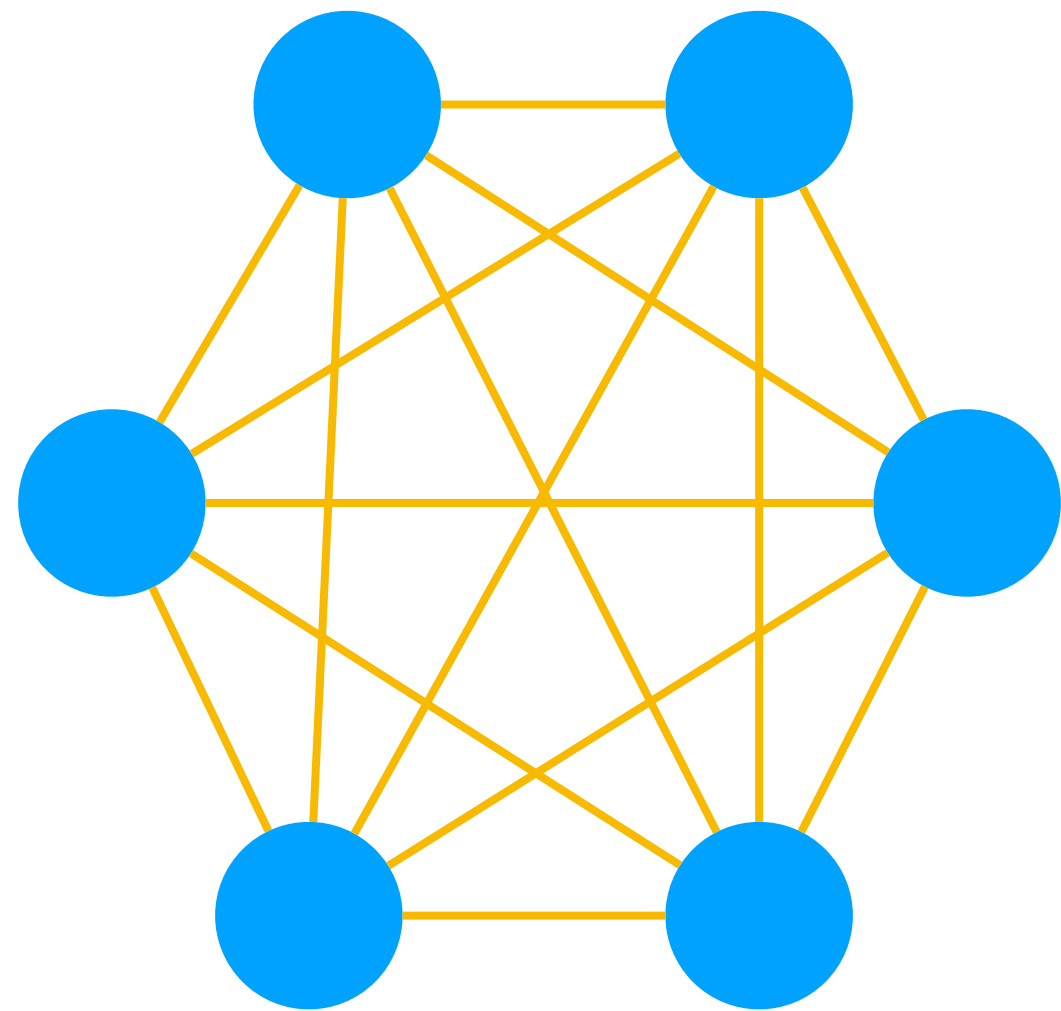




# Treewidth

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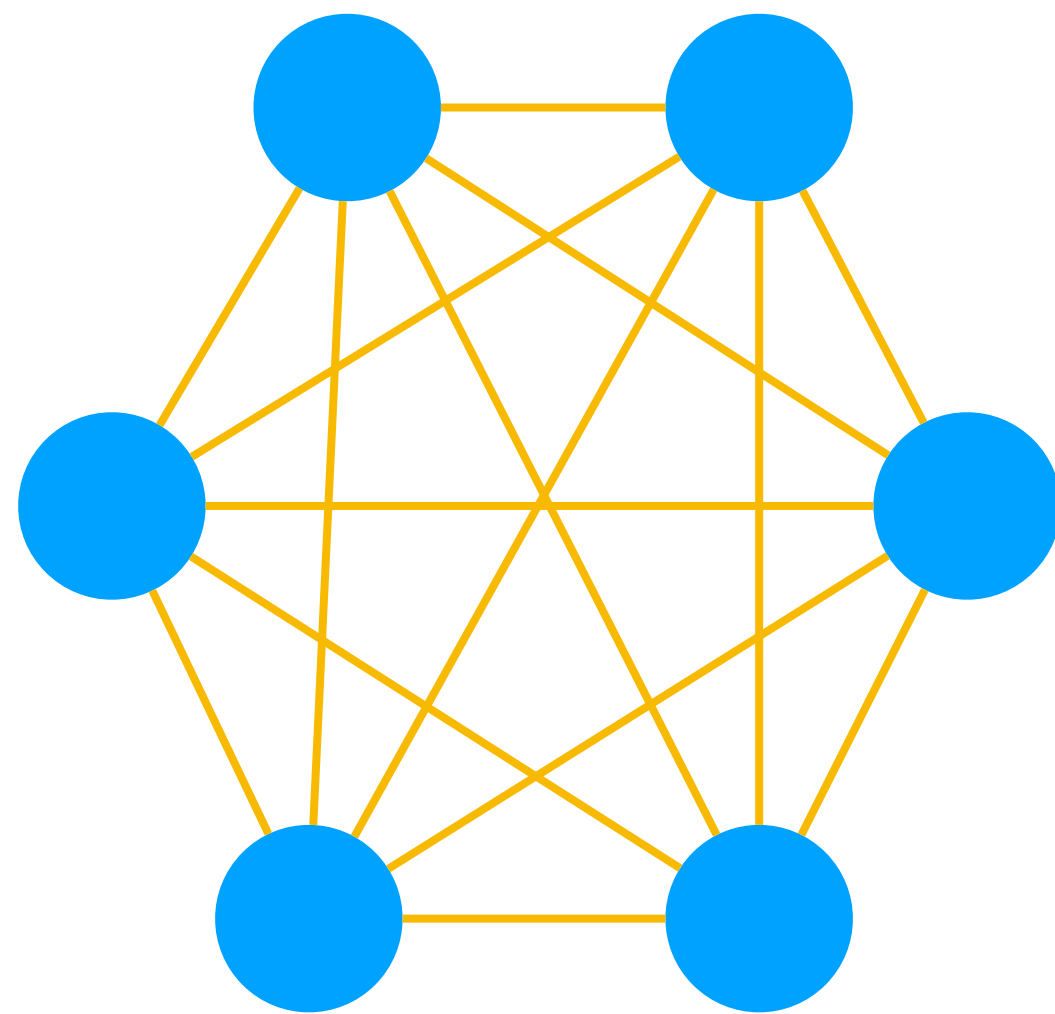
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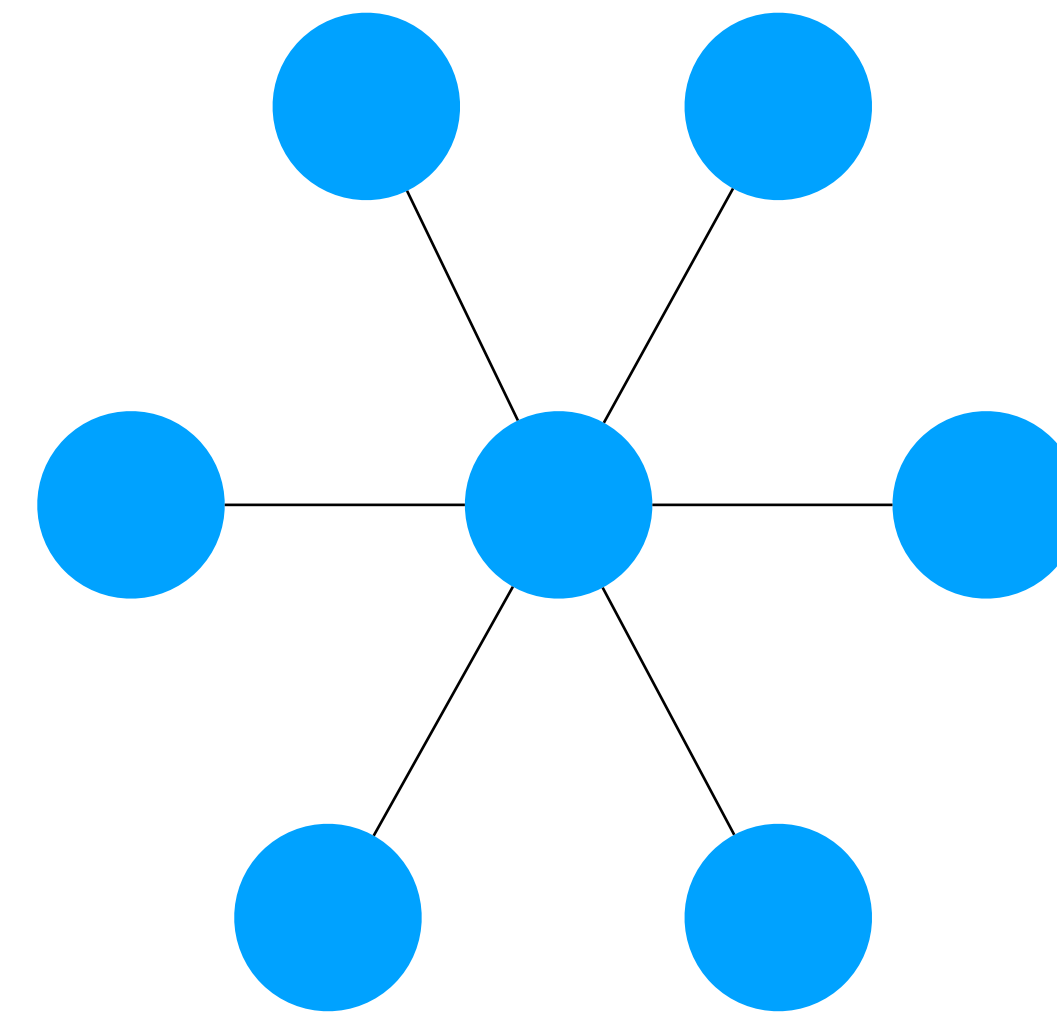
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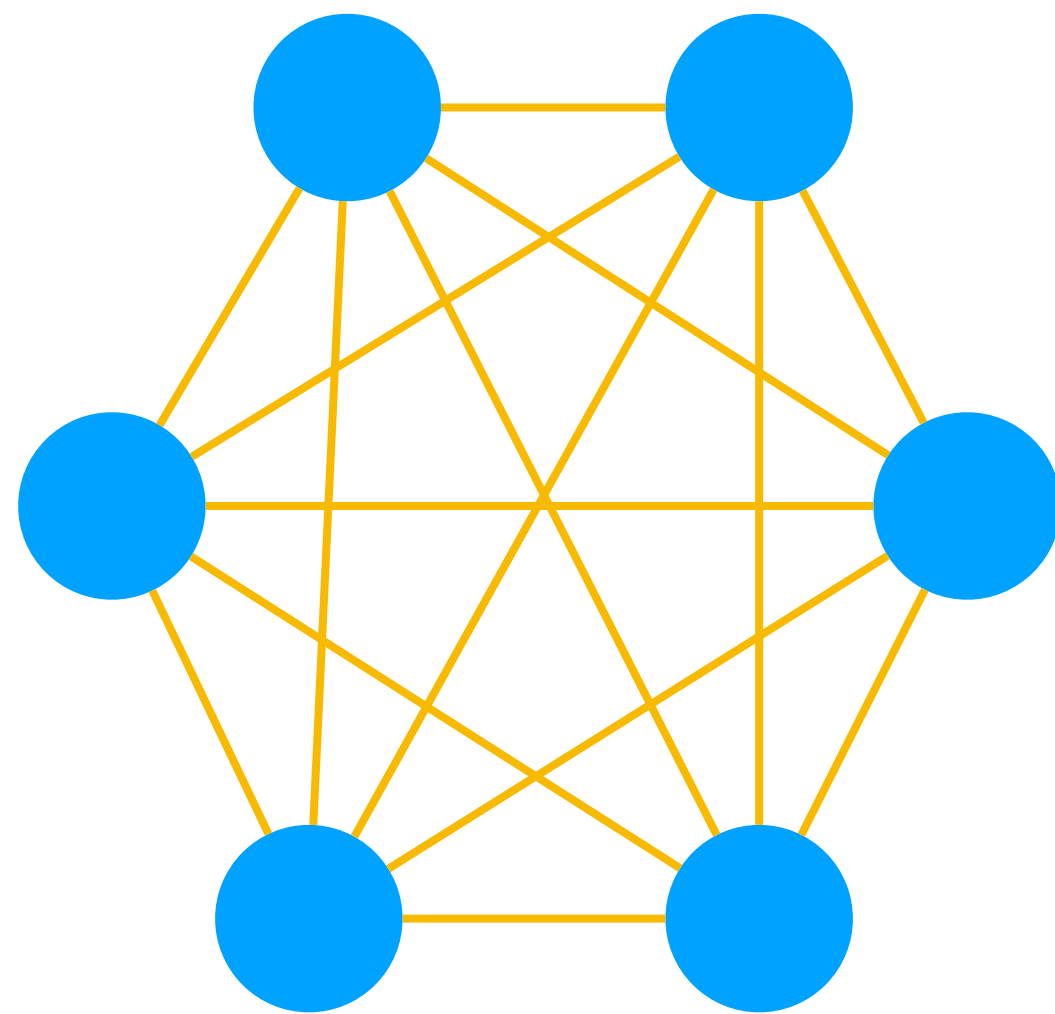
**width 6**



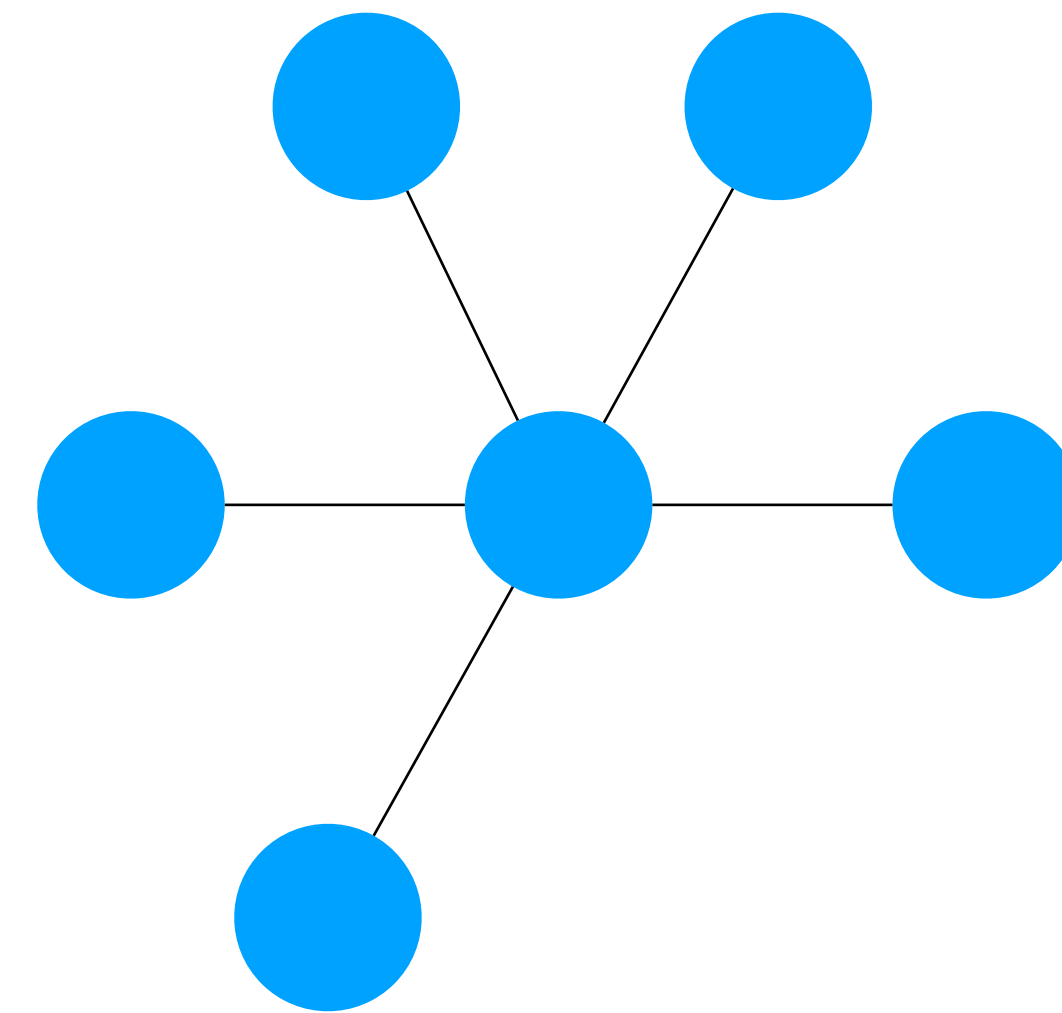
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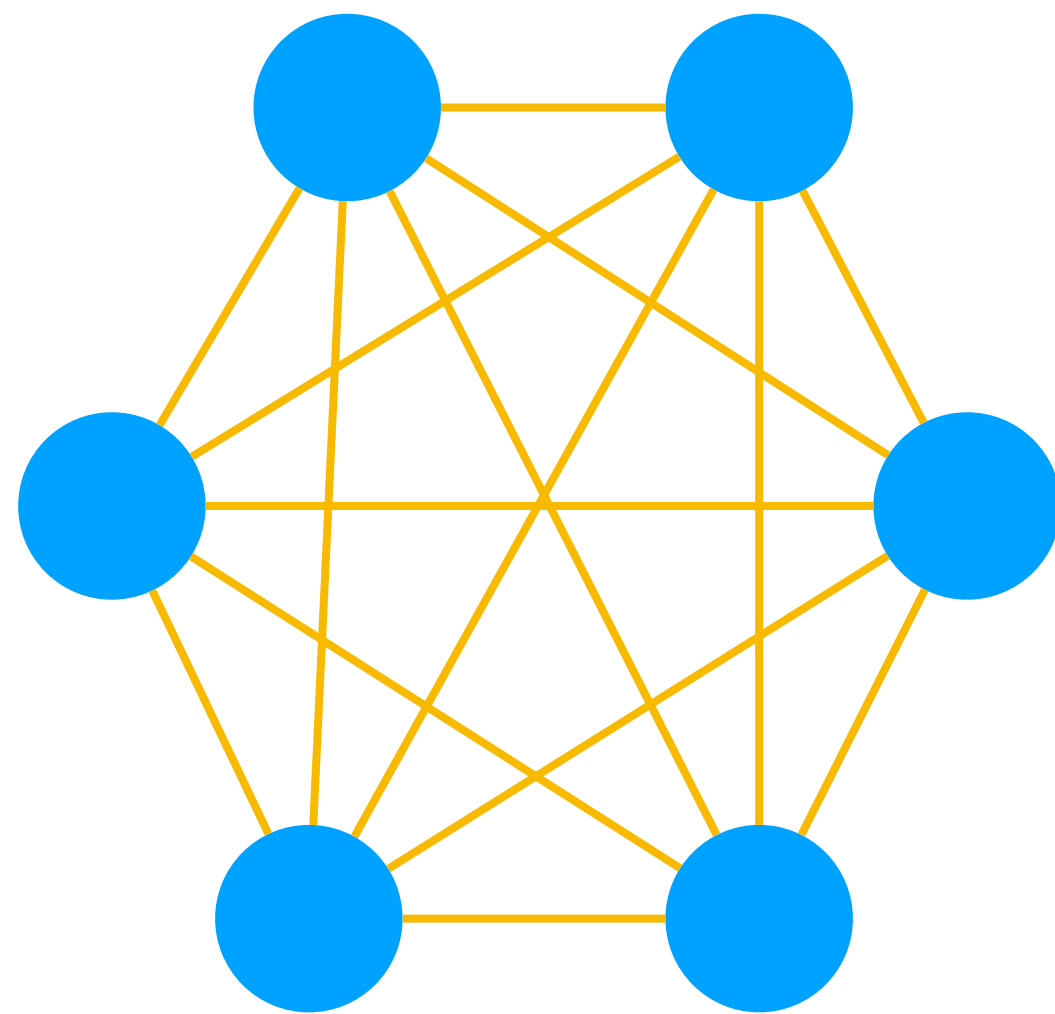
**width 6**



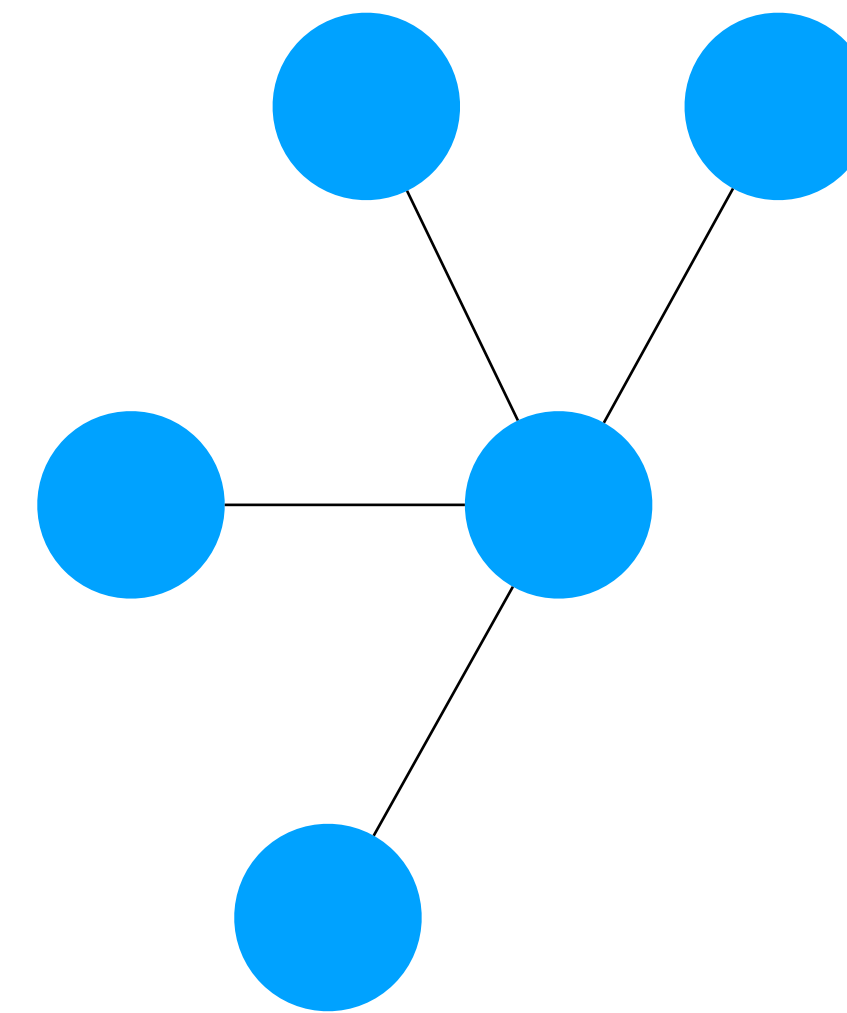
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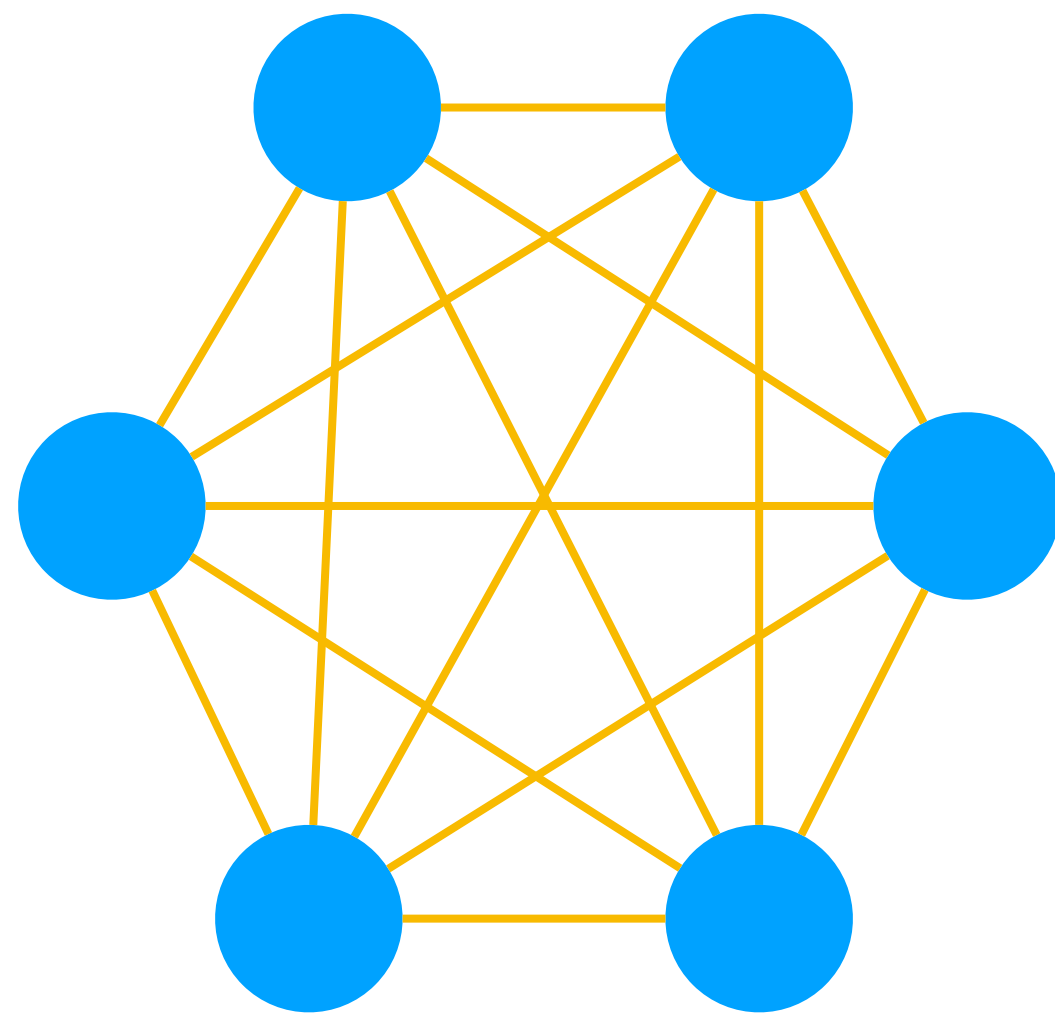
**width 6**



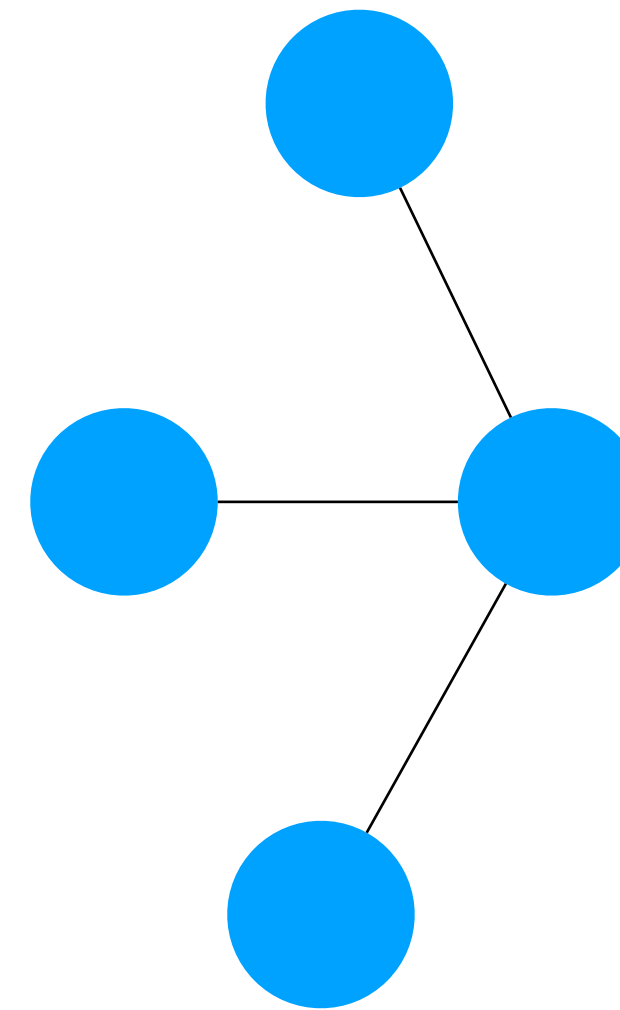
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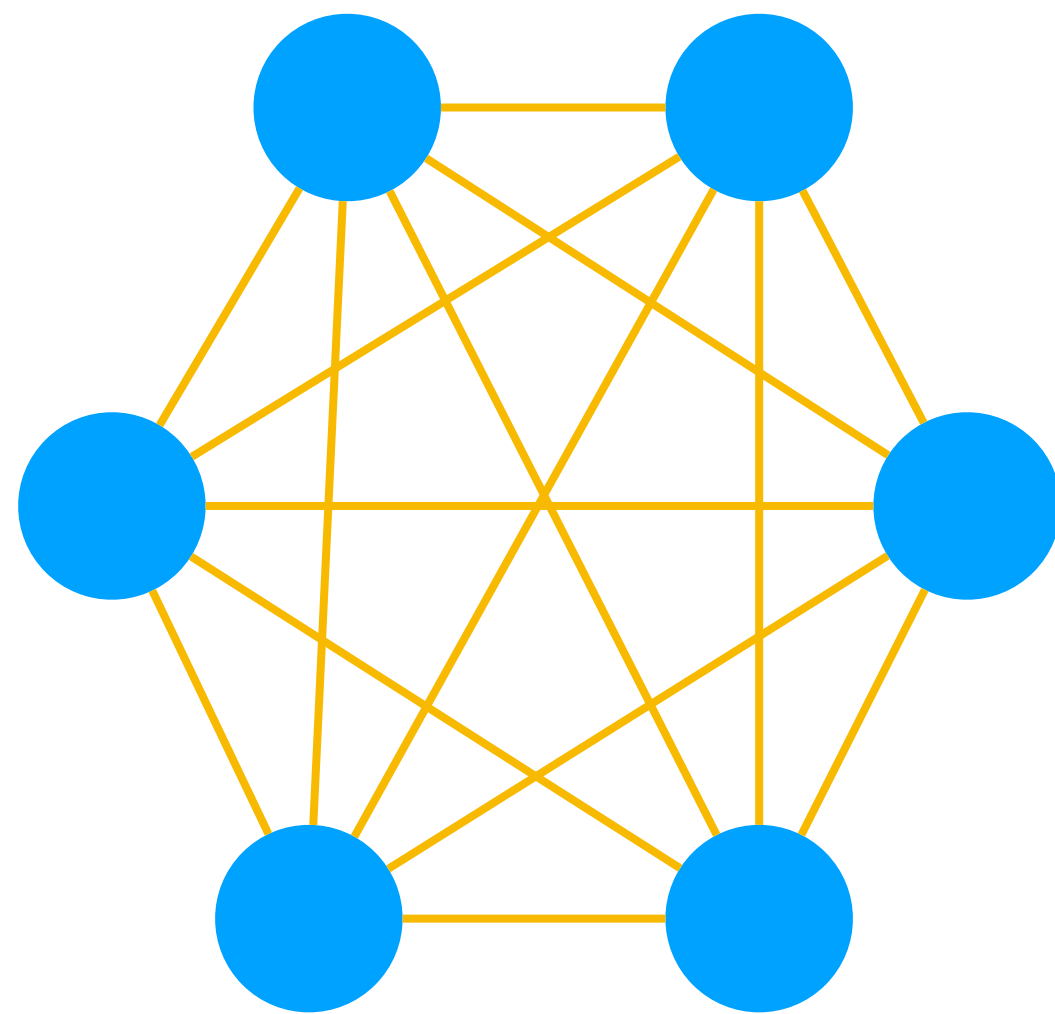
**width 6**



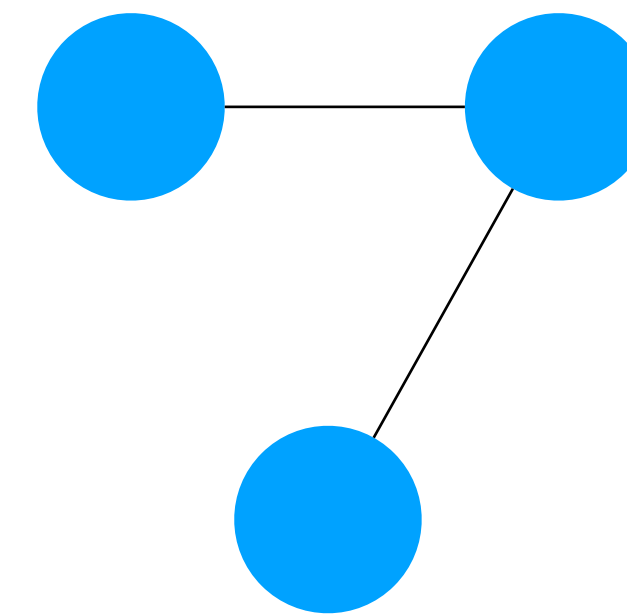
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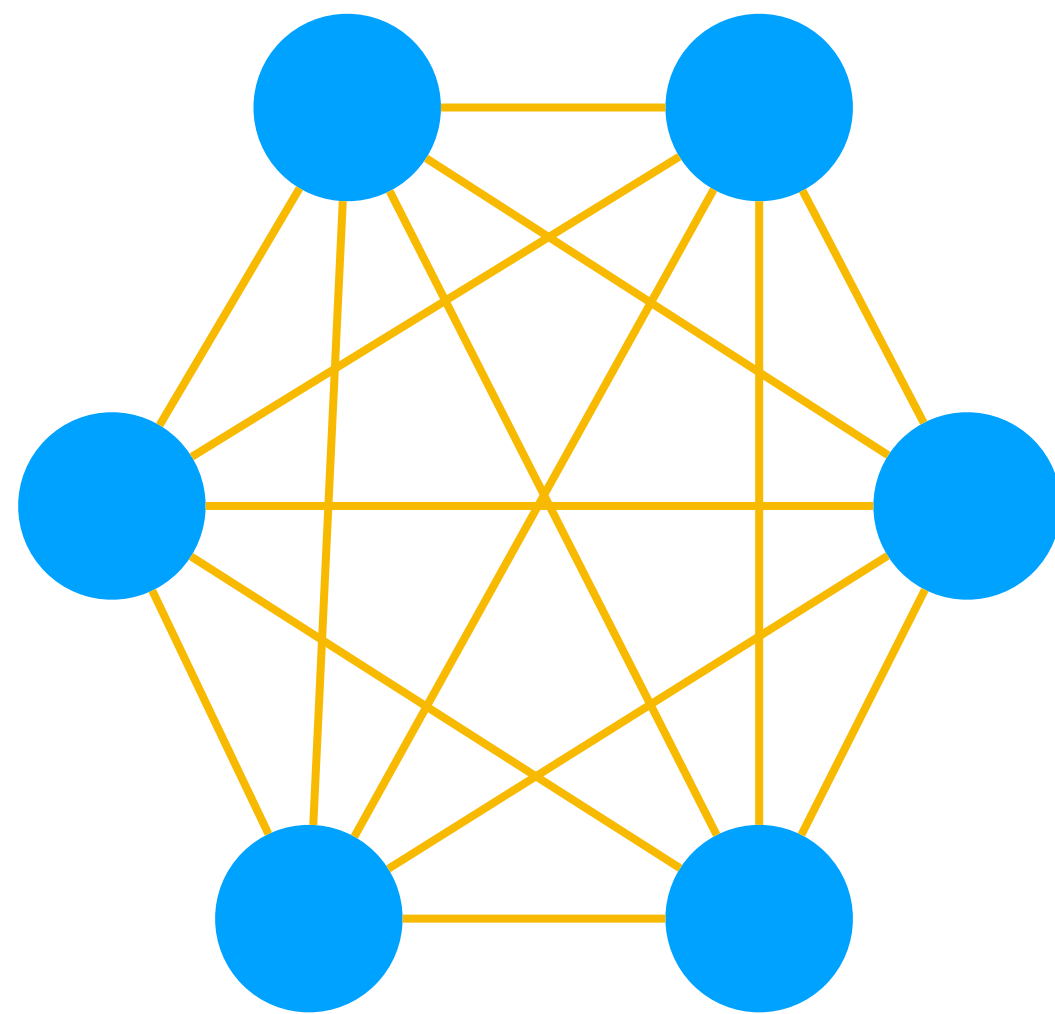
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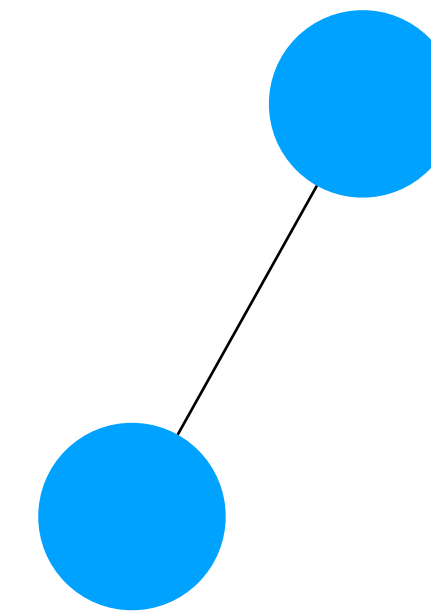
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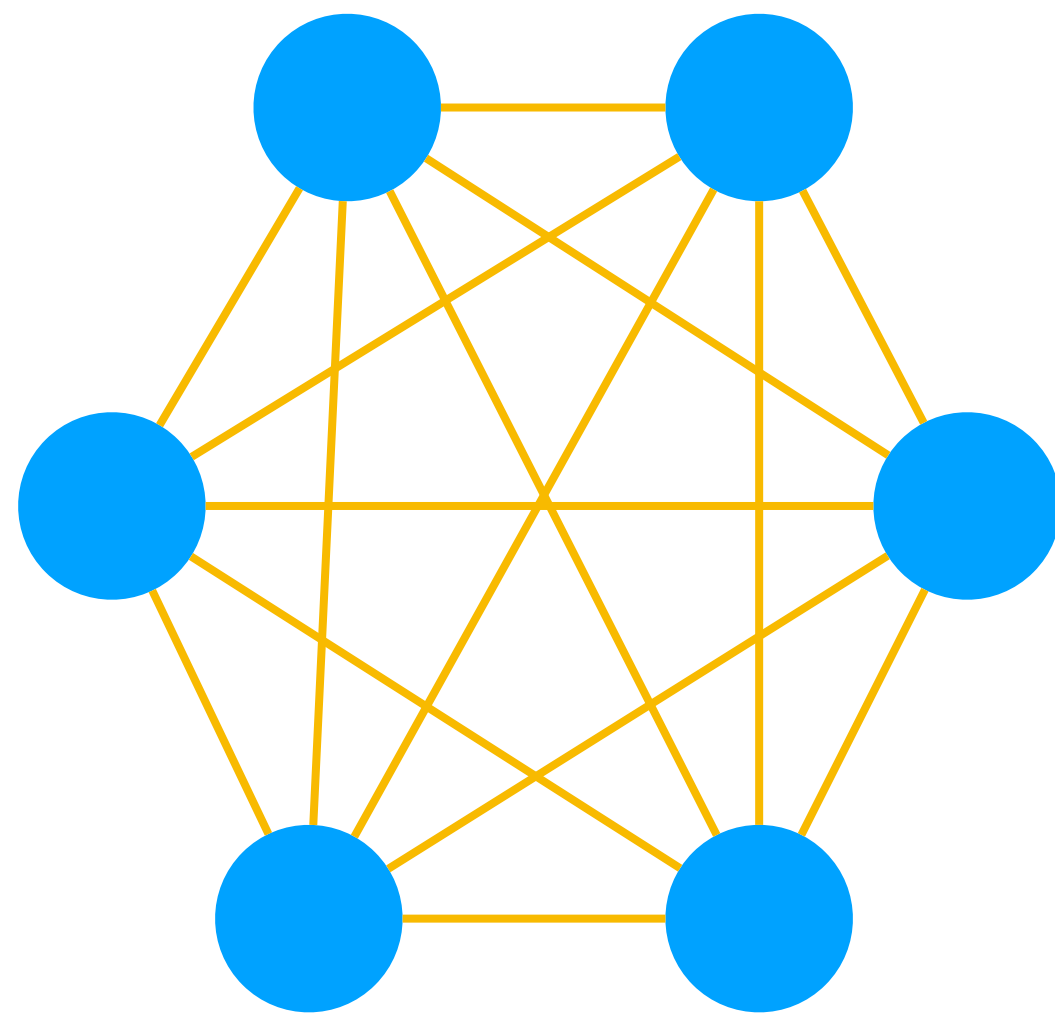
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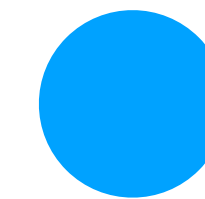
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**width 6**

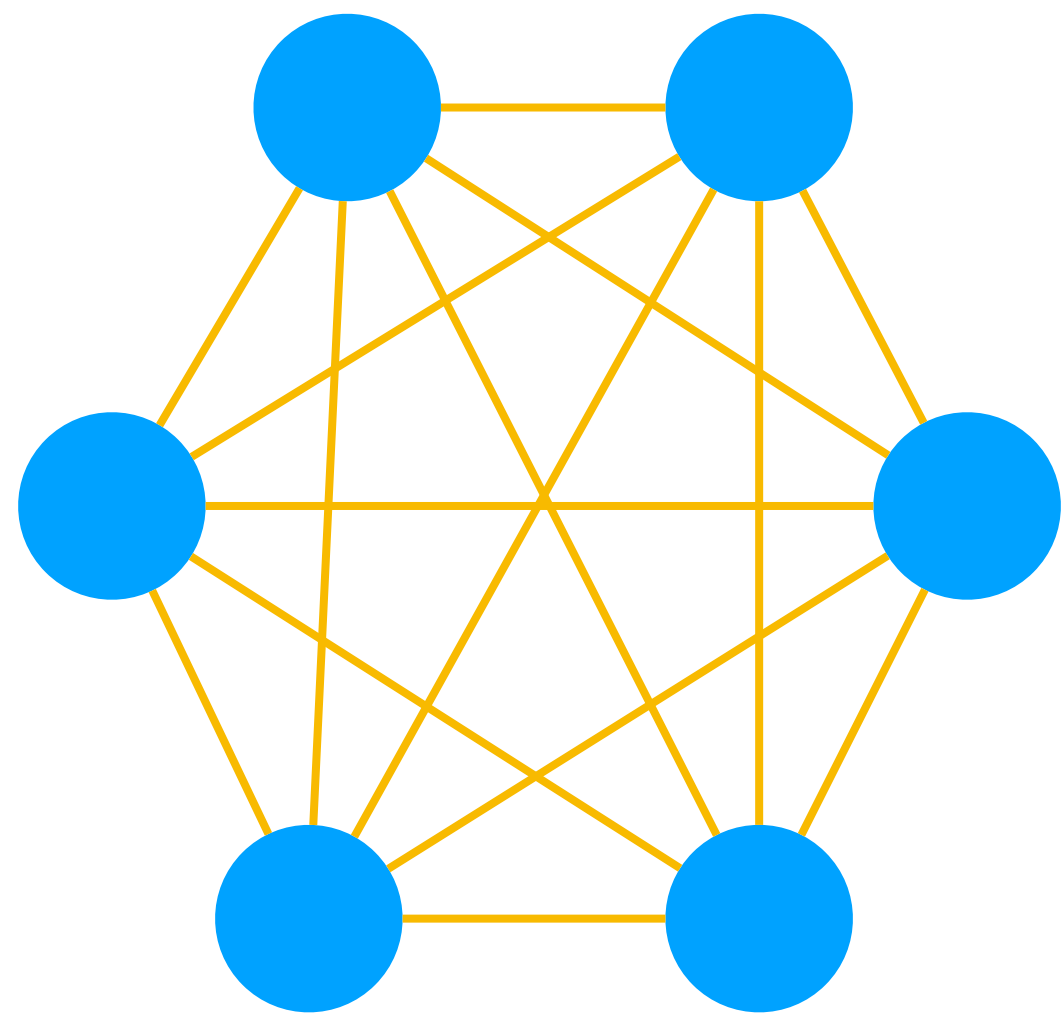




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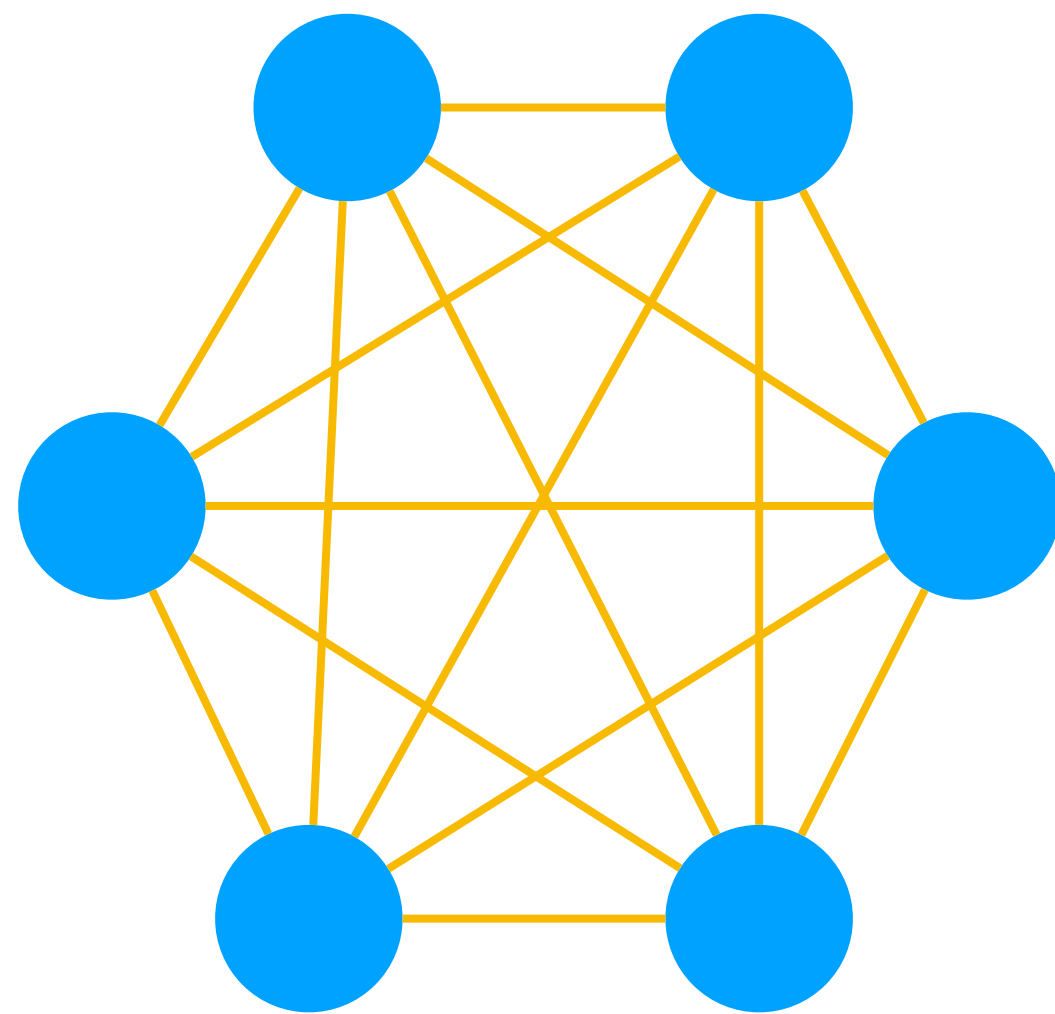


**width 6**

# Treewidth

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The **treewidth** of a graph  $G$  is the minimum width of an elimination ordering of  $G$ .



**width 6**

**width 1**

# Degree and Treewidth

## Proposition

If  $G$  has minimum degree  $d$  then  $d \leq \mathbf{tw}(G)$ .

# Degree and Treewidth

## Proposition

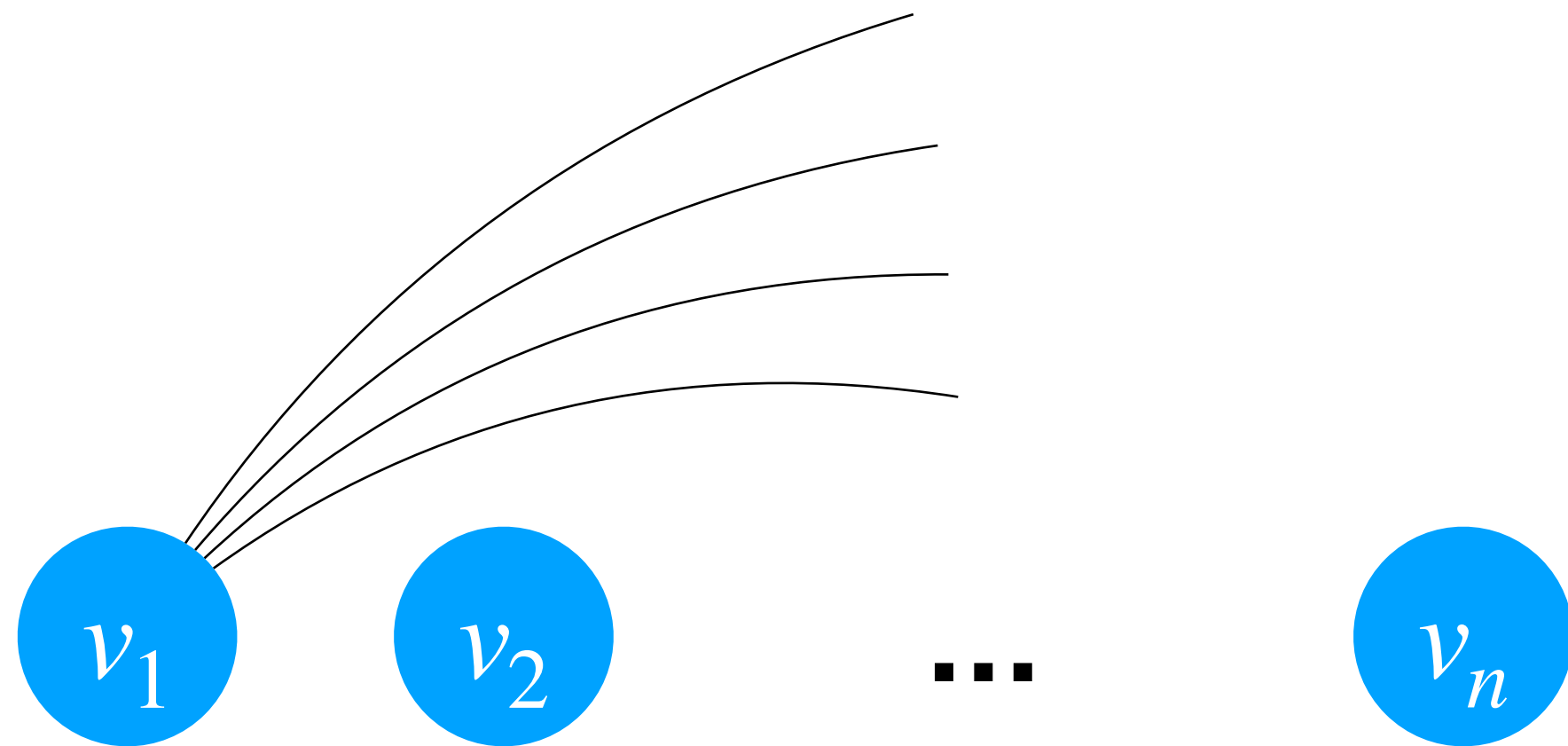
If  $G$  has minimum degree  $d$  then  $d \leq \mathbf{tw}(G)$ .



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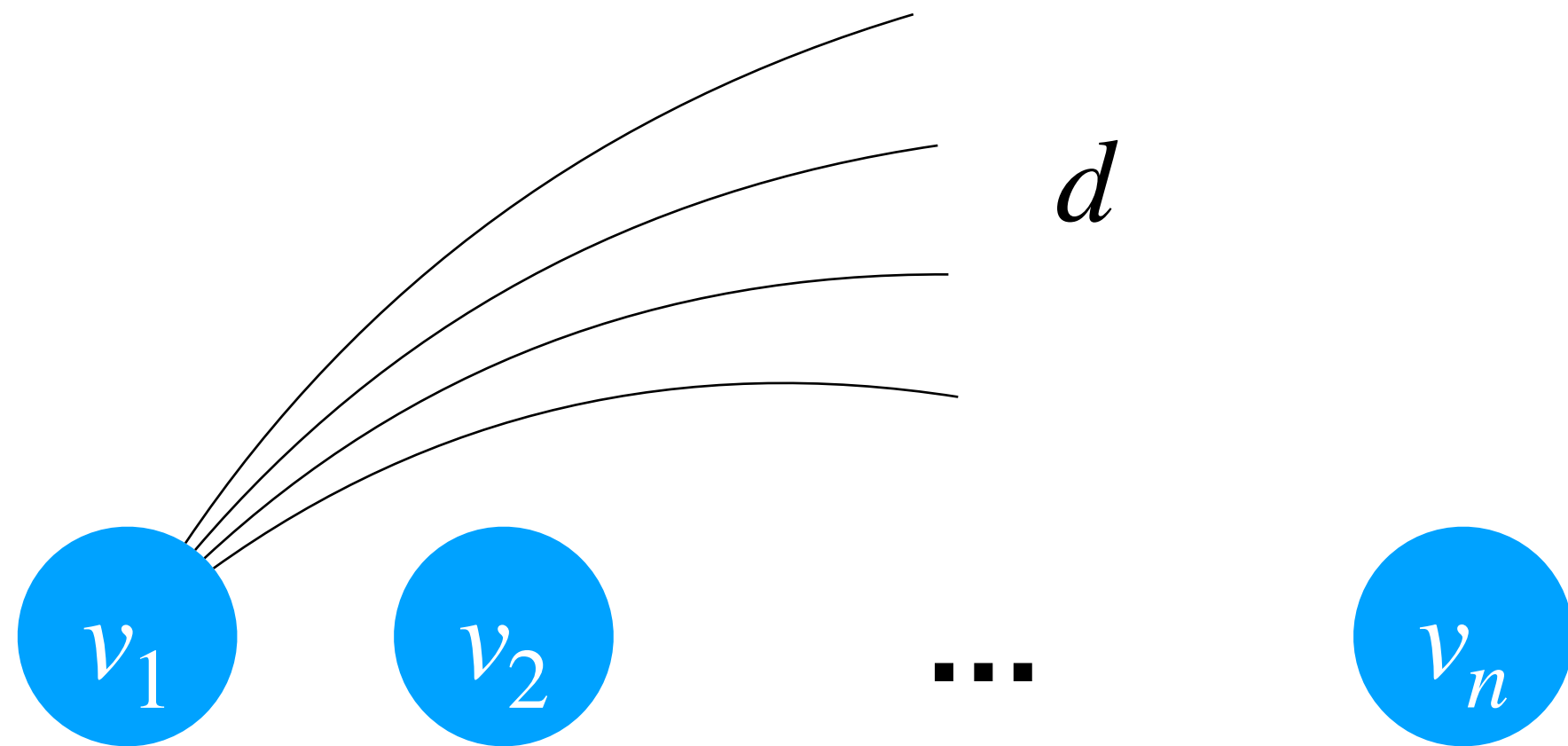
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# Degree and Treewidth

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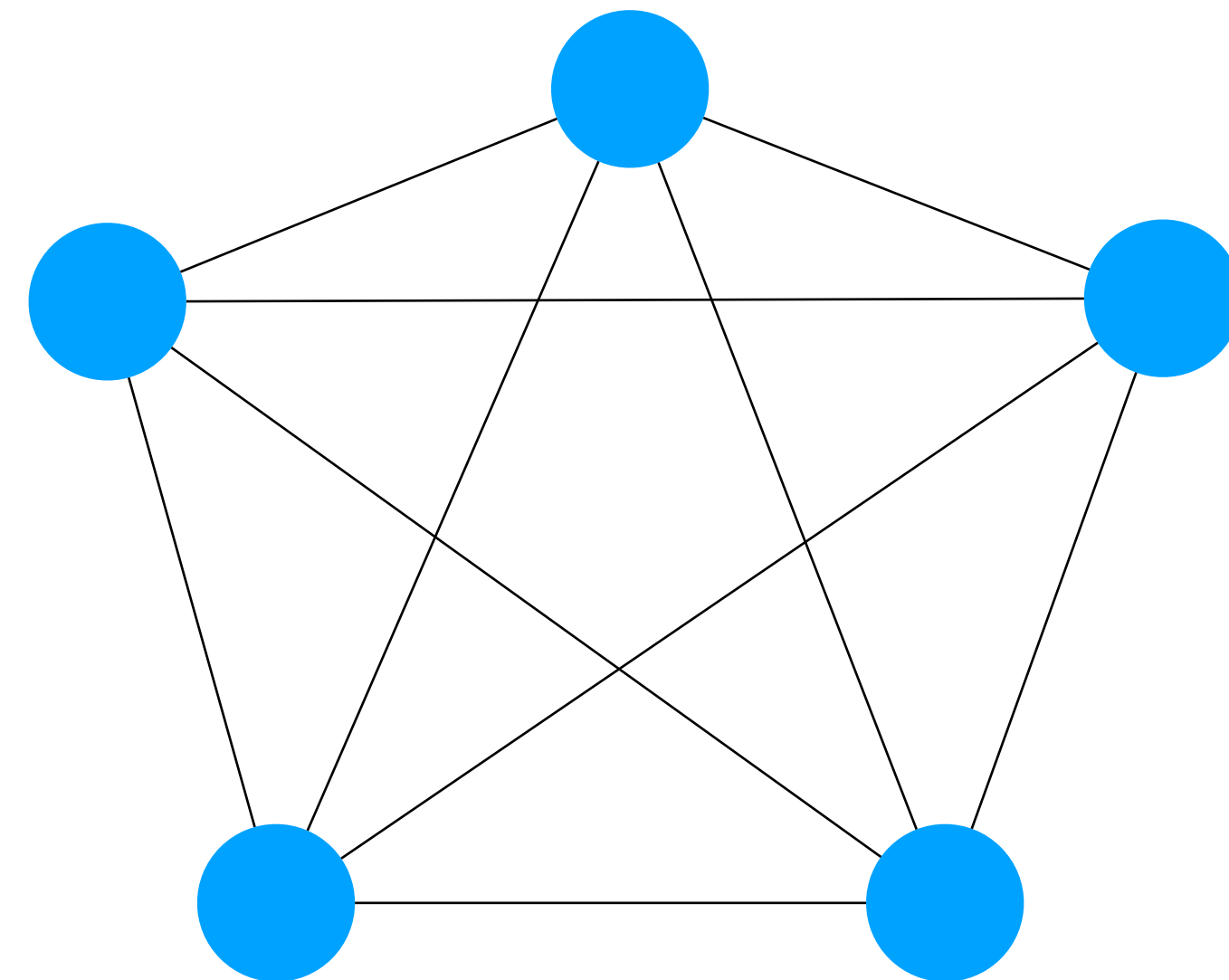
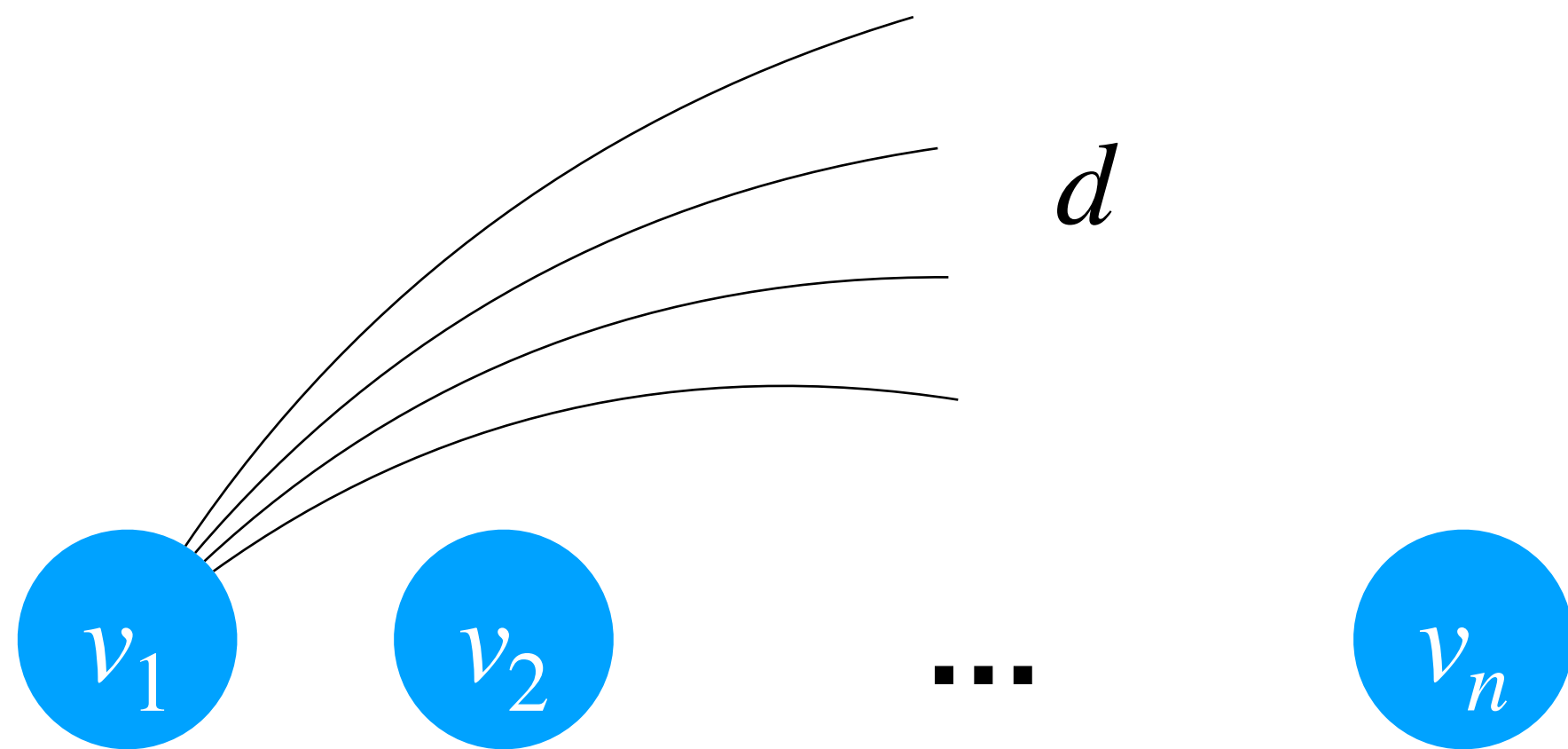
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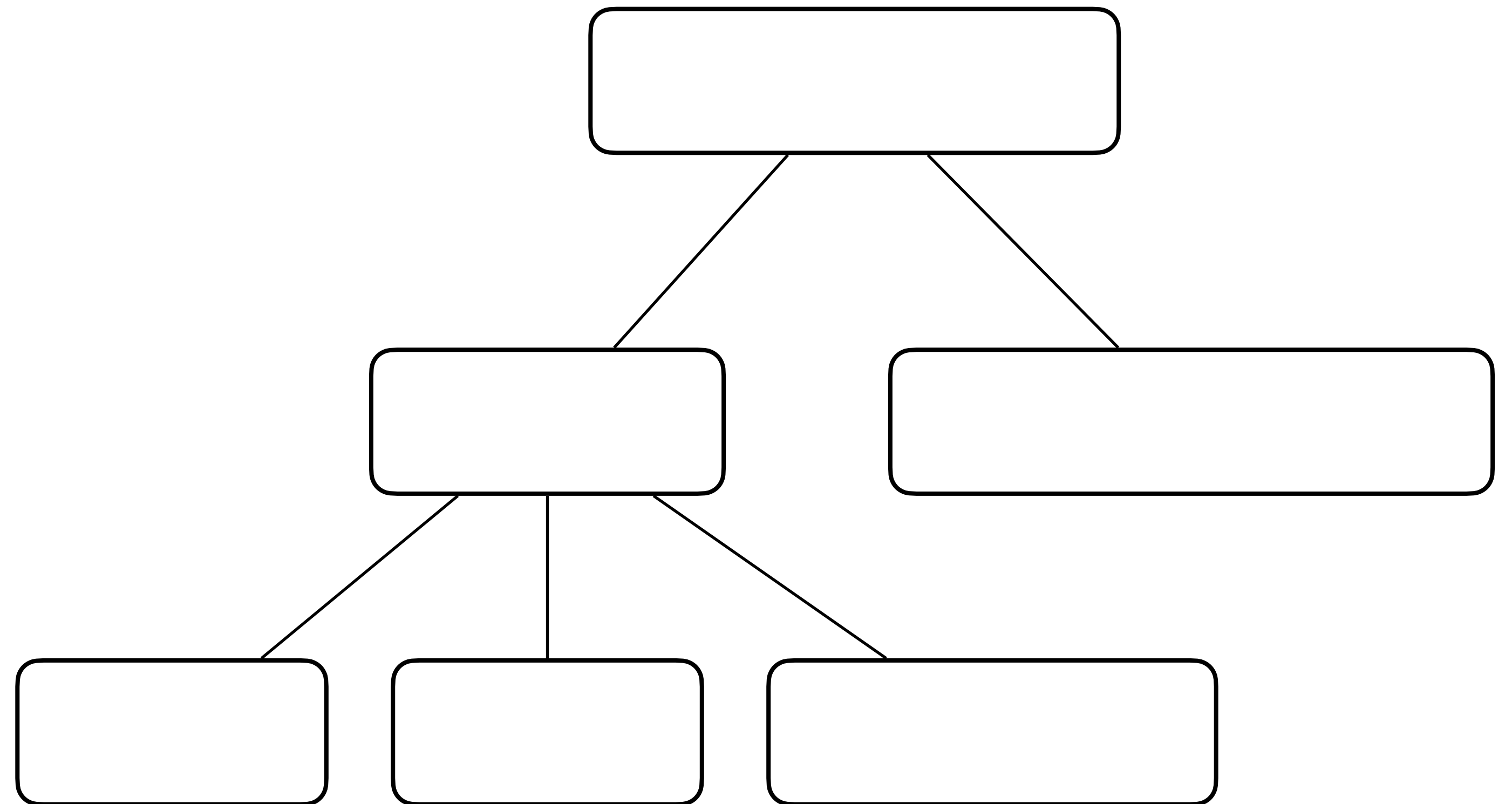
$K_n$  treewidth  $n - 1$

# Tree Decompositions

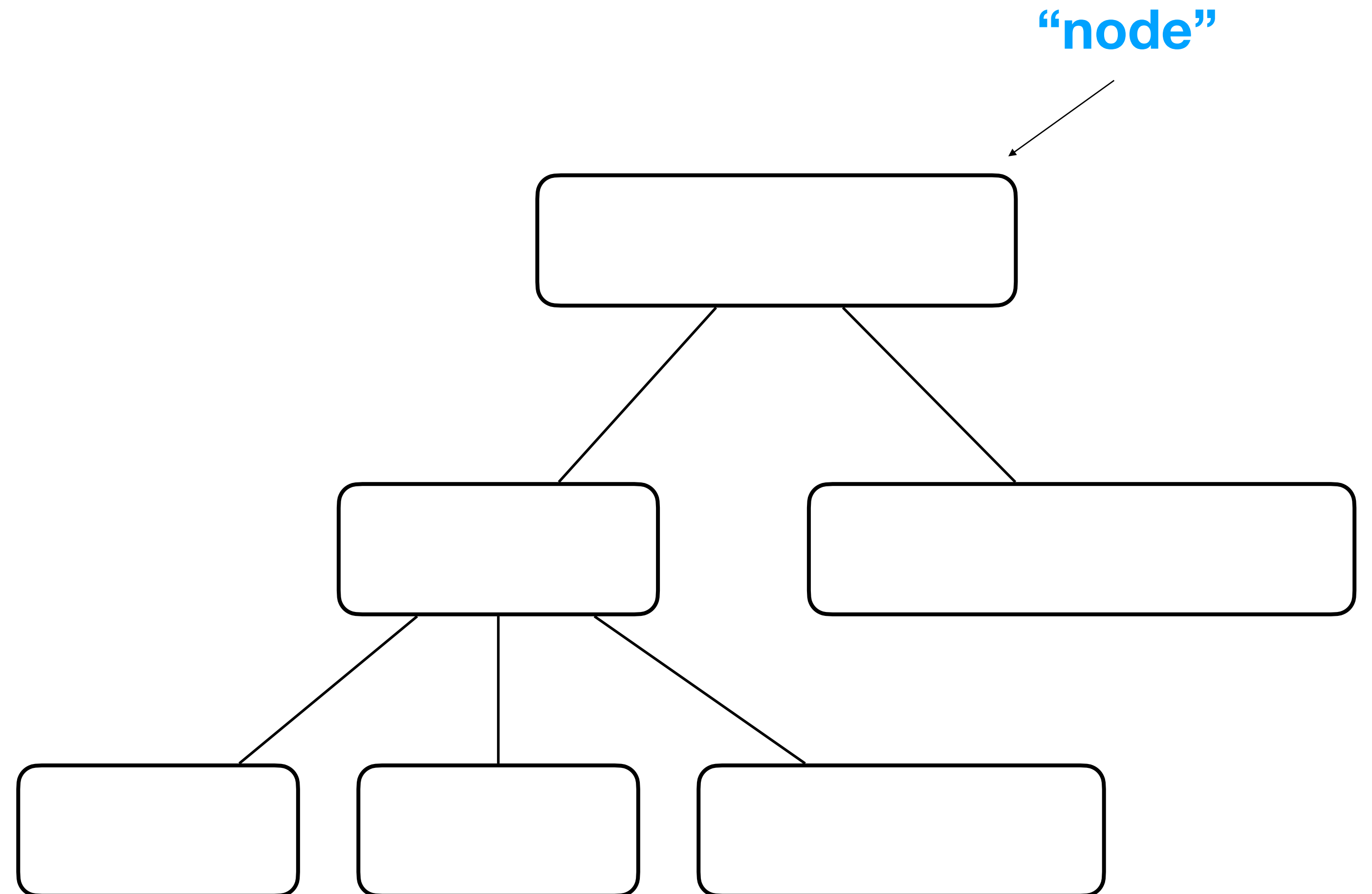


# Tree Decomposition

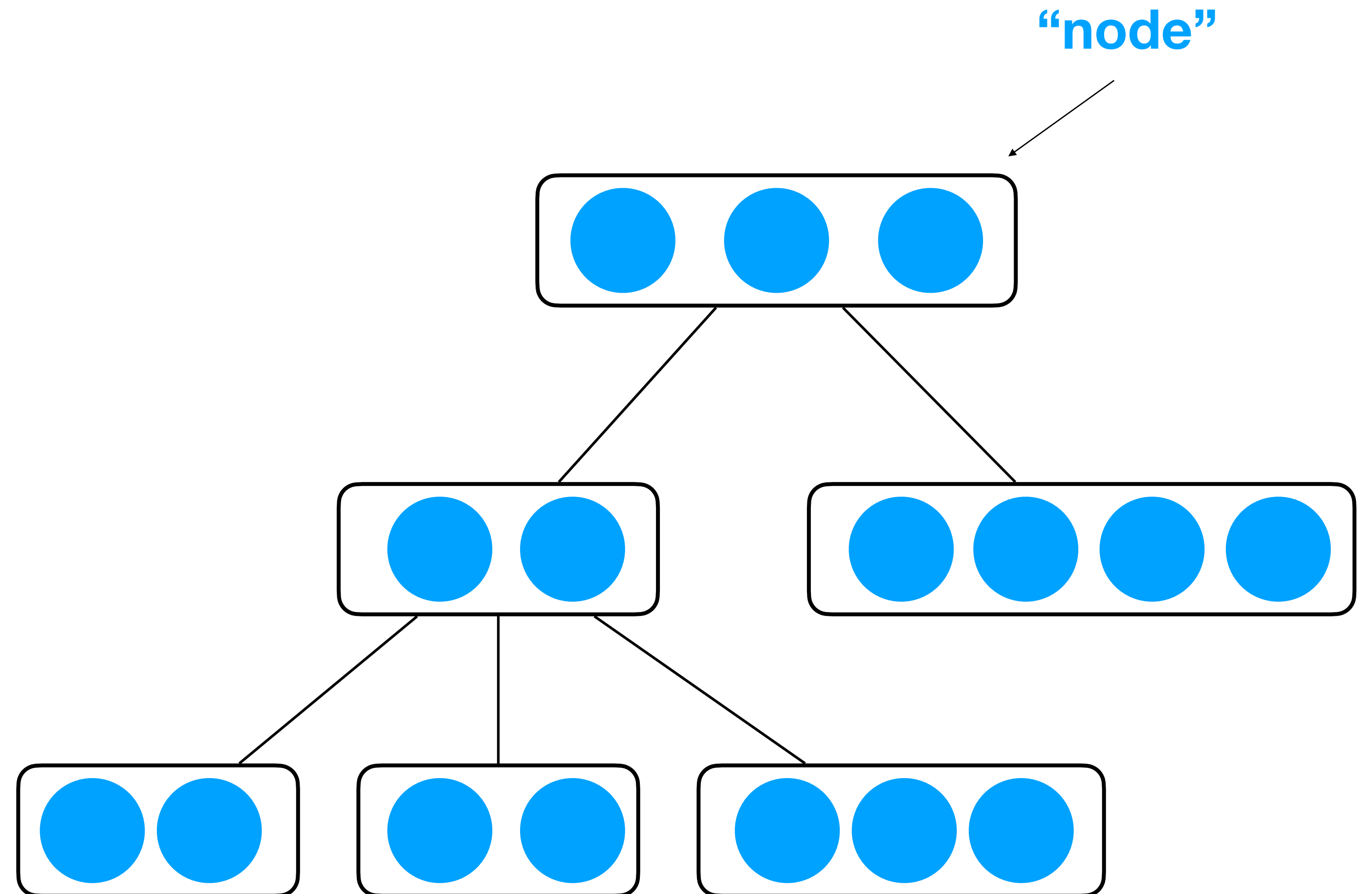
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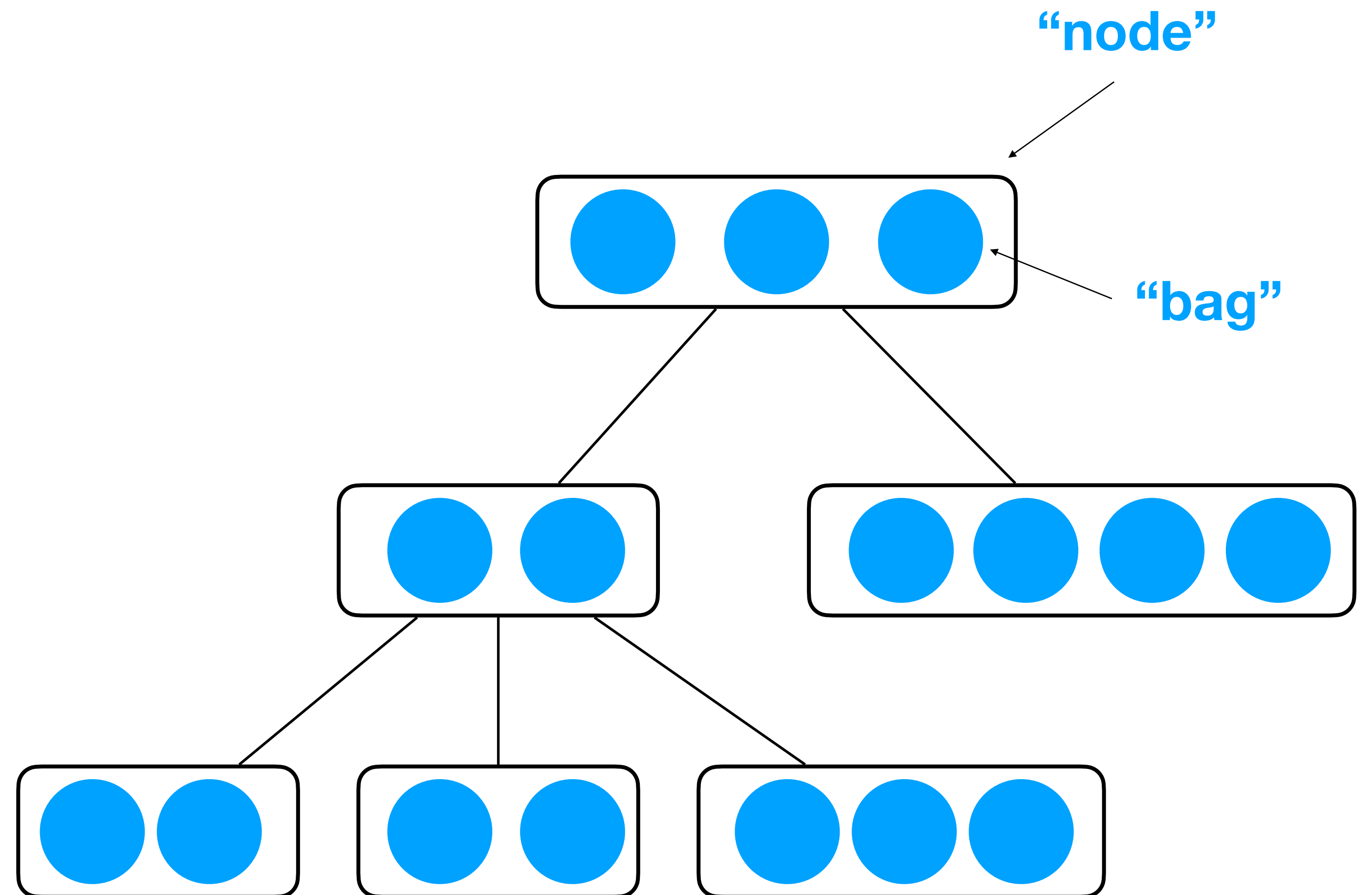
# Tree Decomposition



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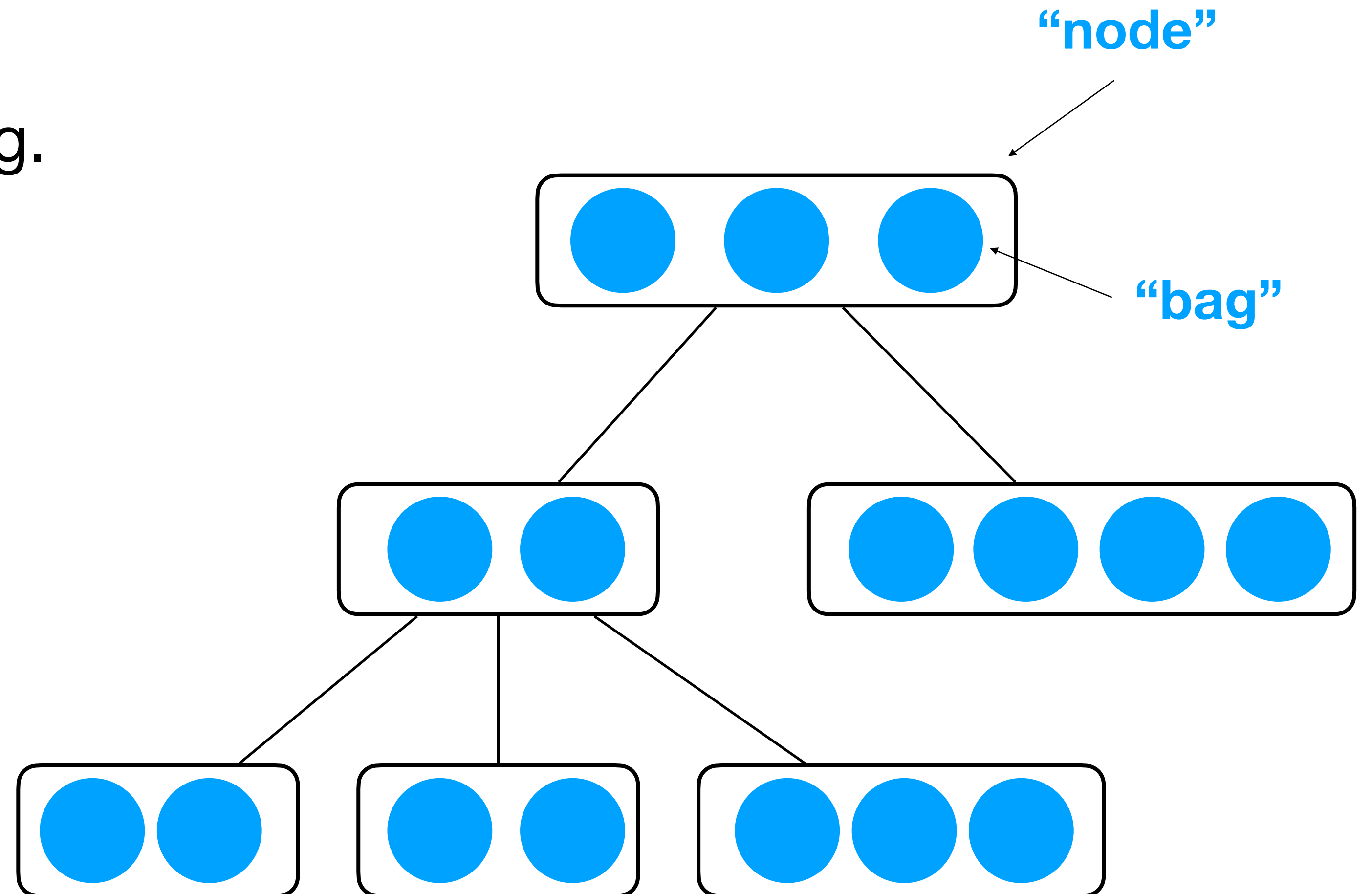


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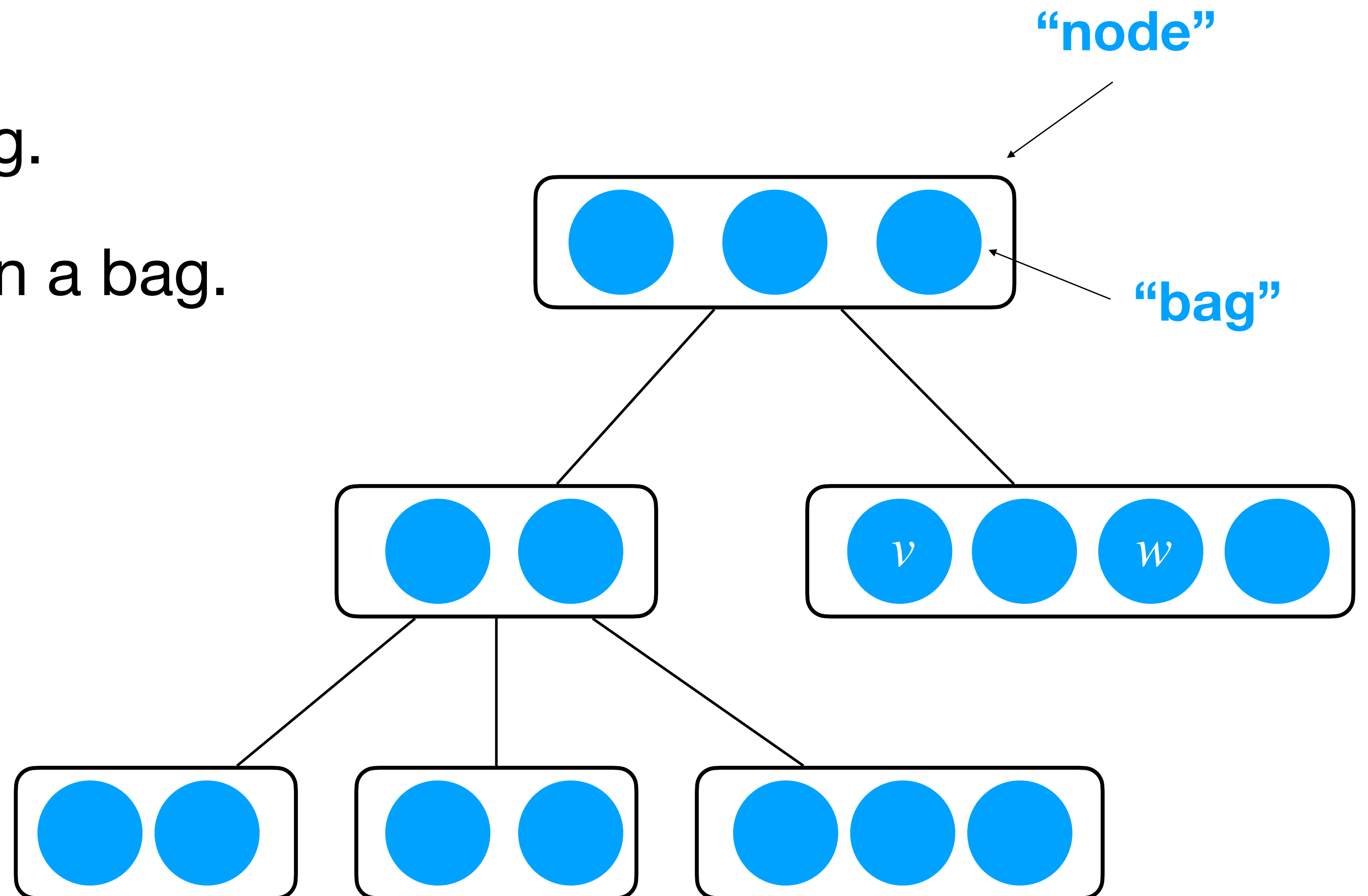
# Tree Decomposition

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



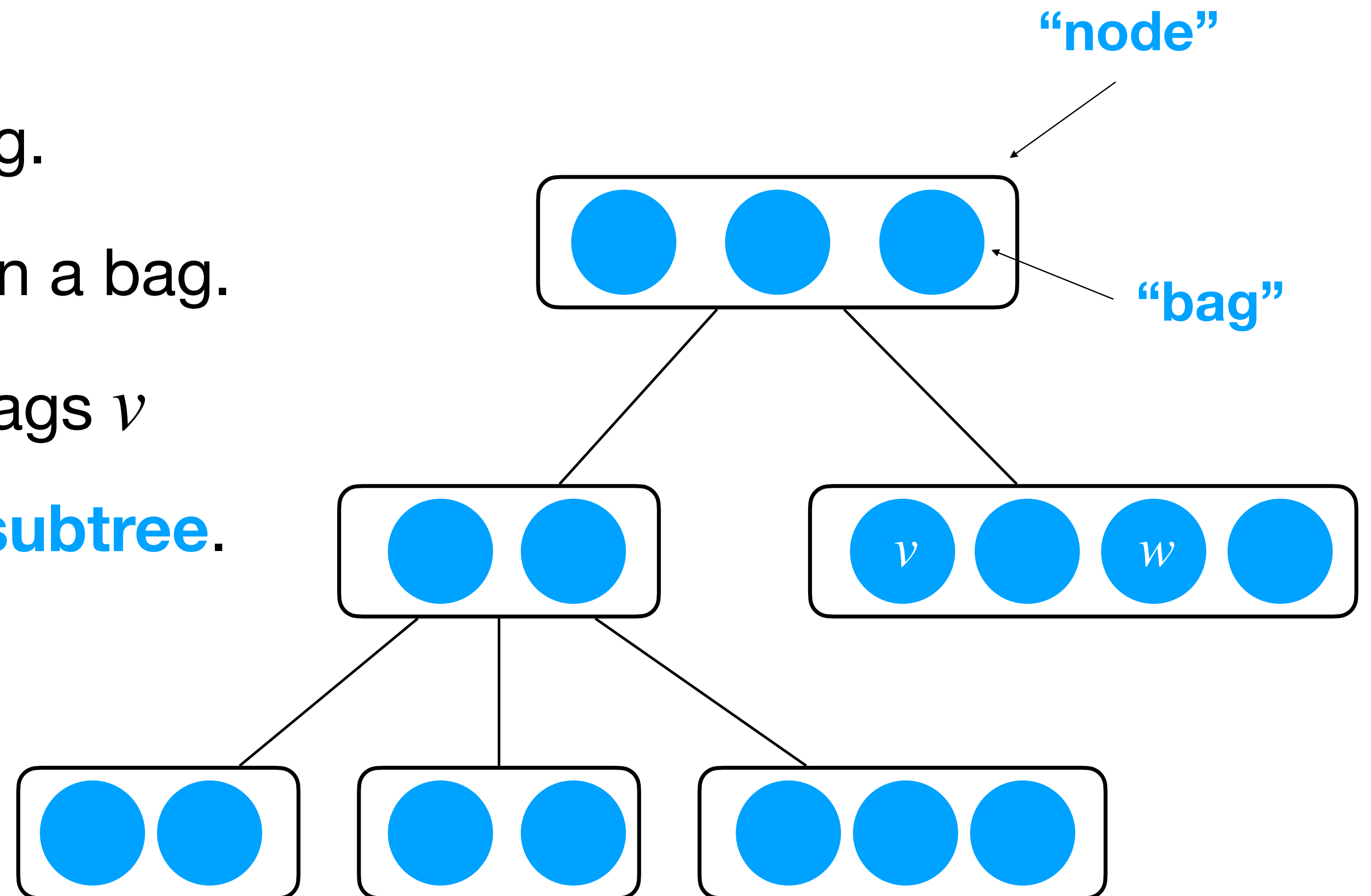
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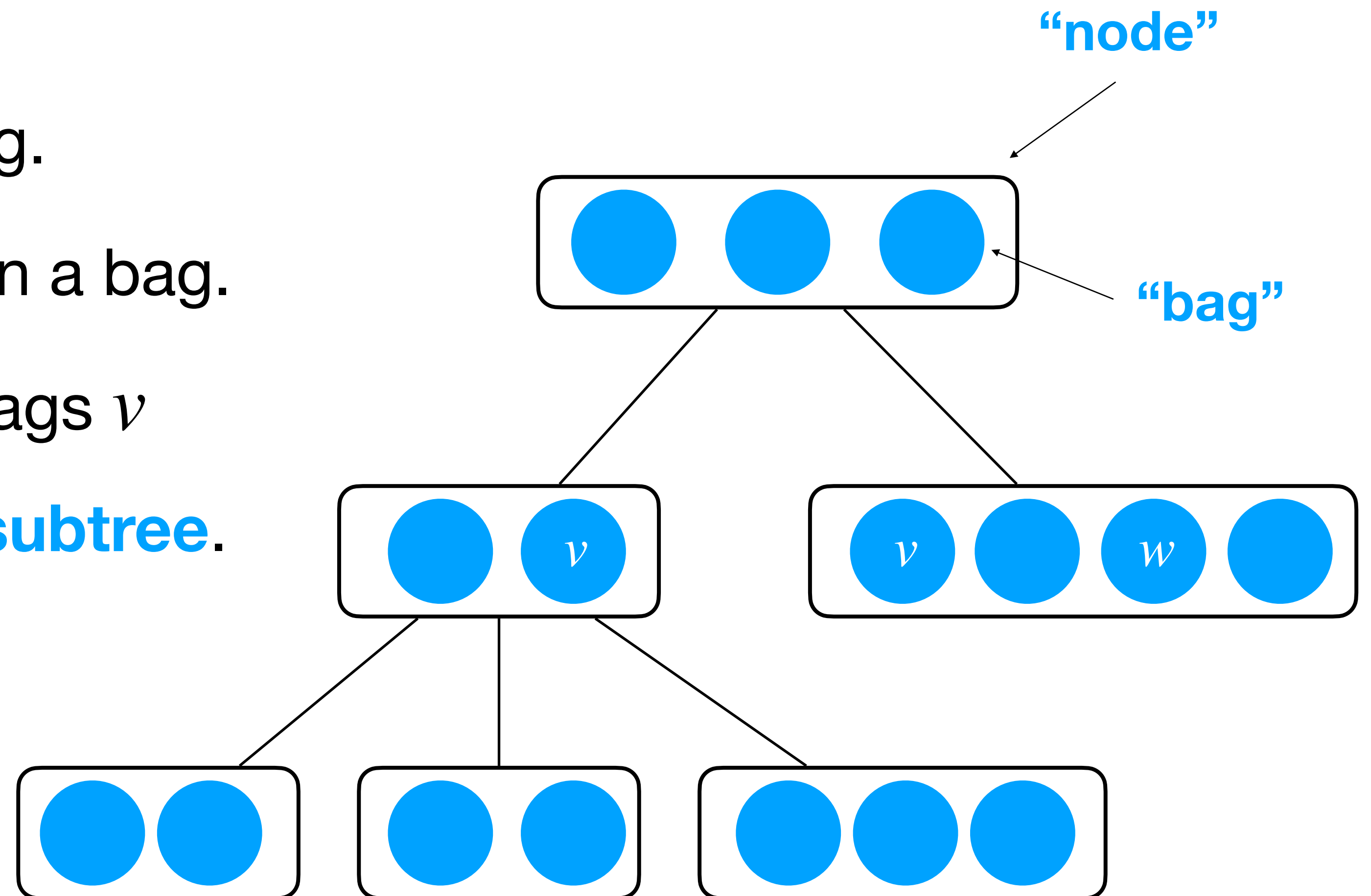
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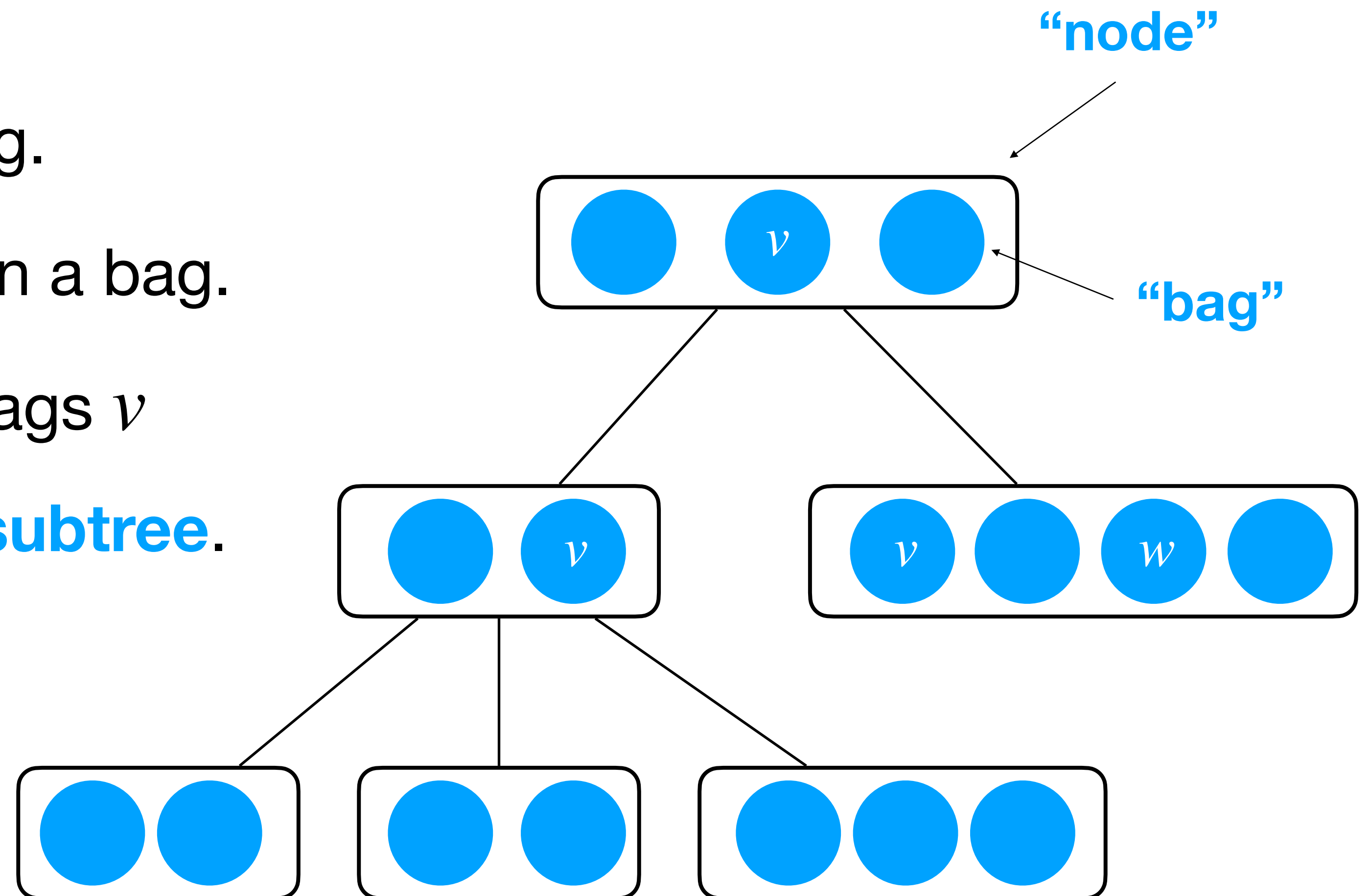
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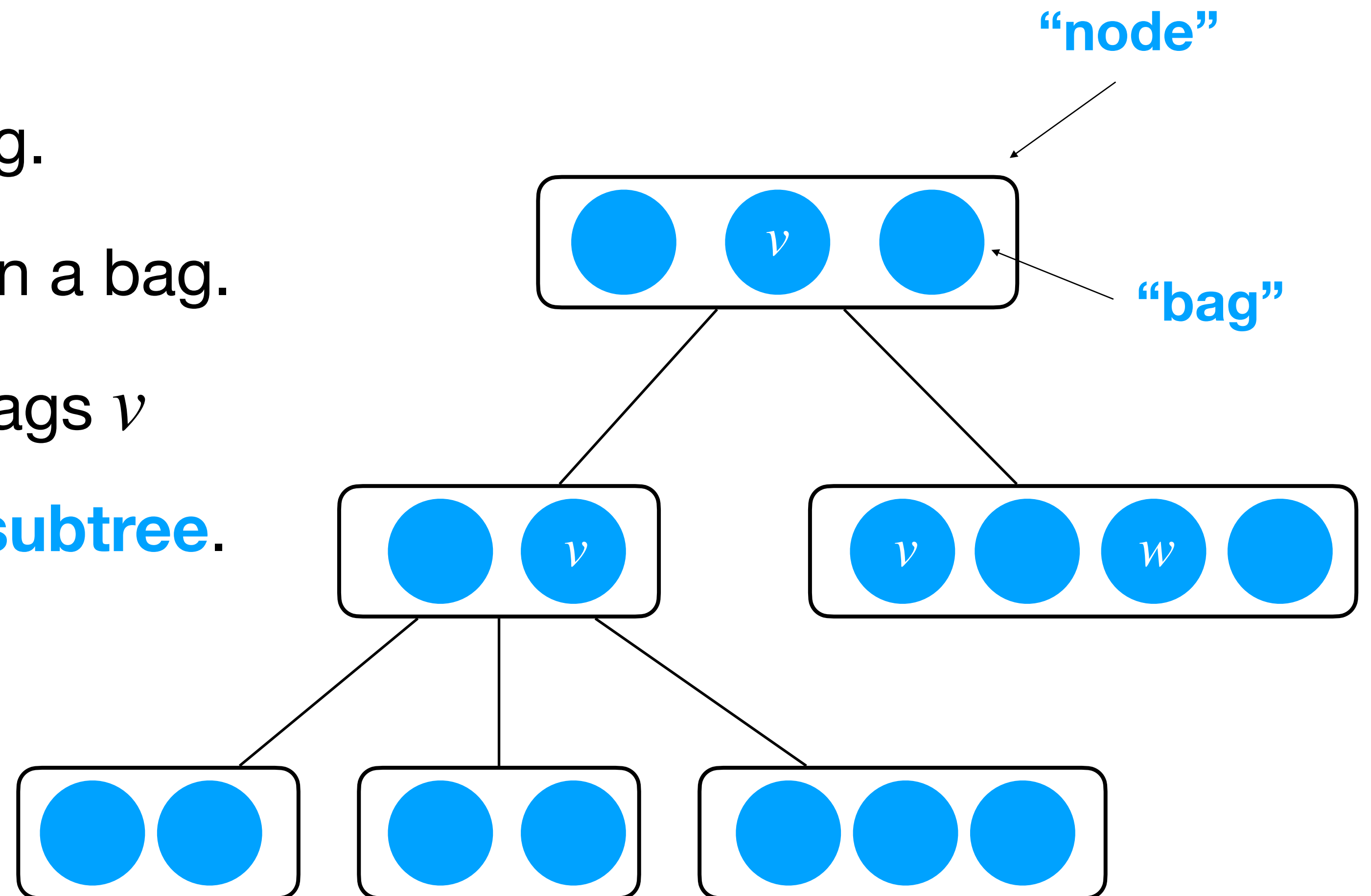
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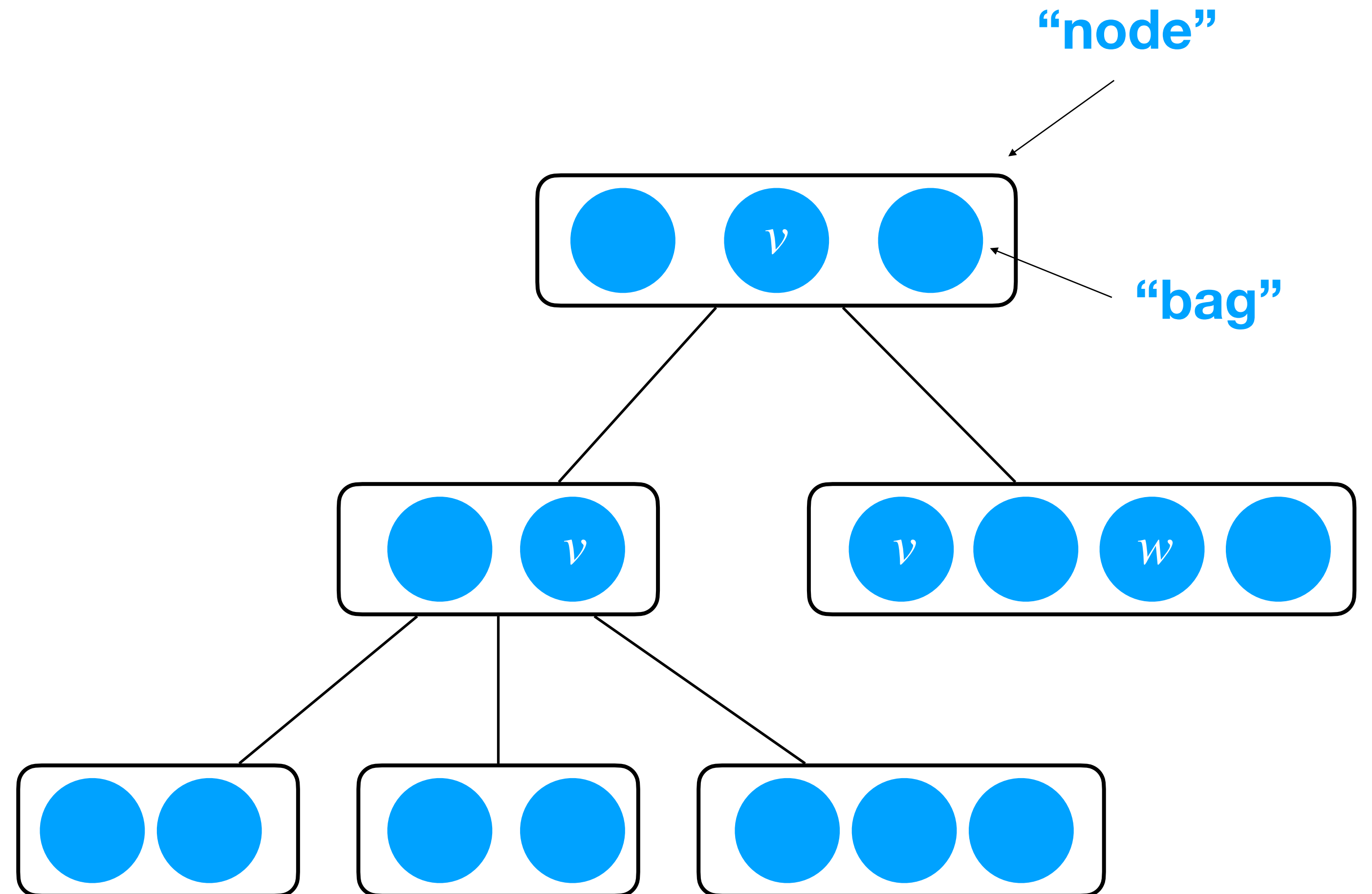
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The **width** of a tree decomposition is the size of its largest bag - 1.

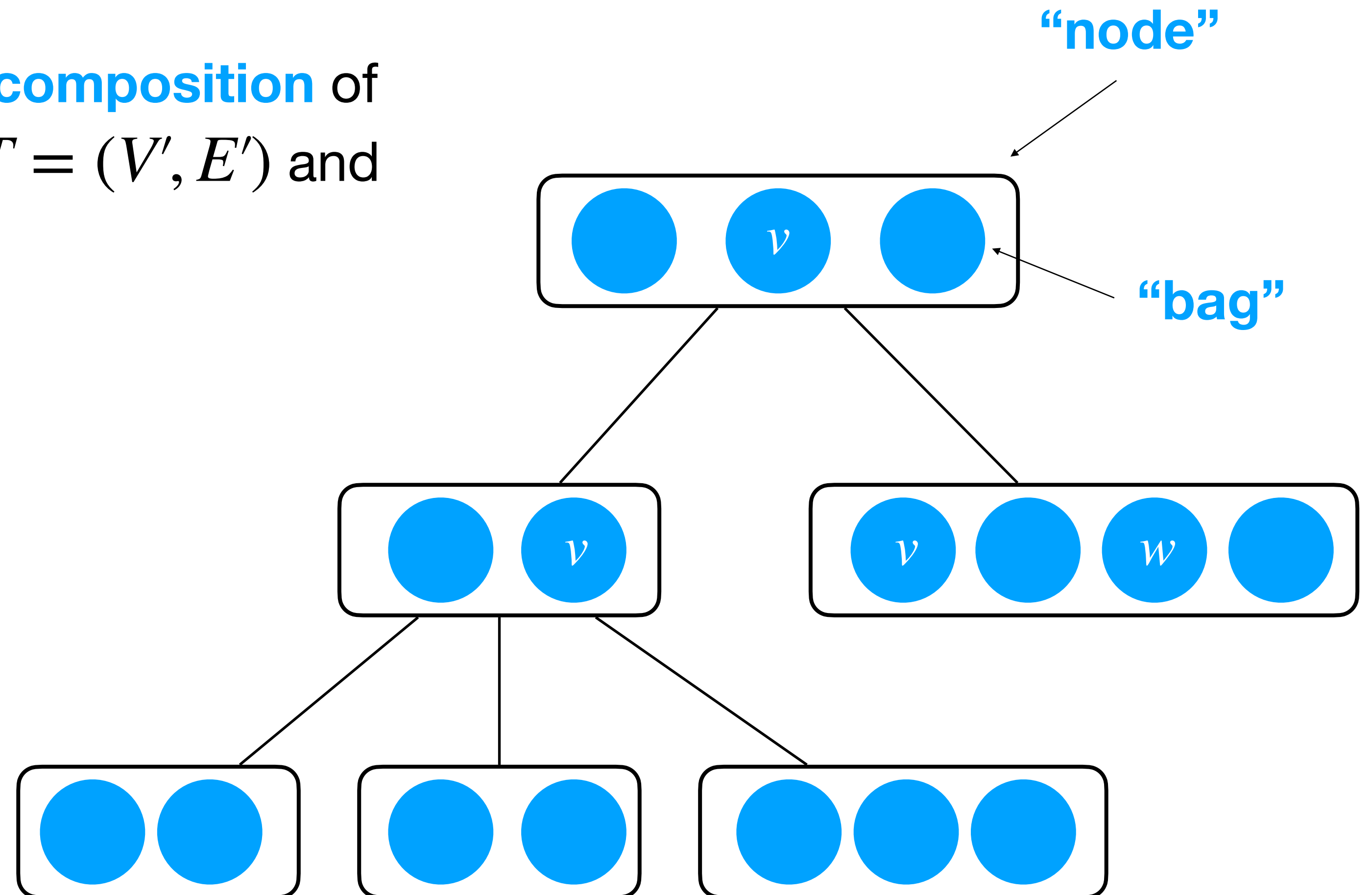
# Tree Decomposition



# Tree Decomposition

## Definition

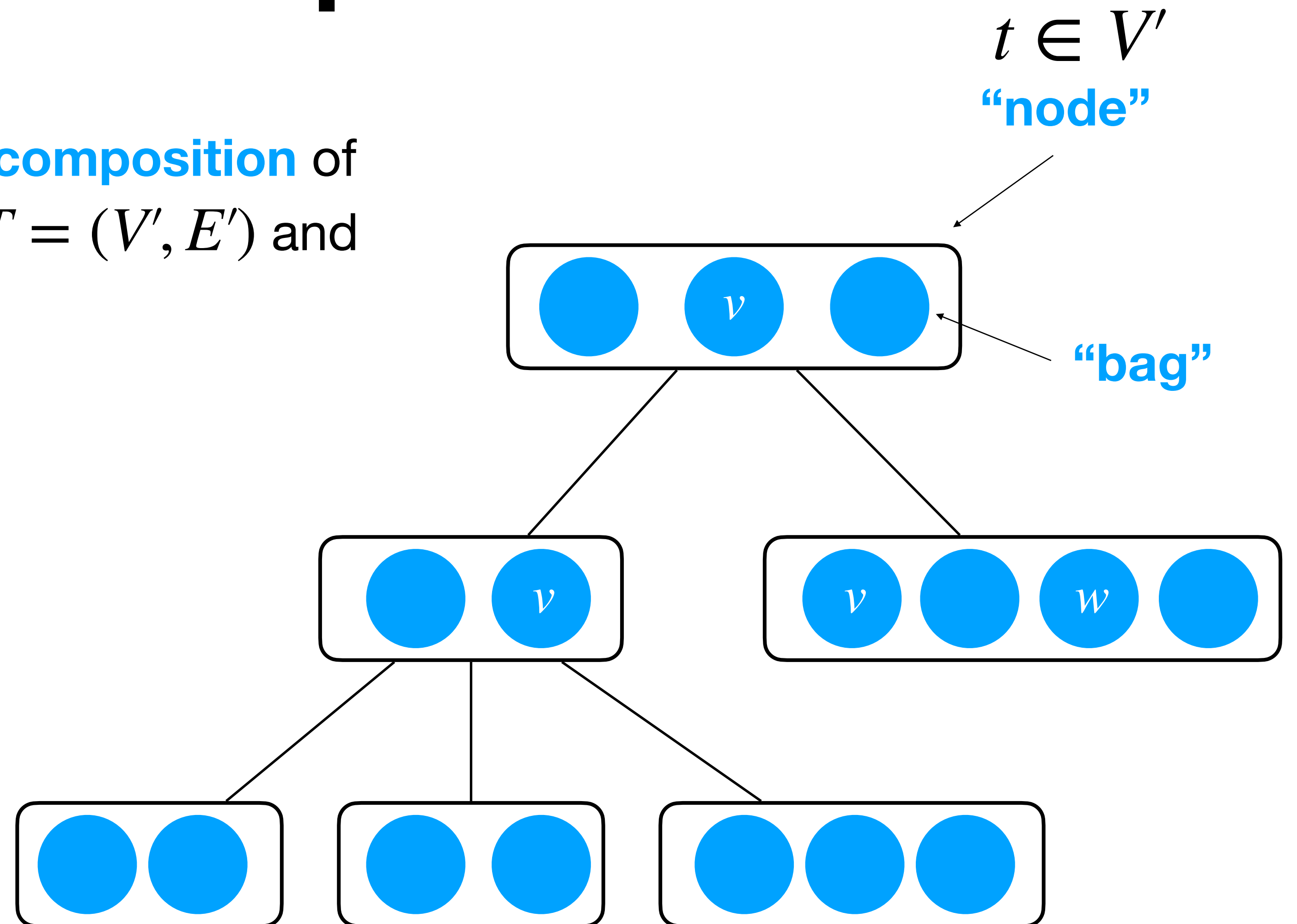
Let  $G = (V, E)$  be a graph. A **tree decomposition** of  $G$  is a pair  $(T, \chi)$  consisting of a tree  $T = (V', E')$  and a mapping  $\chi : V' \rightarrow 2^V$  such that



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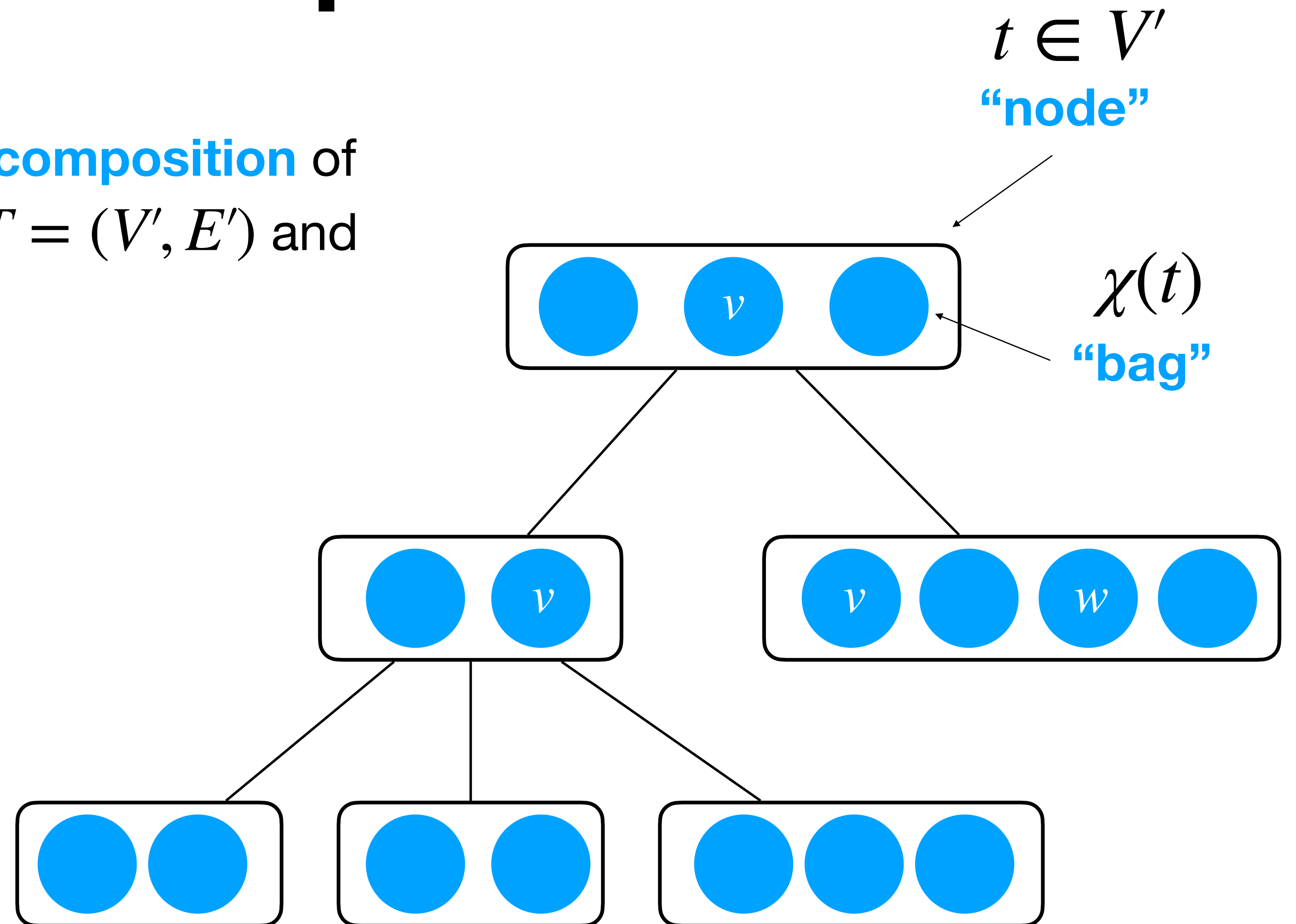
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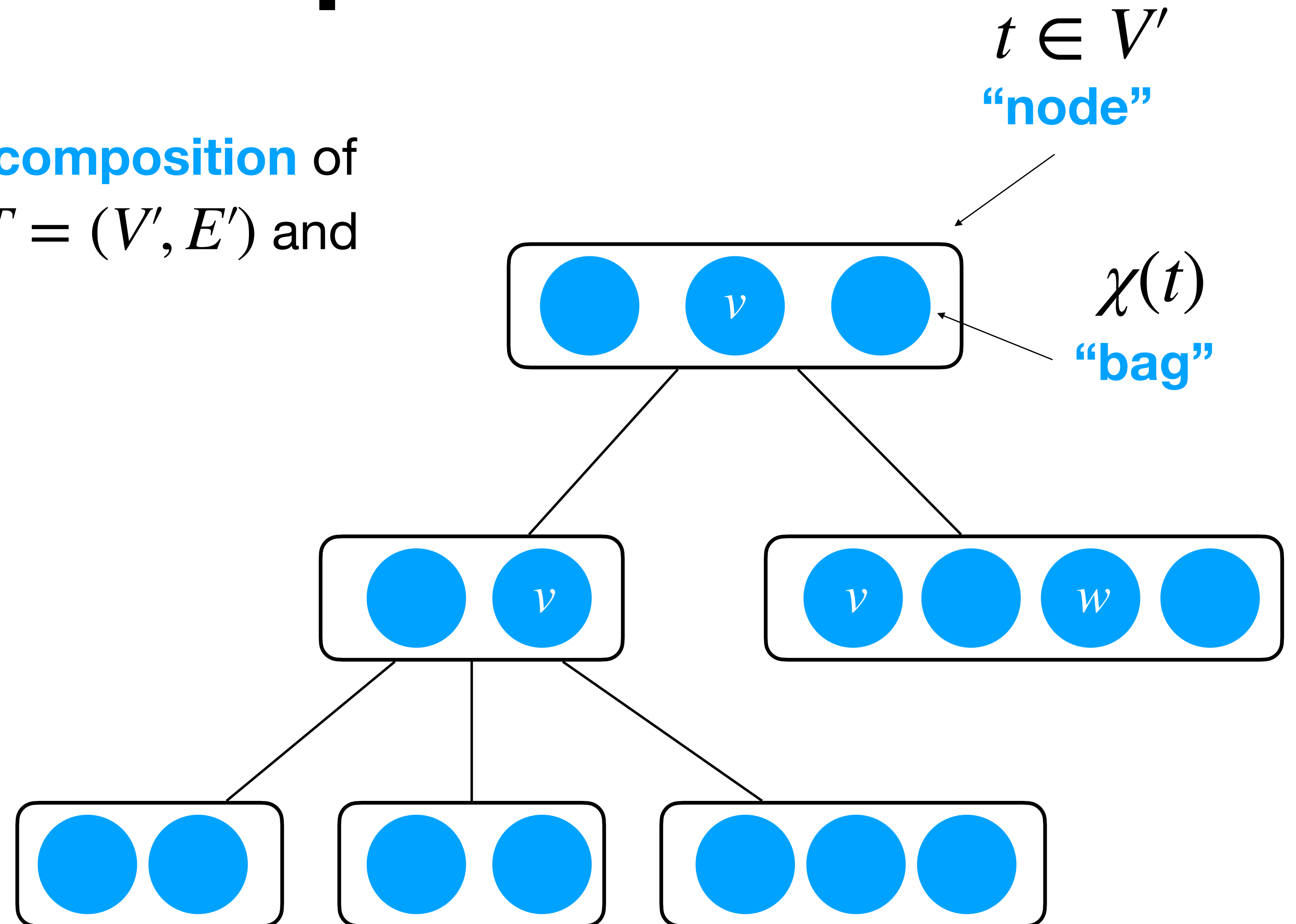


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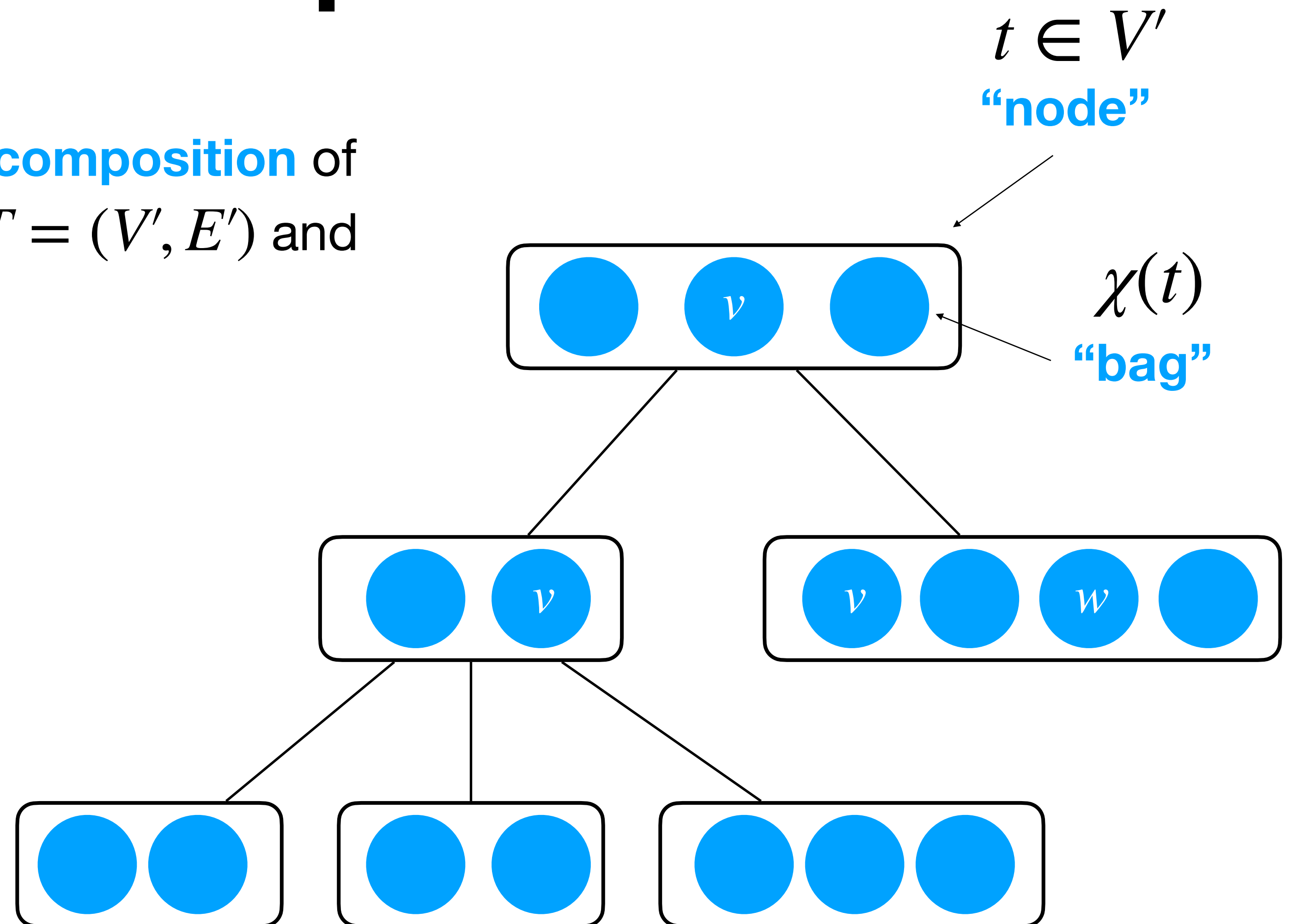


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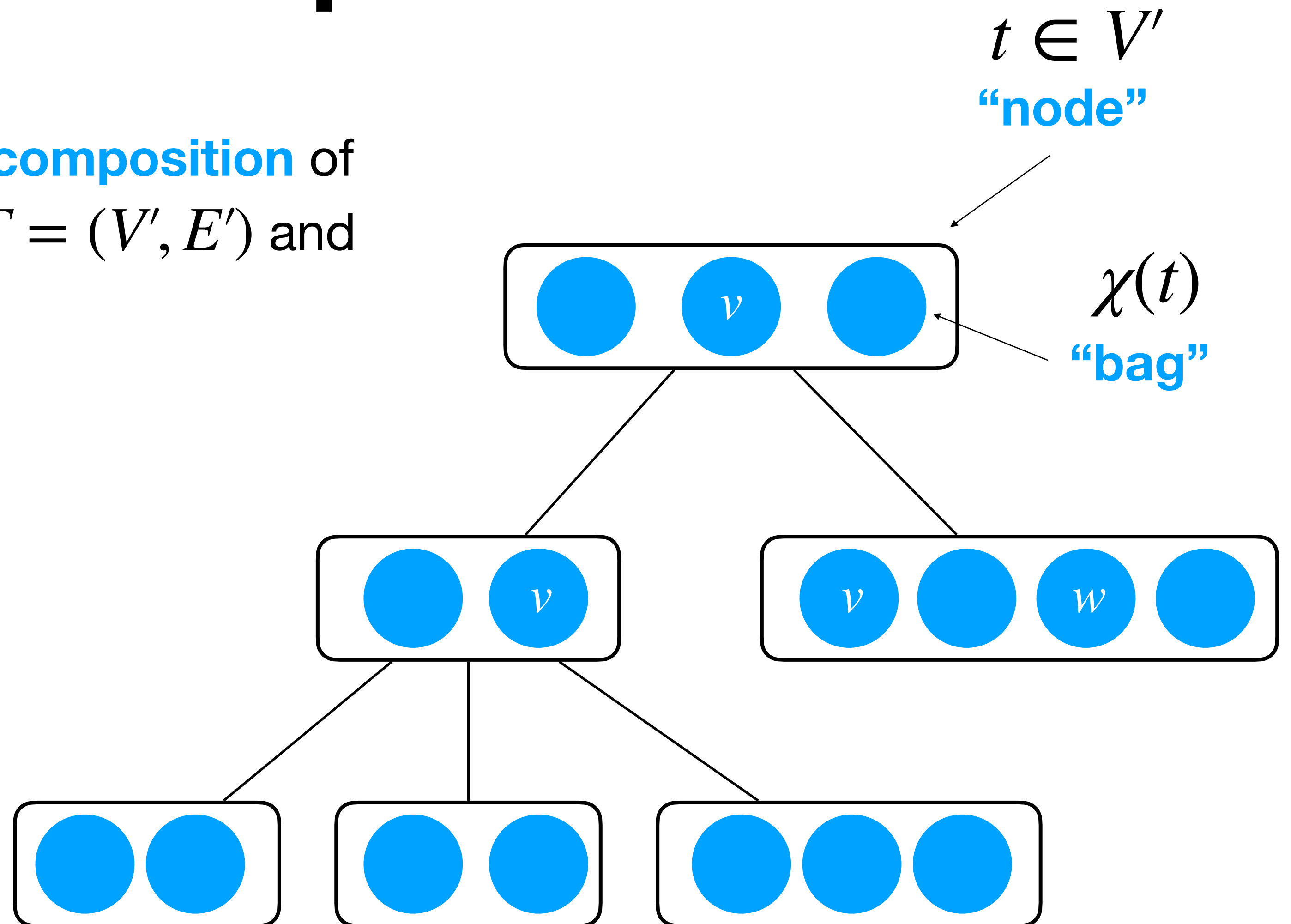


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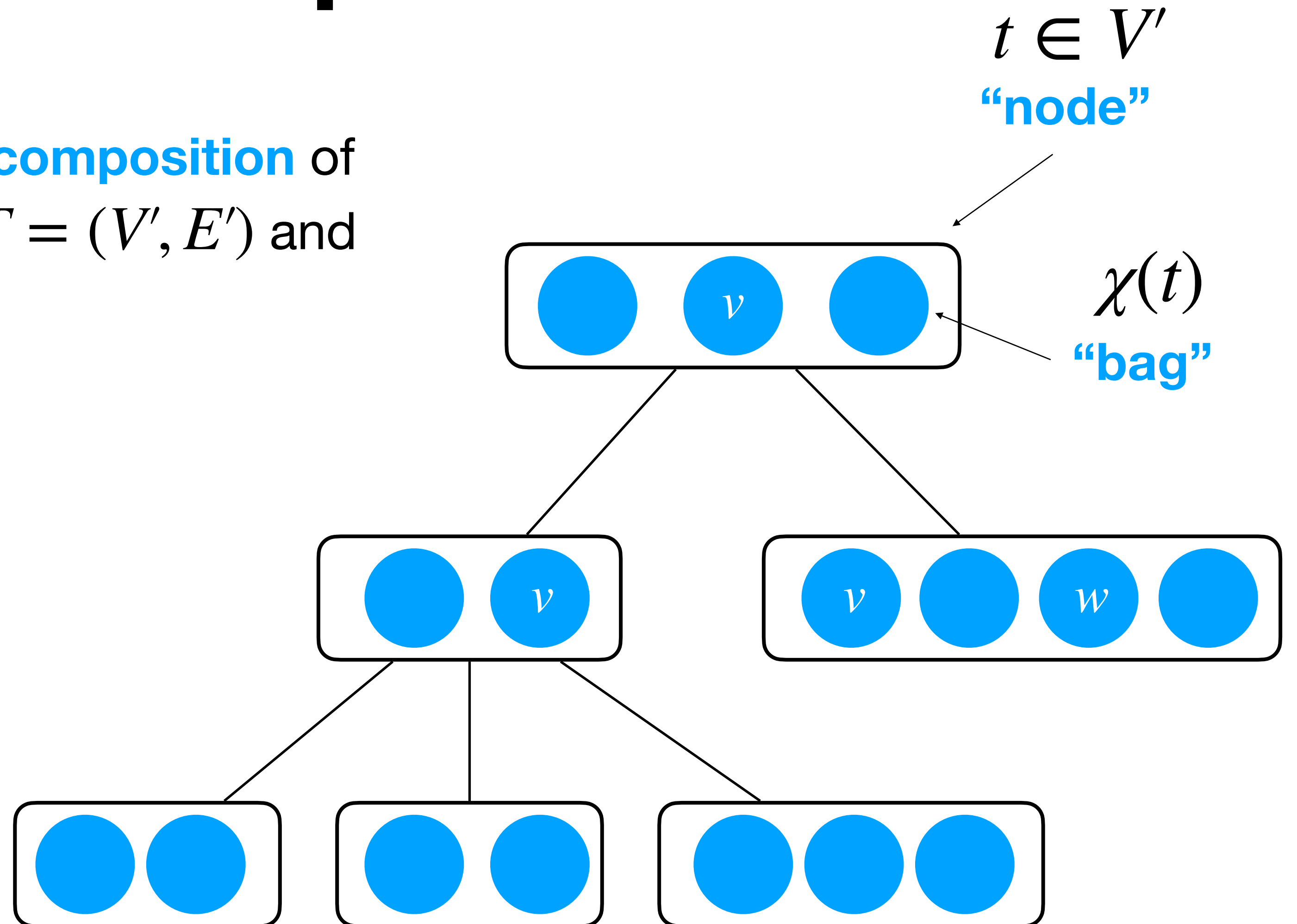
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The **width** of  $(T, \chi)$  is  $\max_{t \in V'} |\chi(t)| - 1$ .



# Tree Decompositions and Treewidth

# Tree Decompositions and Treewidth

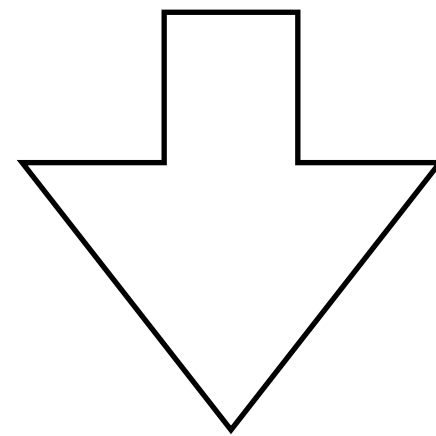
## Fact

A graph has a tree decomposition of width  $k$  if, and only if, it has an elimination ordering of width  $k$ .

# Tree Decompositions and Treewidth

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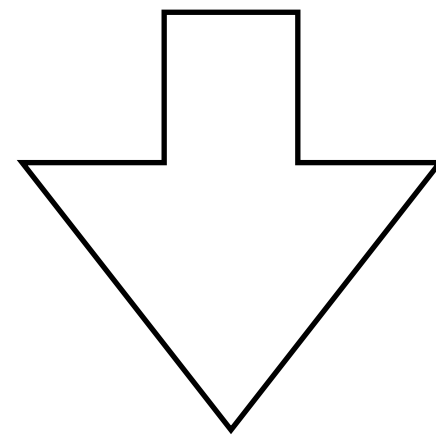
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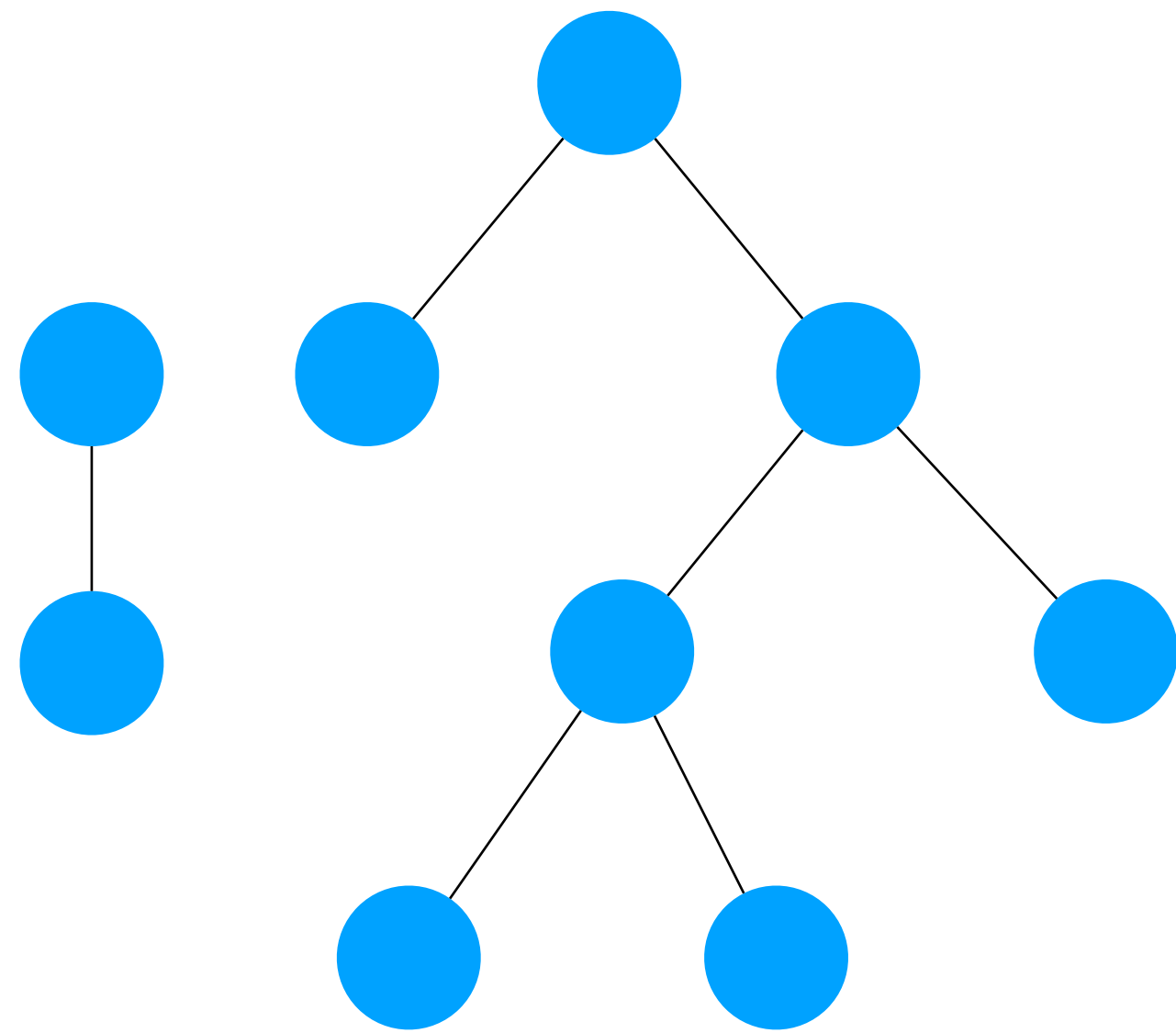
The treewidth of a graph  $G$  is the minimum width of a tree decomposition of  $G$ .

# Examples



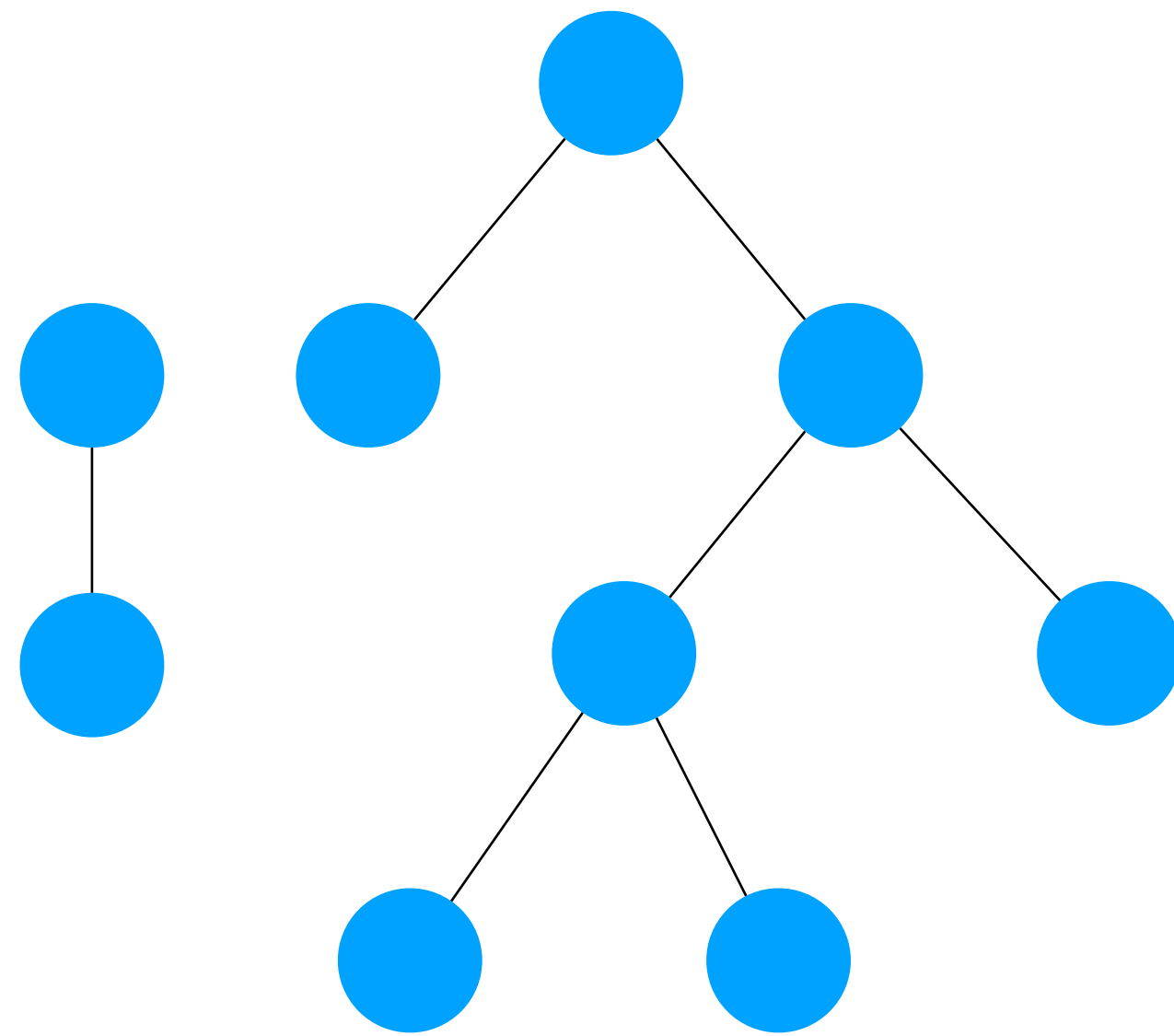
# Examples

## Forests



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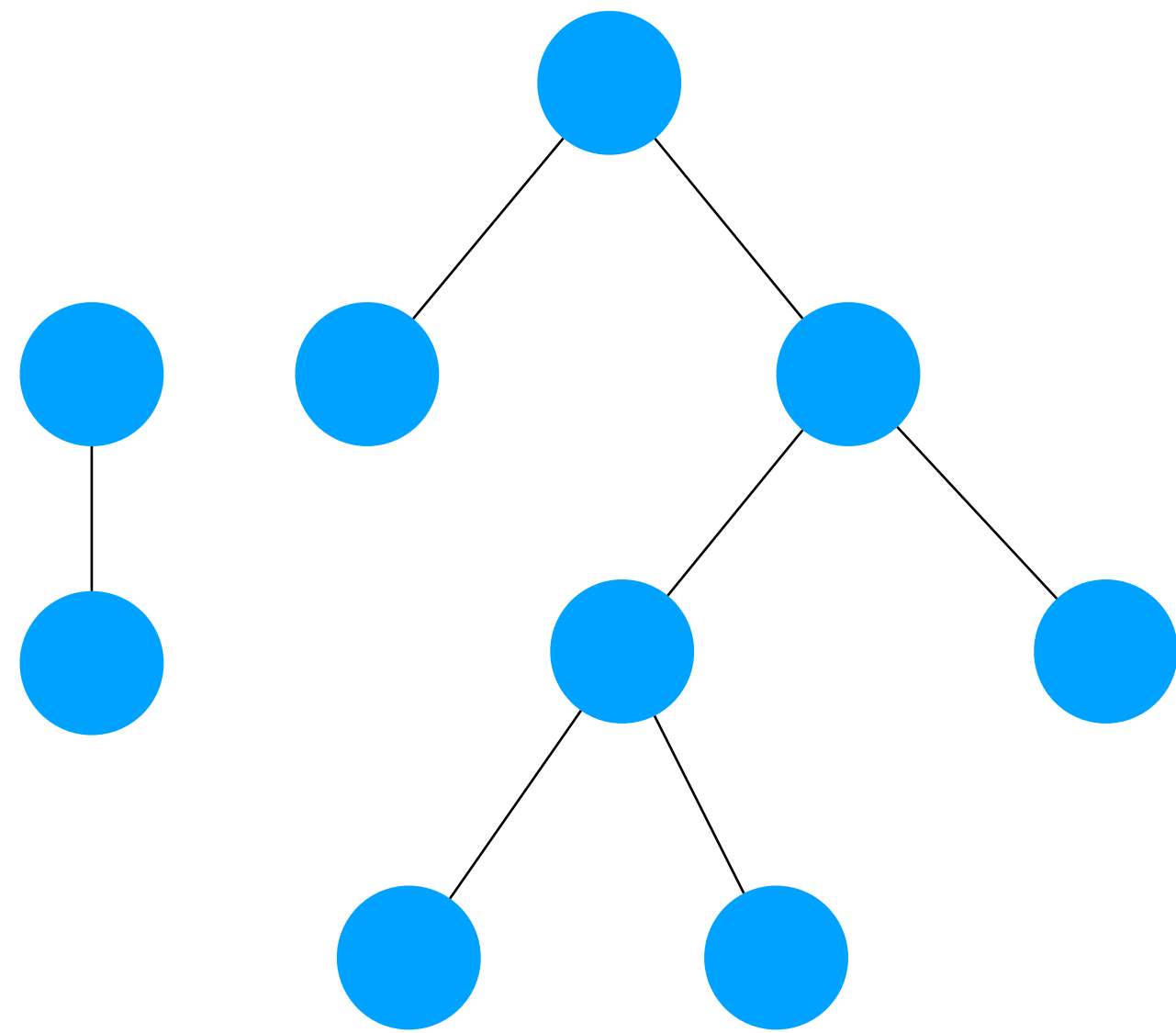
## Forests



treewidth 1

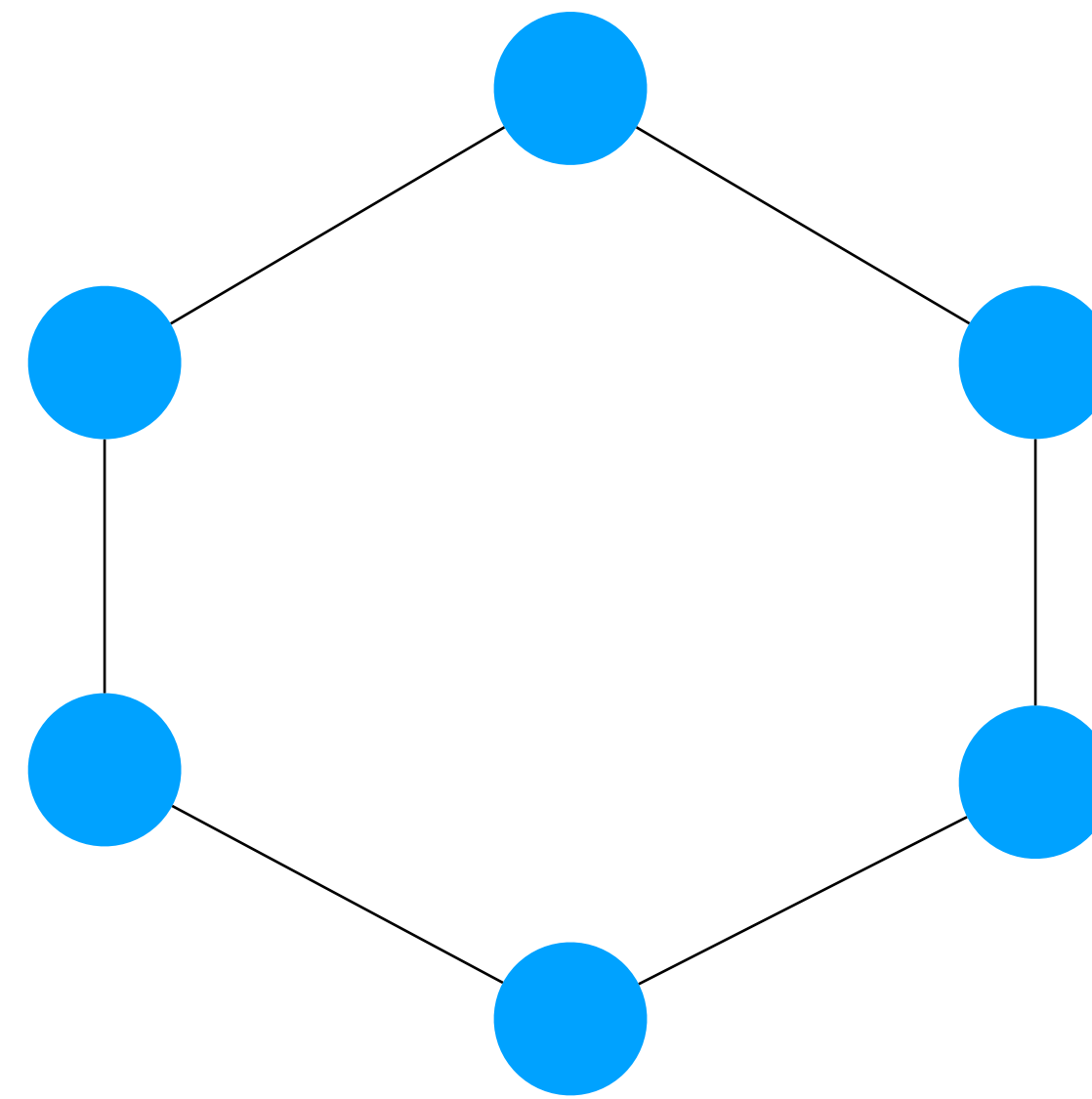
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Forests



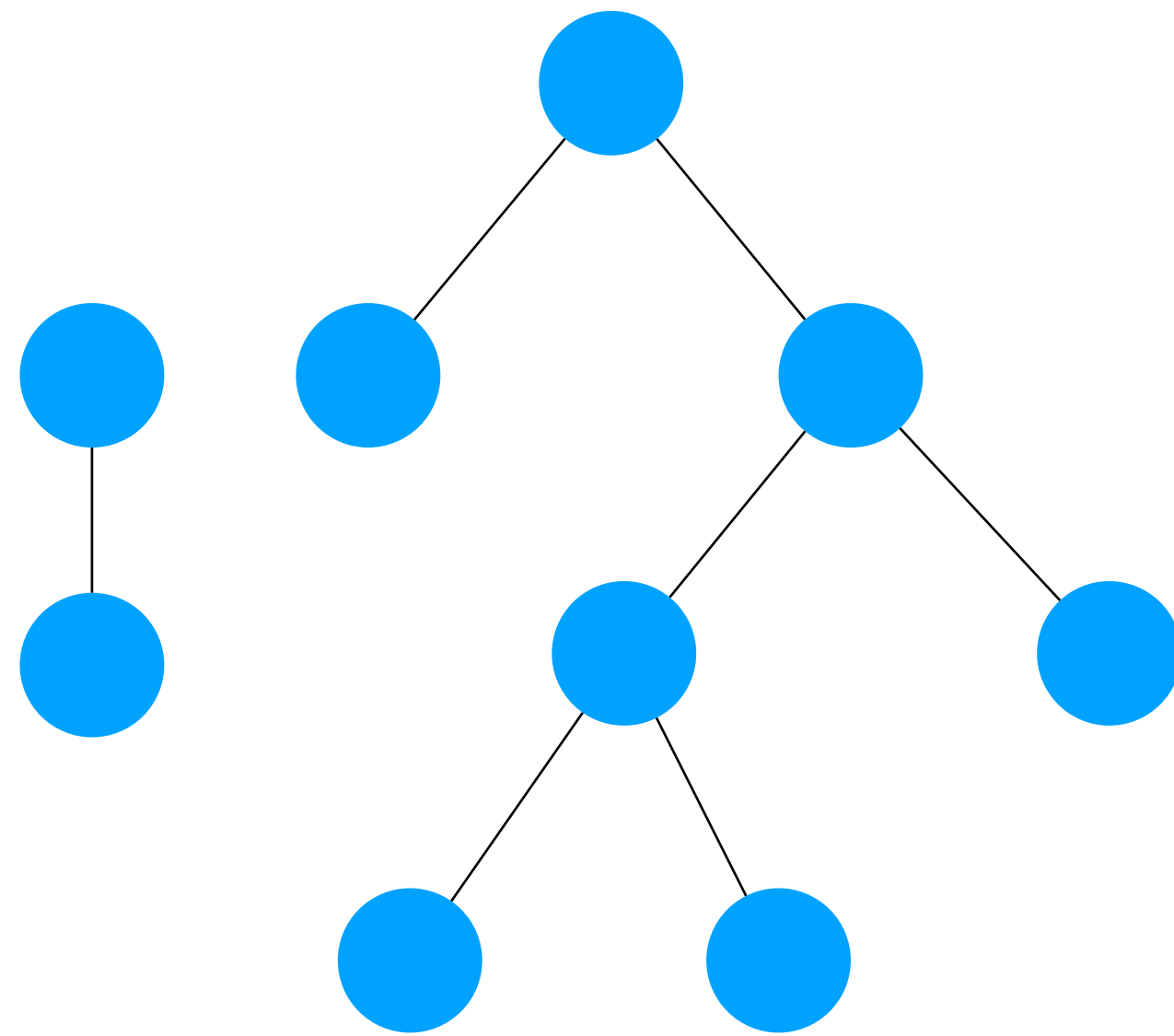
treewidth 1

Cycles



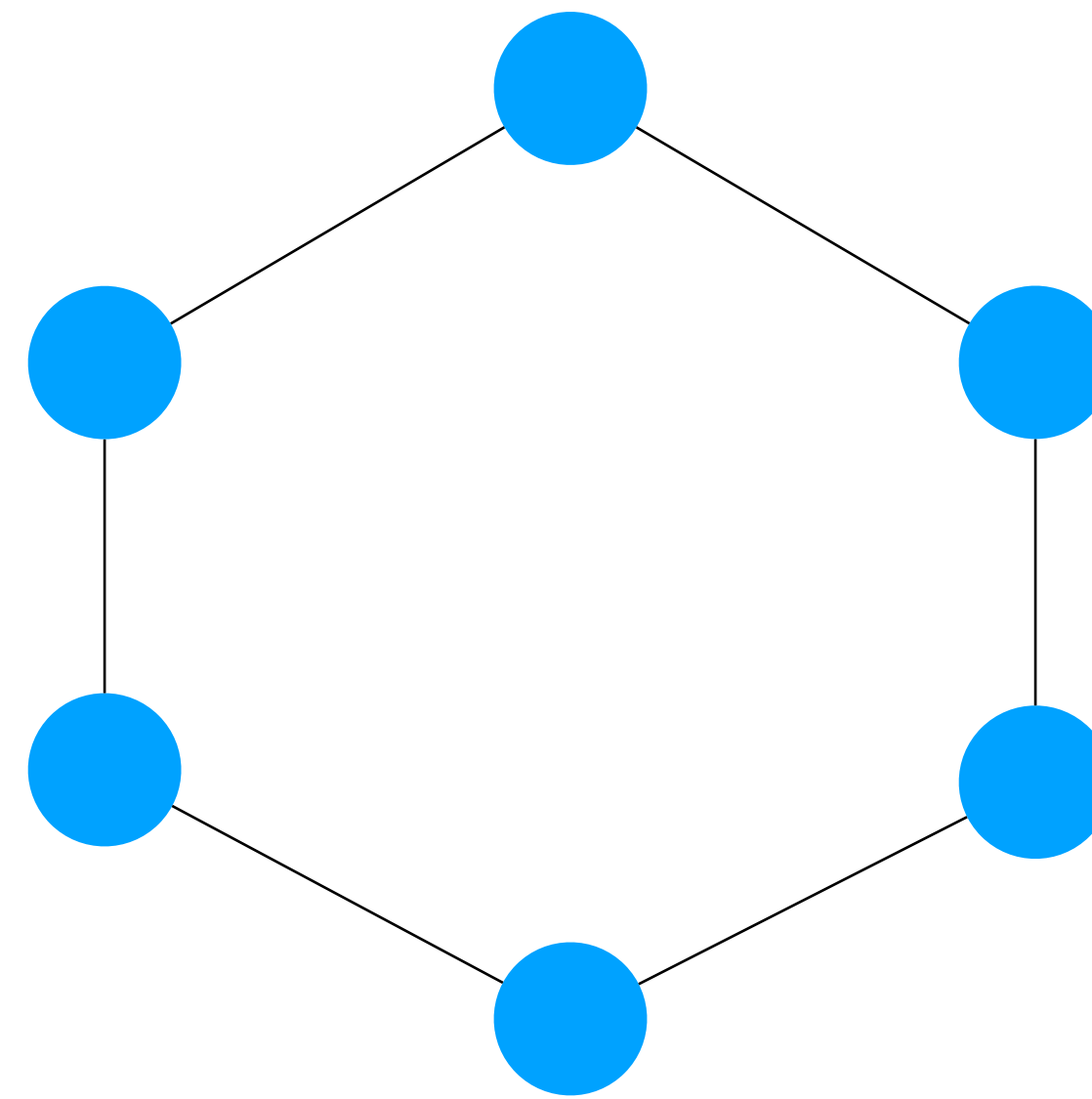
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Forests



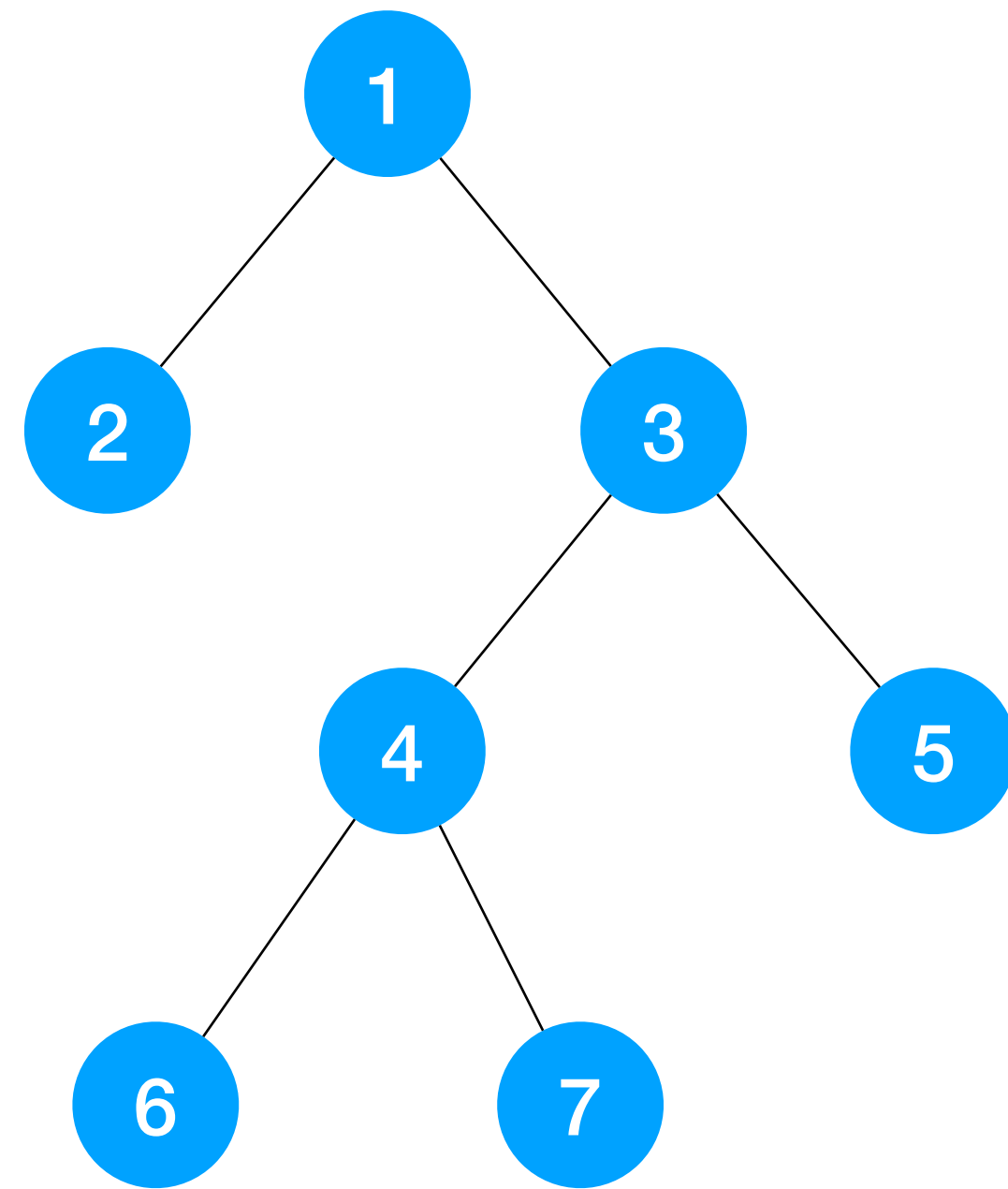
treewidth 1

Cycles

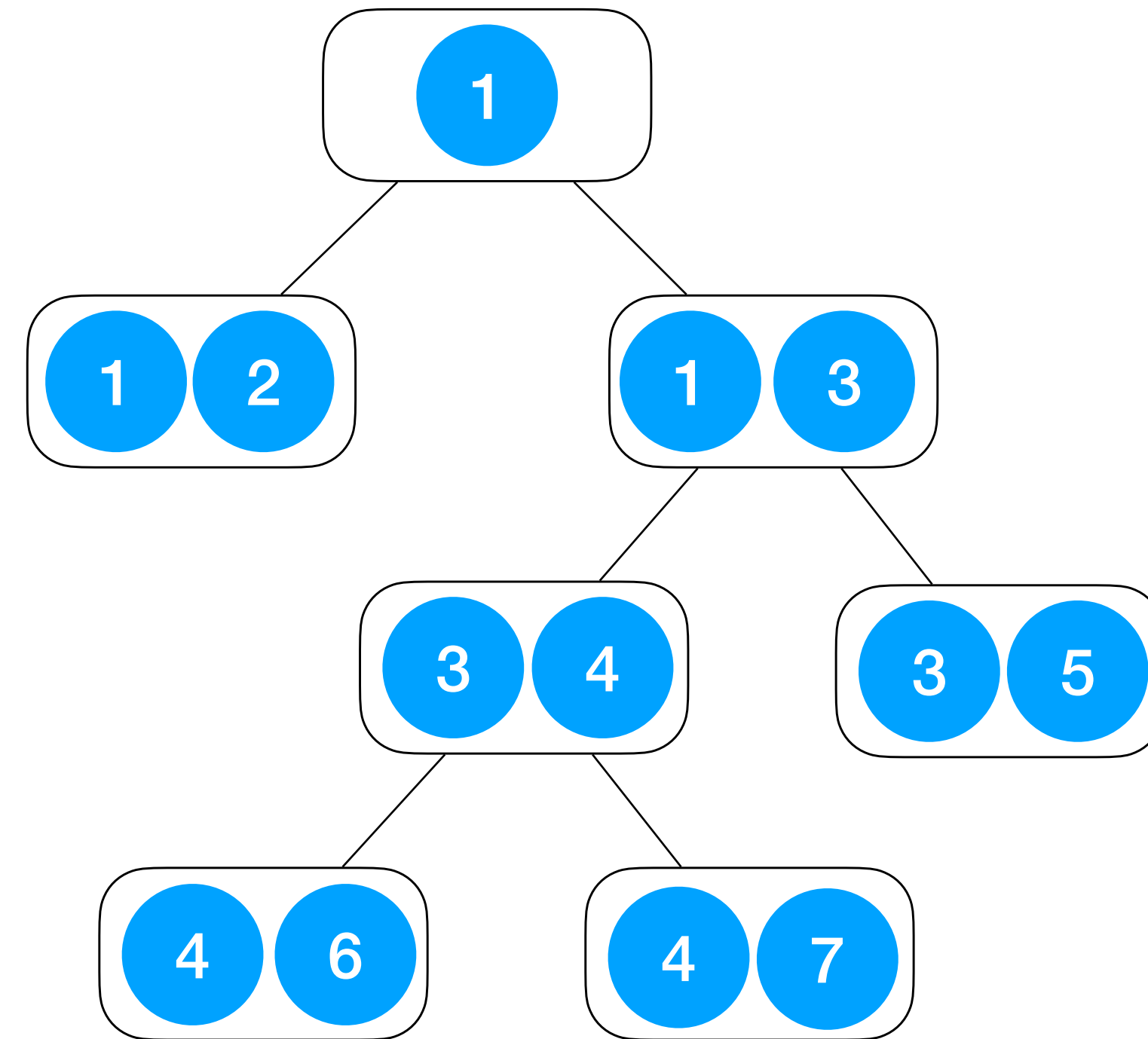
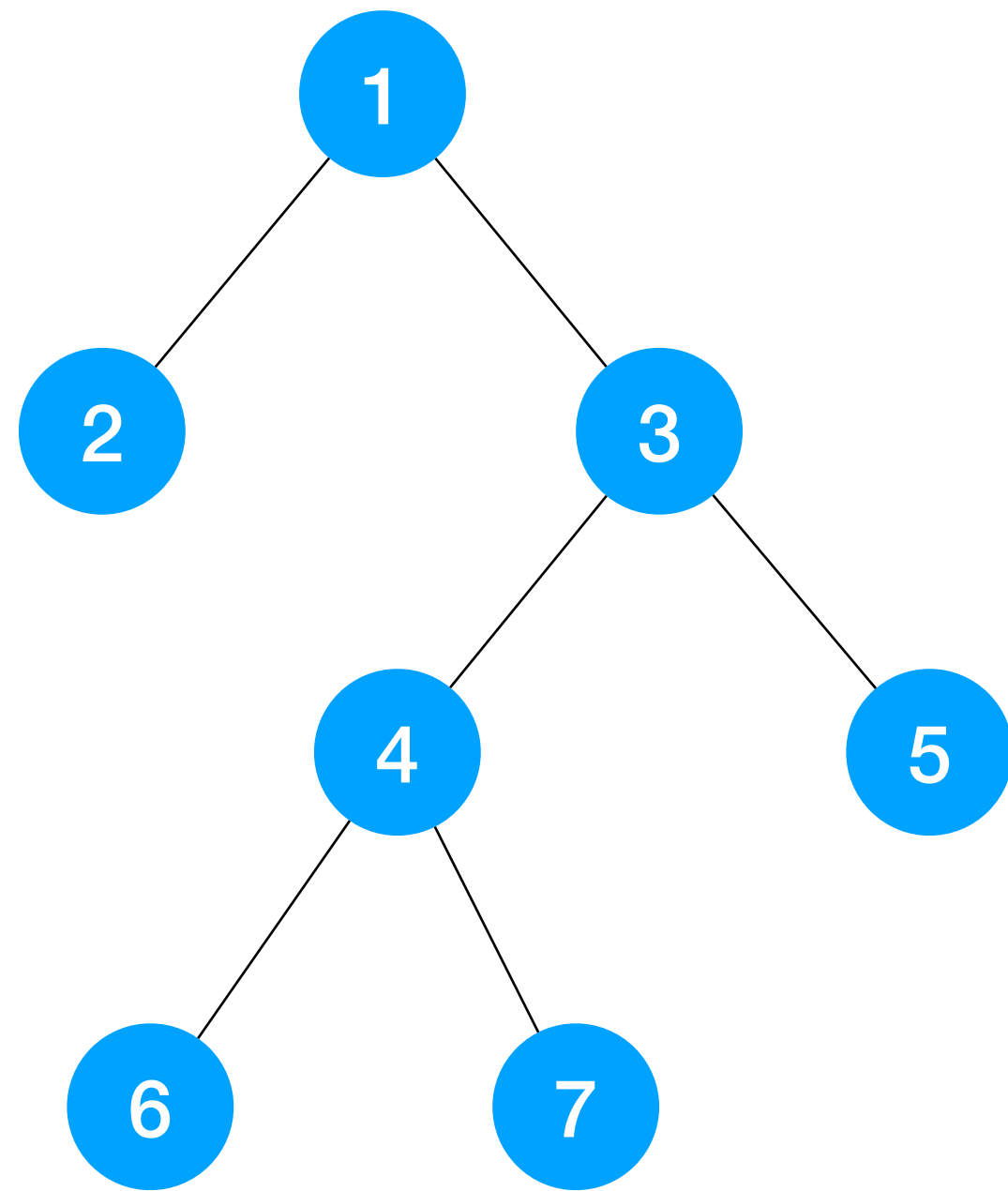


treewidth 2

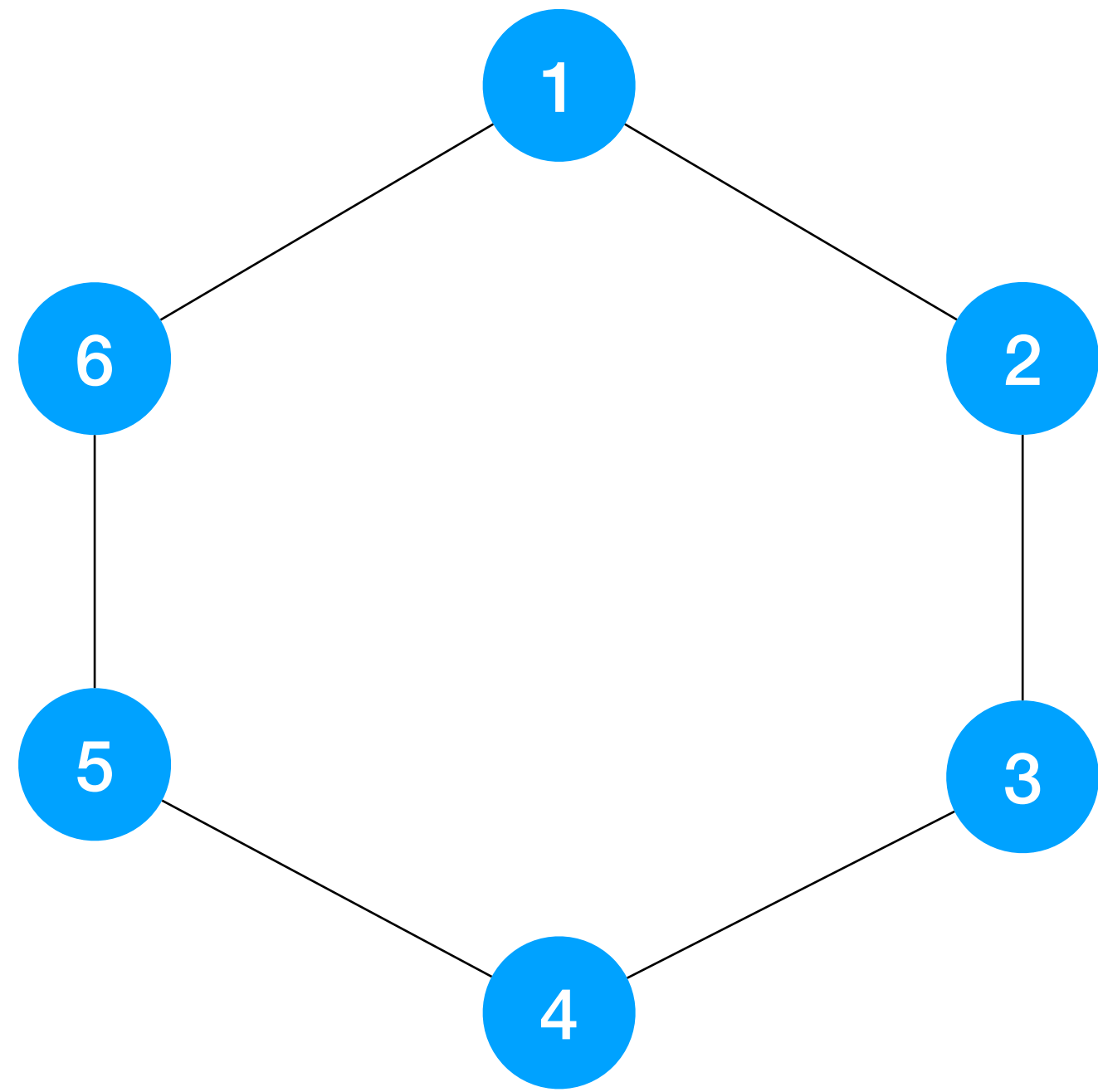
# Trees



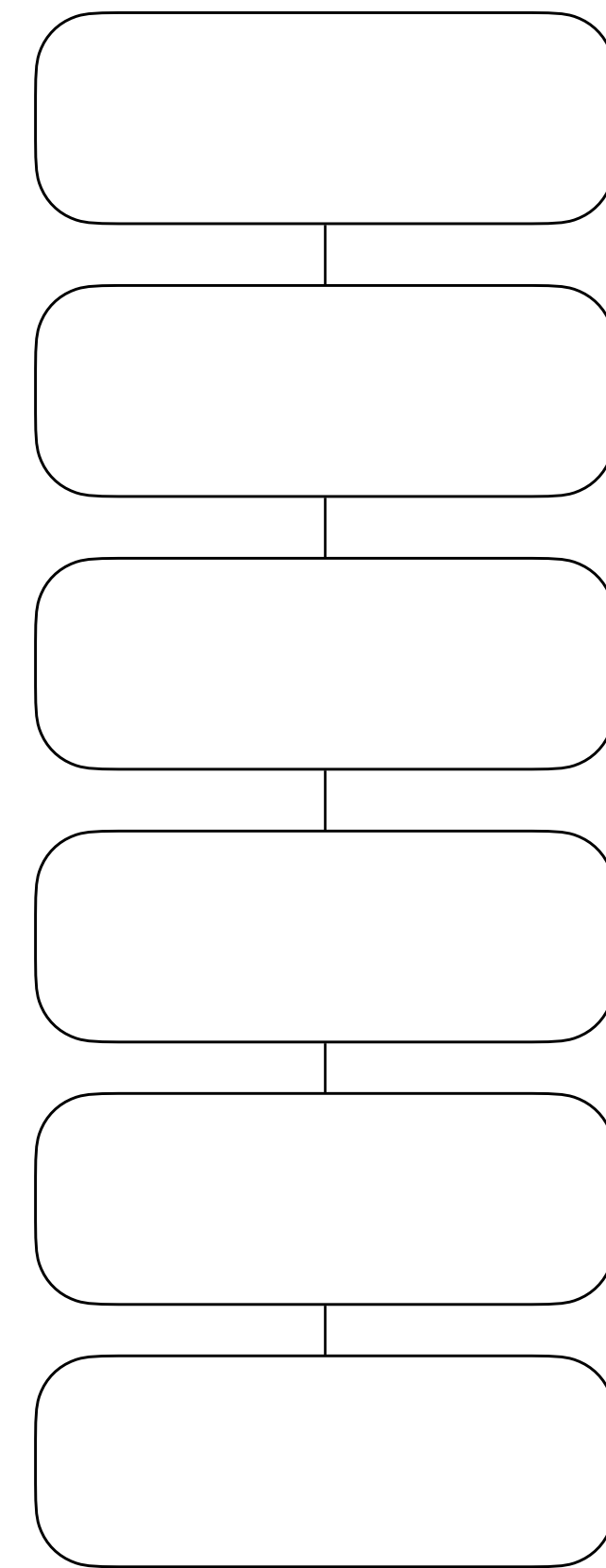
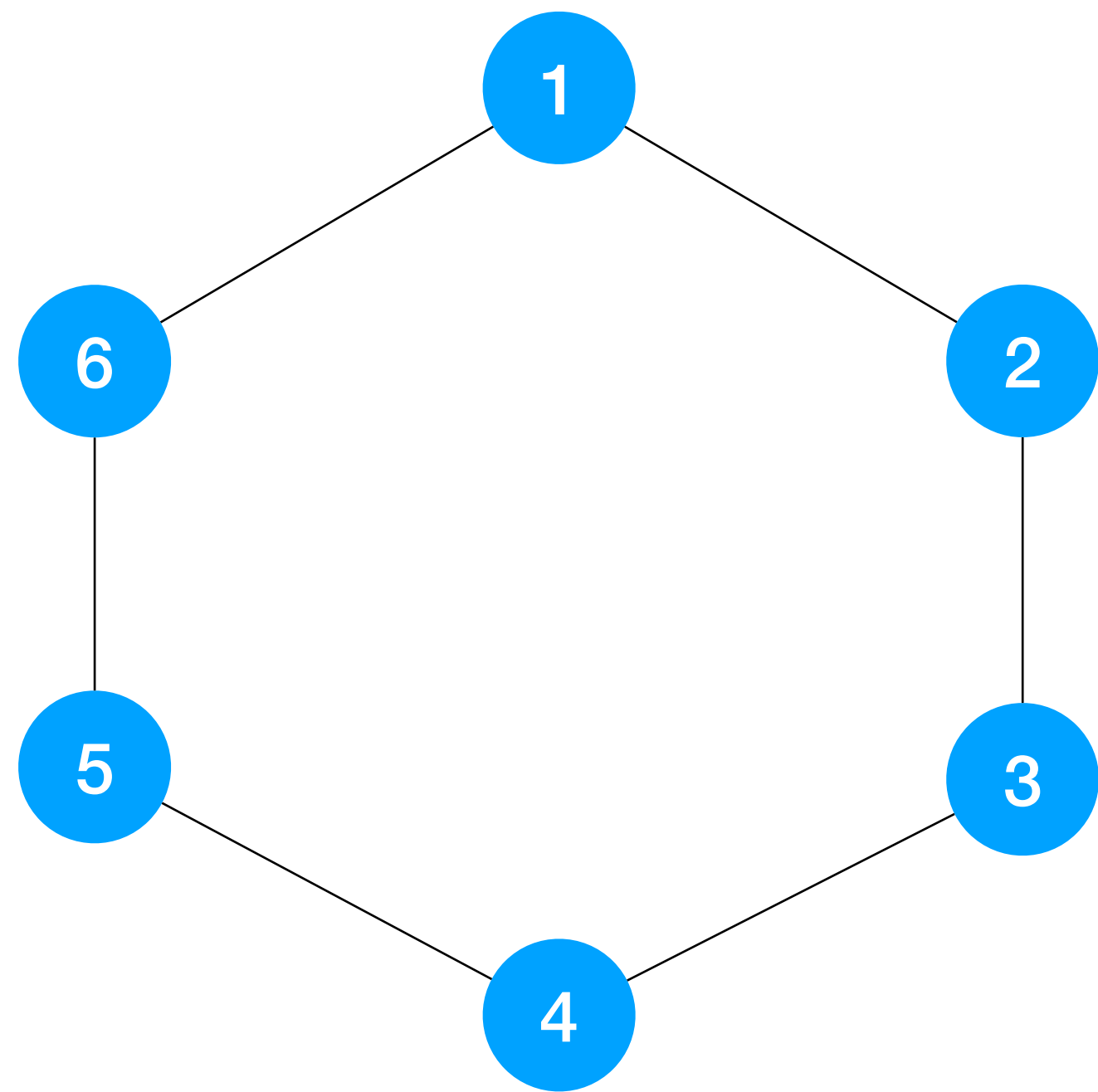
# Trees



# Cycles

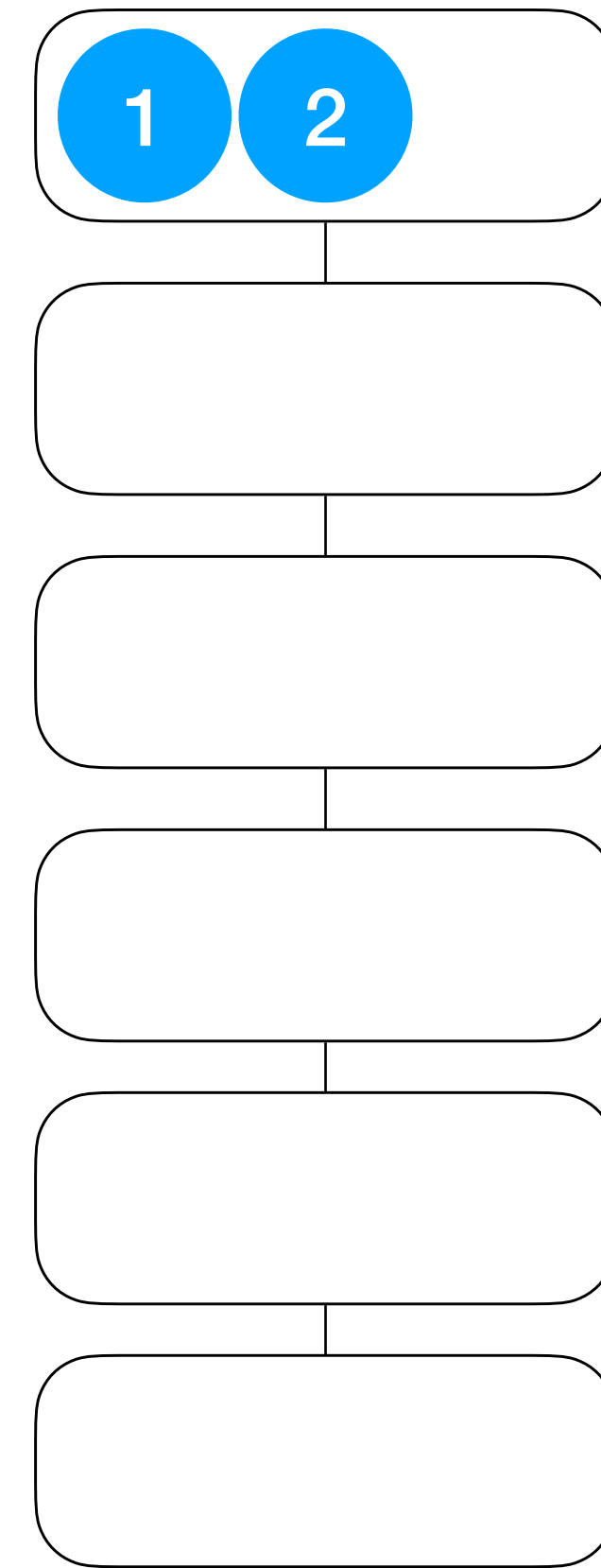
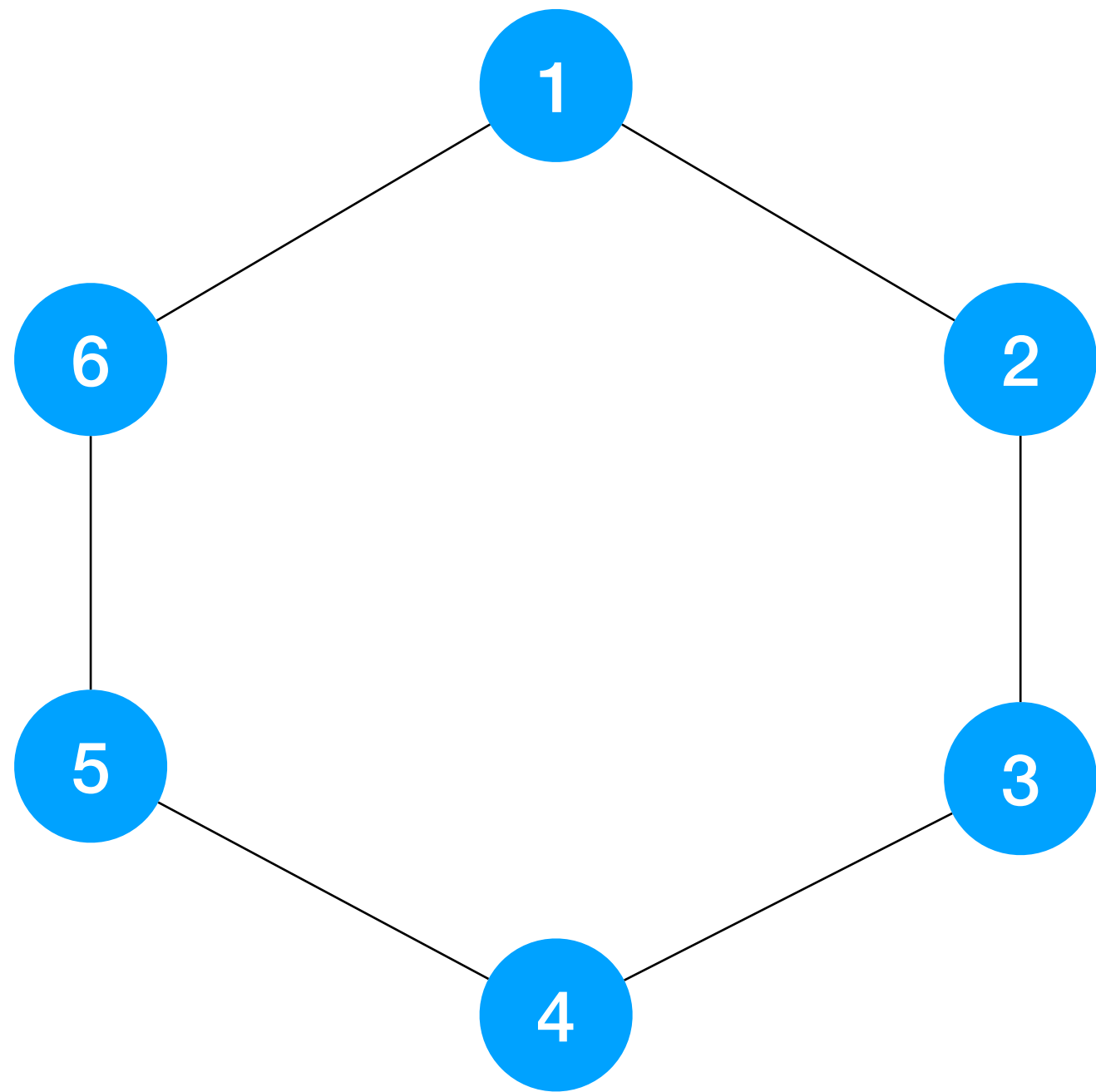


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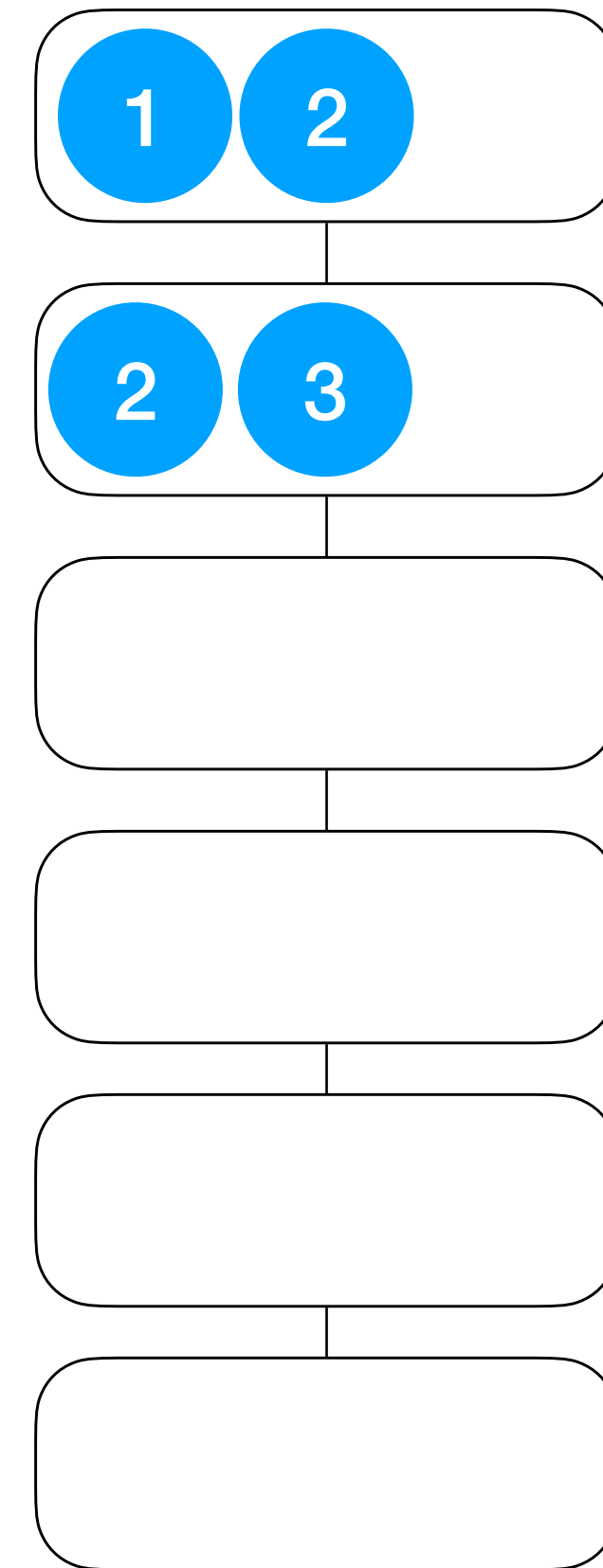
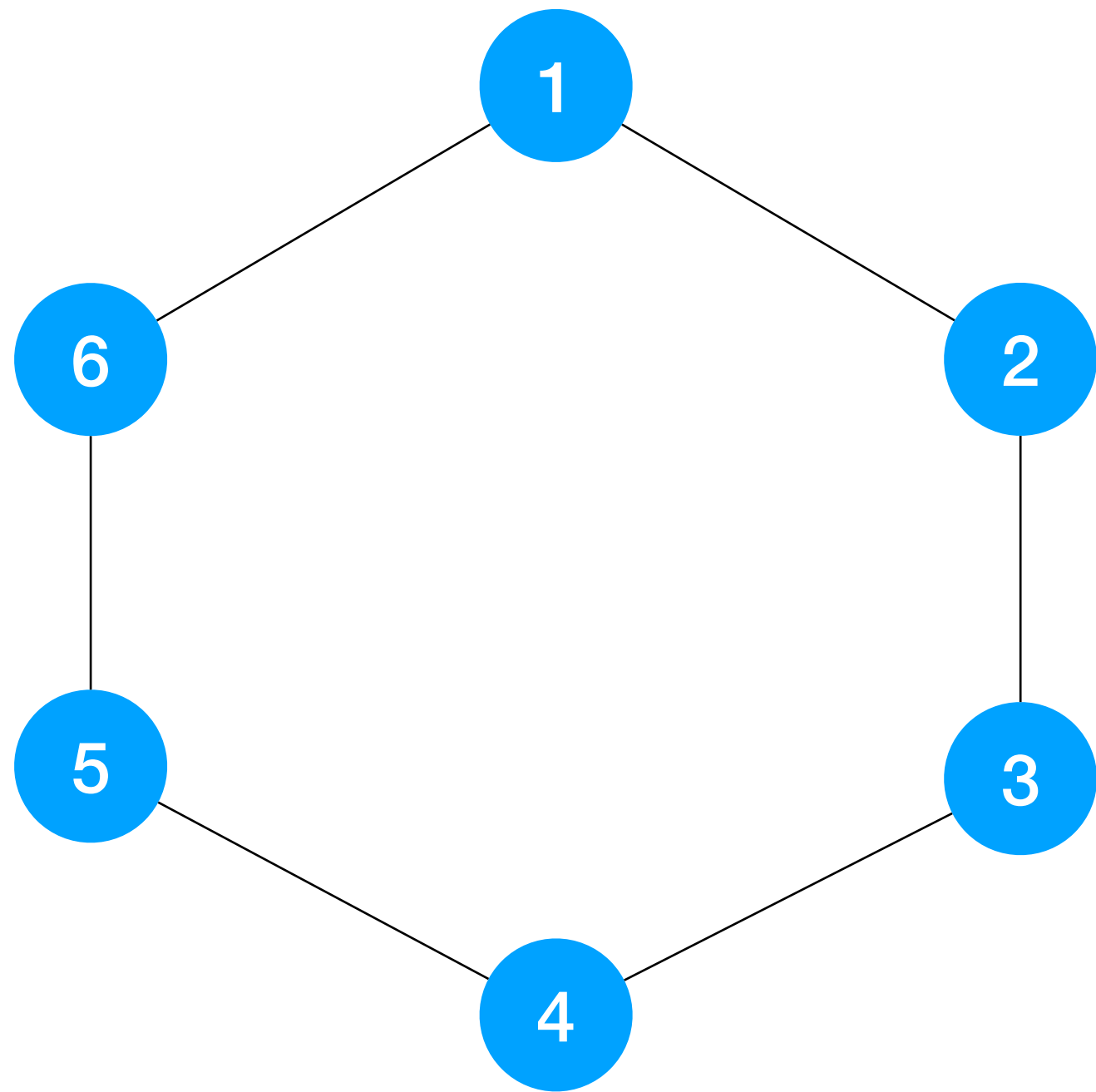




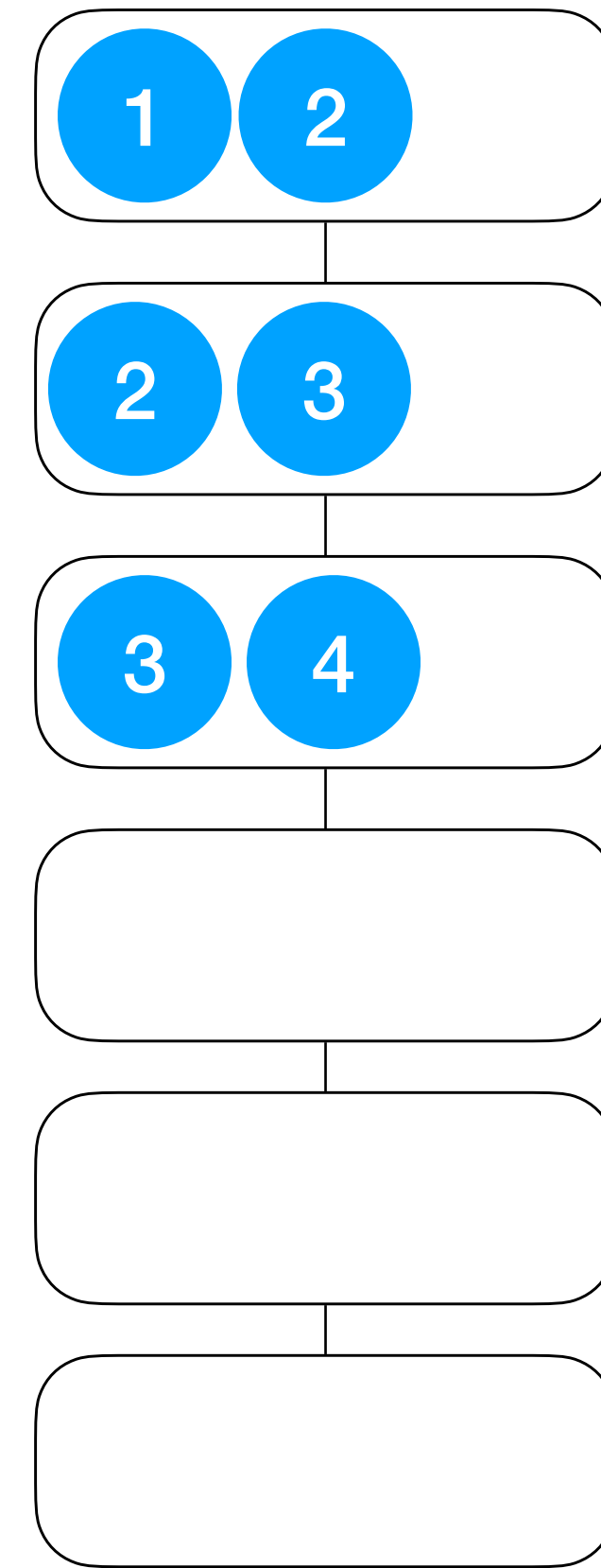
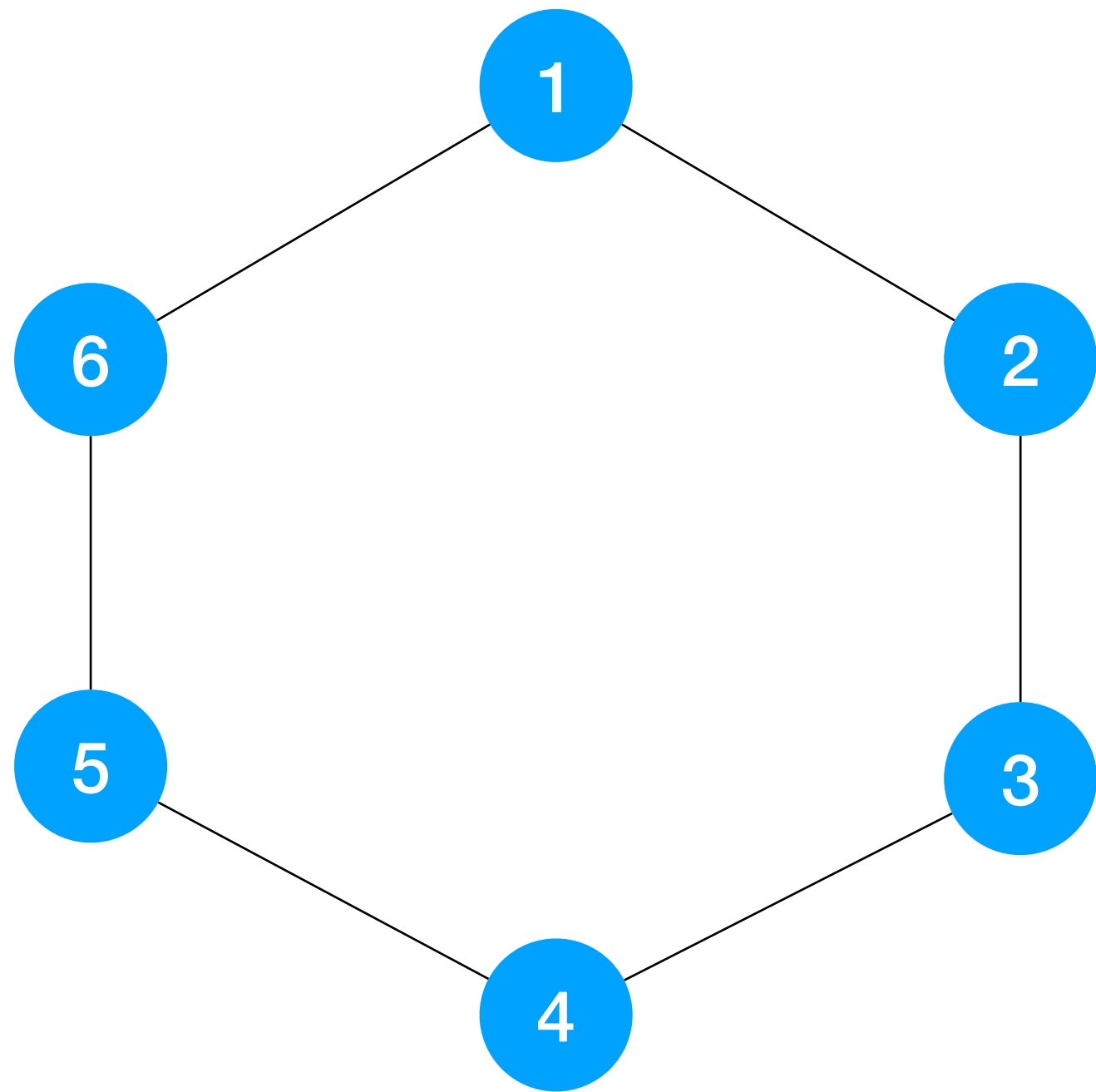
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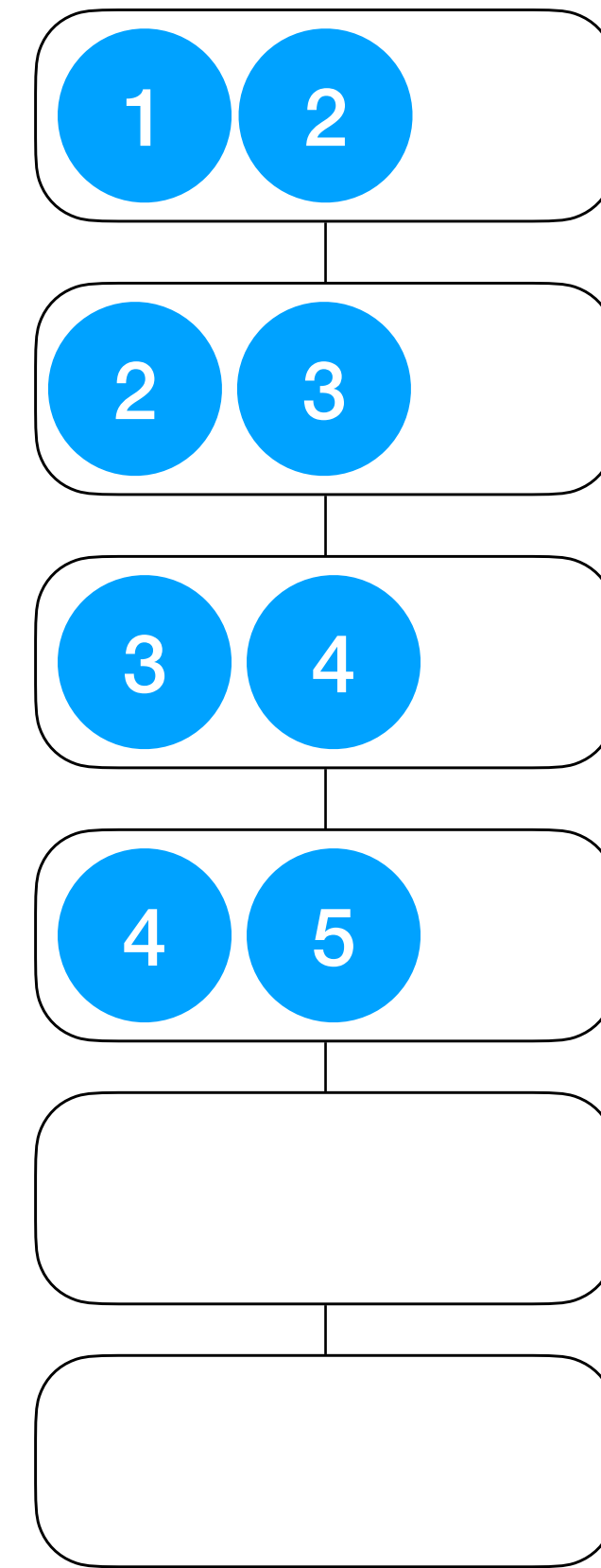
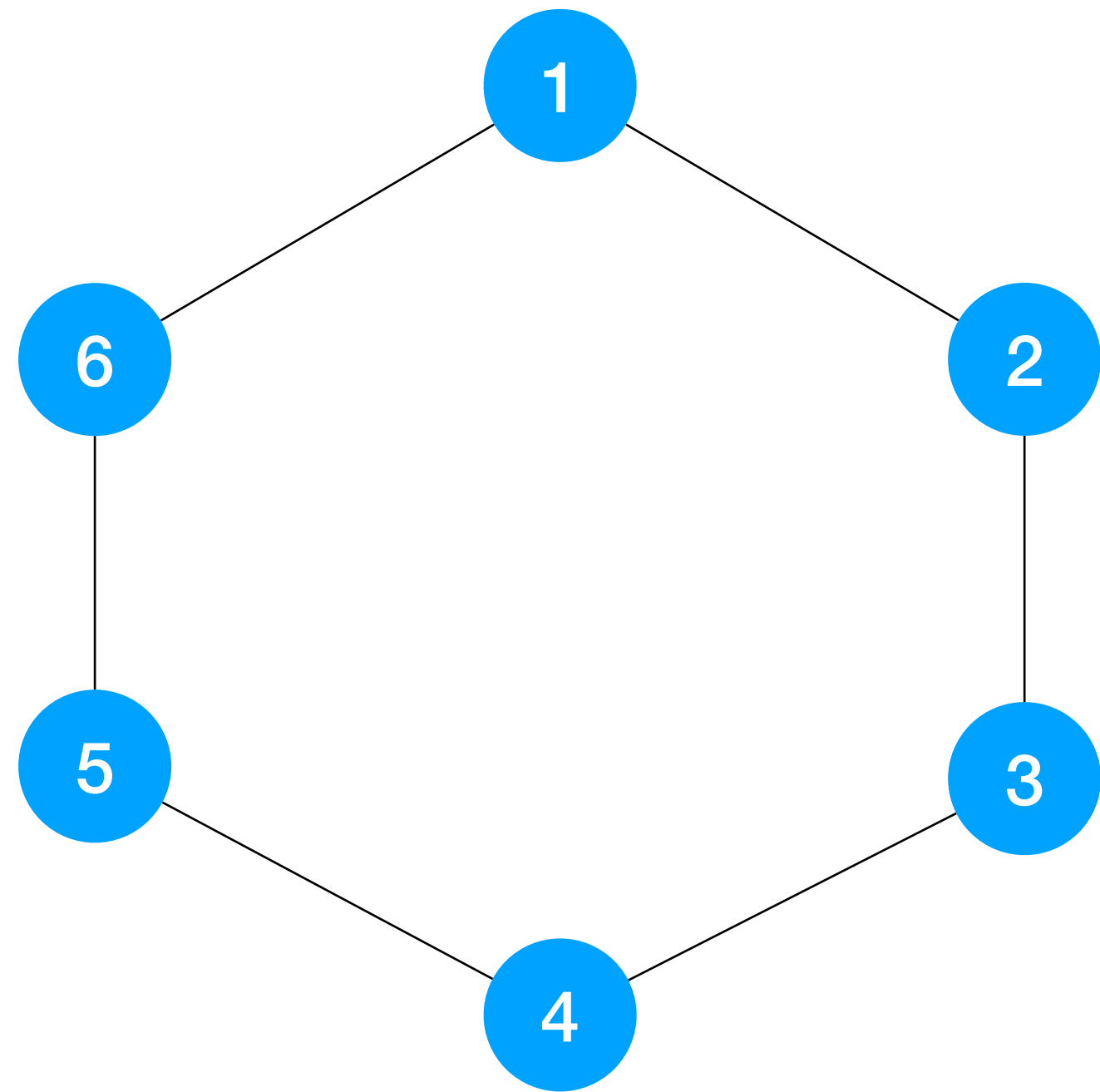
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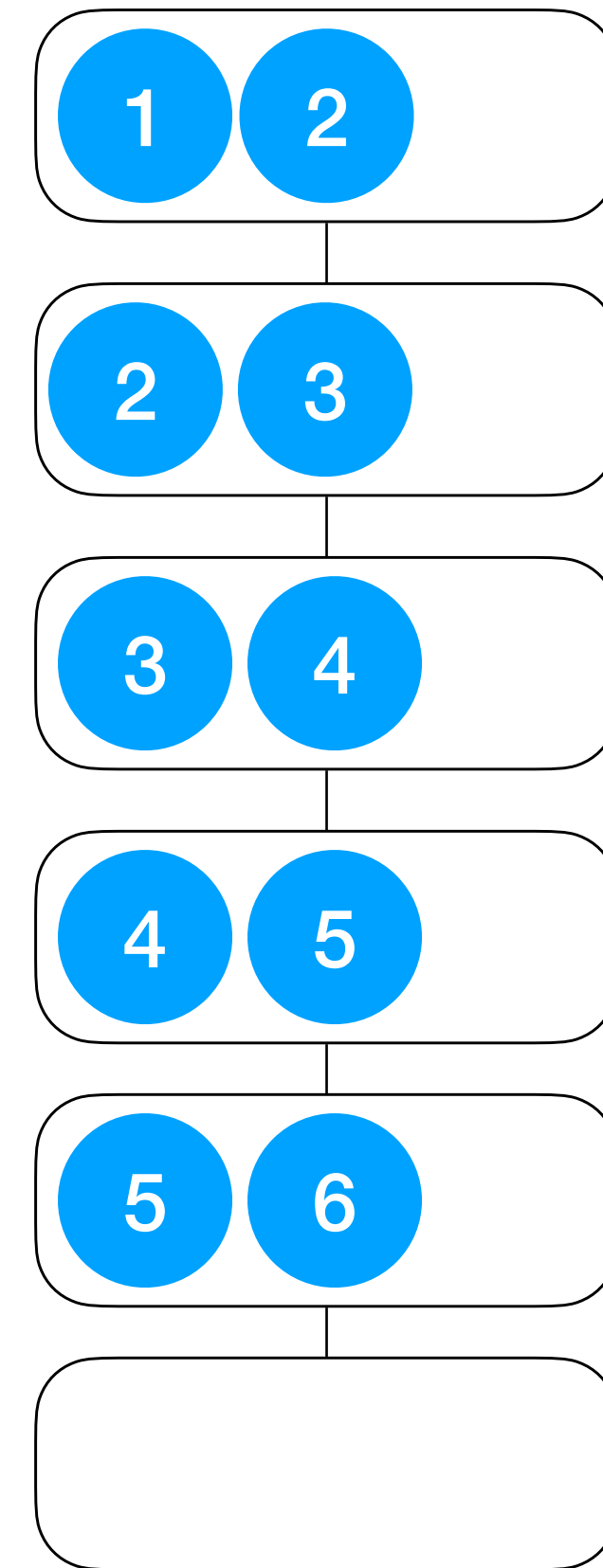
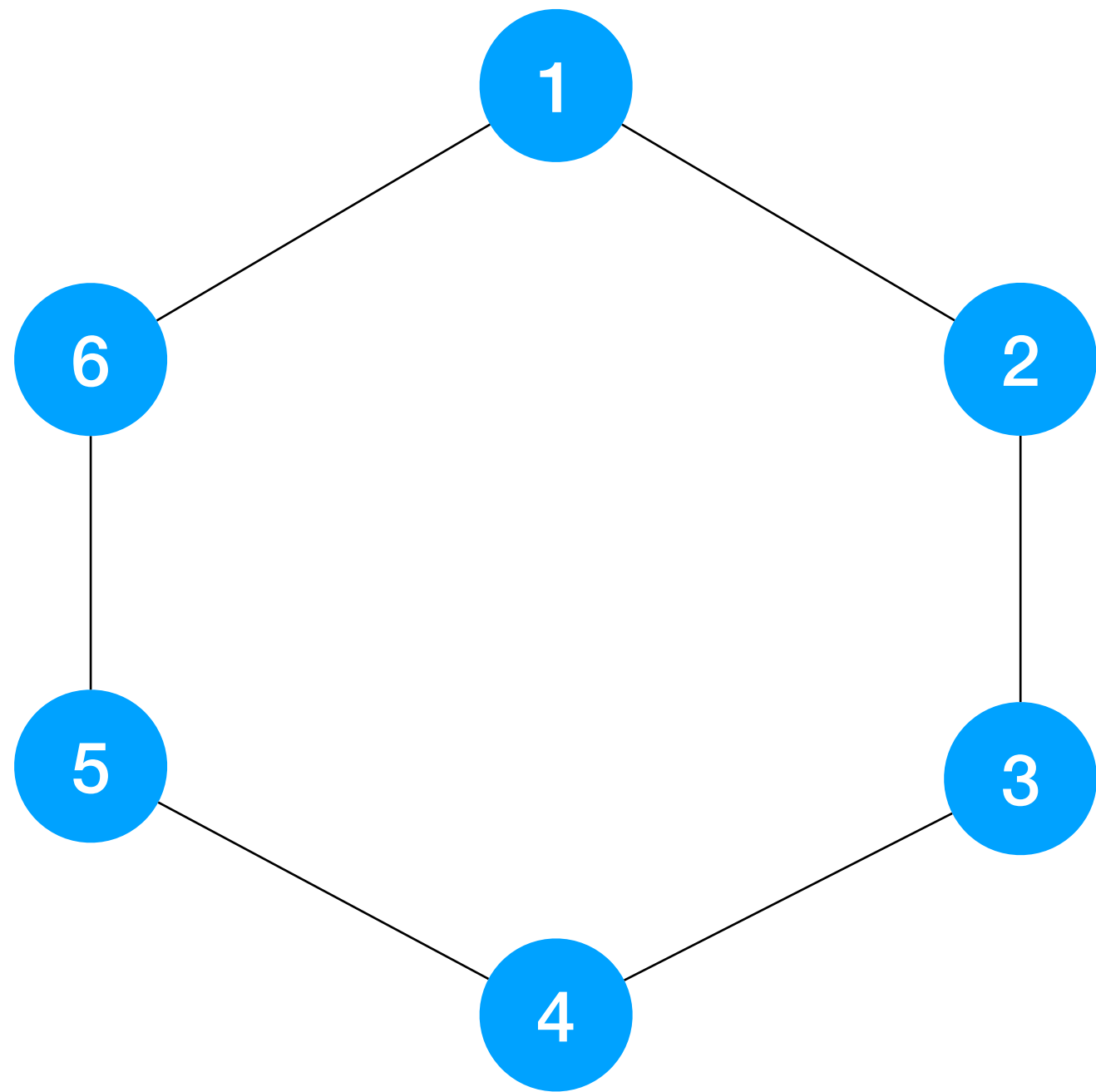
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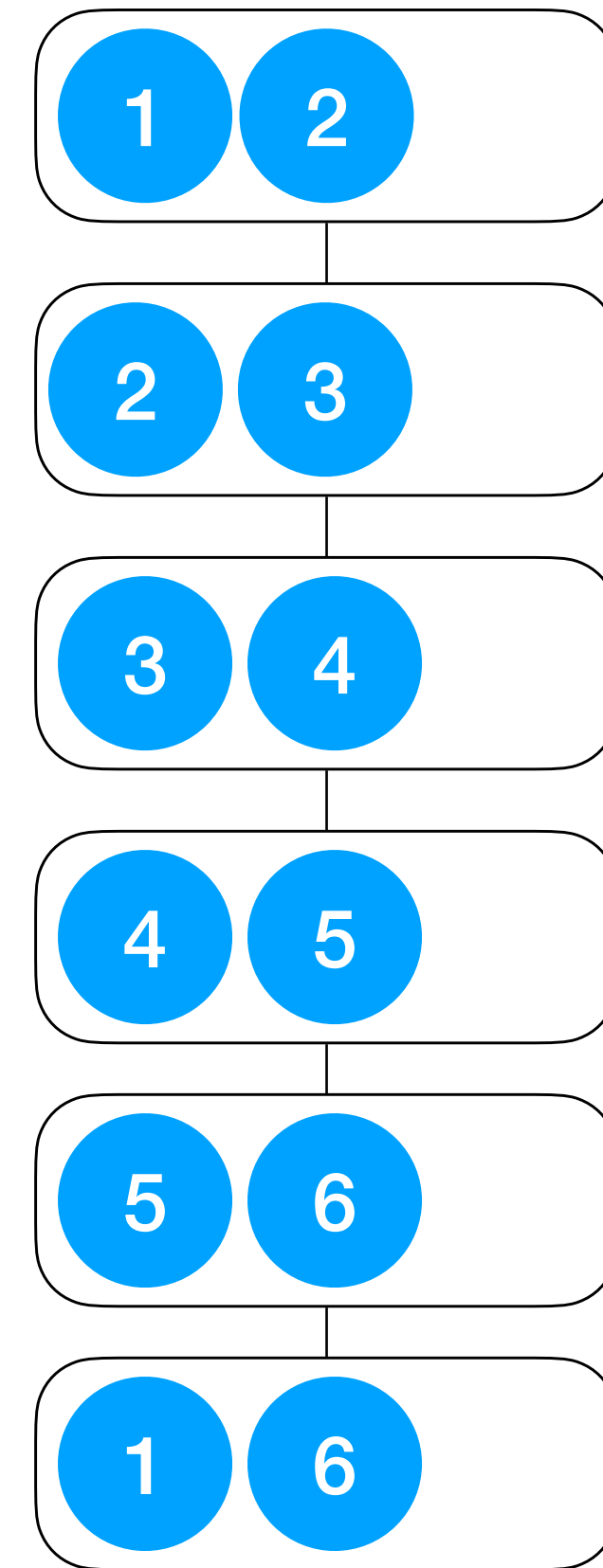
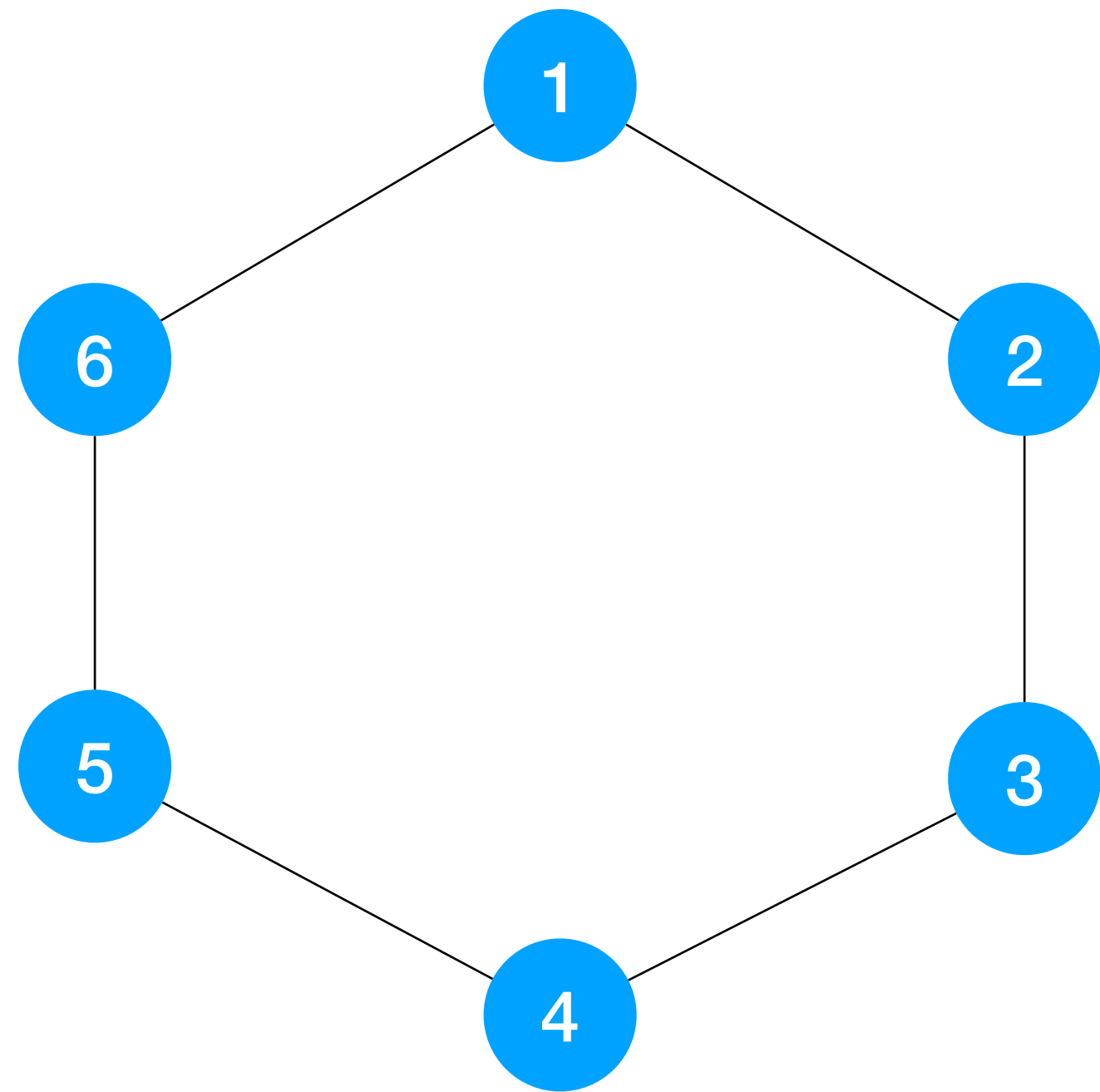
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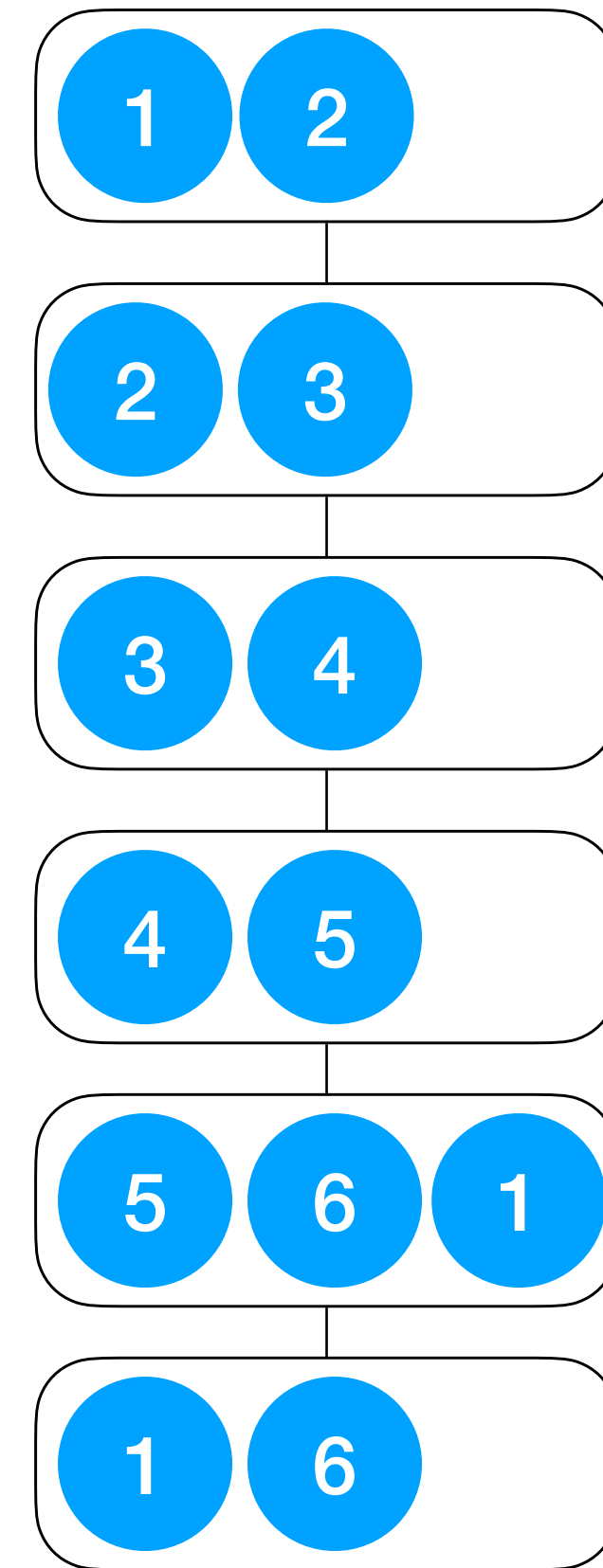
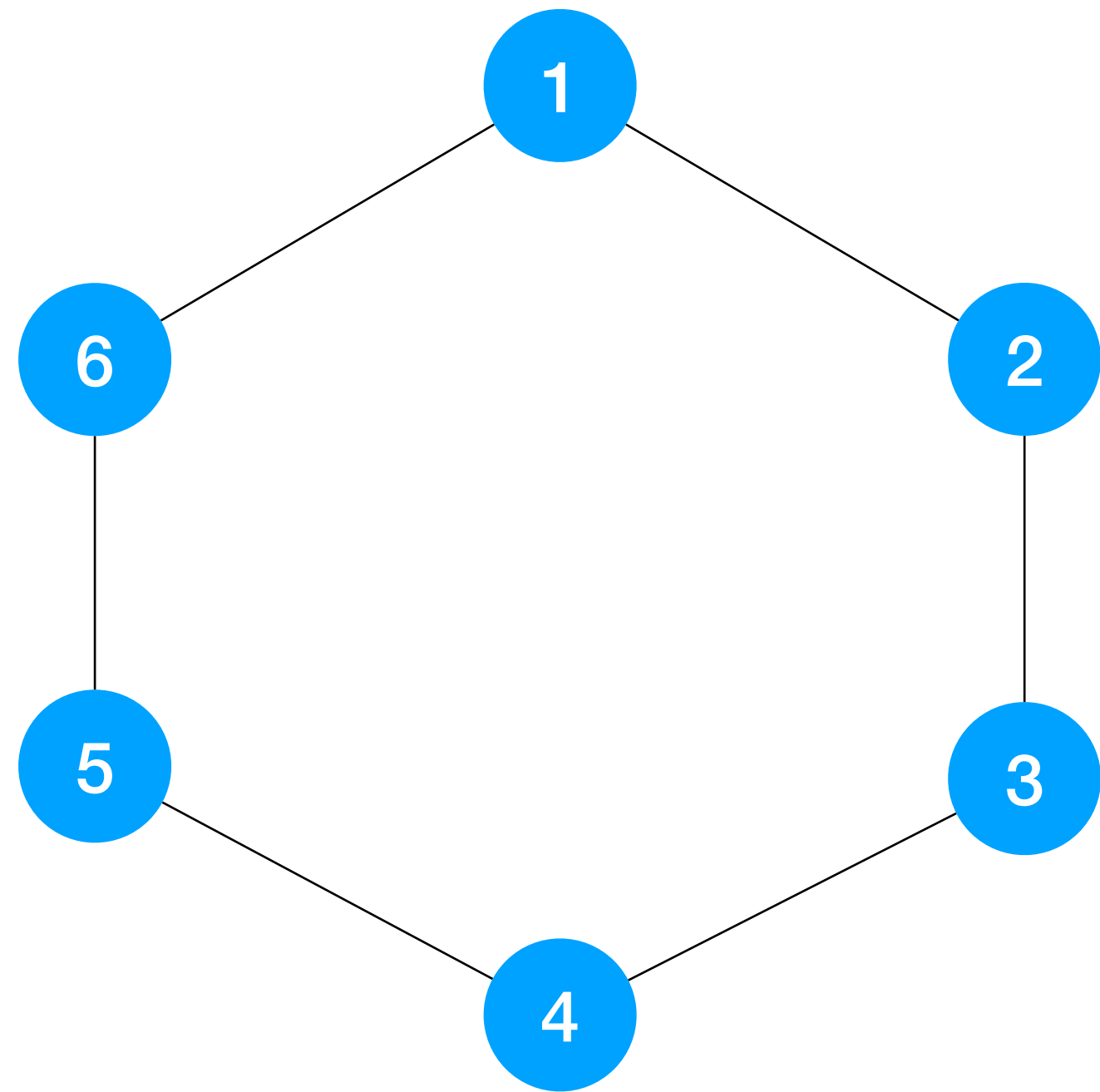
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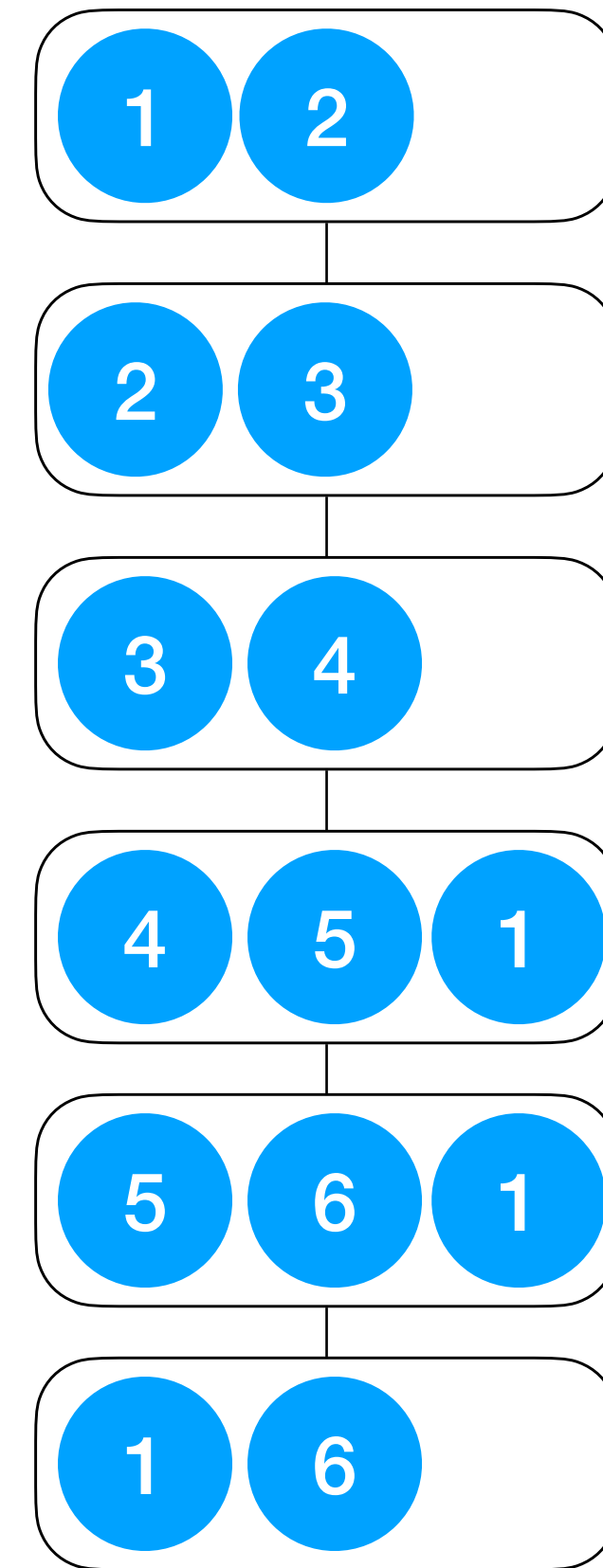
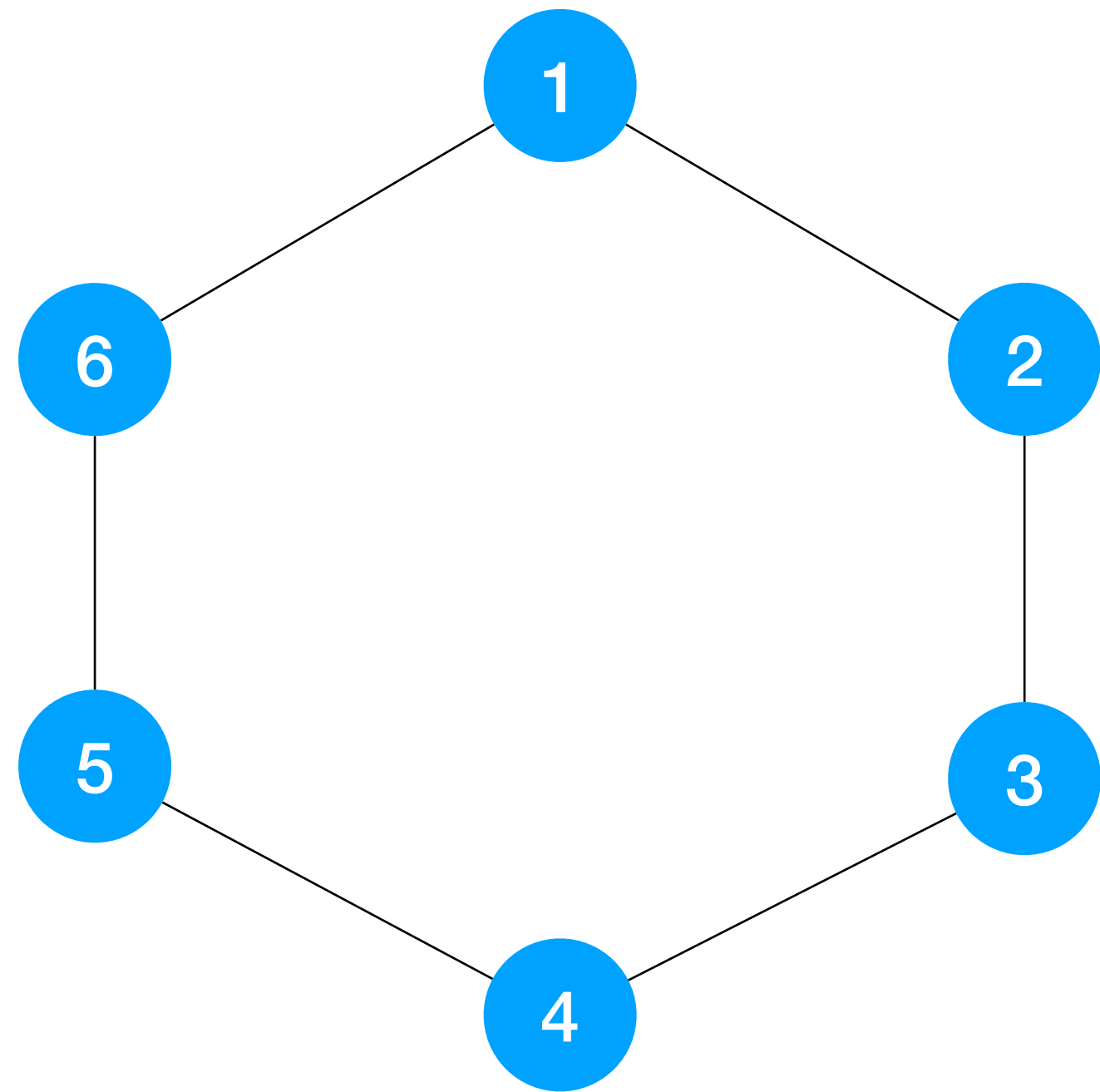
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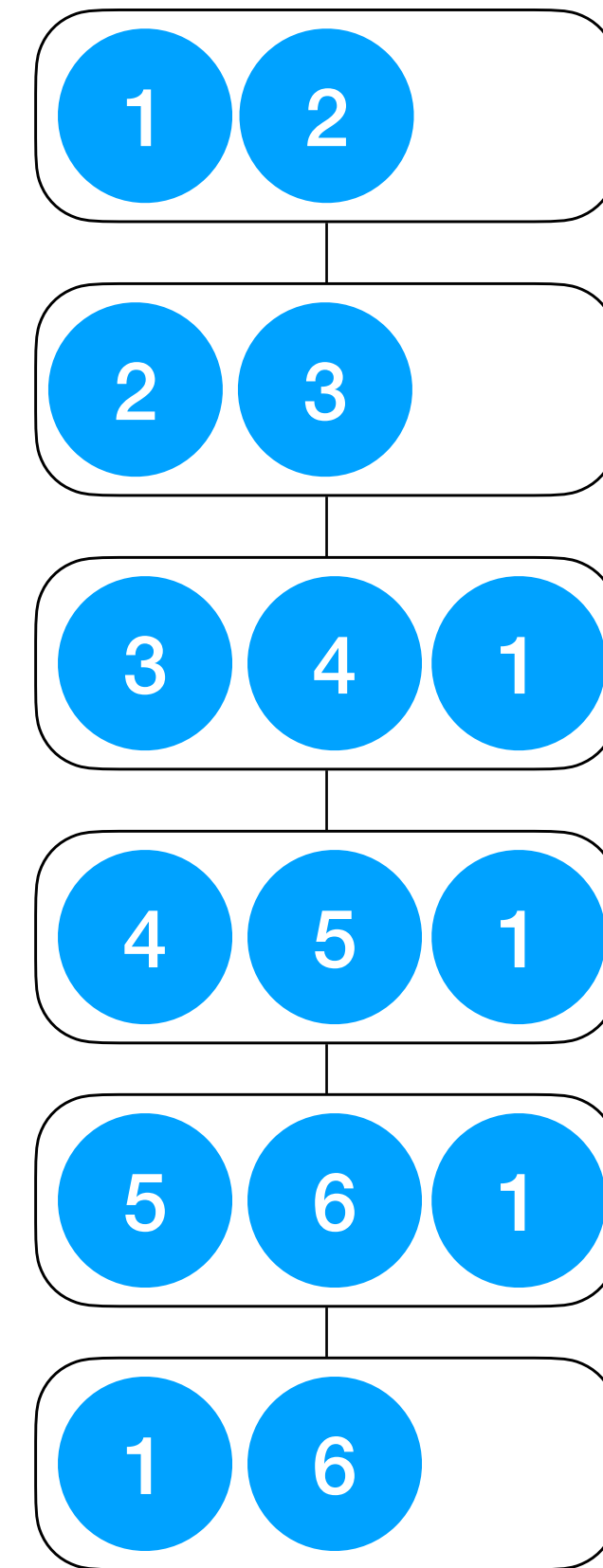
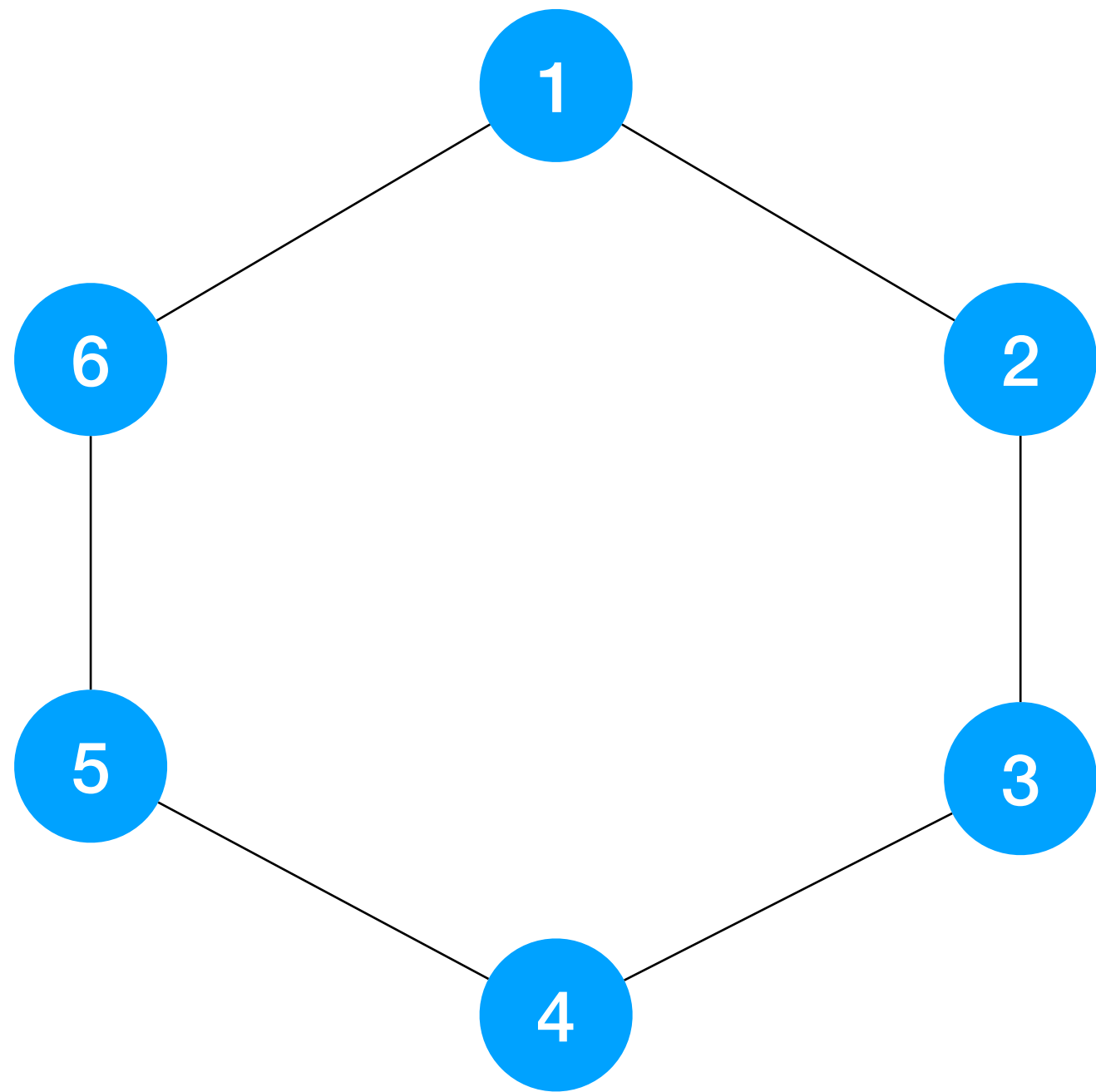


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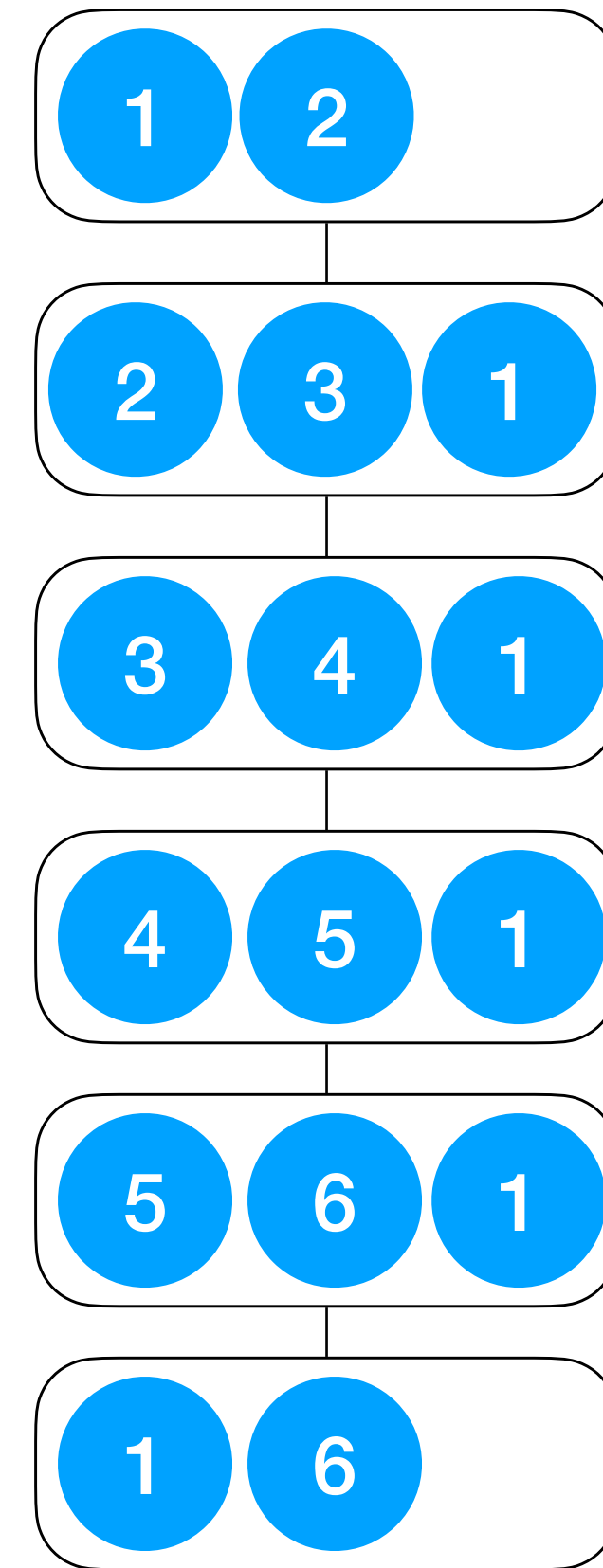
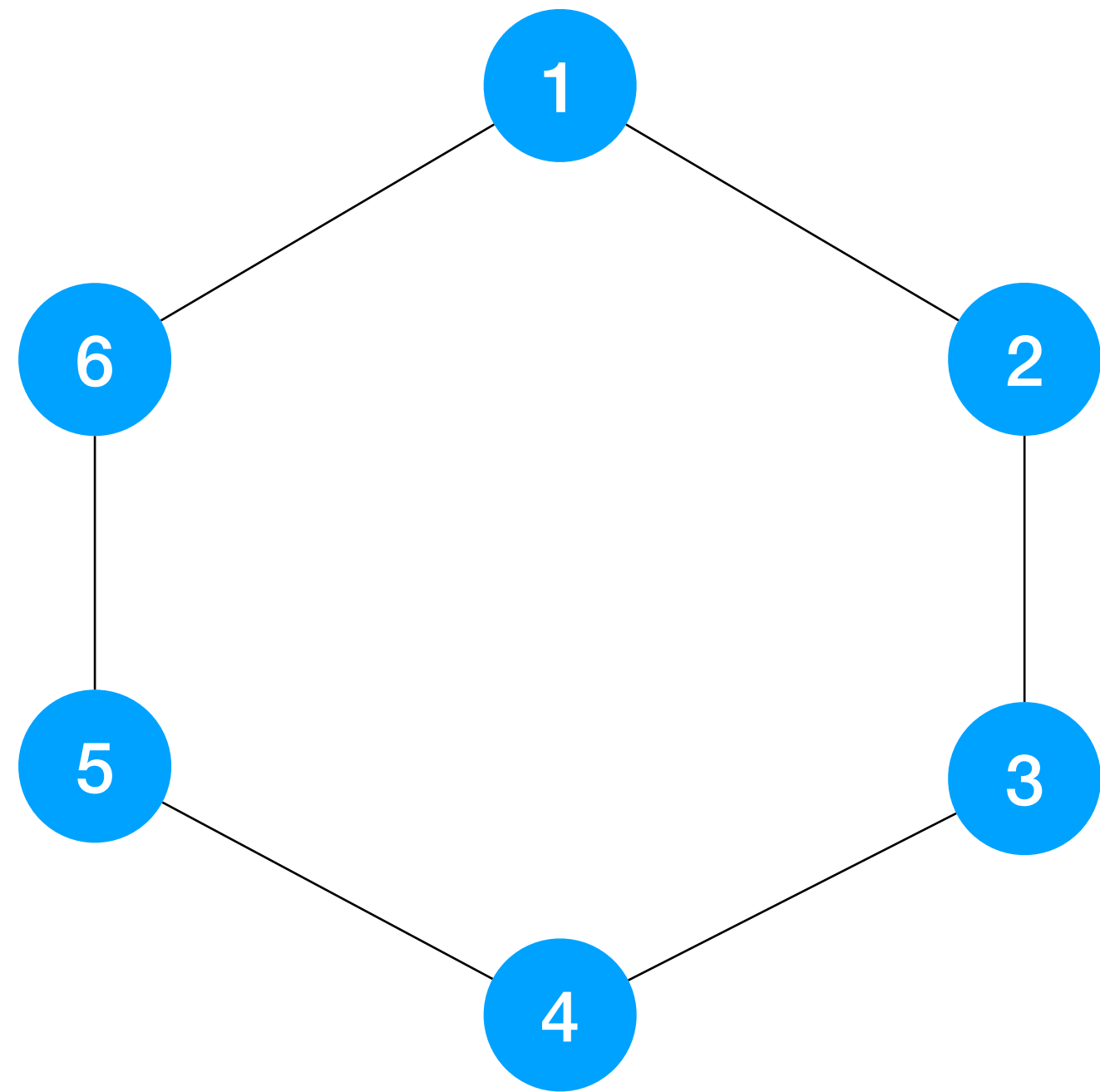




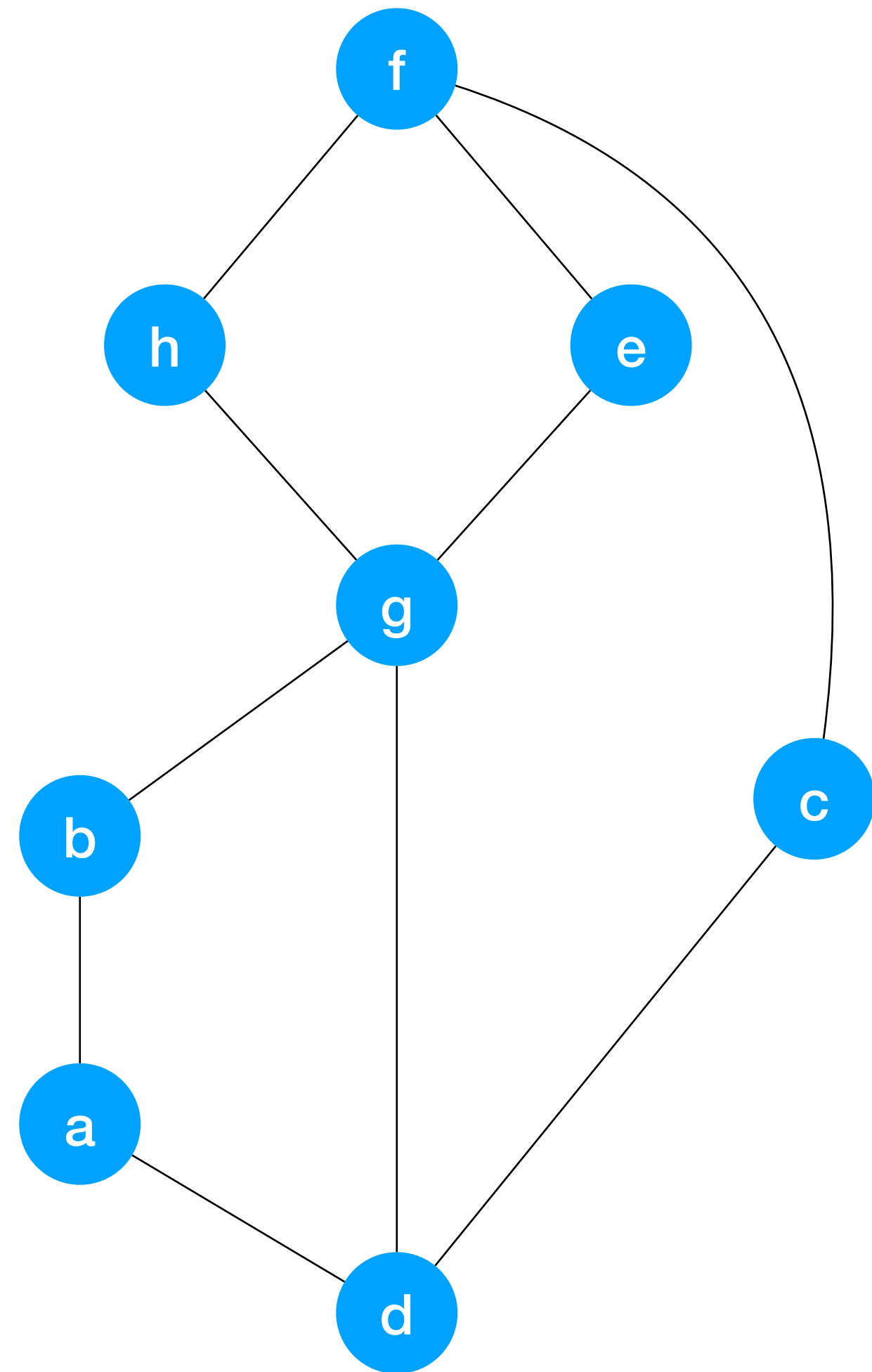
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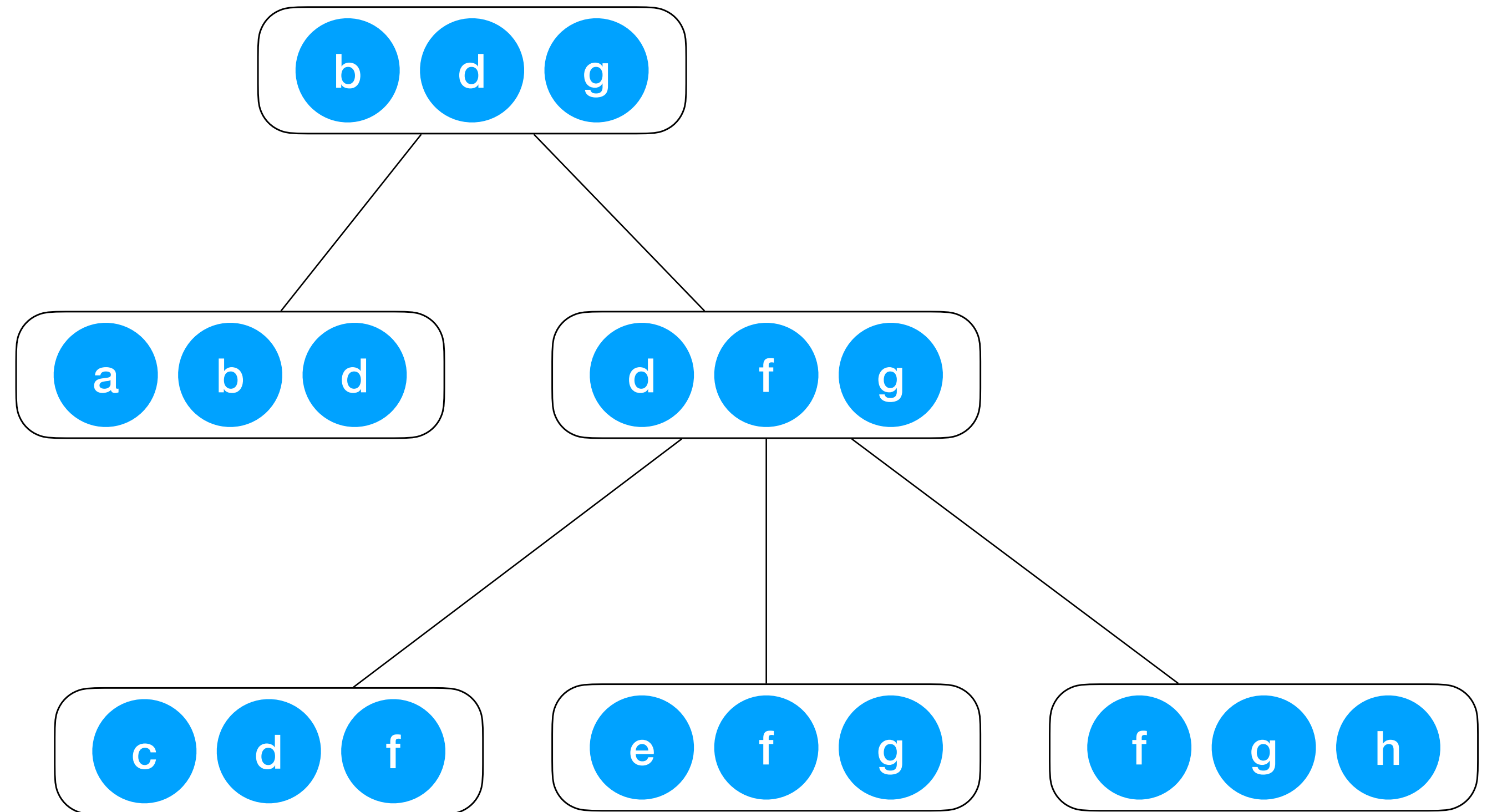
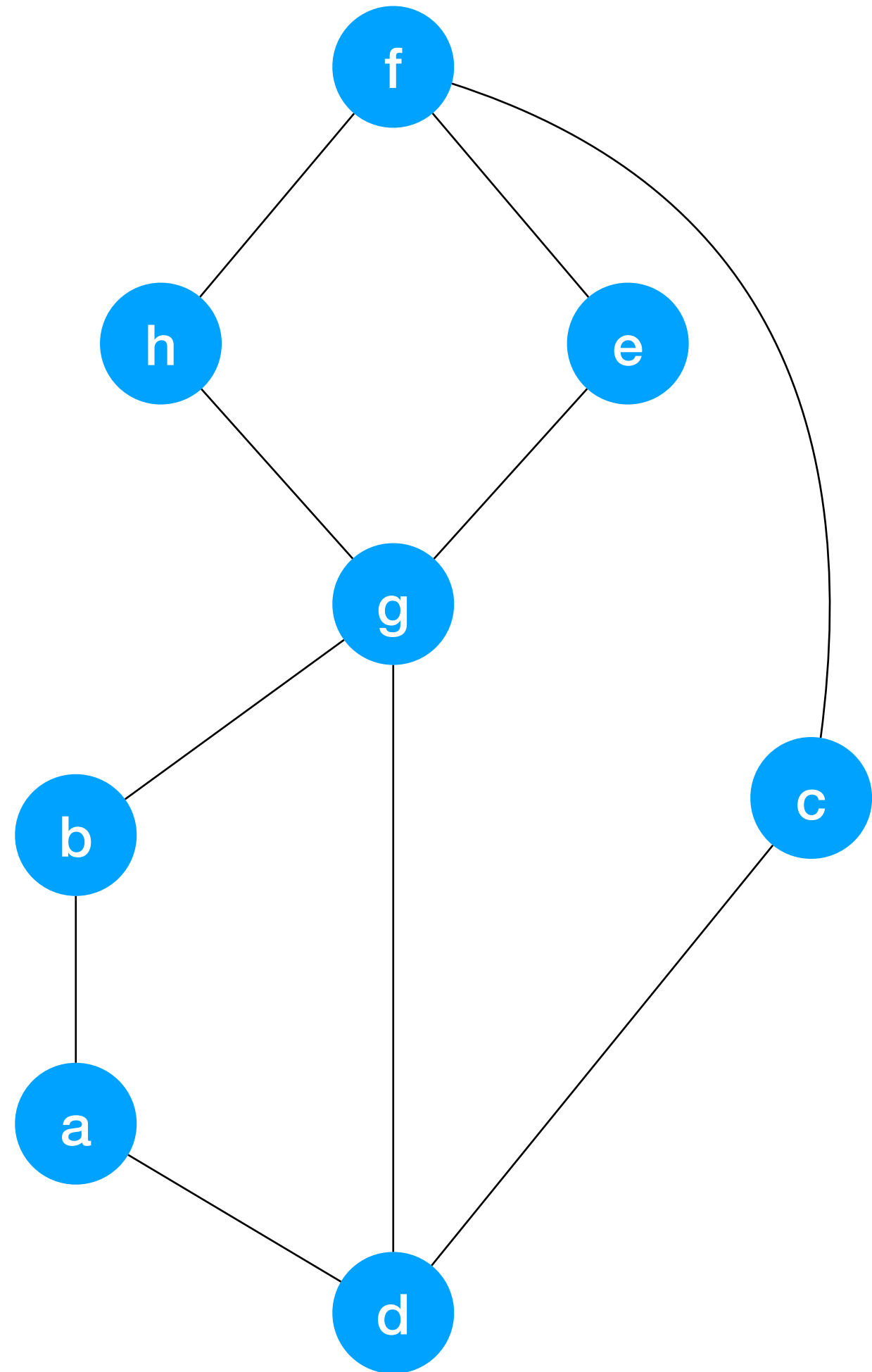
# Cycles



# Another Example



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# Properties of Treewidth

# Subgraphs

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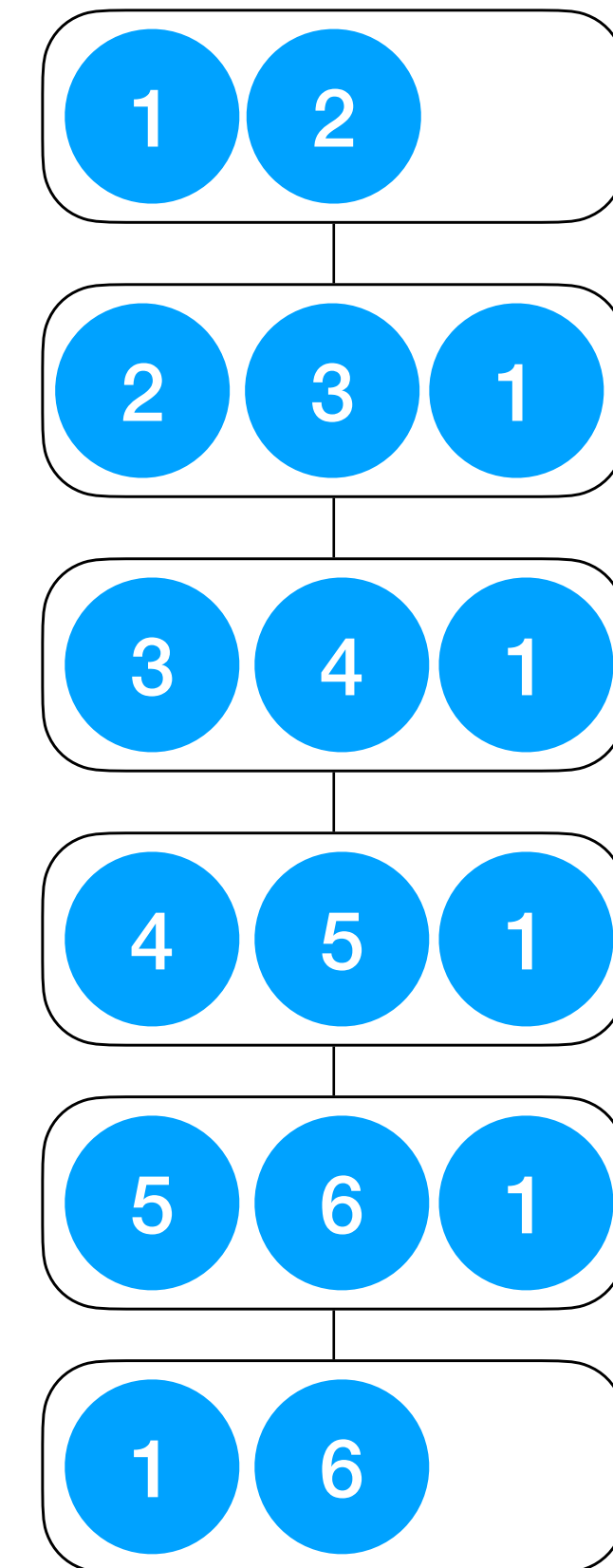
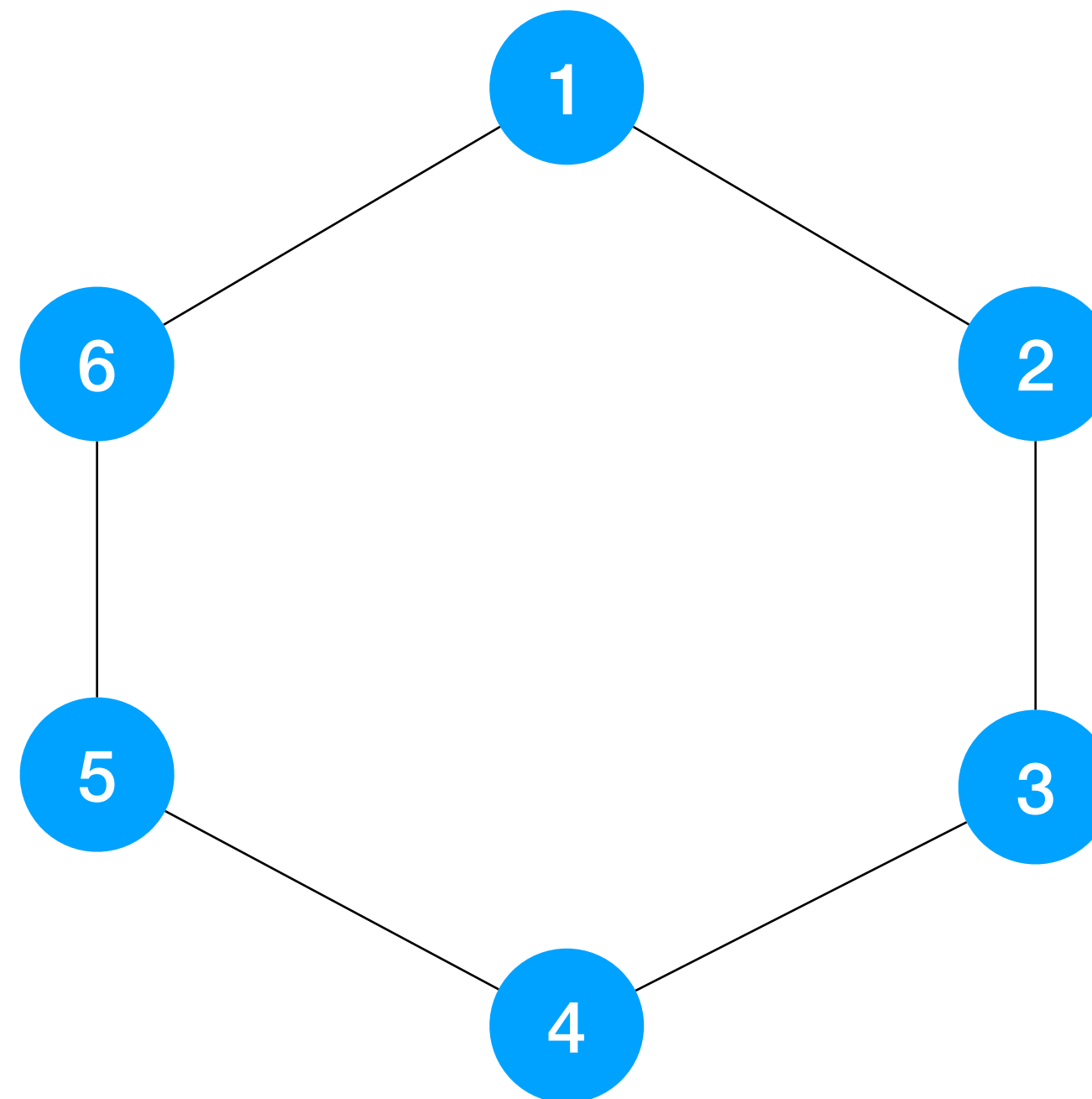


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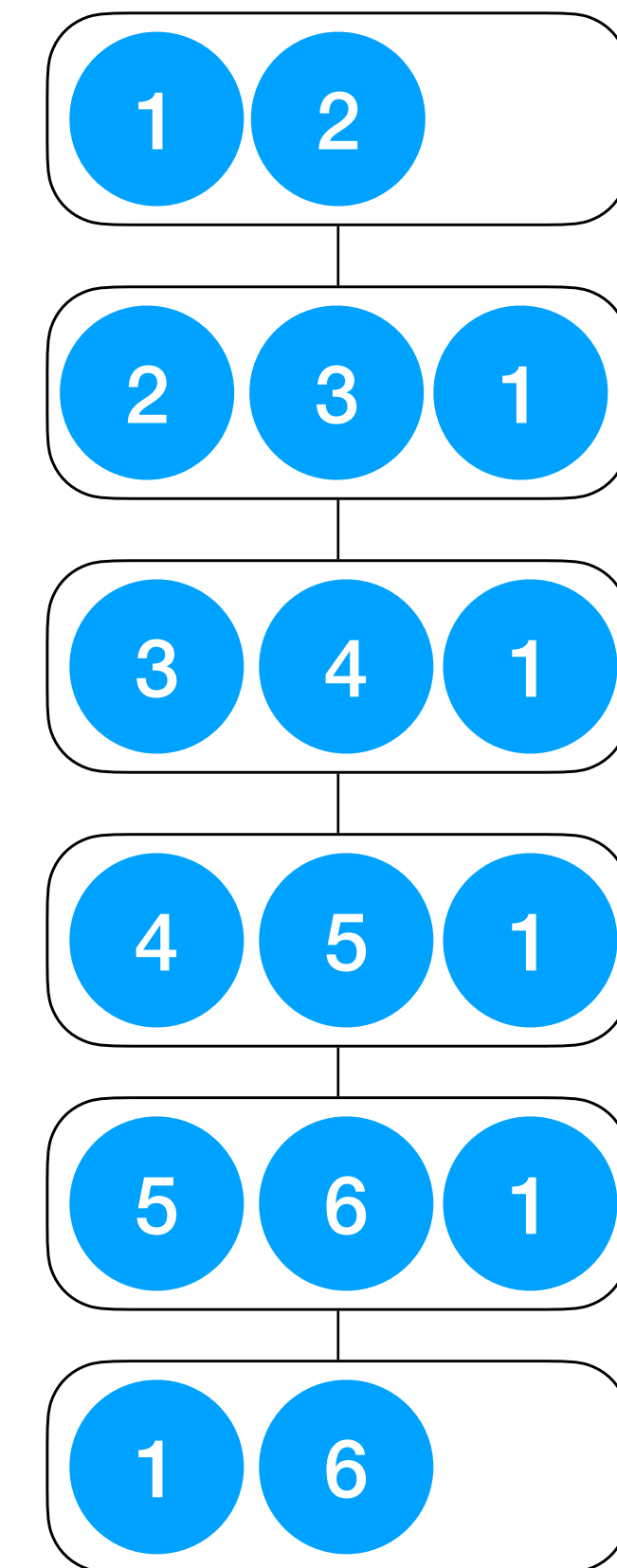
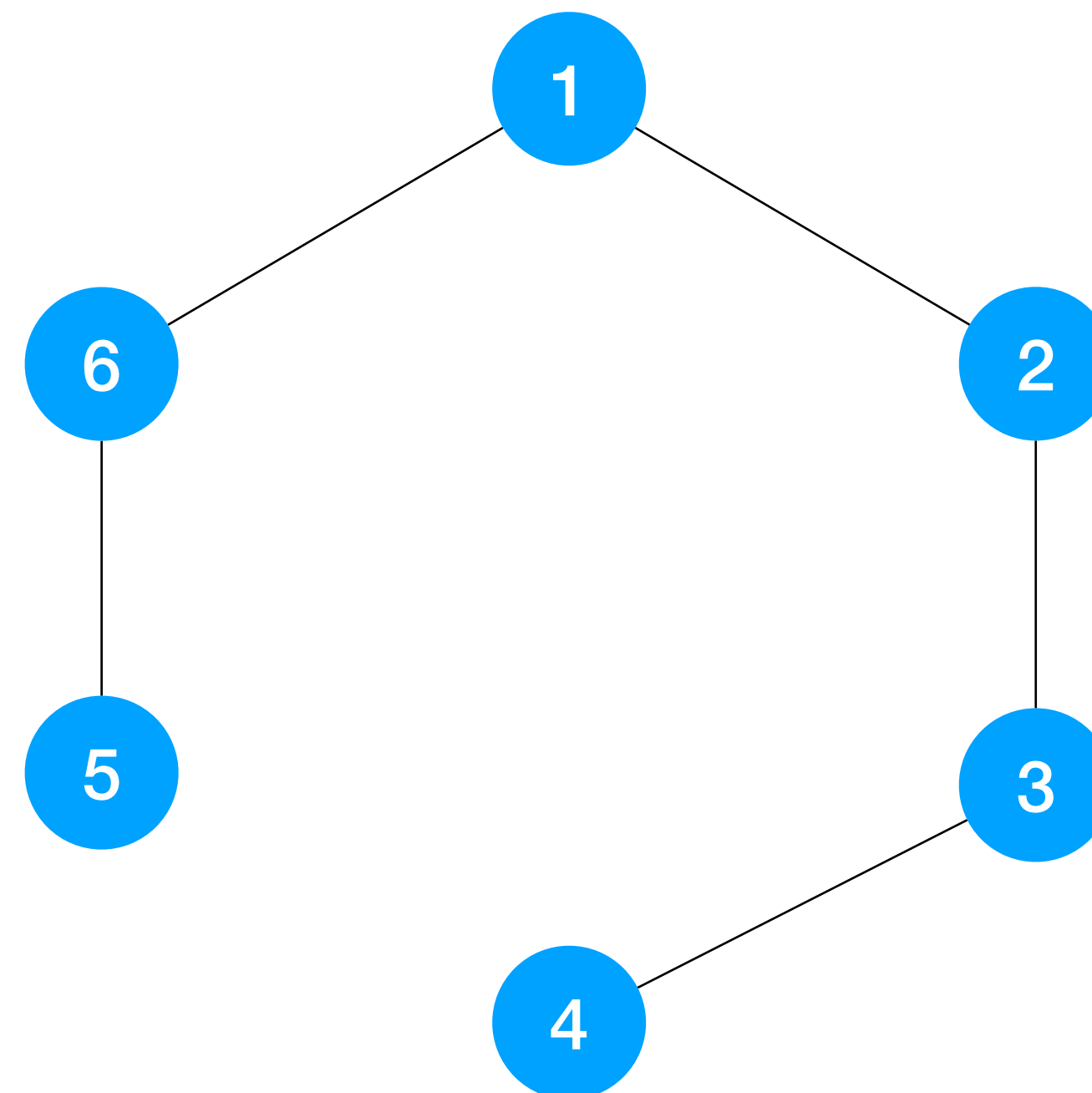


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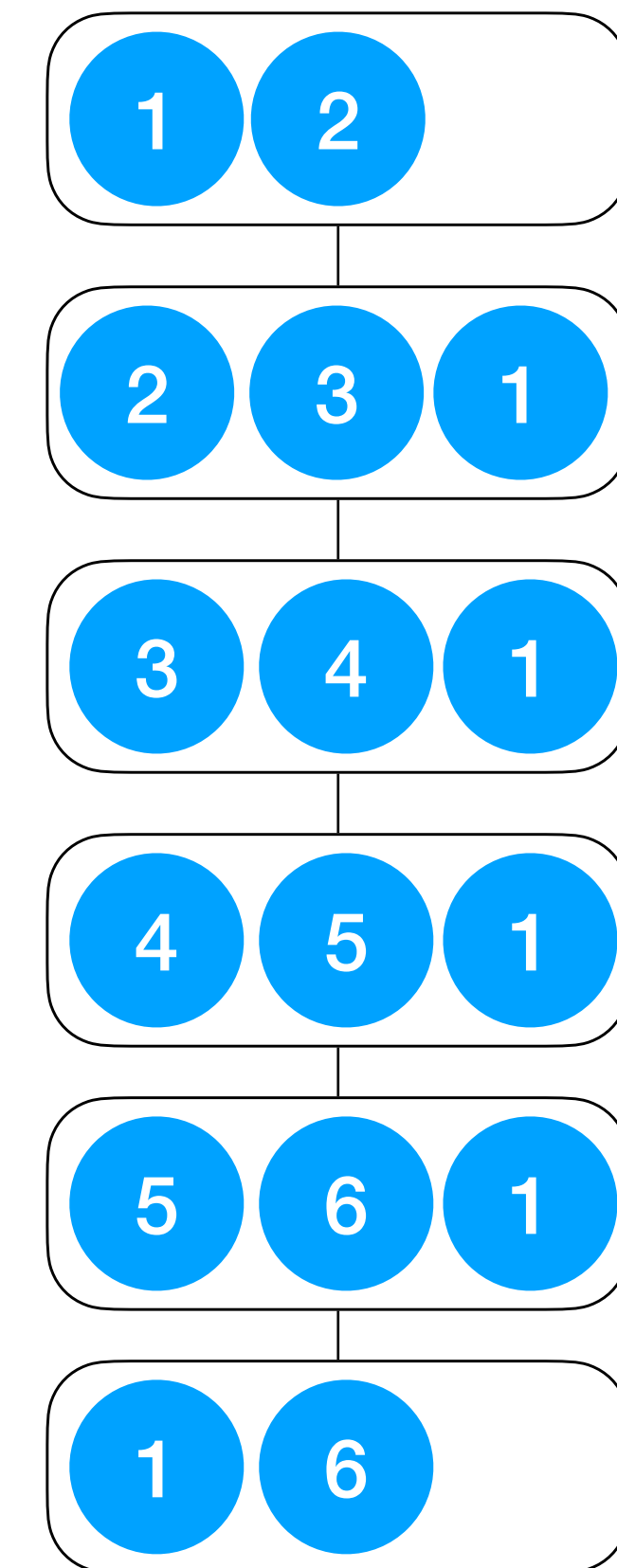
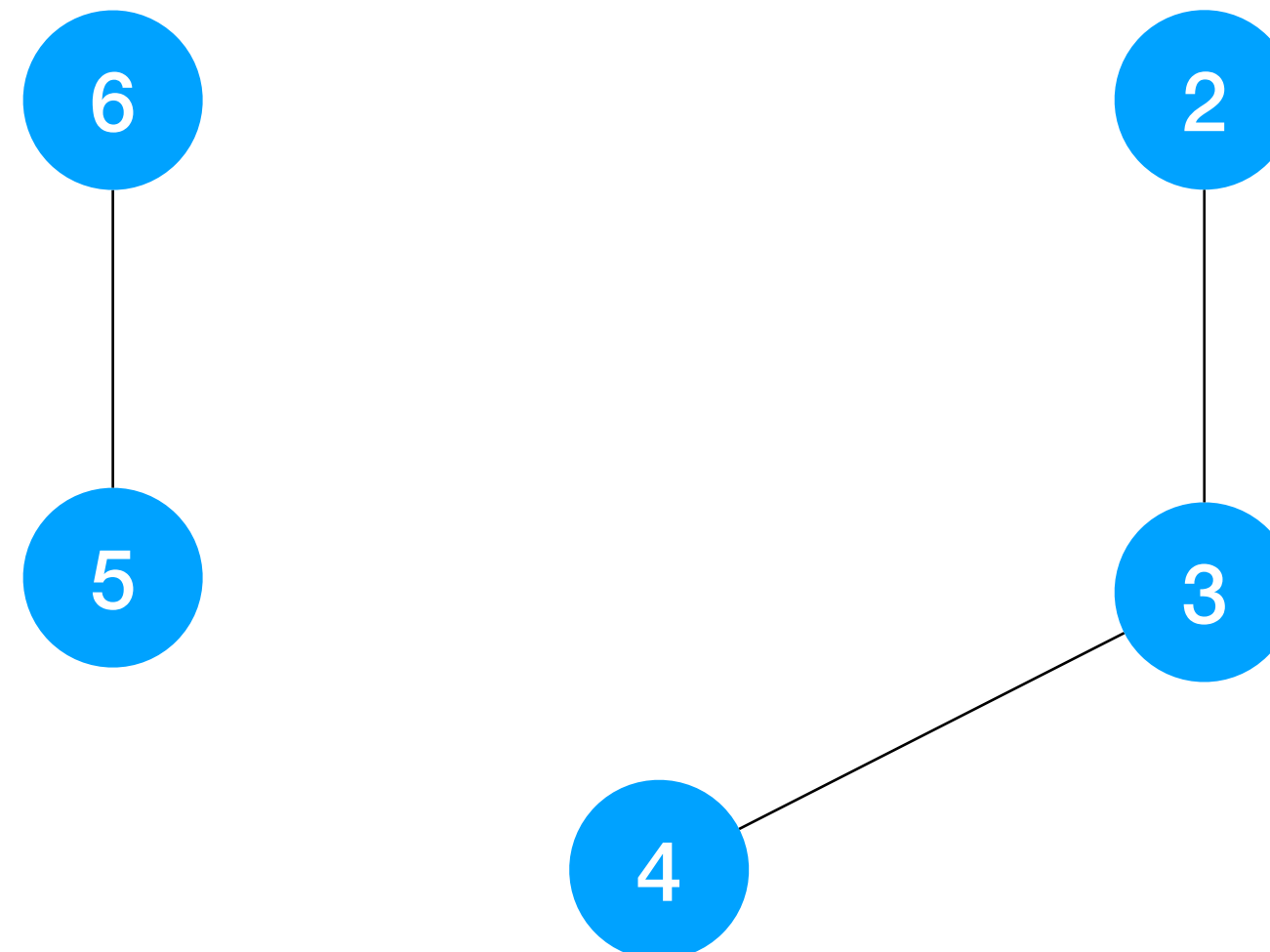


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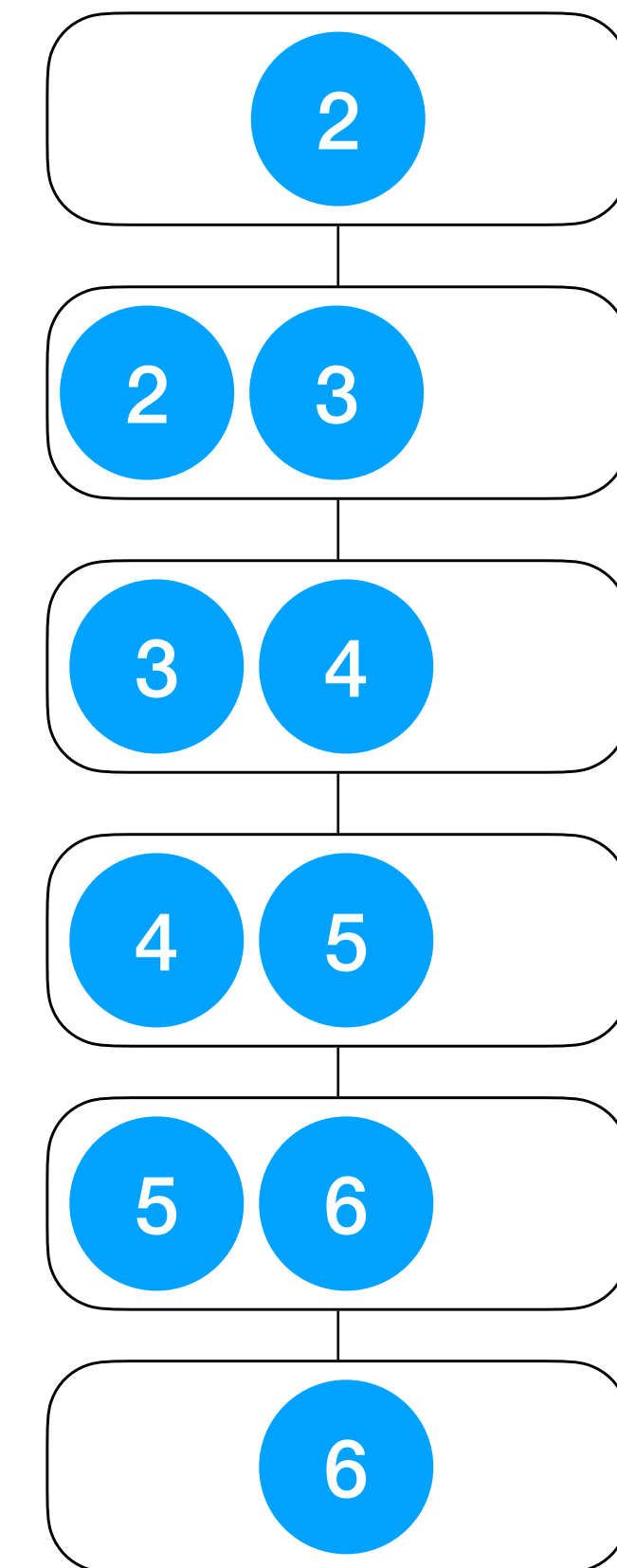
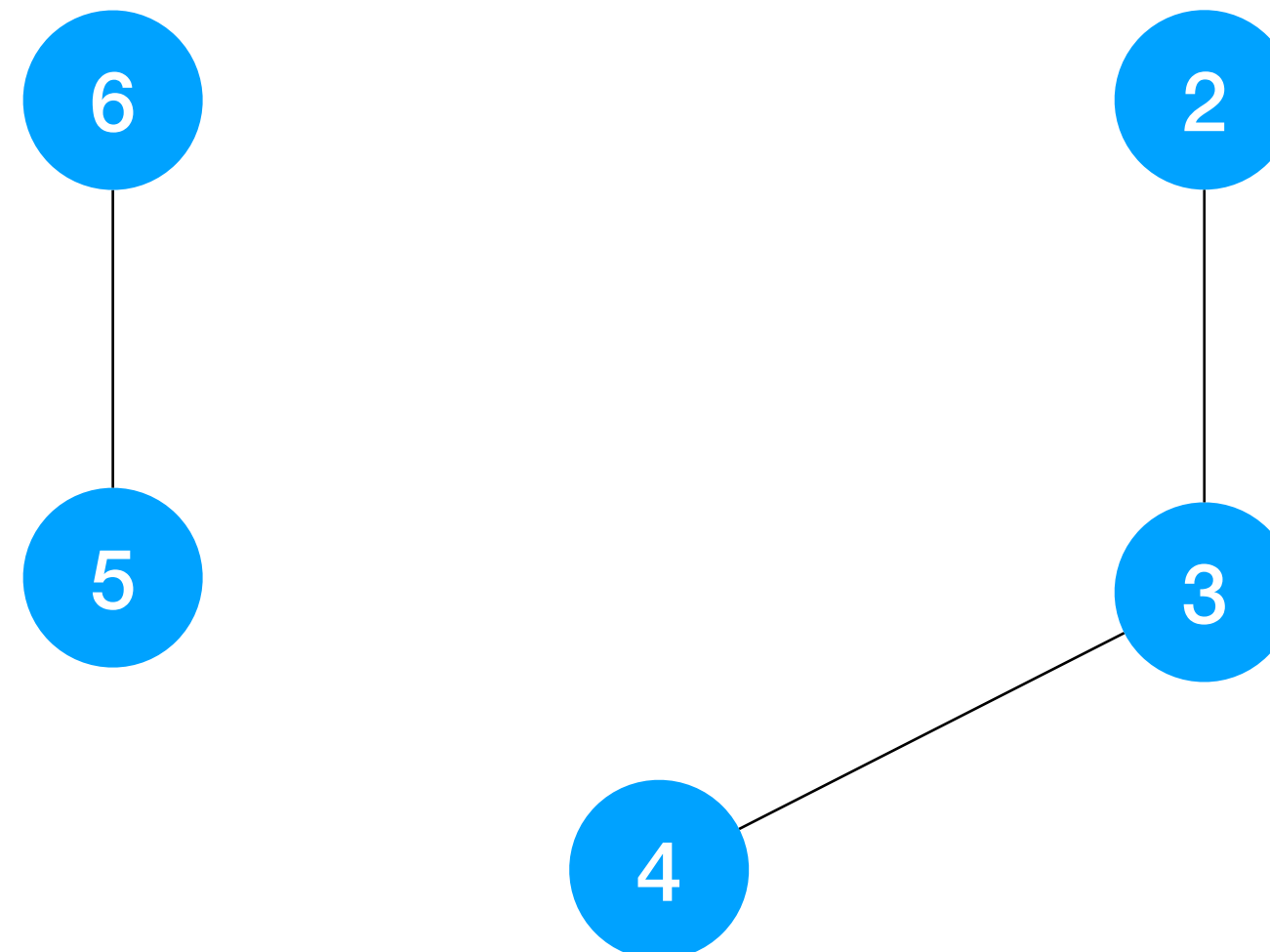


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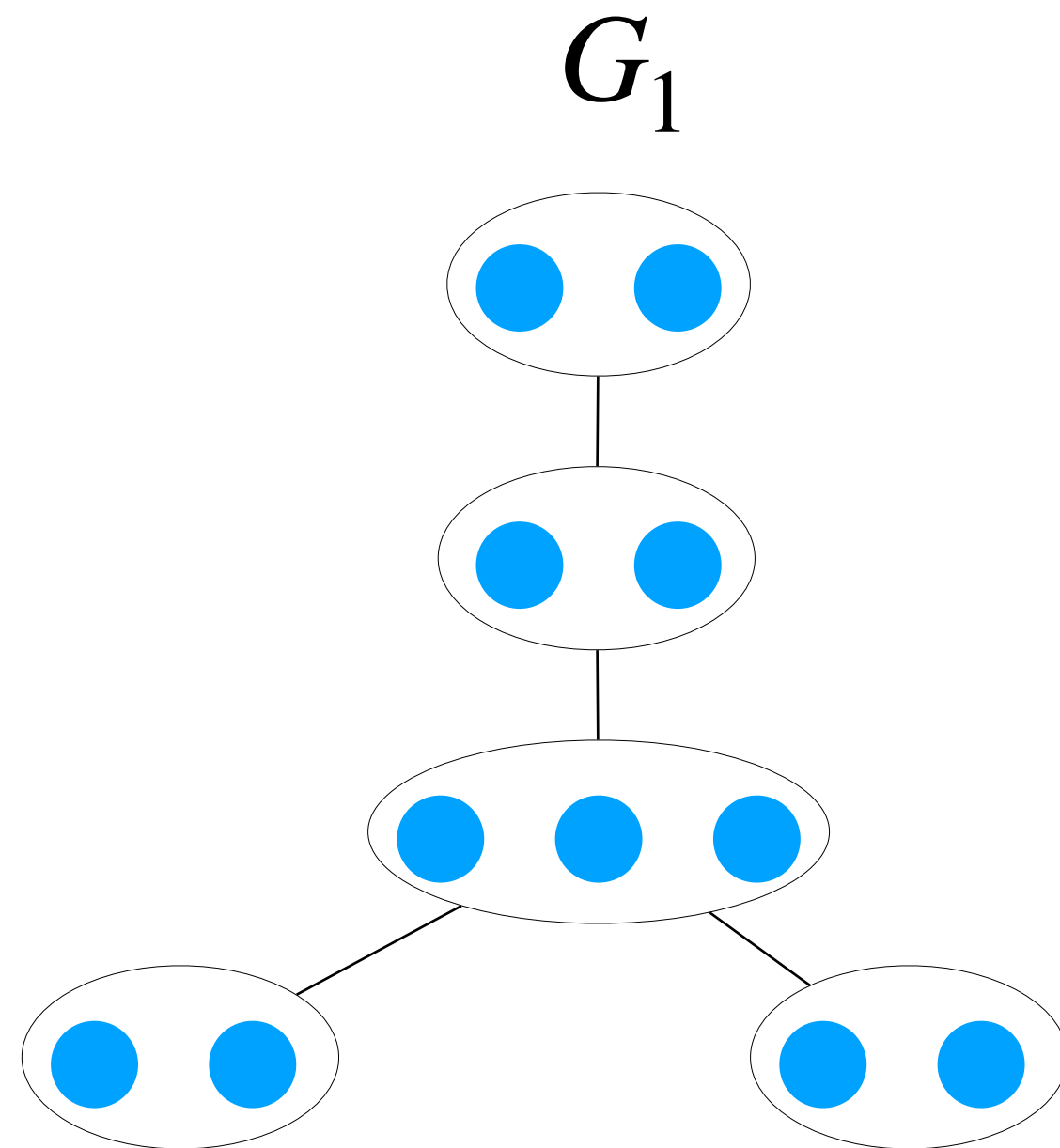
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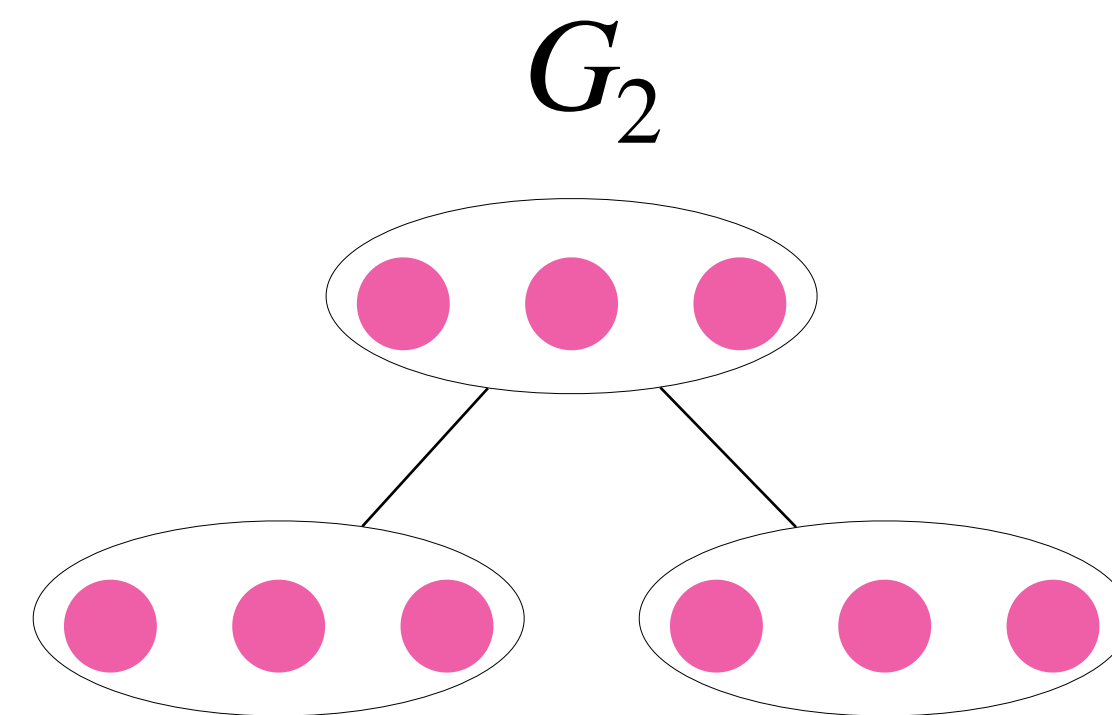
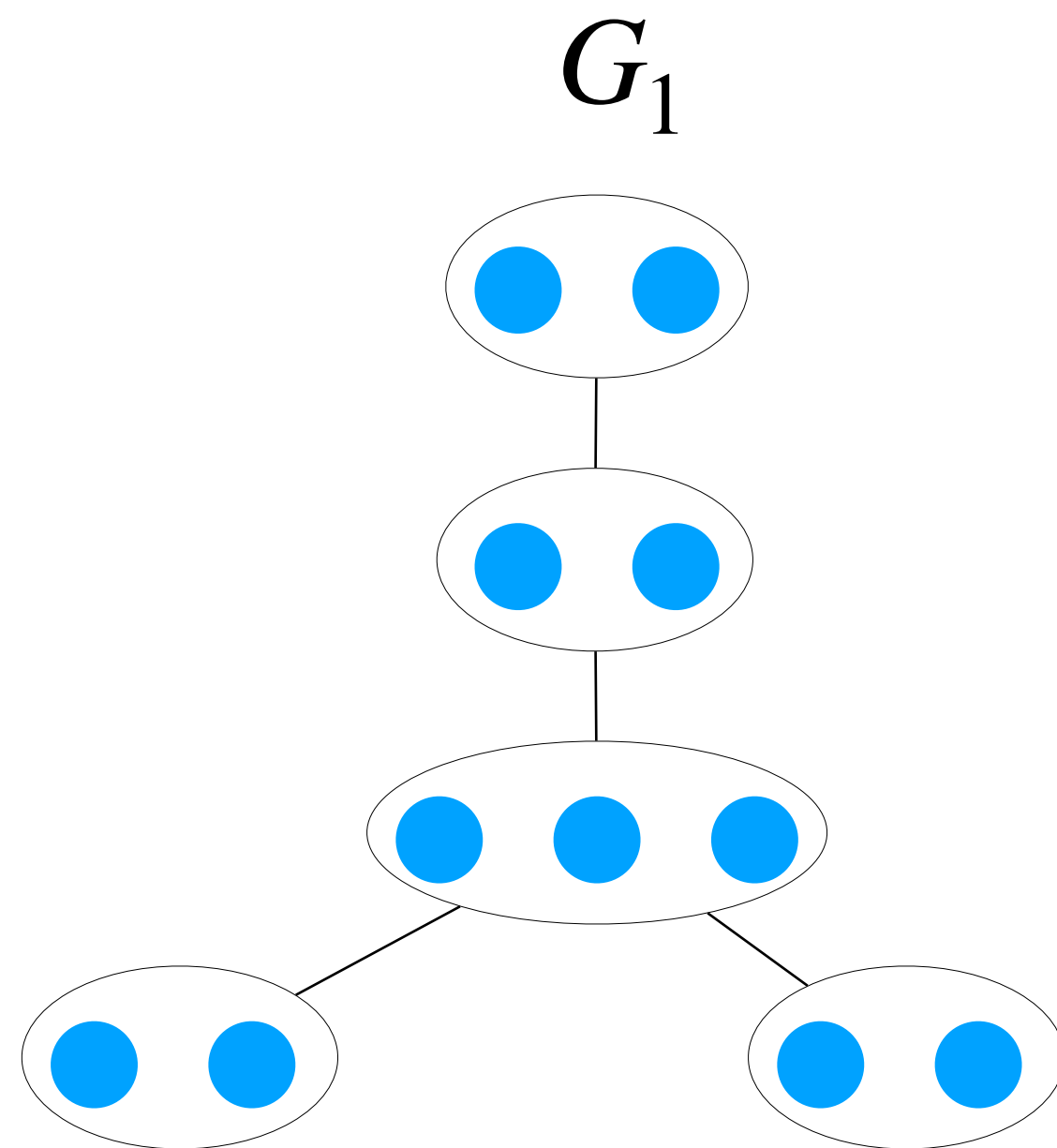
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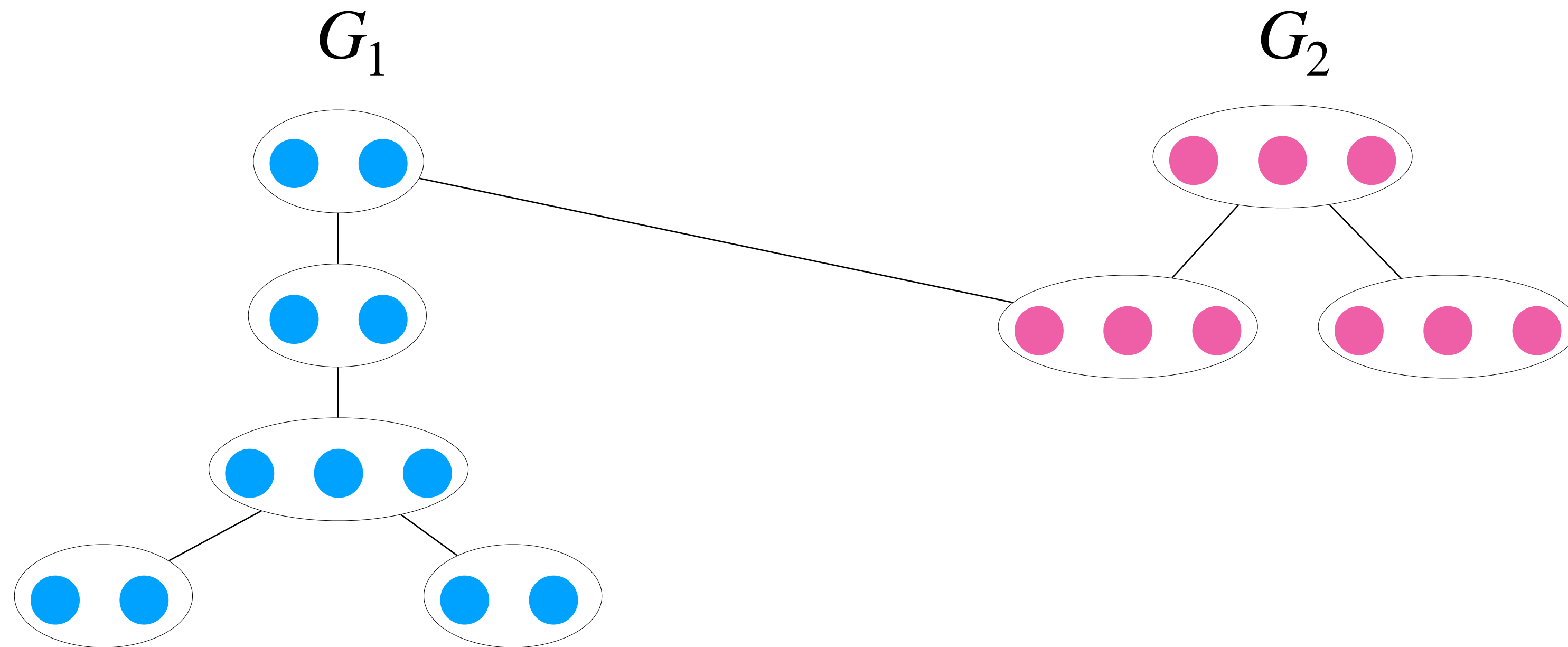
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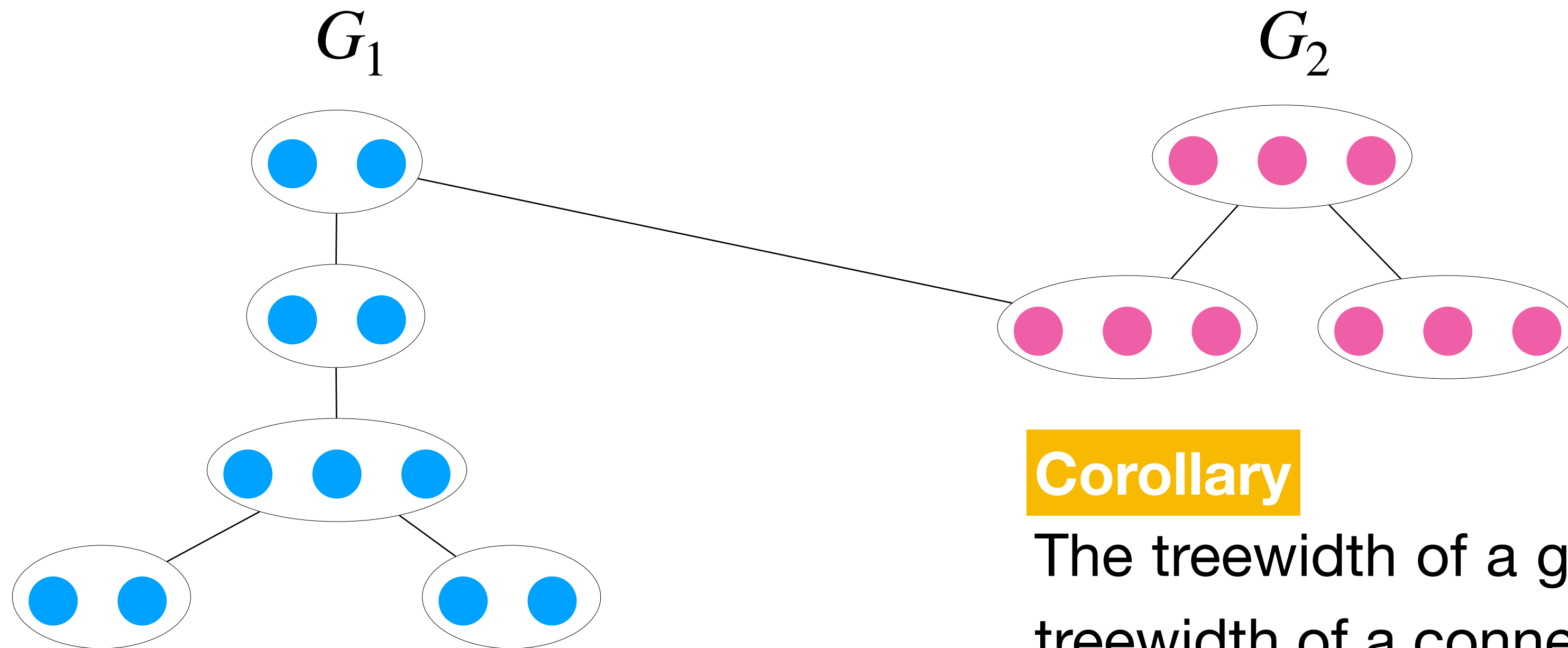




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## Corollary

The treewidth of a graph  $G$  is the maximum treewidth of a connected component of  $G$ .

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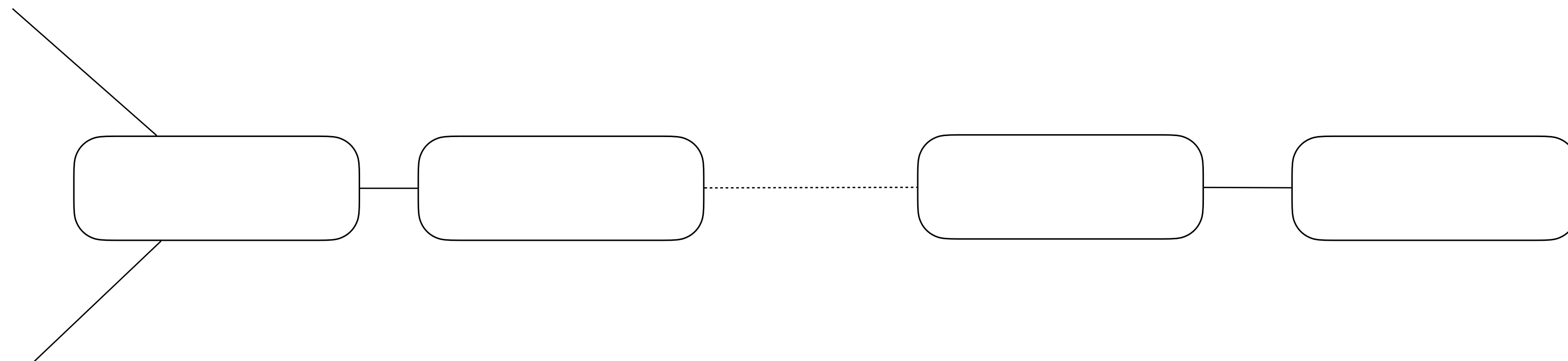
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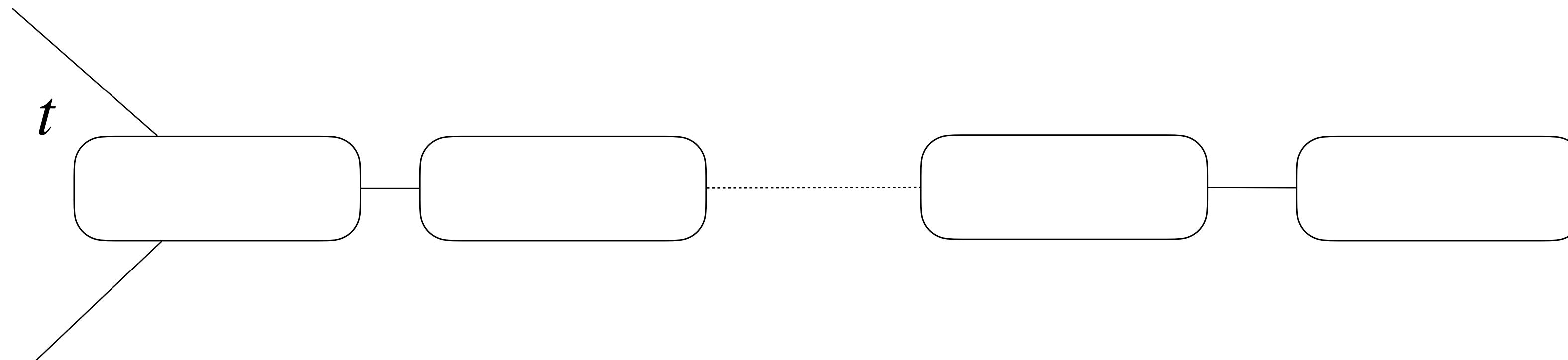
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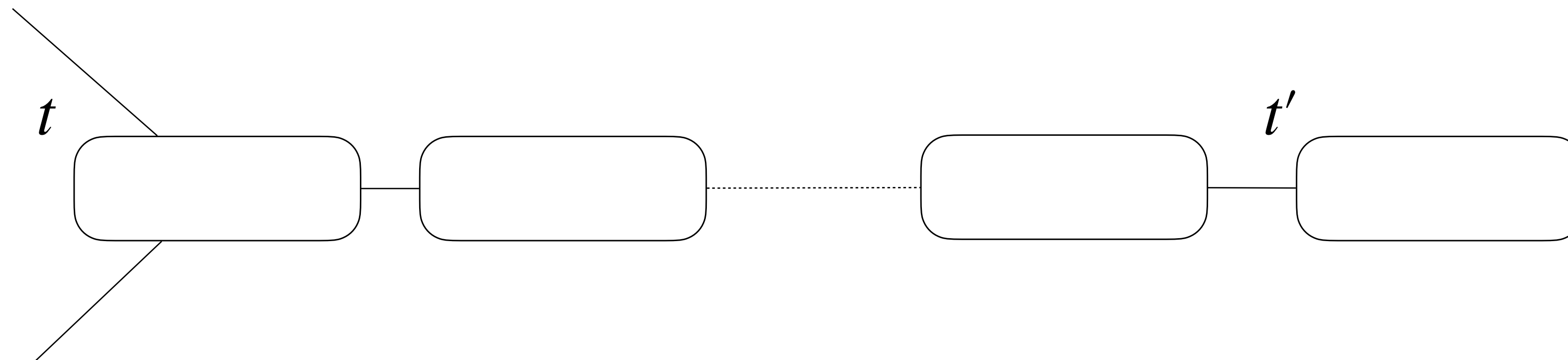
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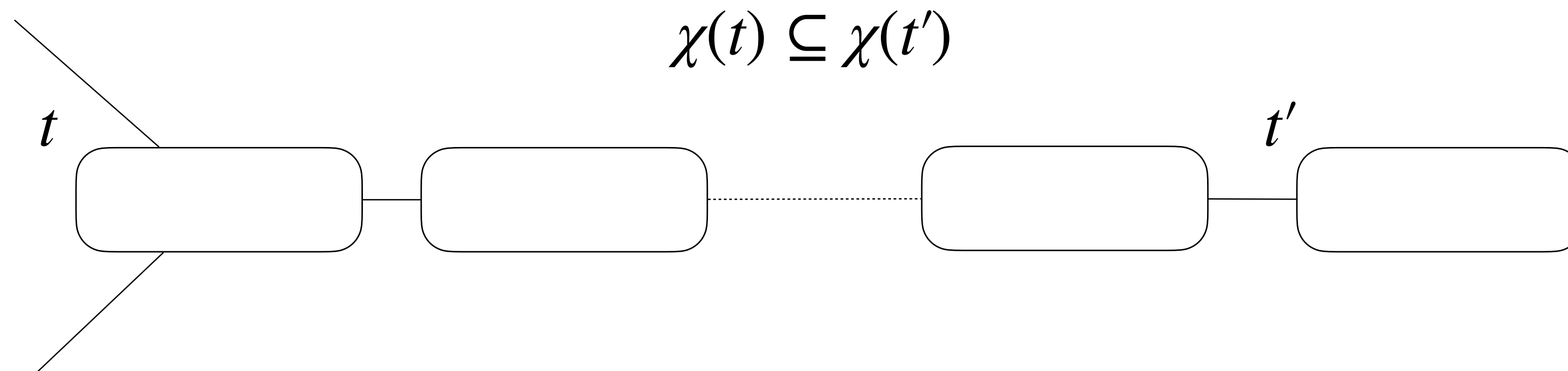
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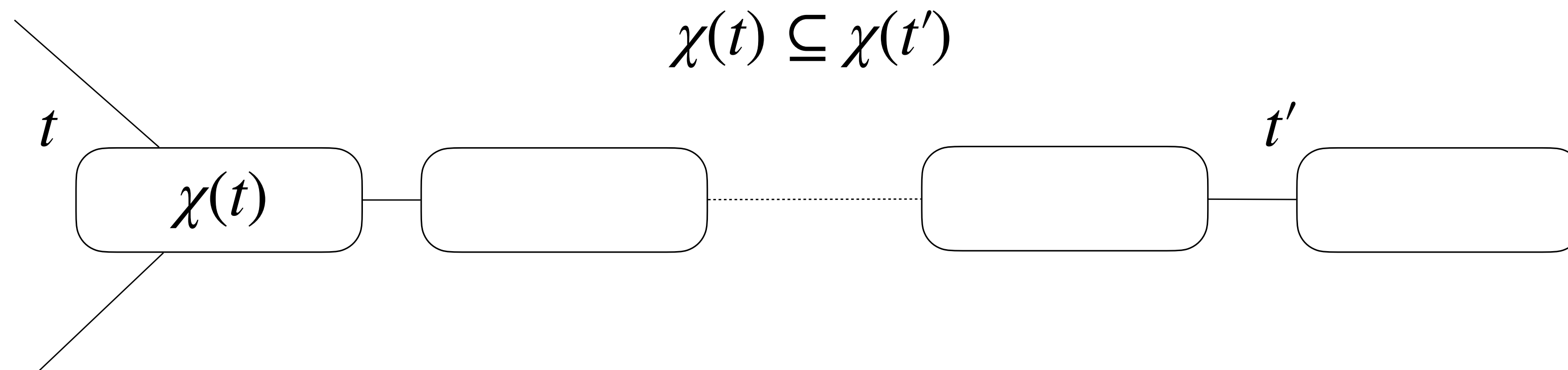
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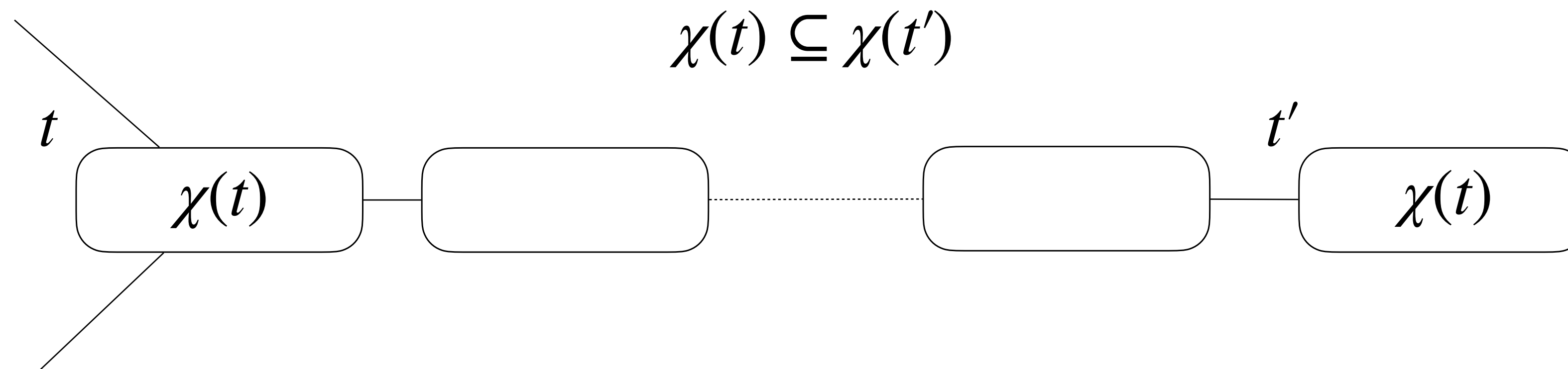
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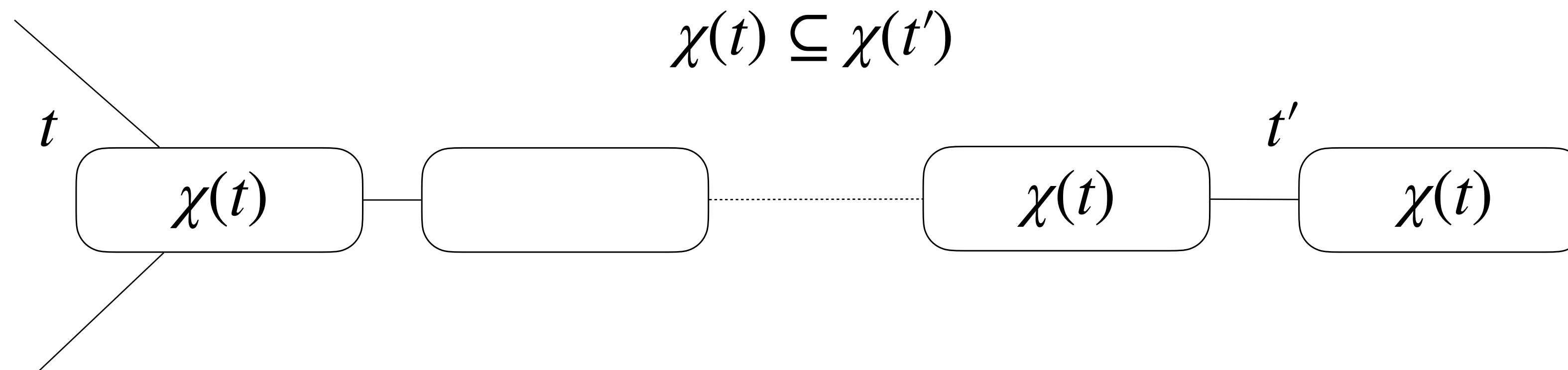
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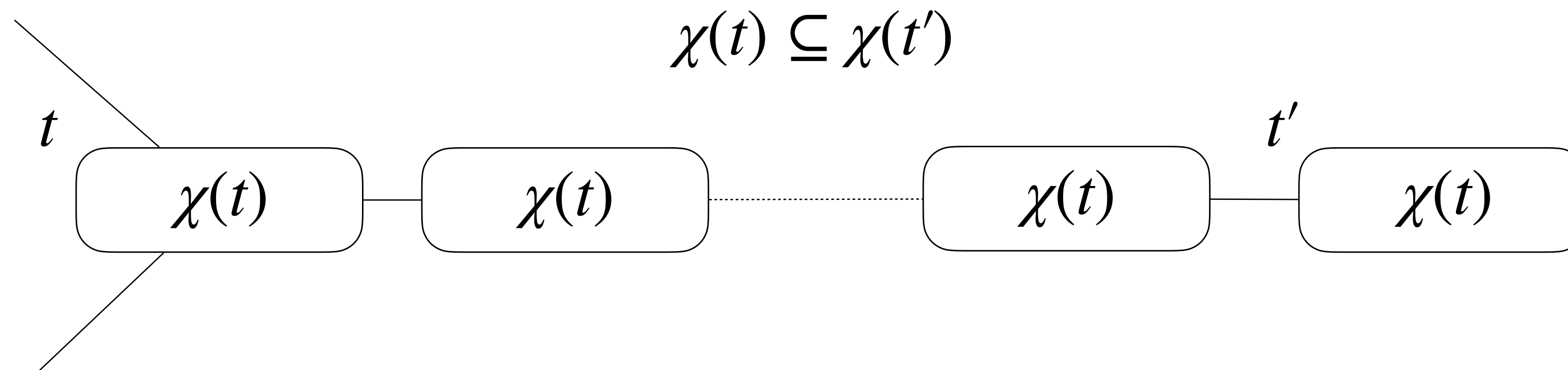
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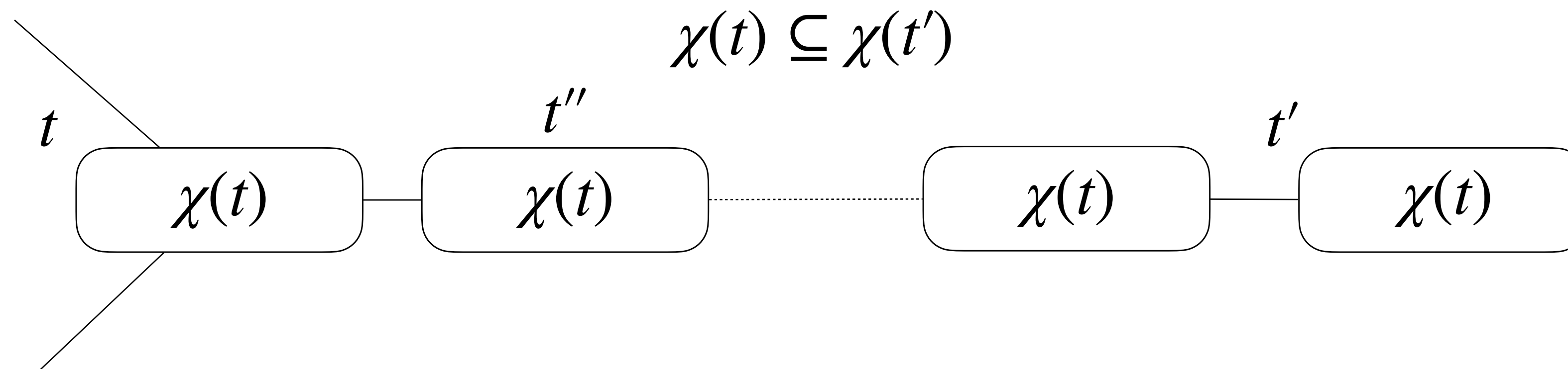
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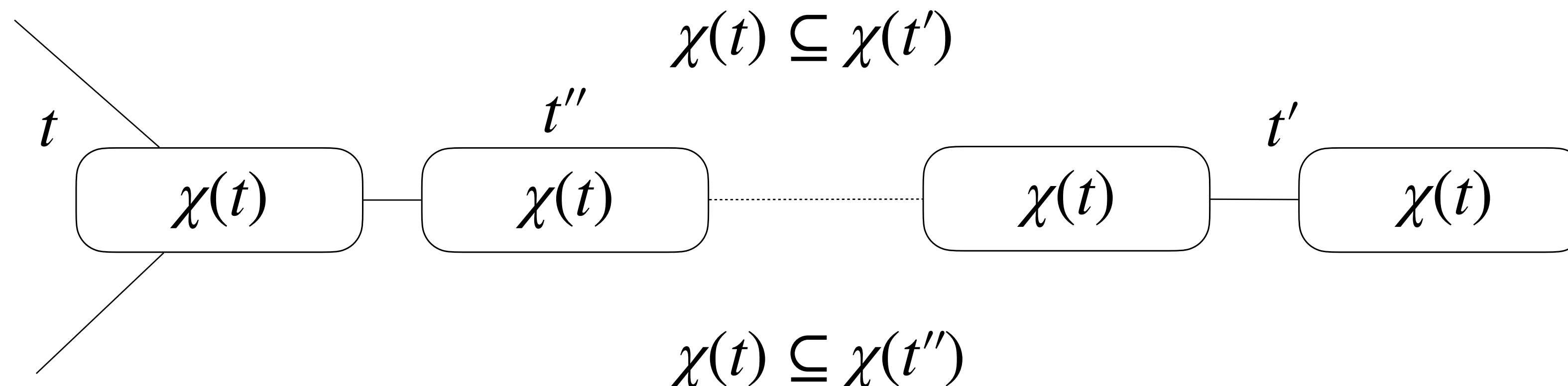
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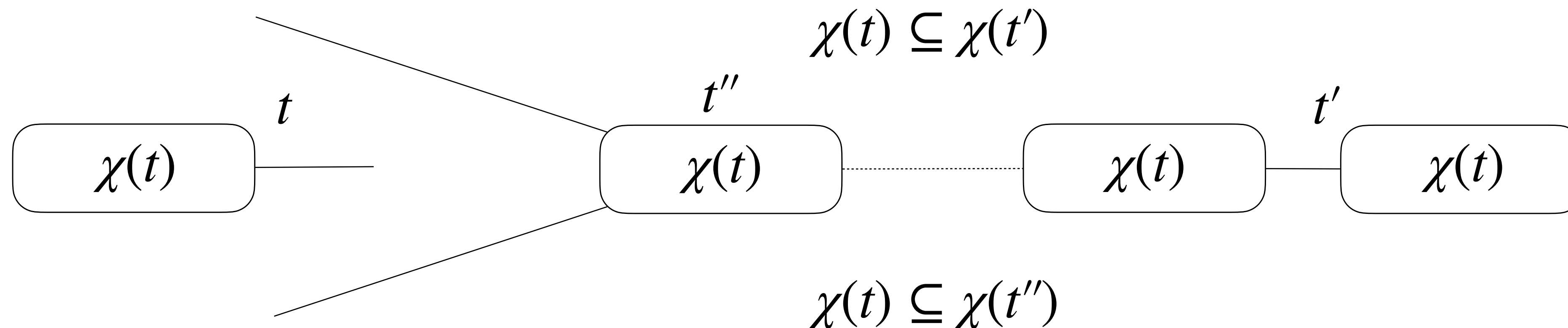
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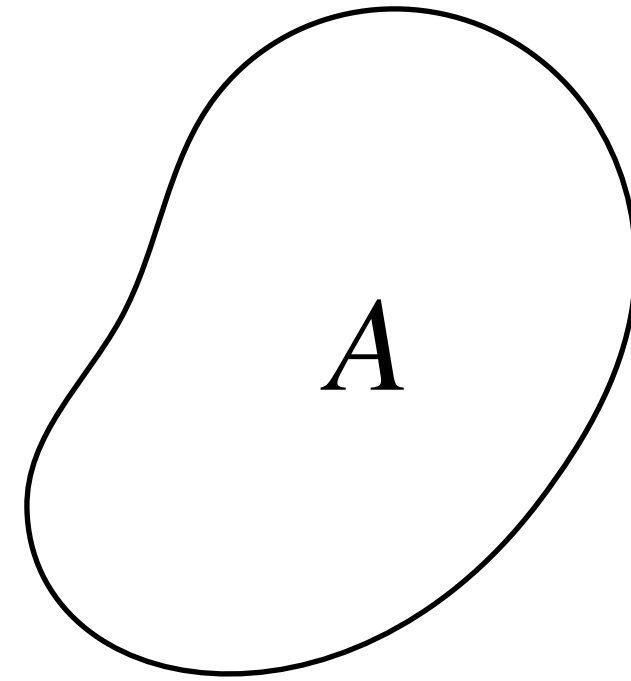
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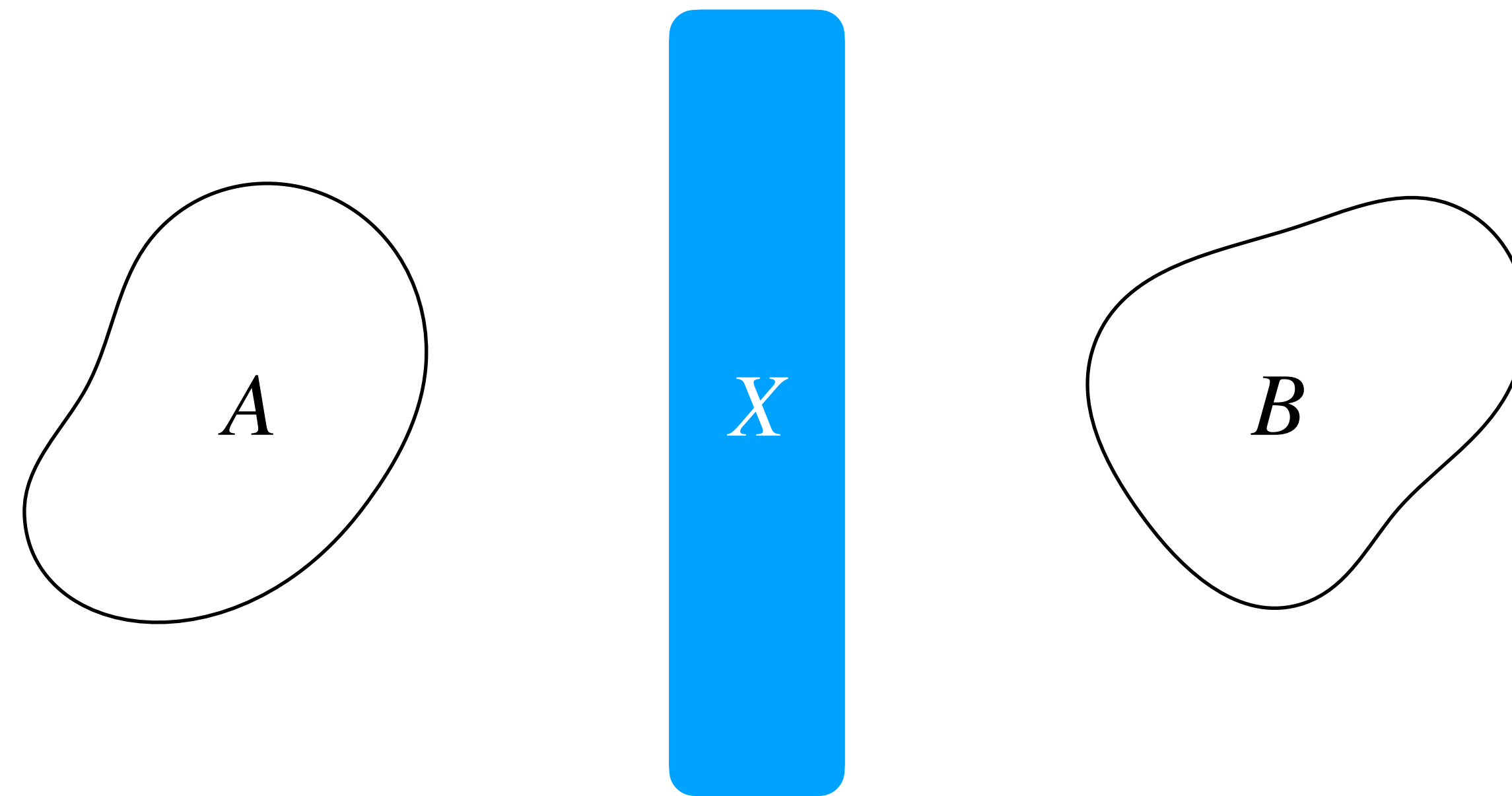
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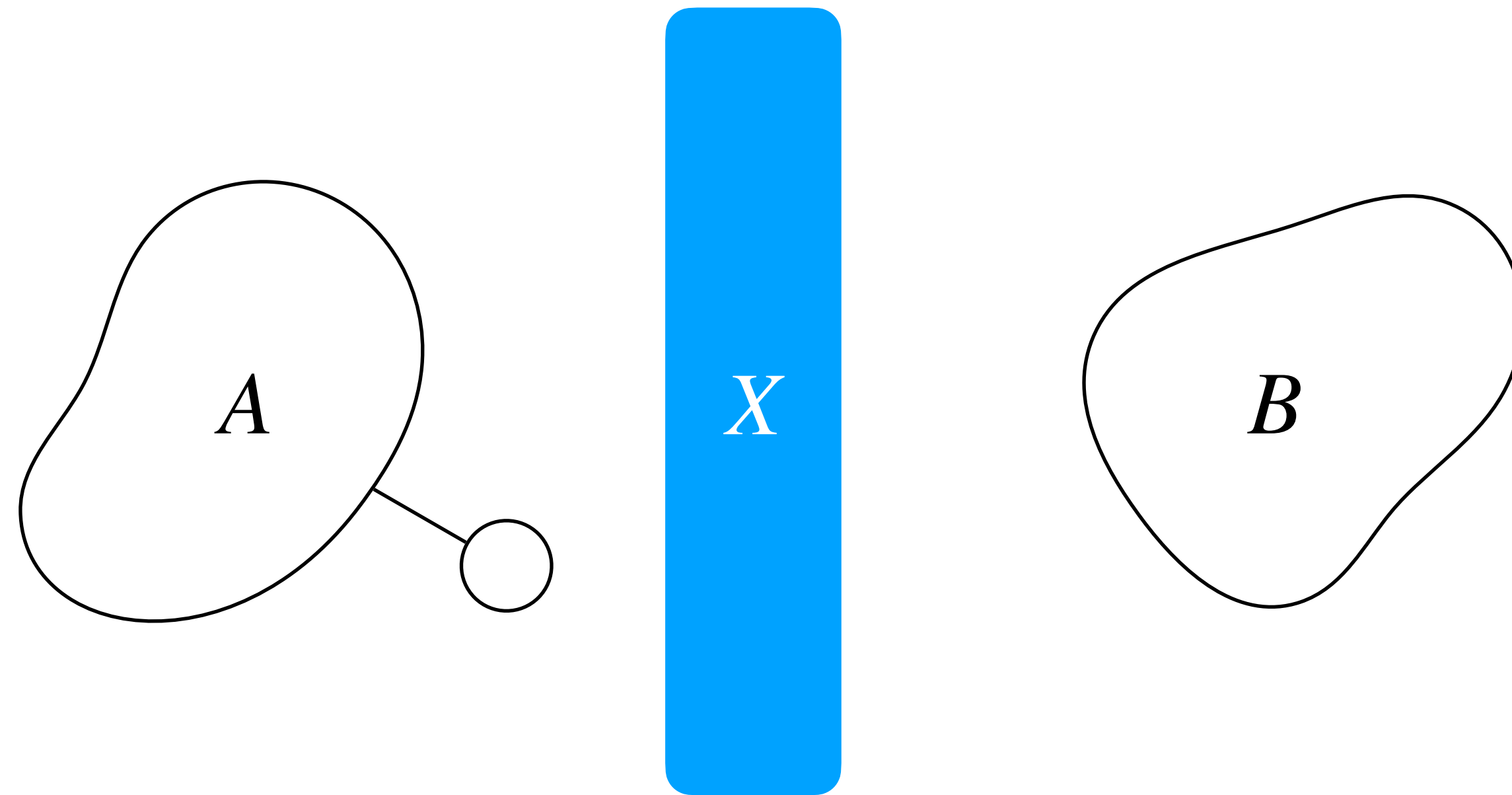
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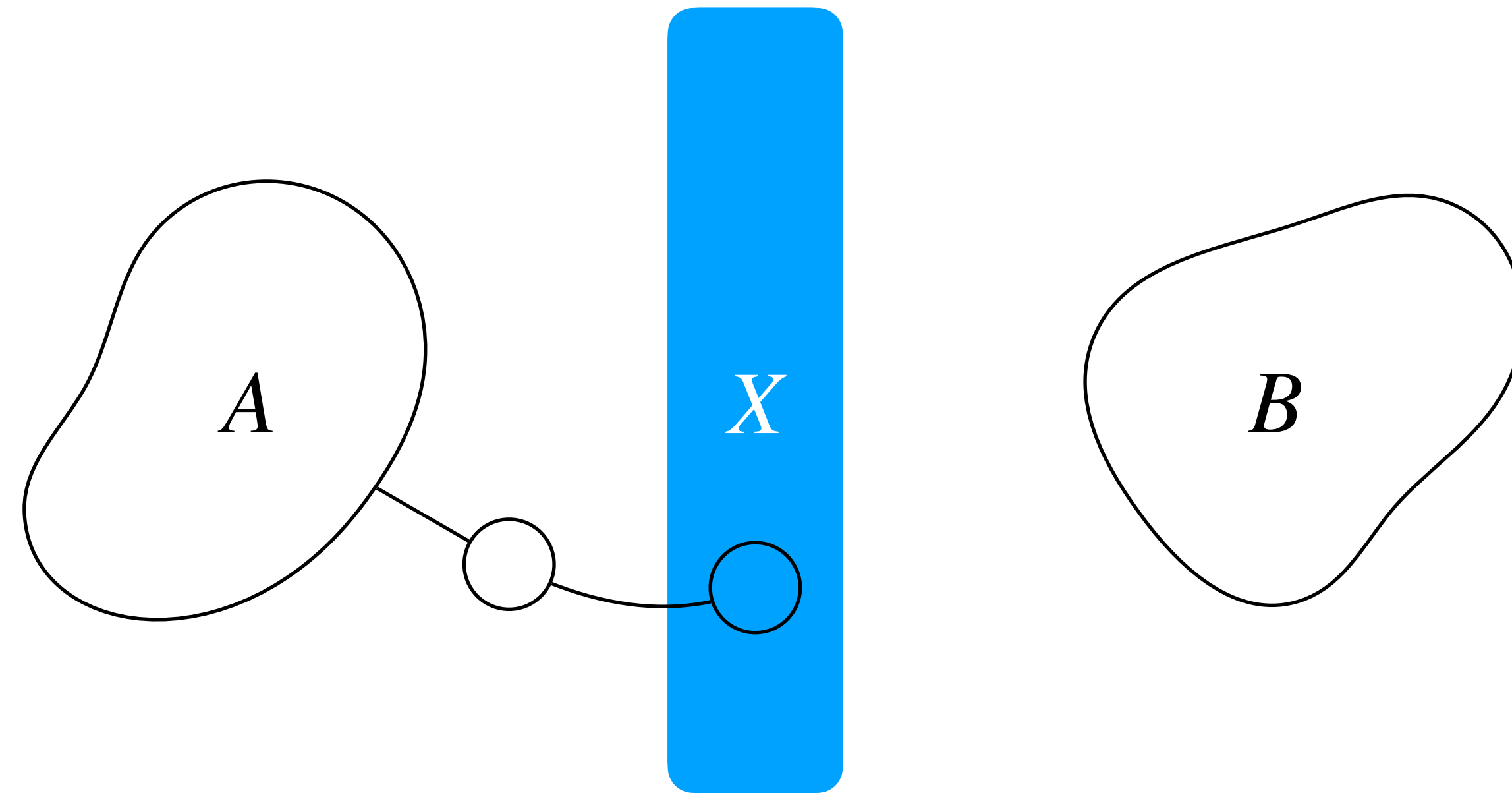
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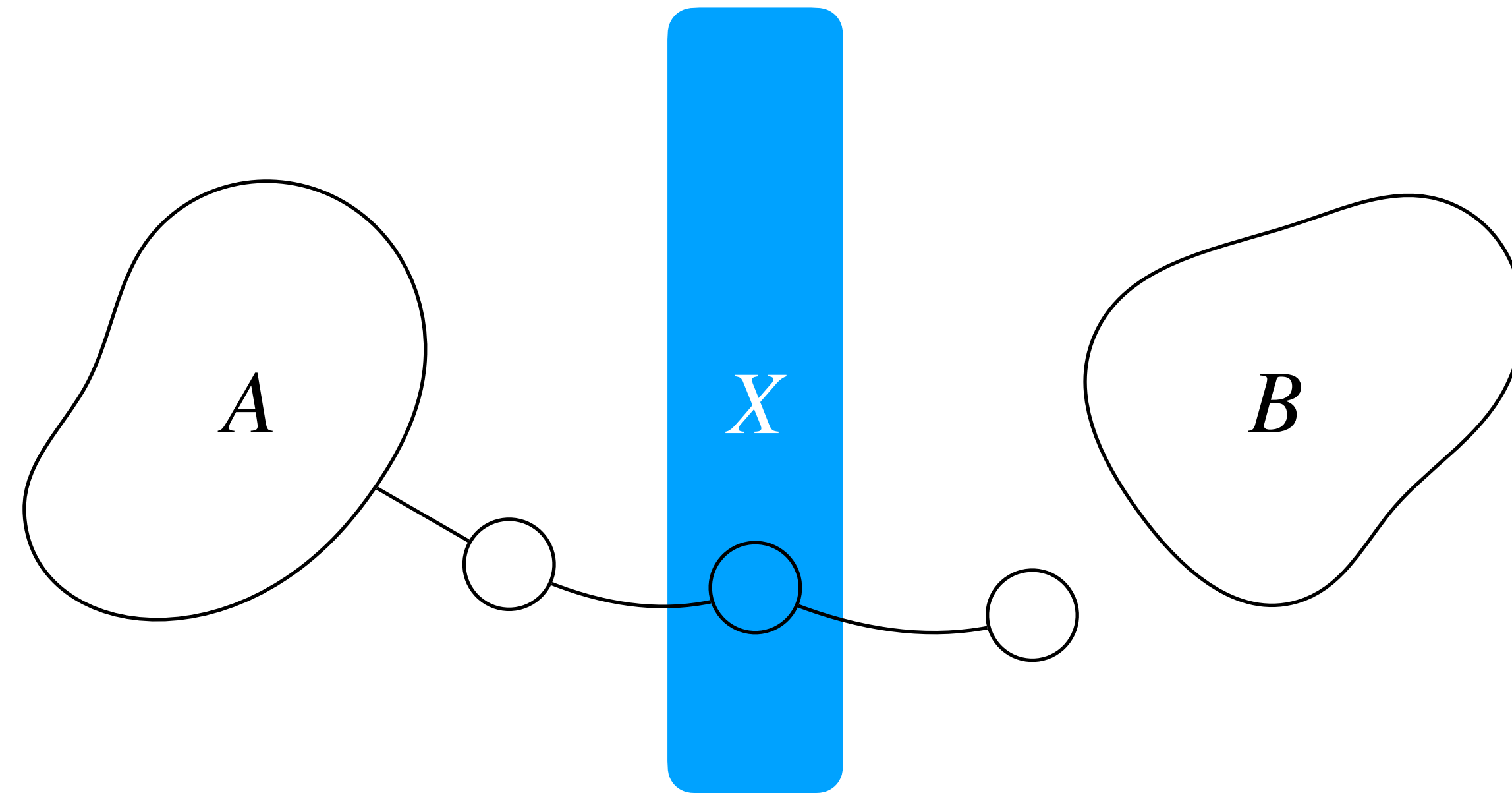
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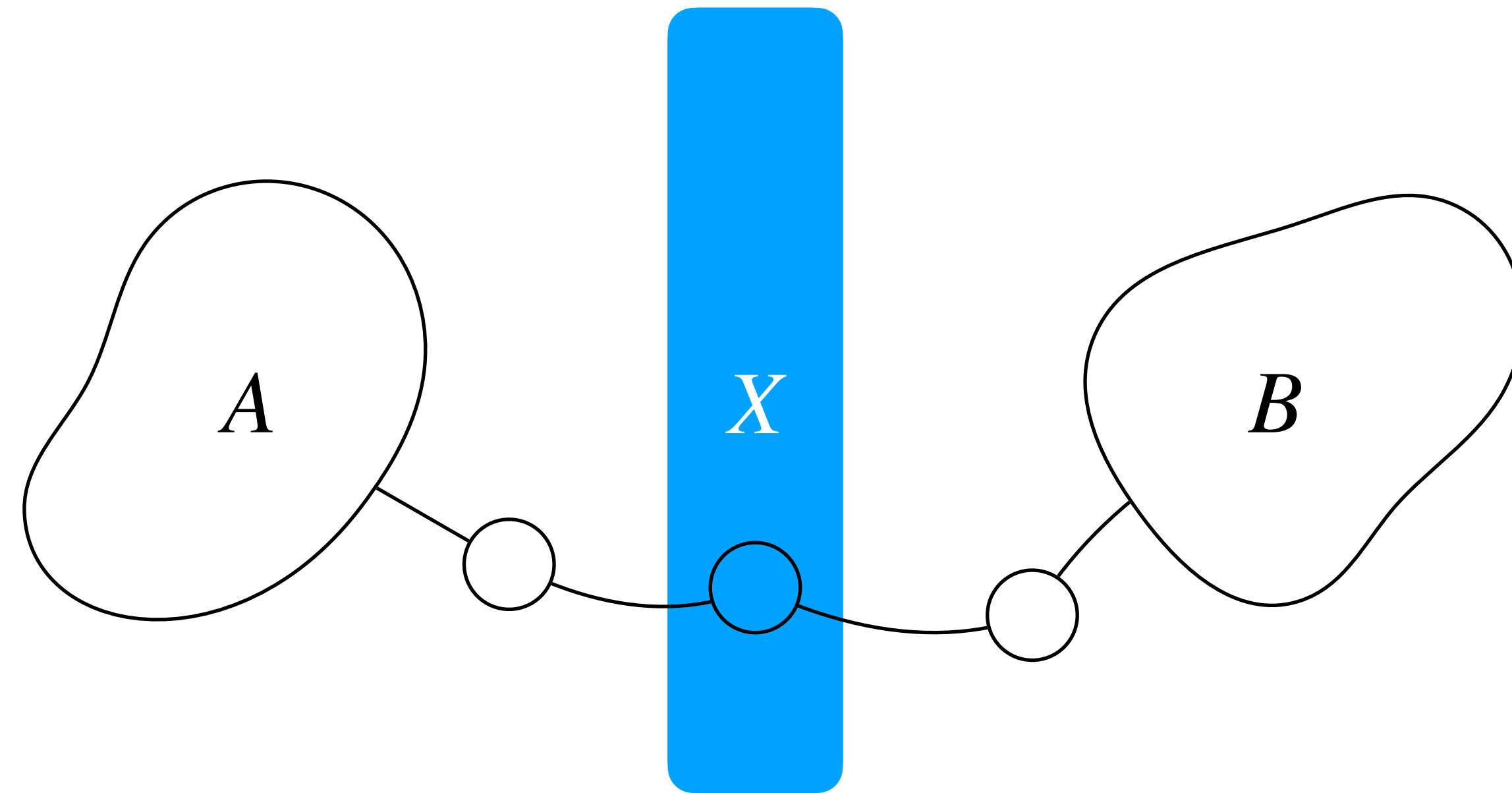
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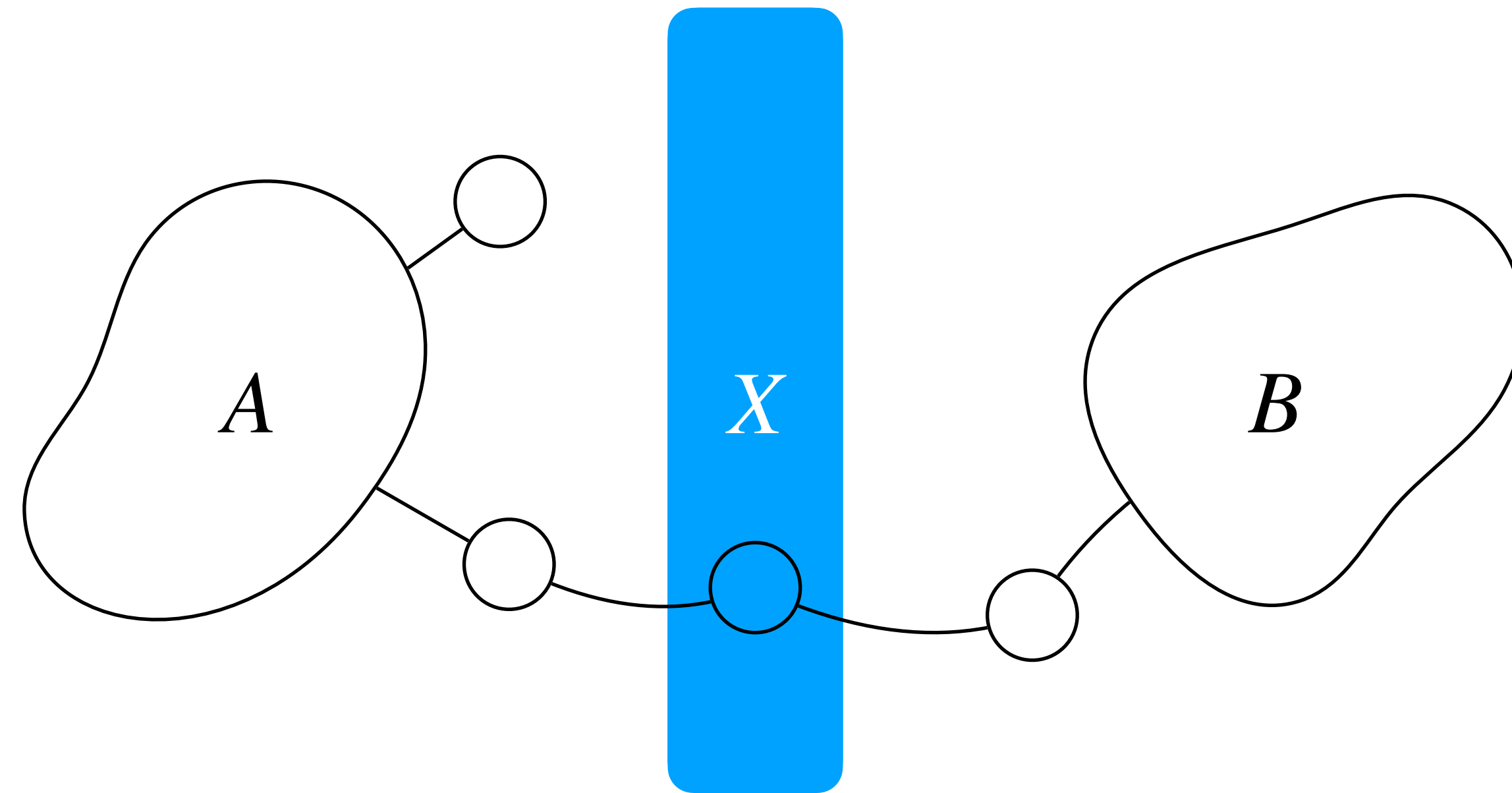
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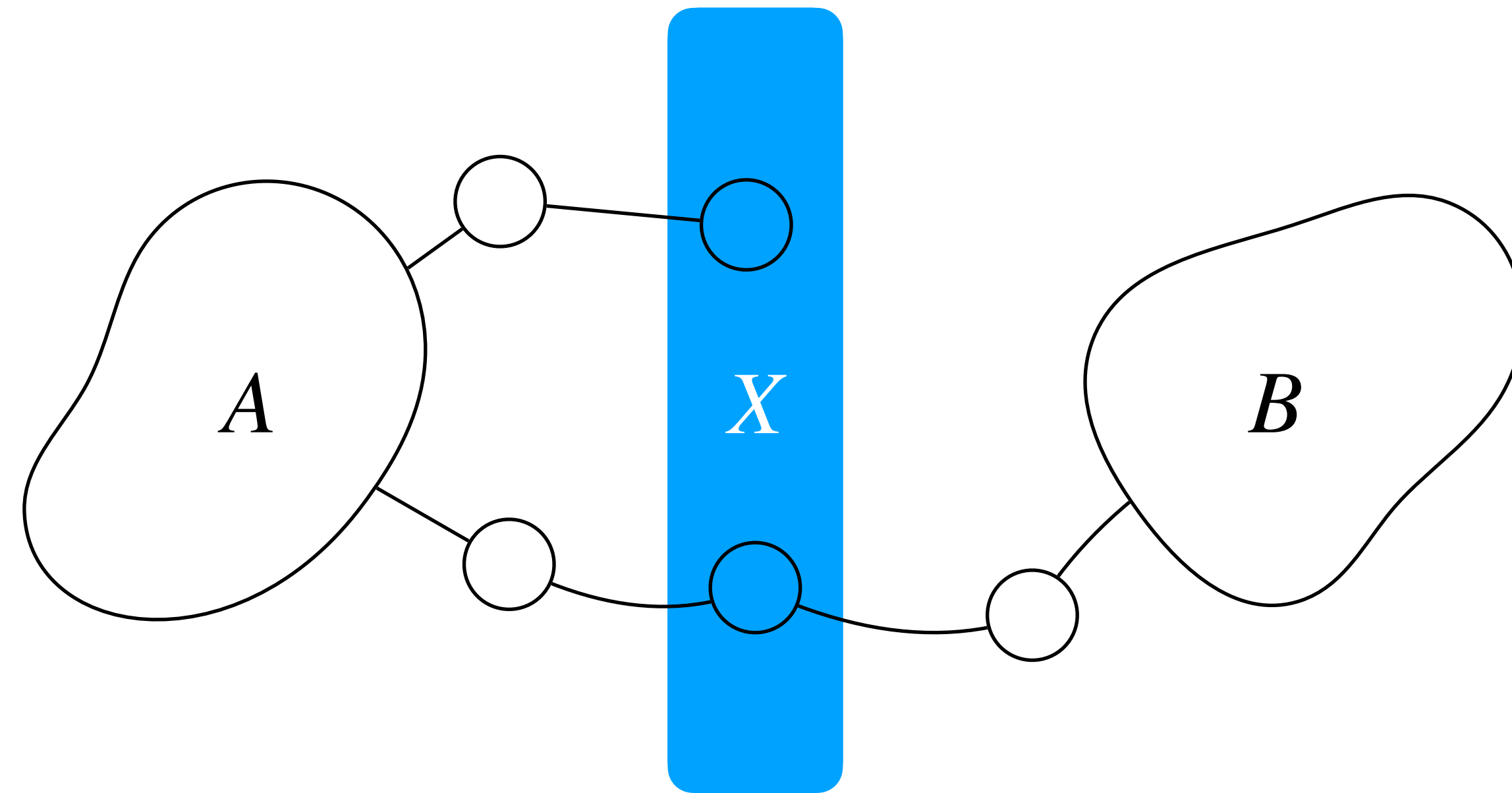
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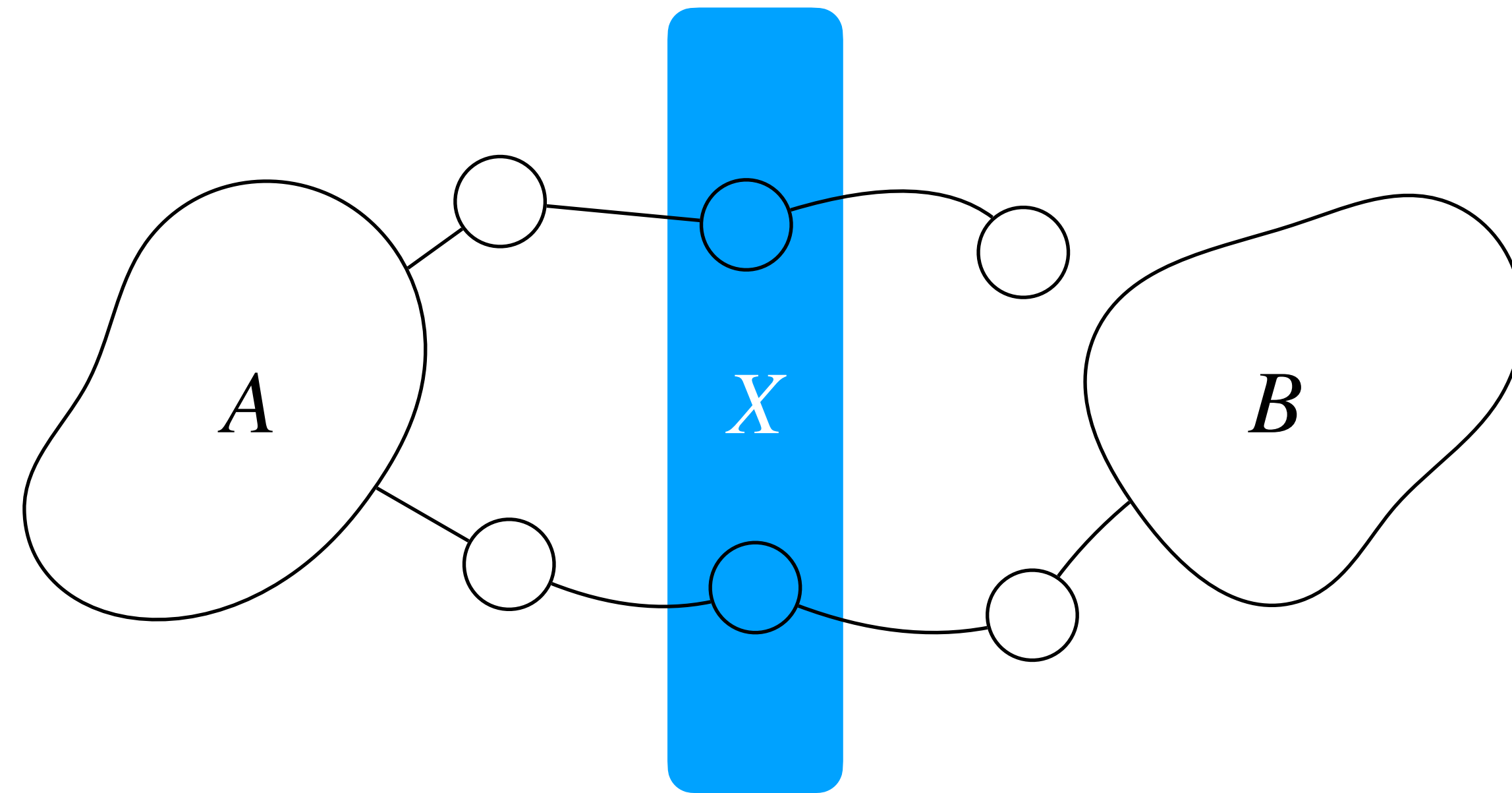
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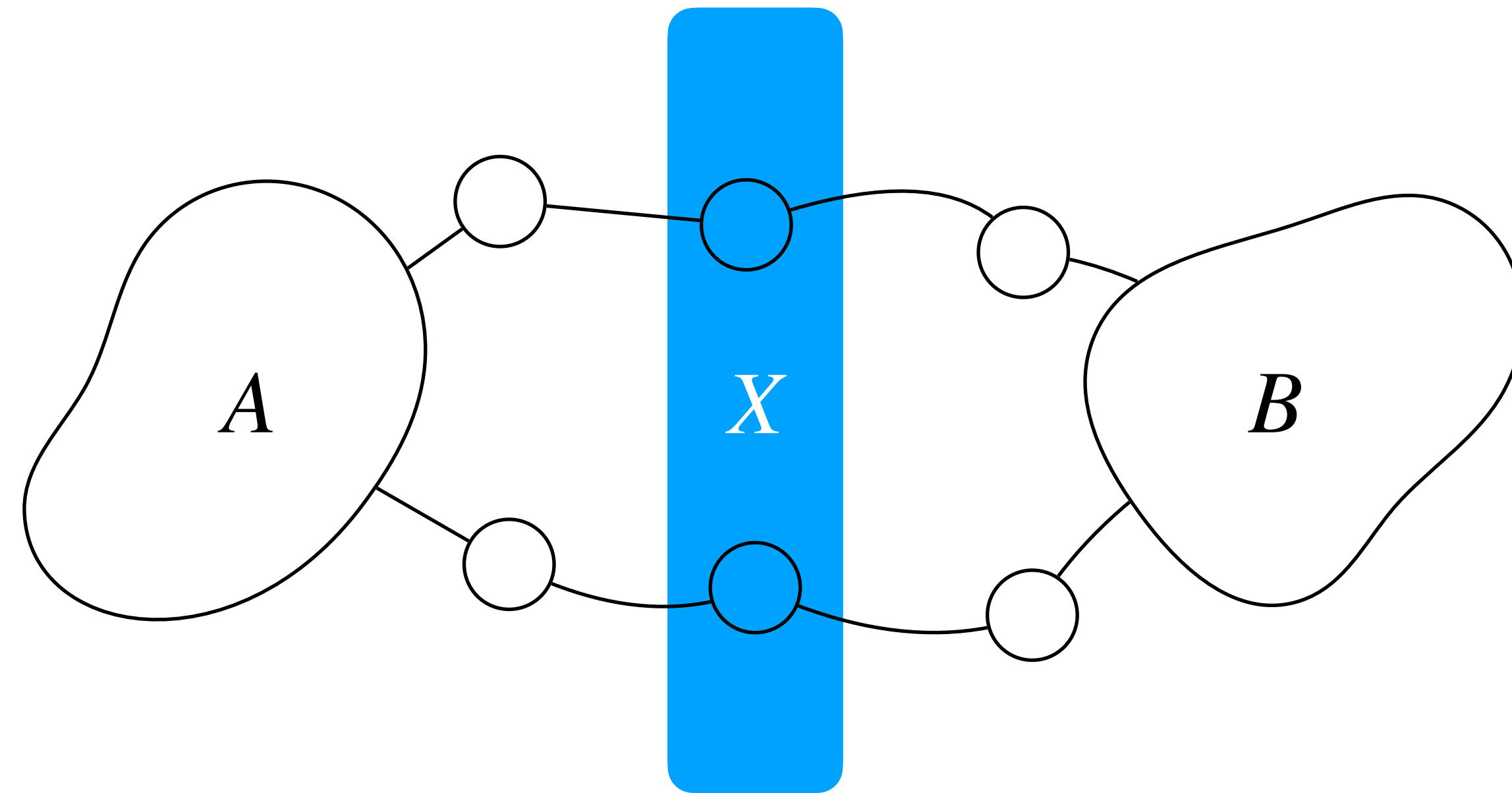
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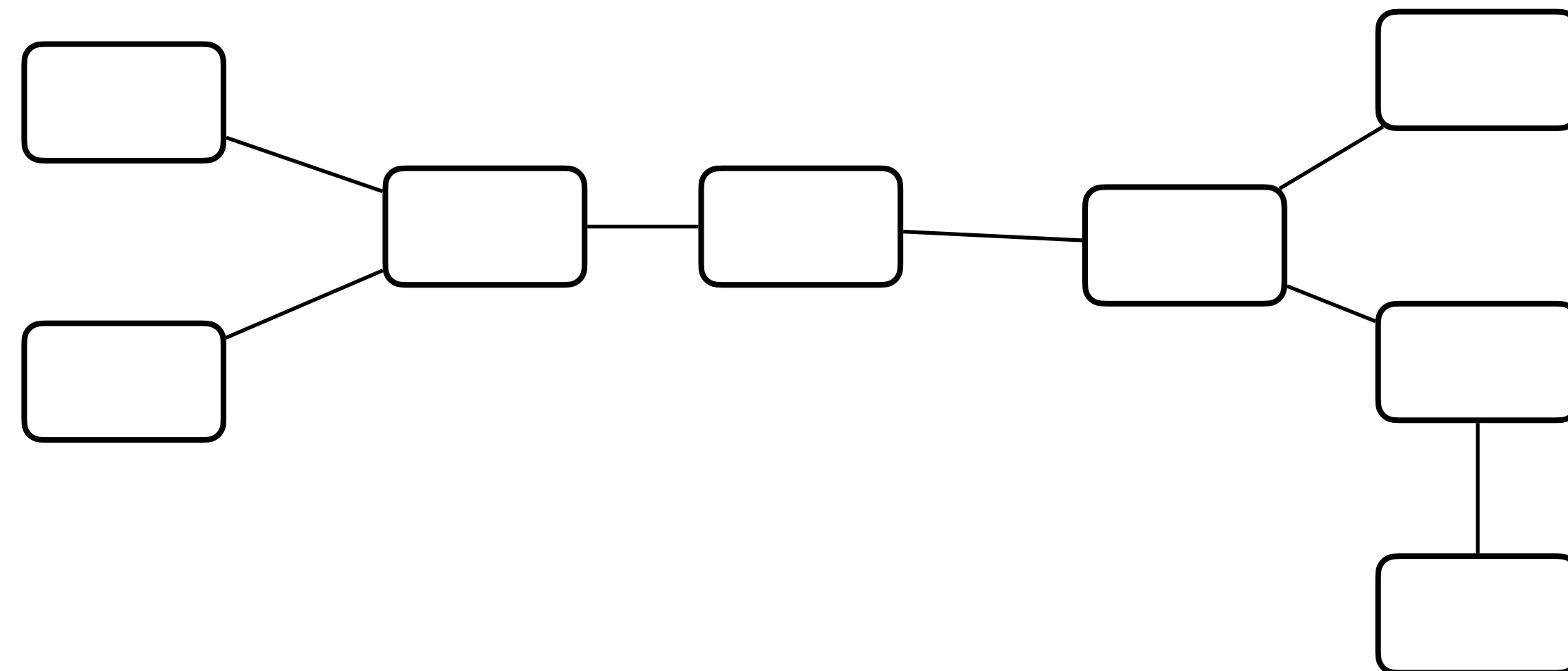
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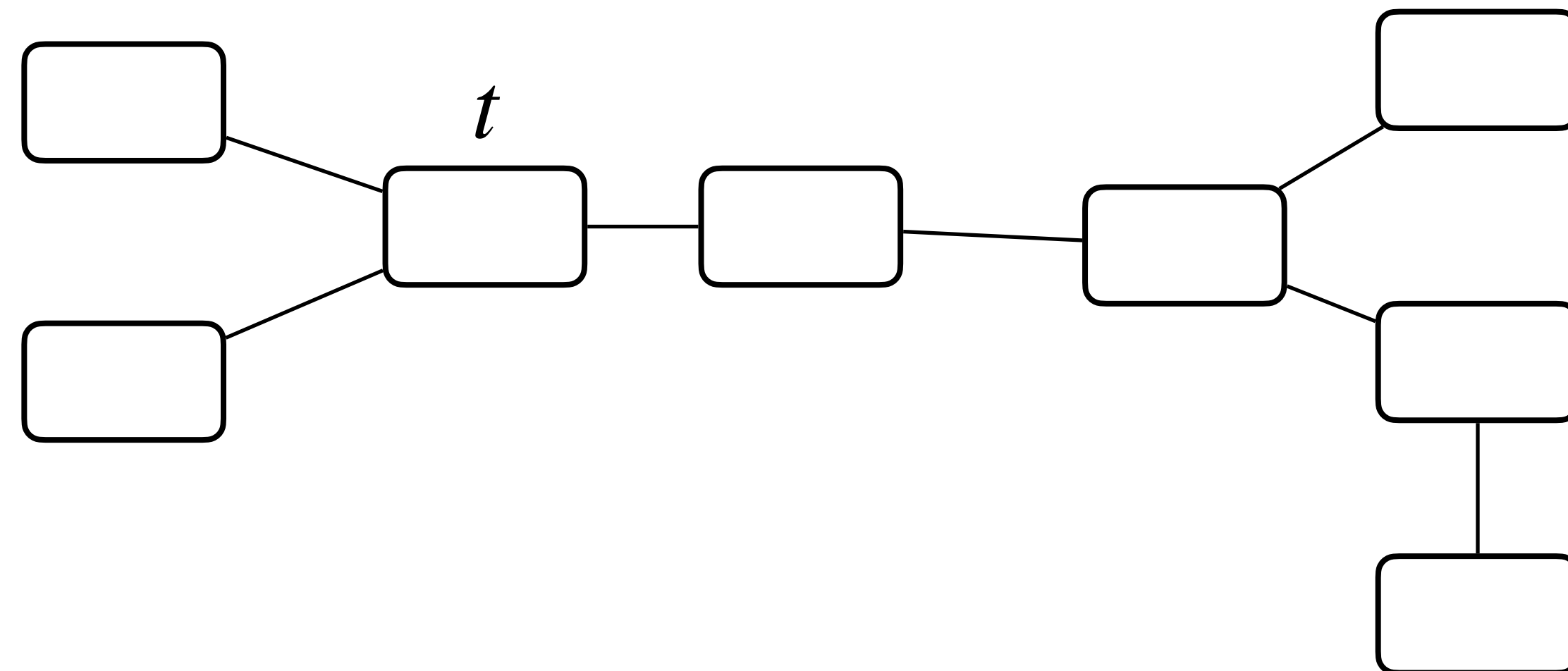


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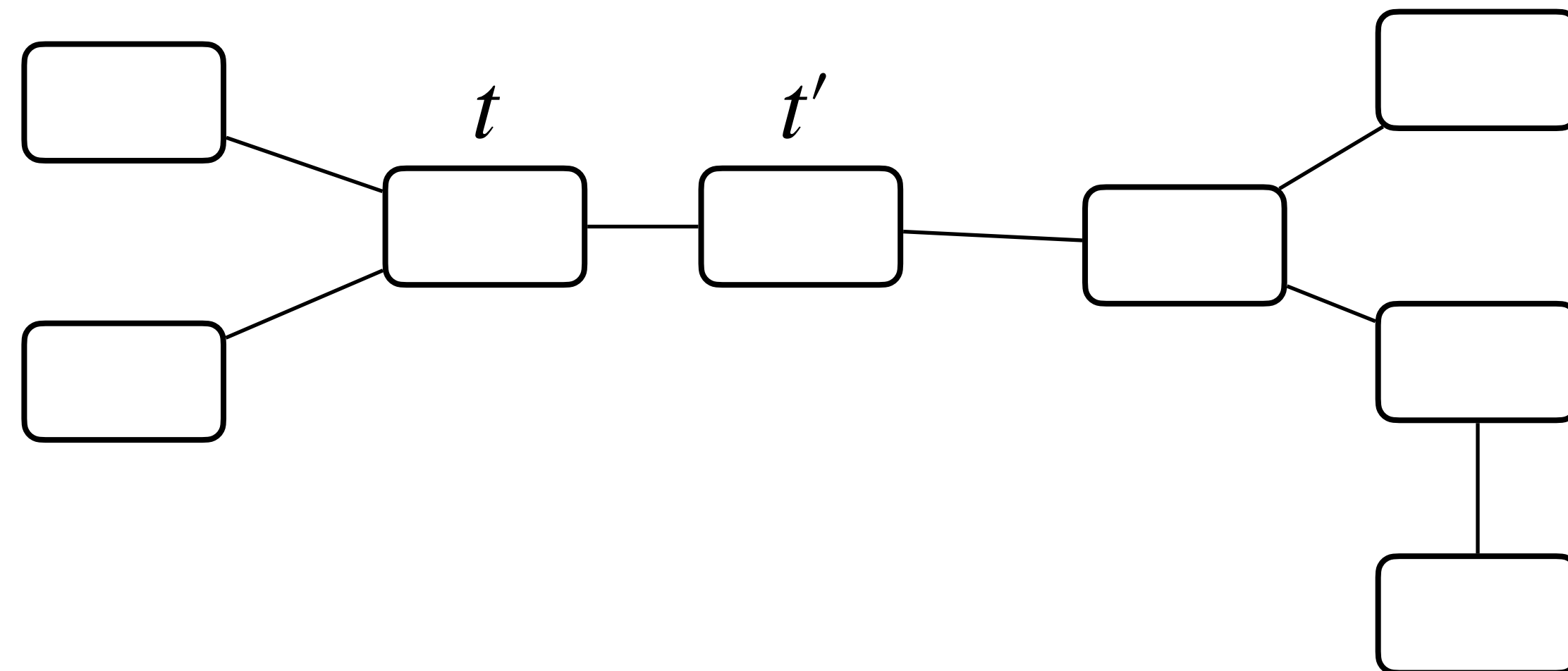


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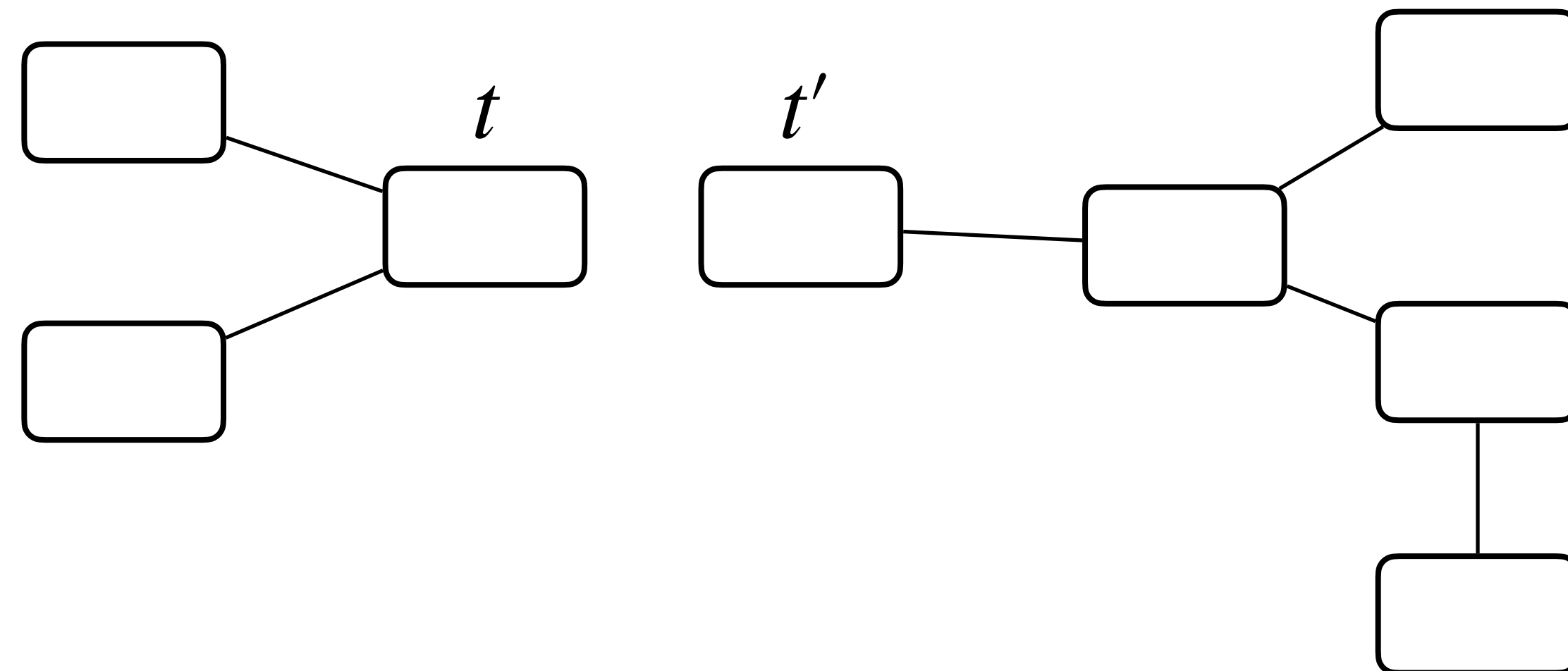


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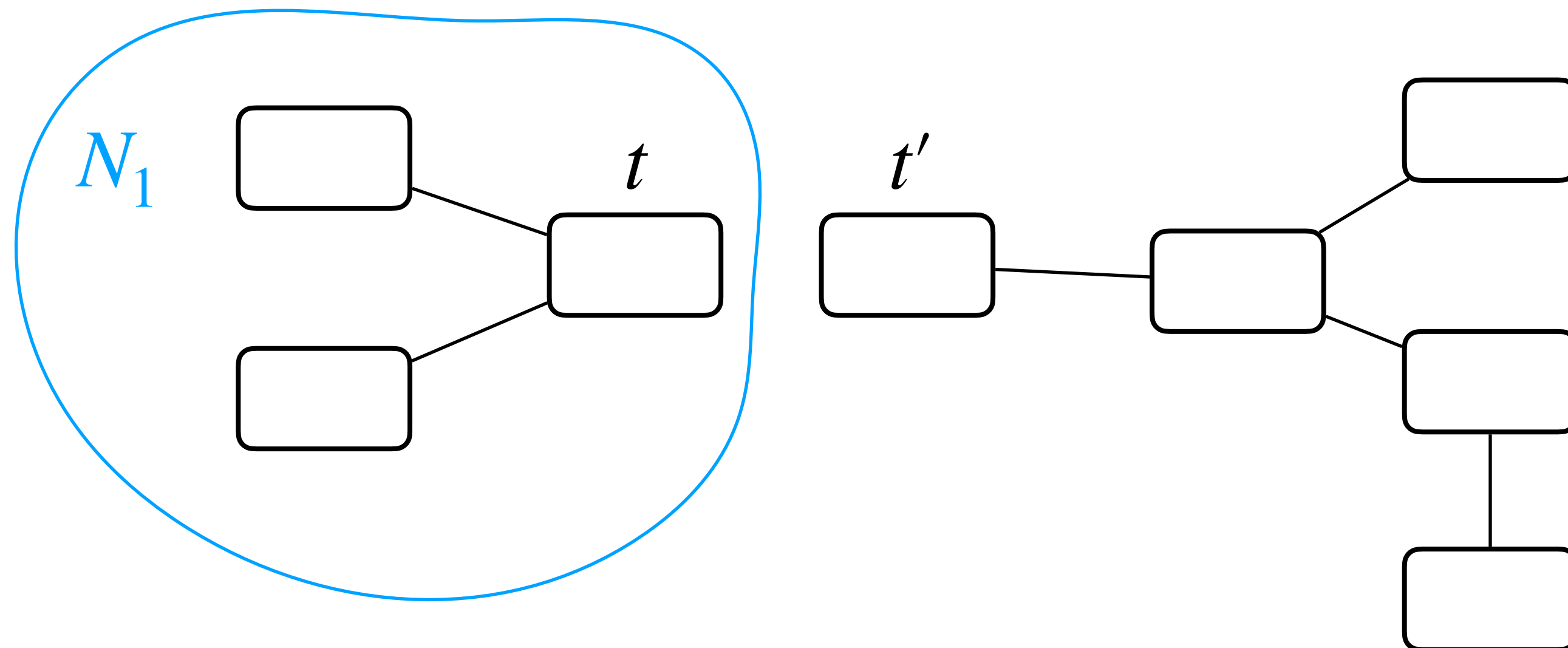


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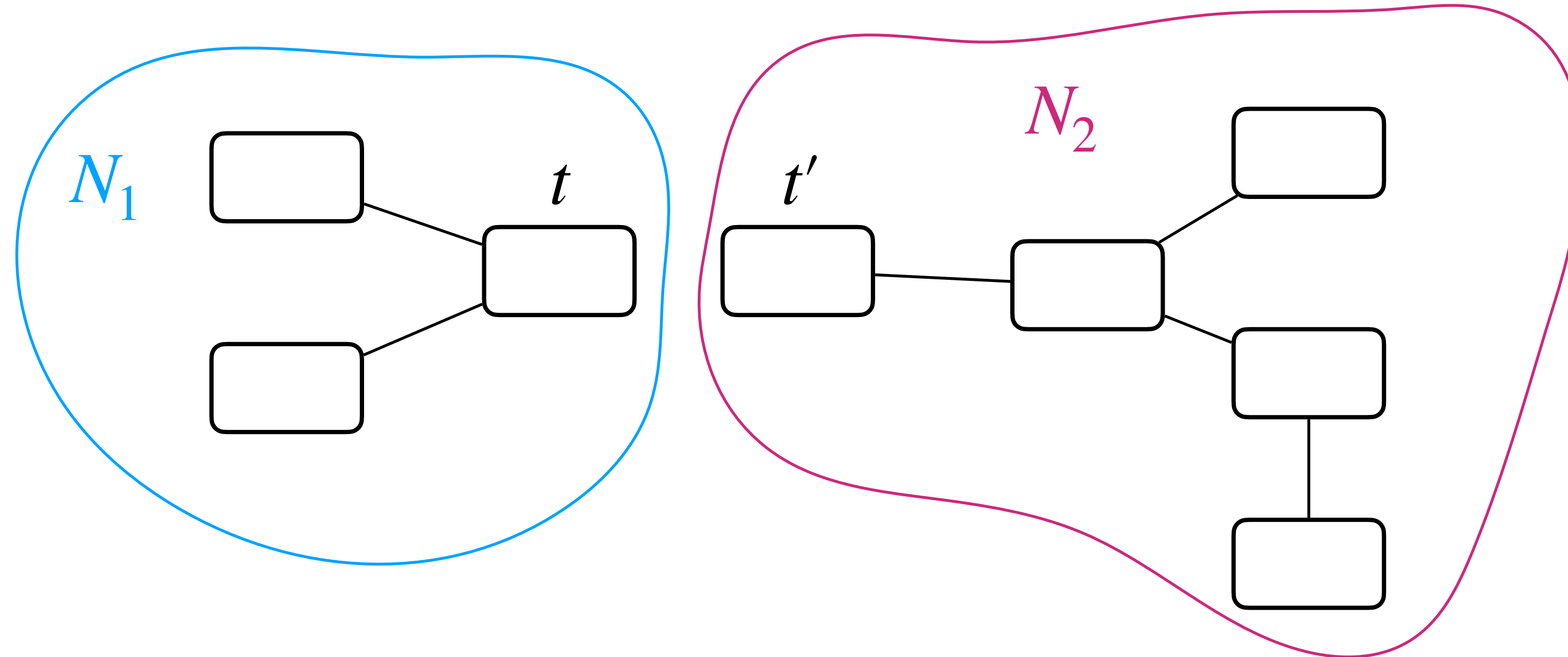


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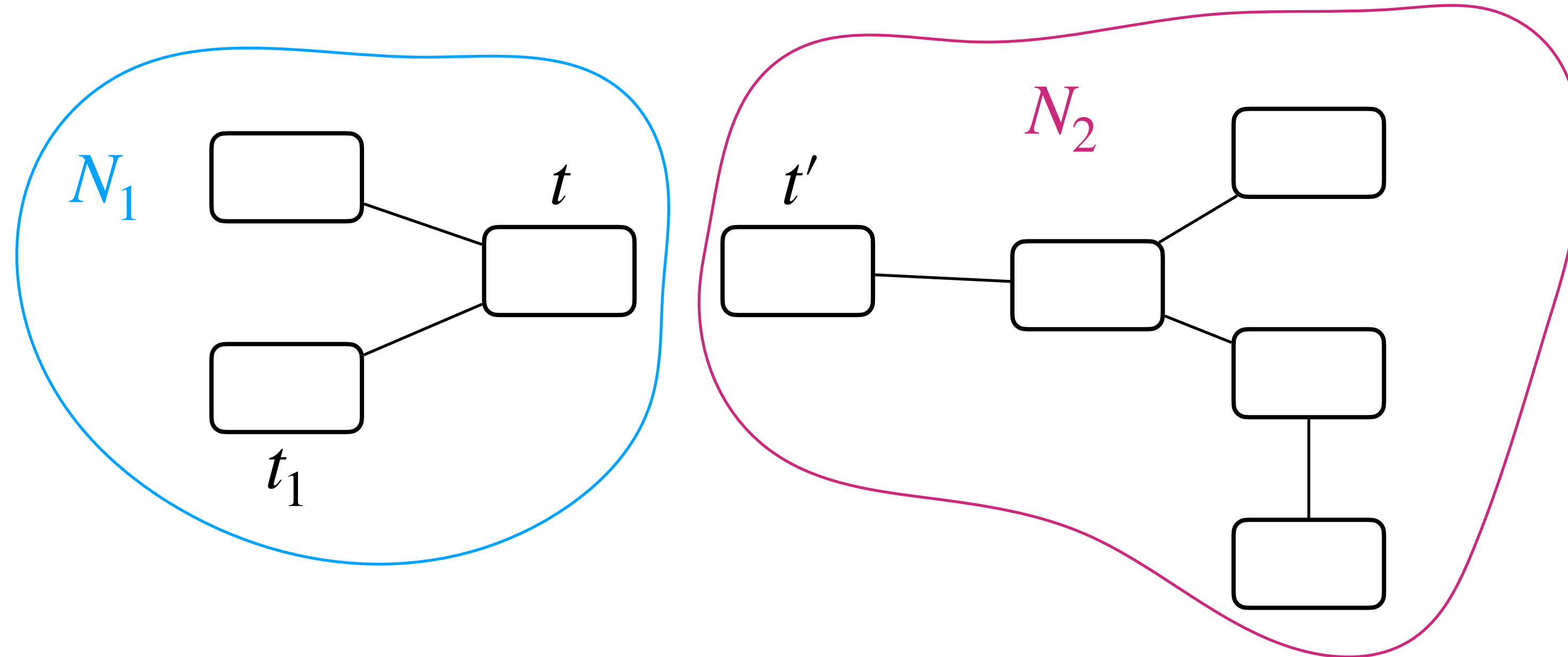


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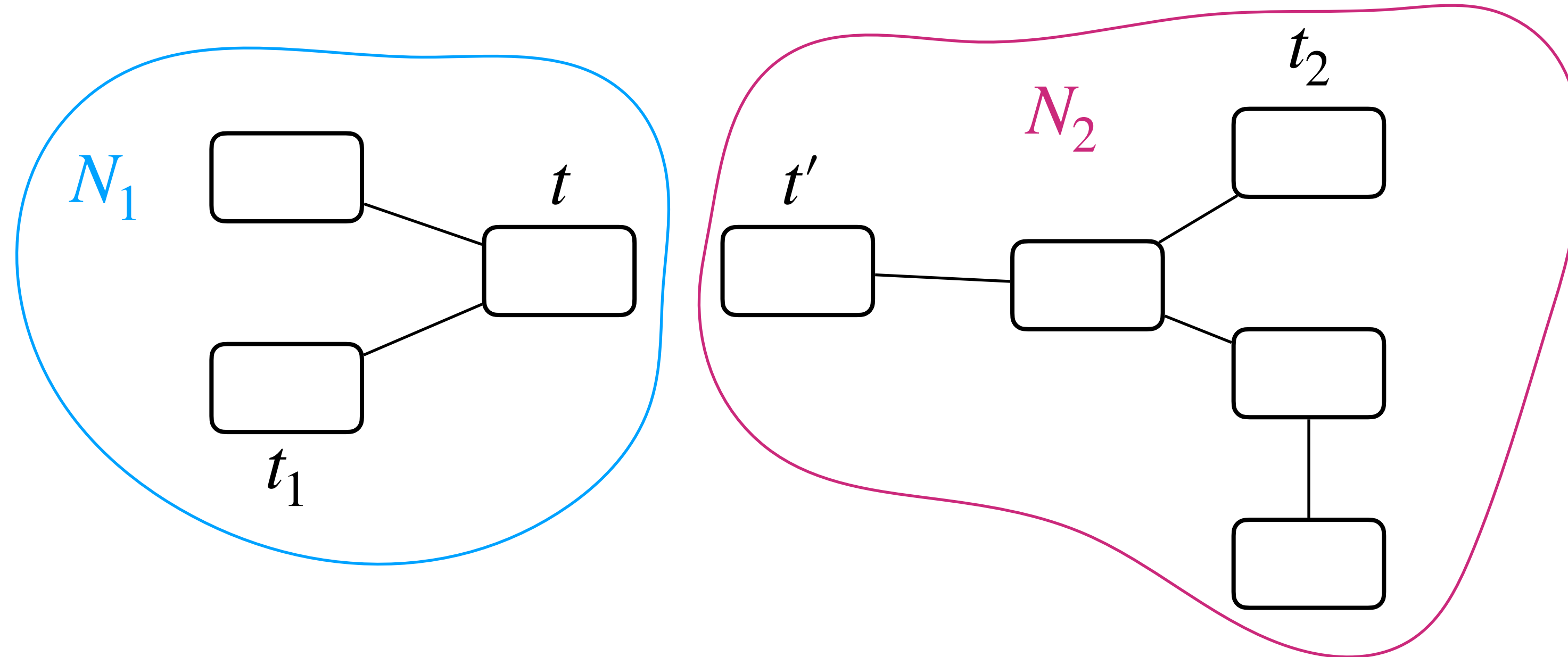


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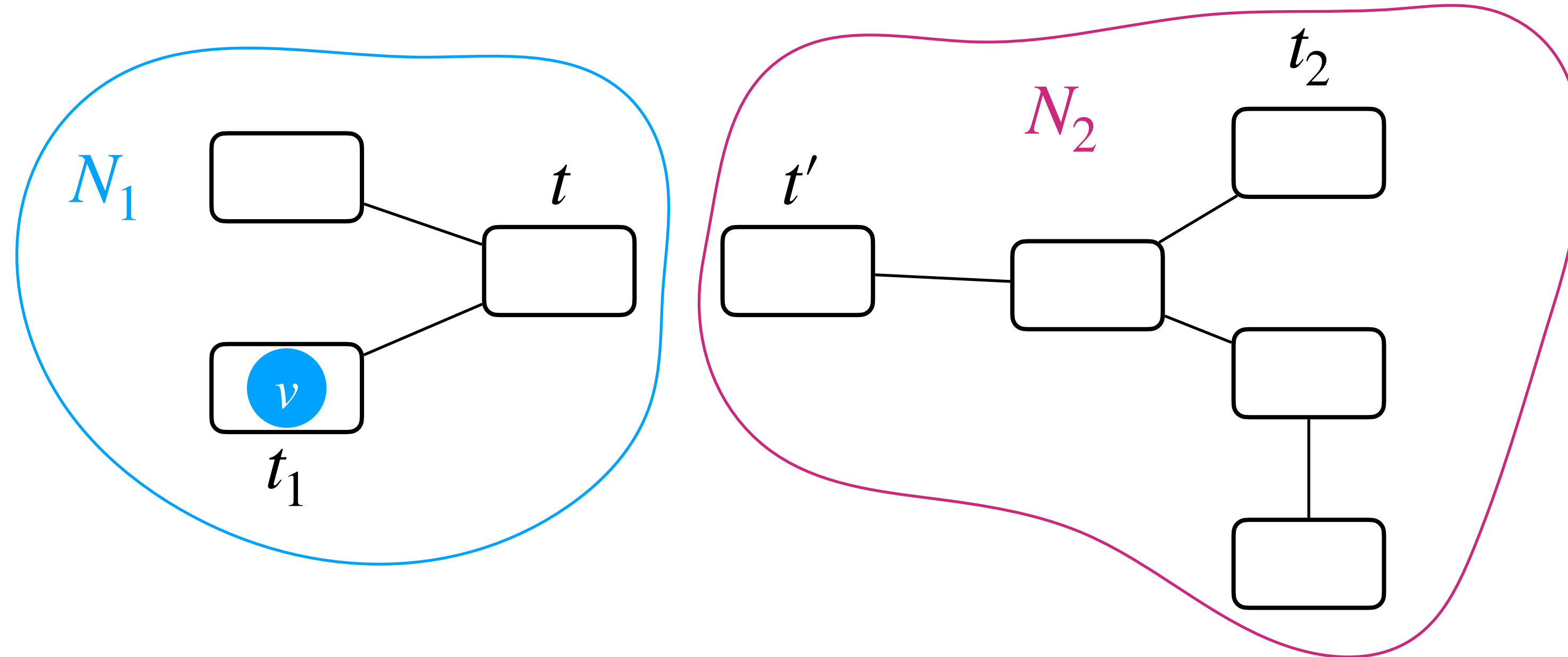


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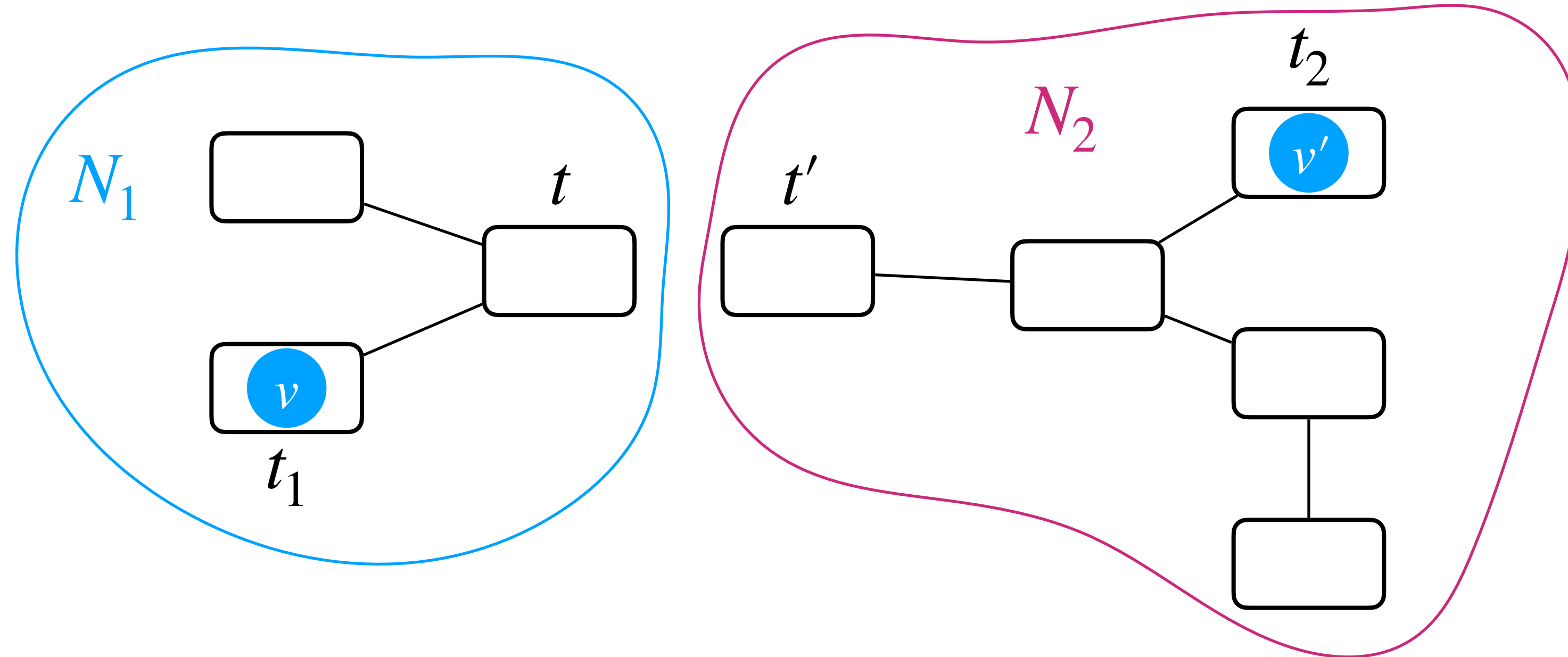


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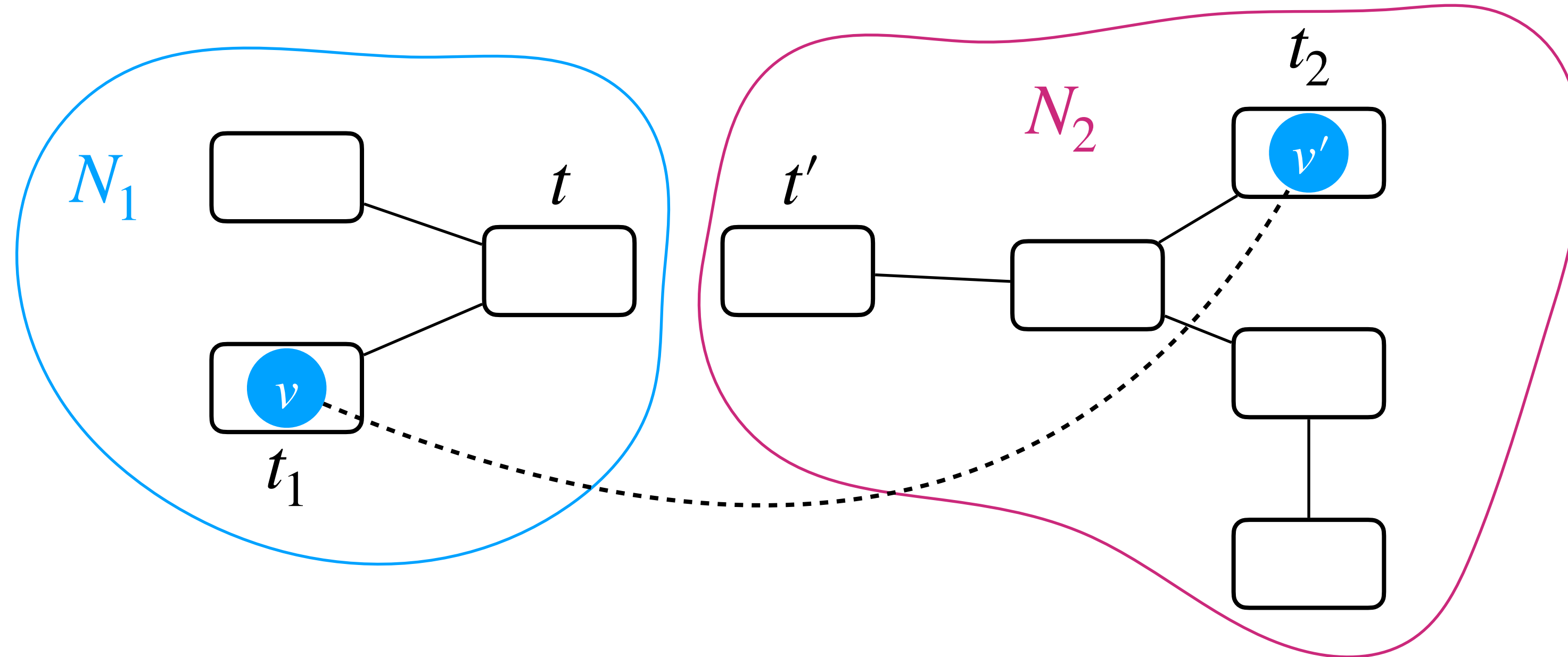


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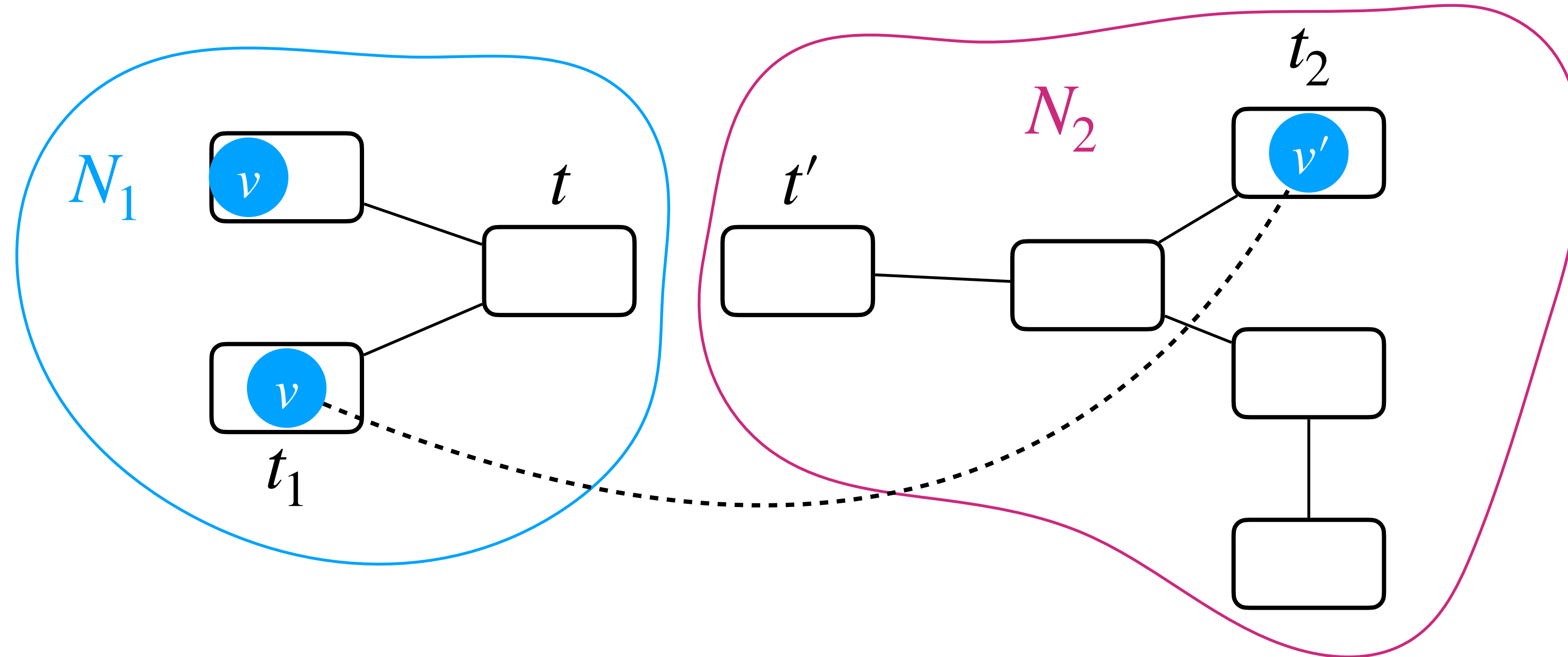


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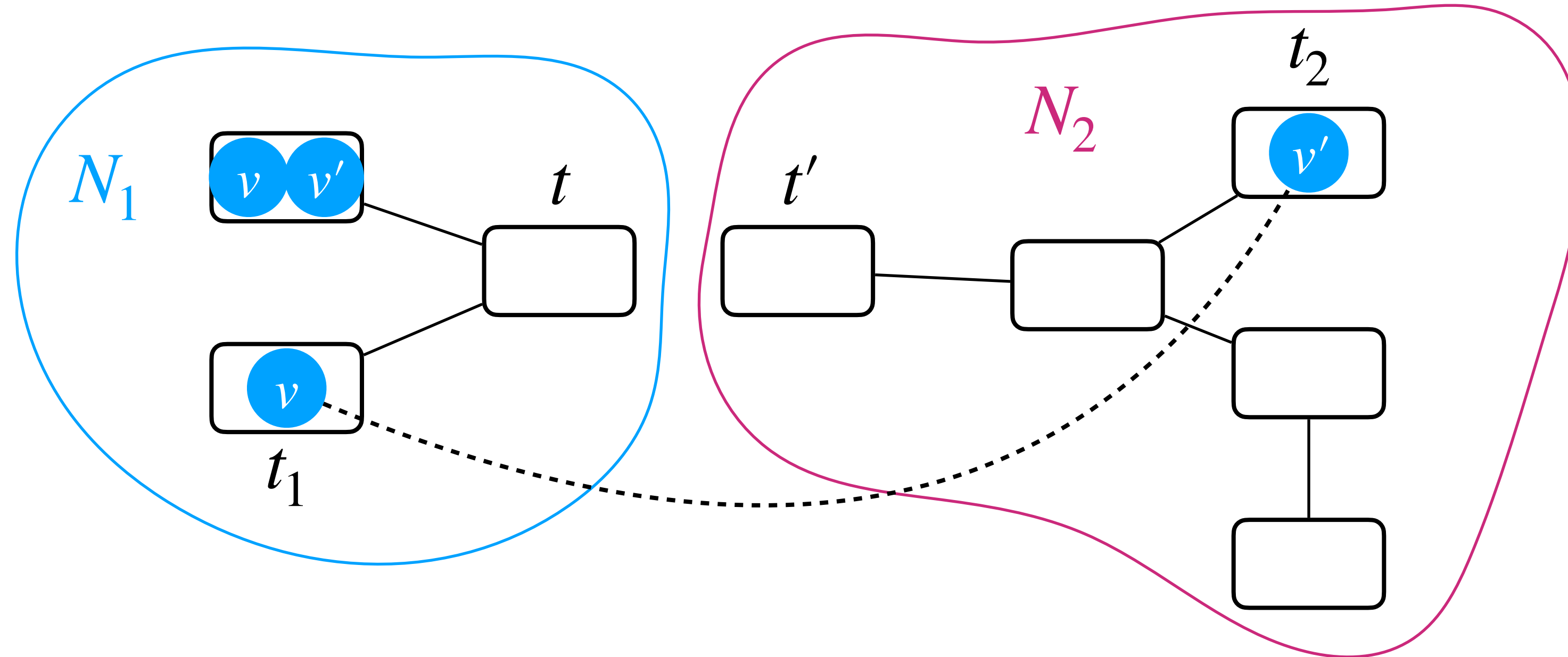


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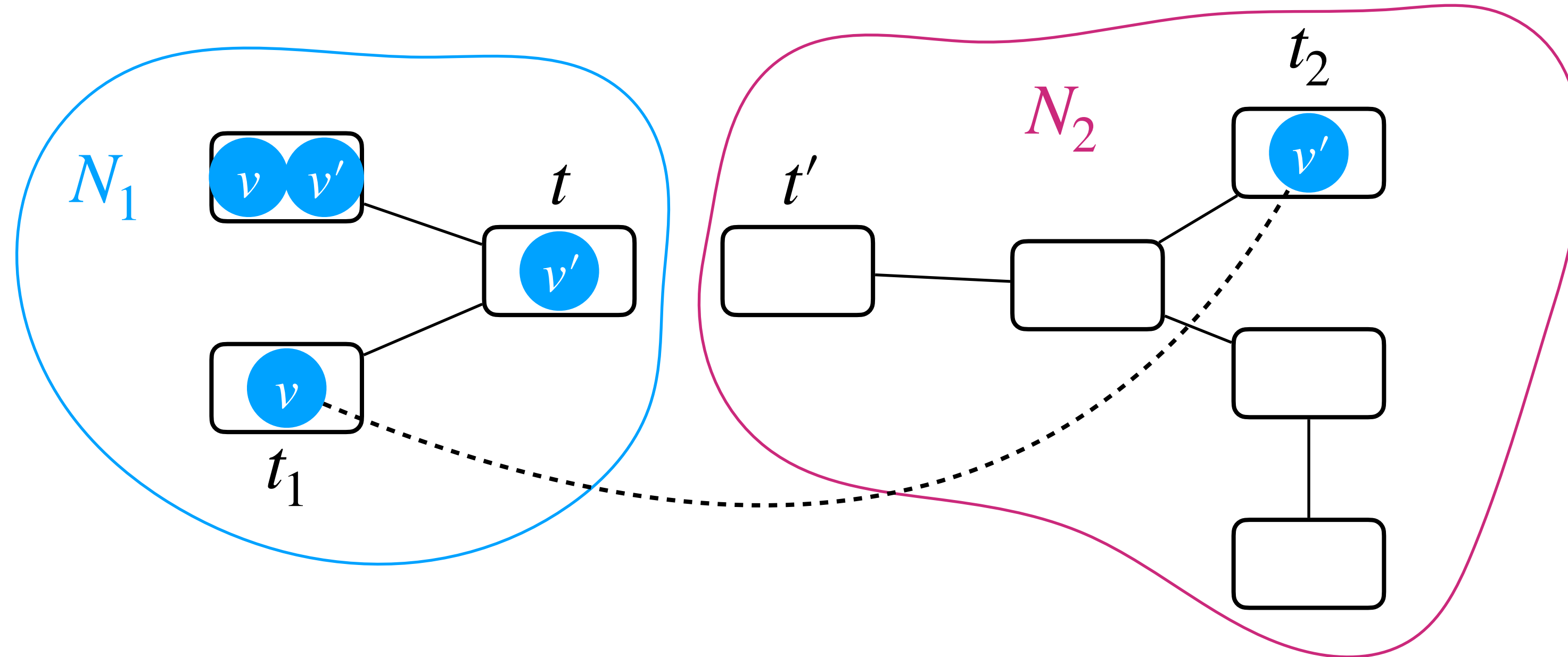


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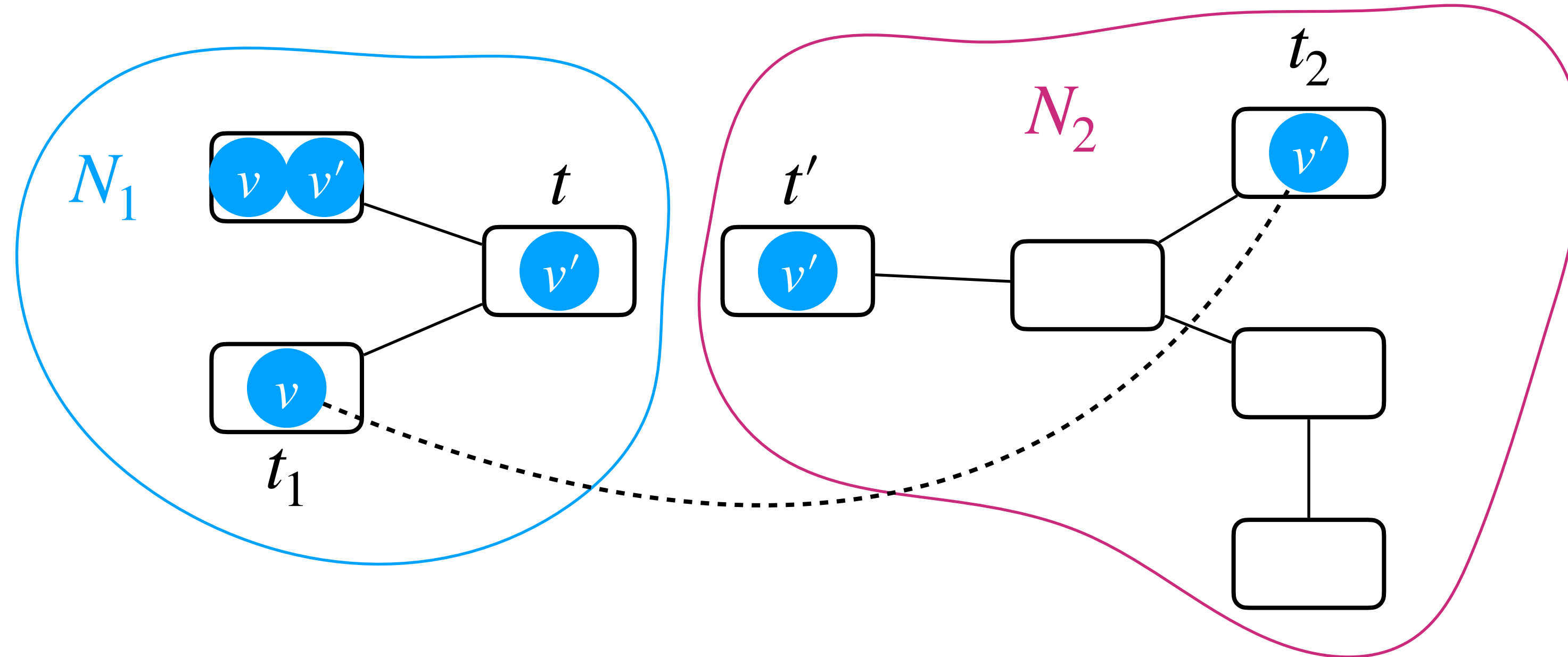


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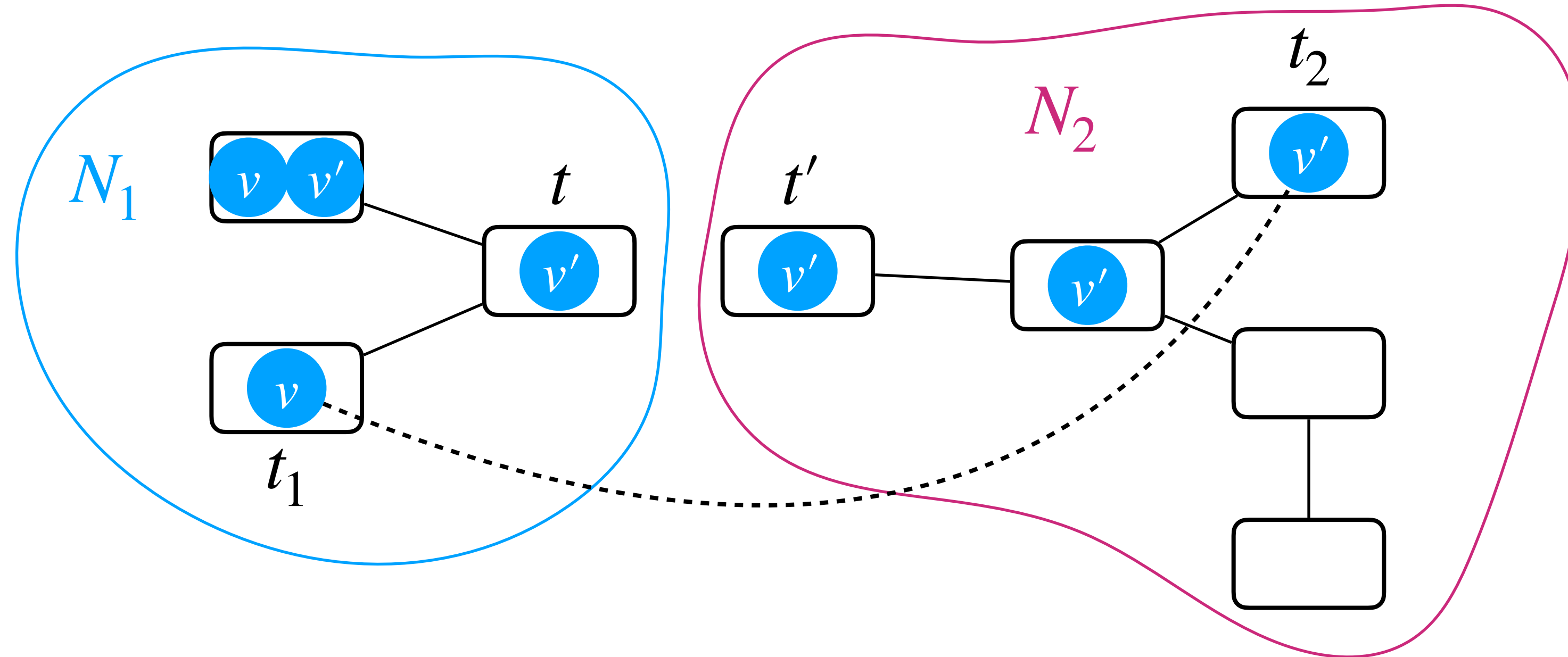


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# Graph Minors and Treewidth



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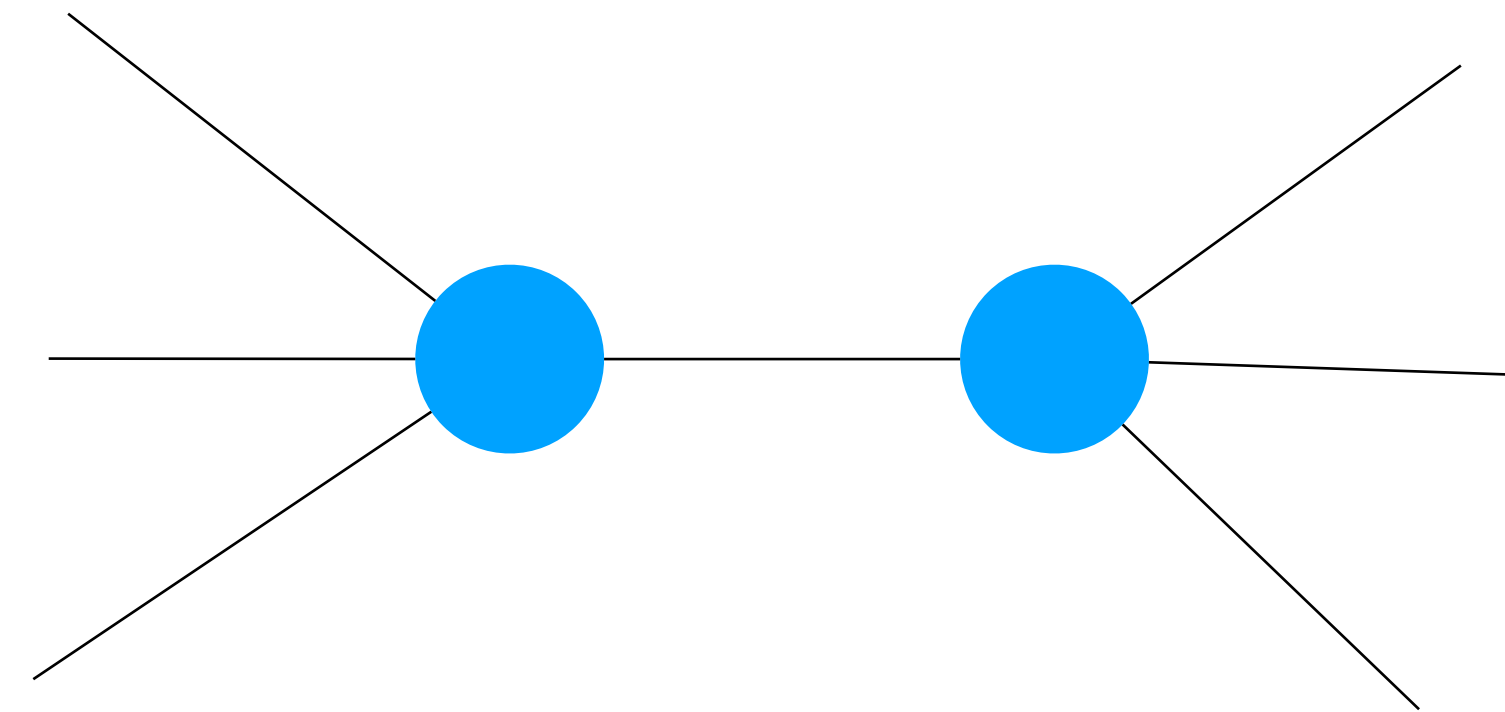
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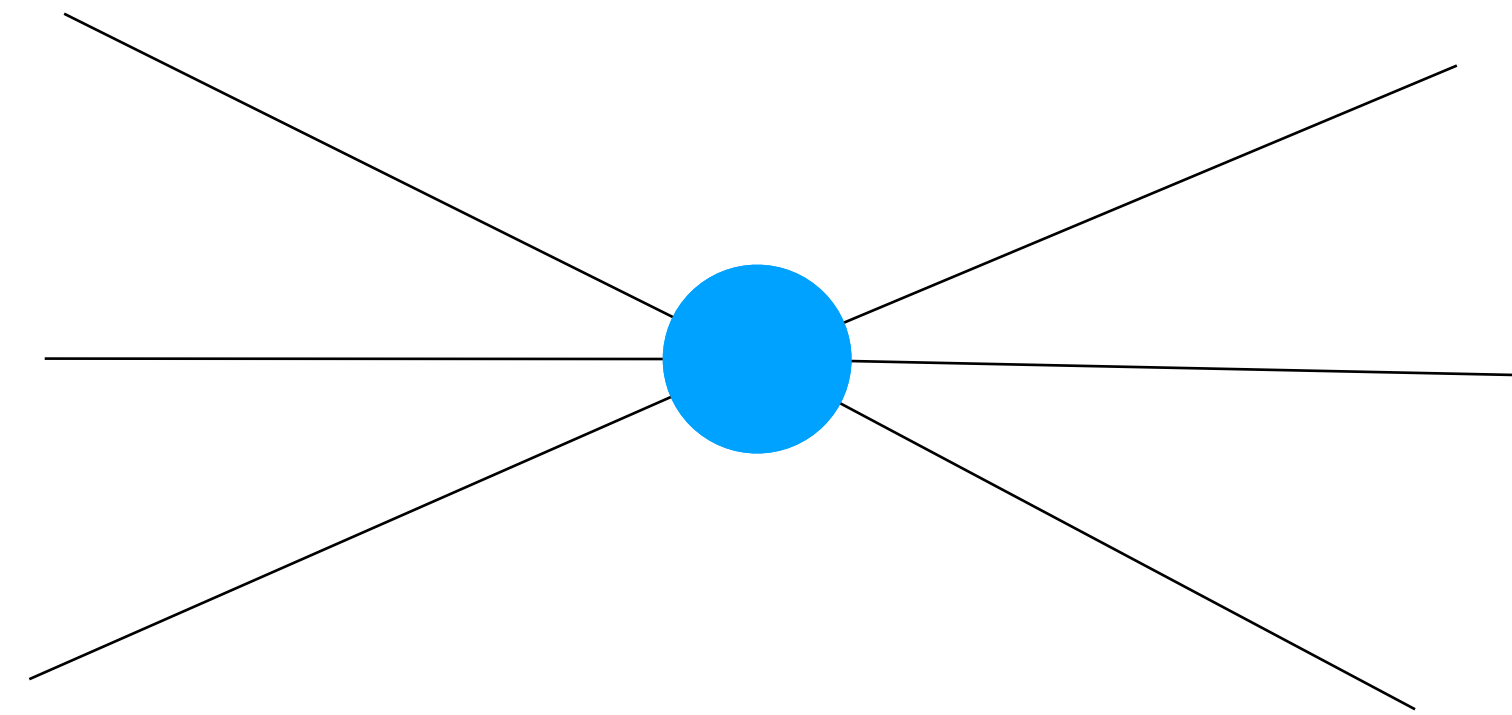


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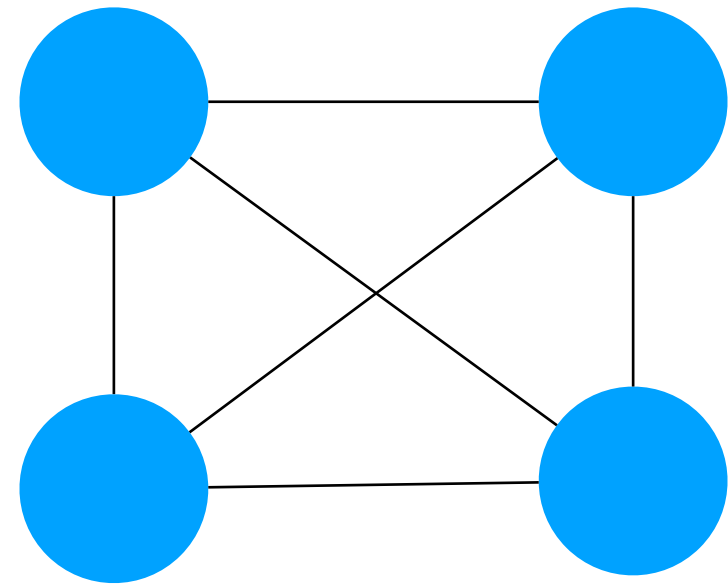
A graph is **planar** if it can be embedded in the plane without edge crossings.



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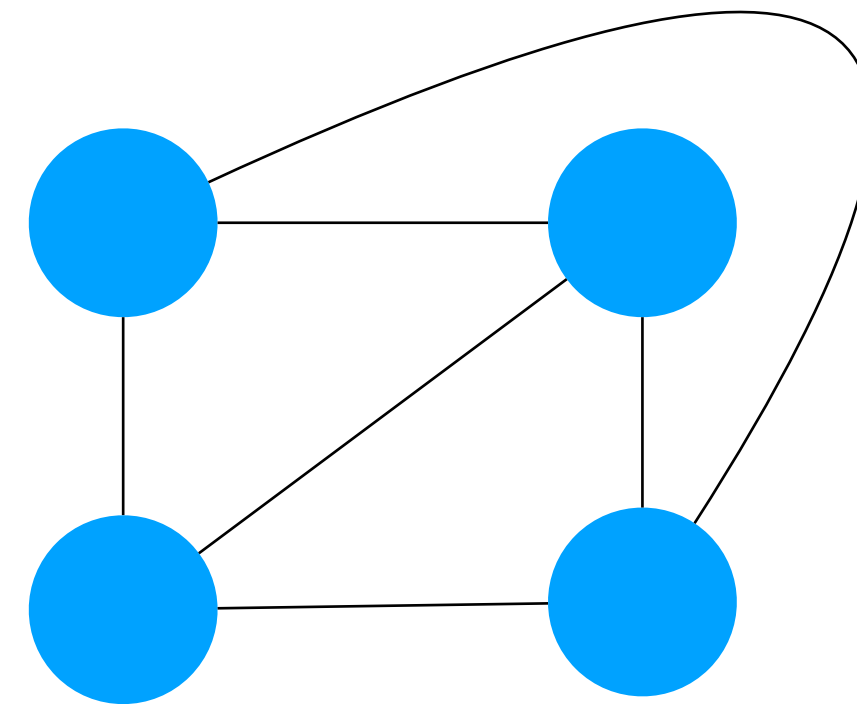
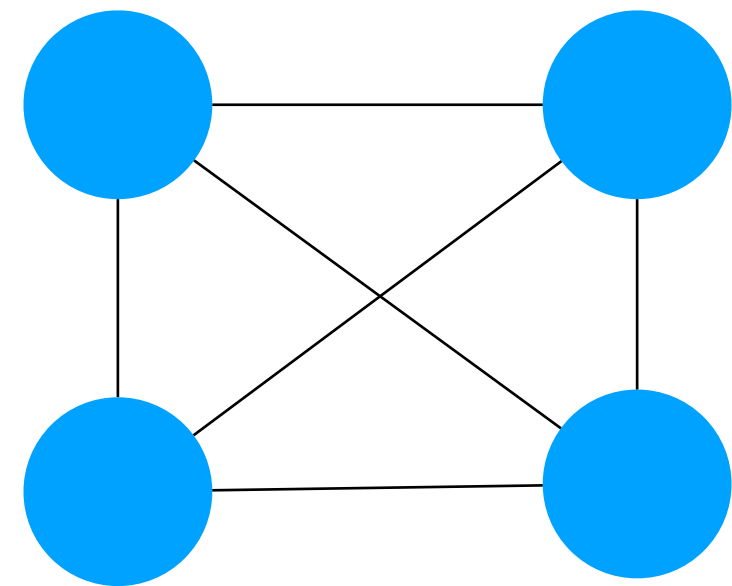
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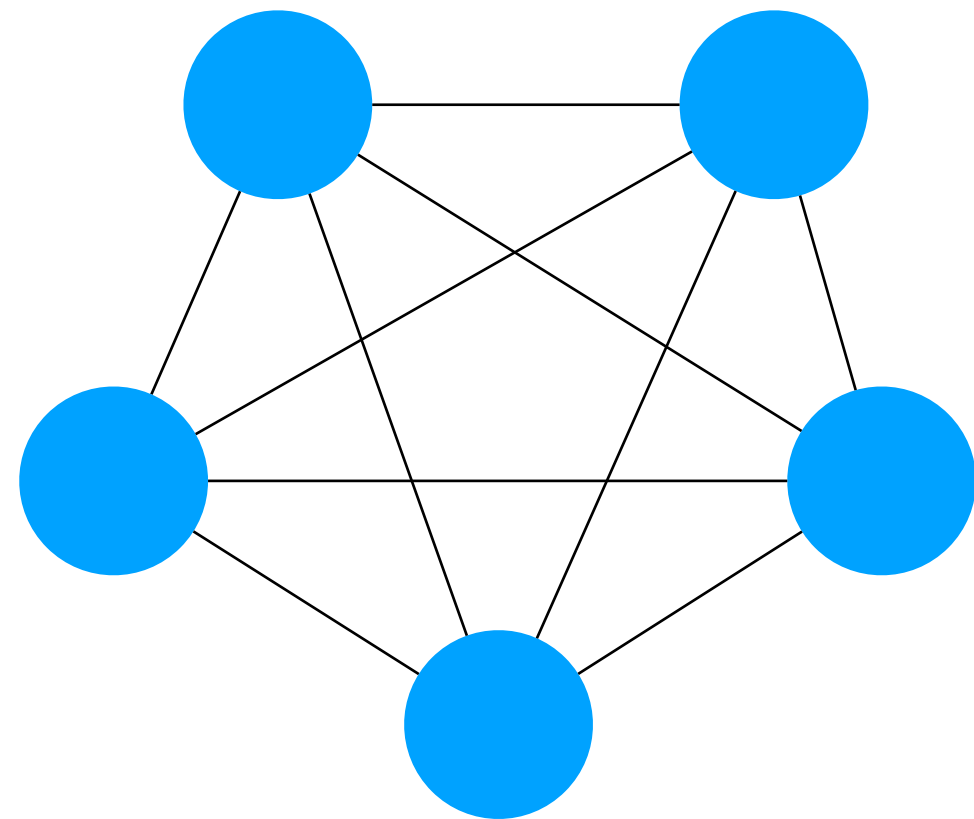
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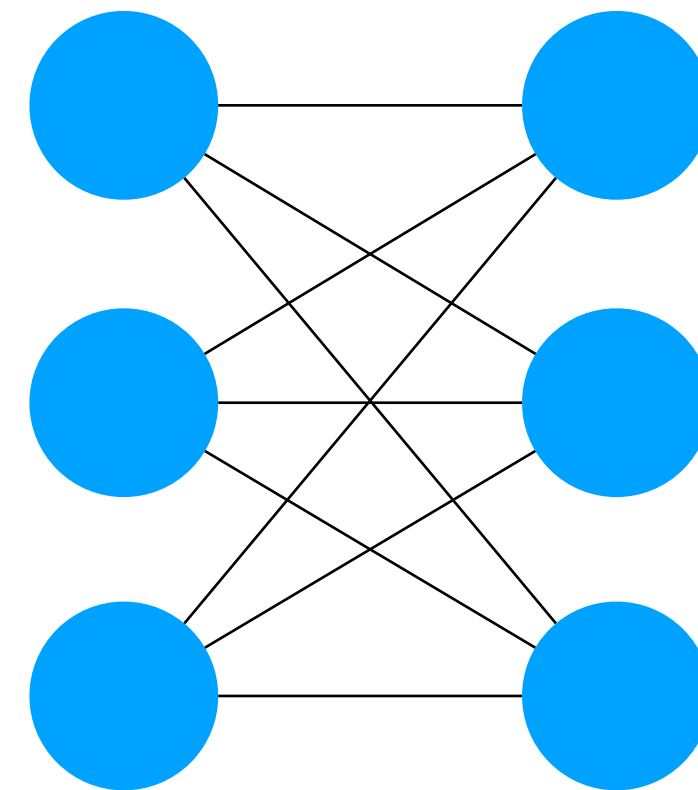
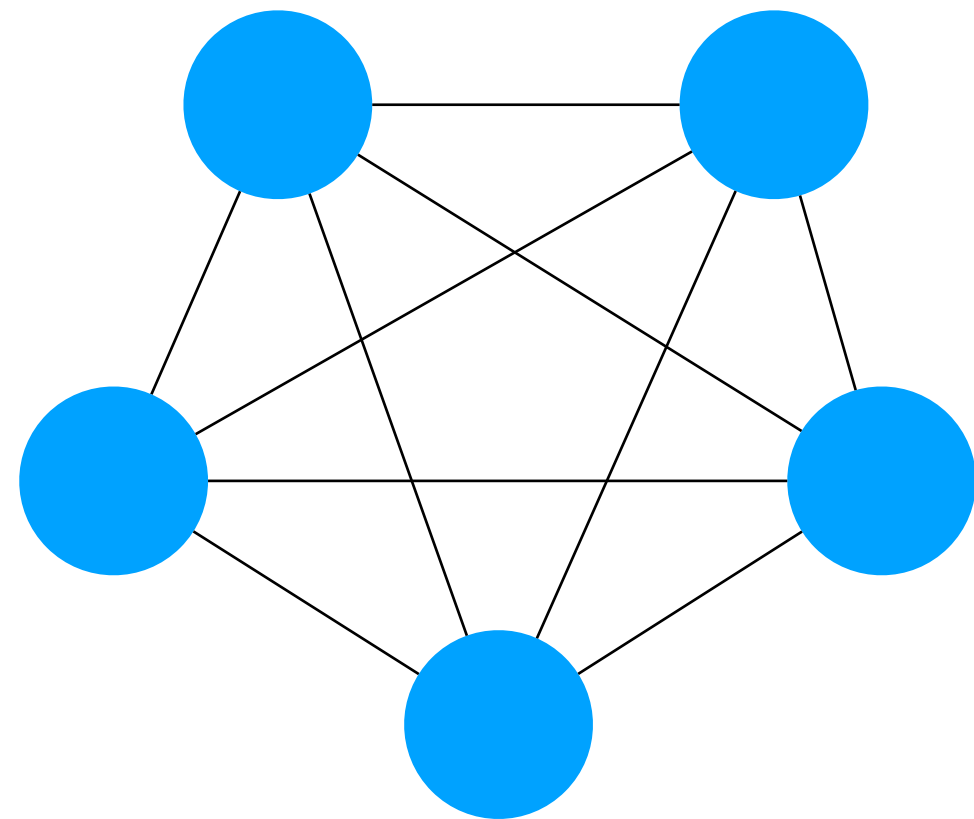
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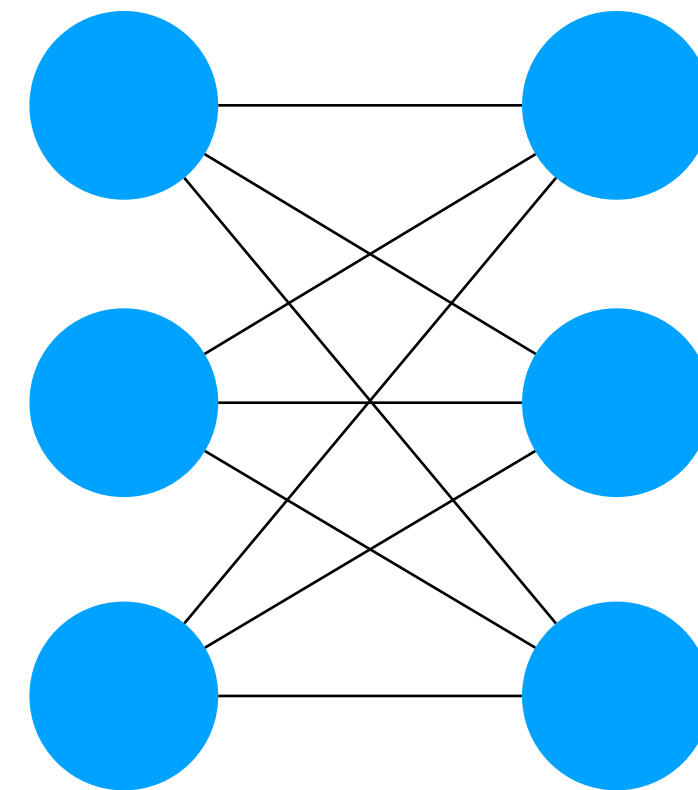
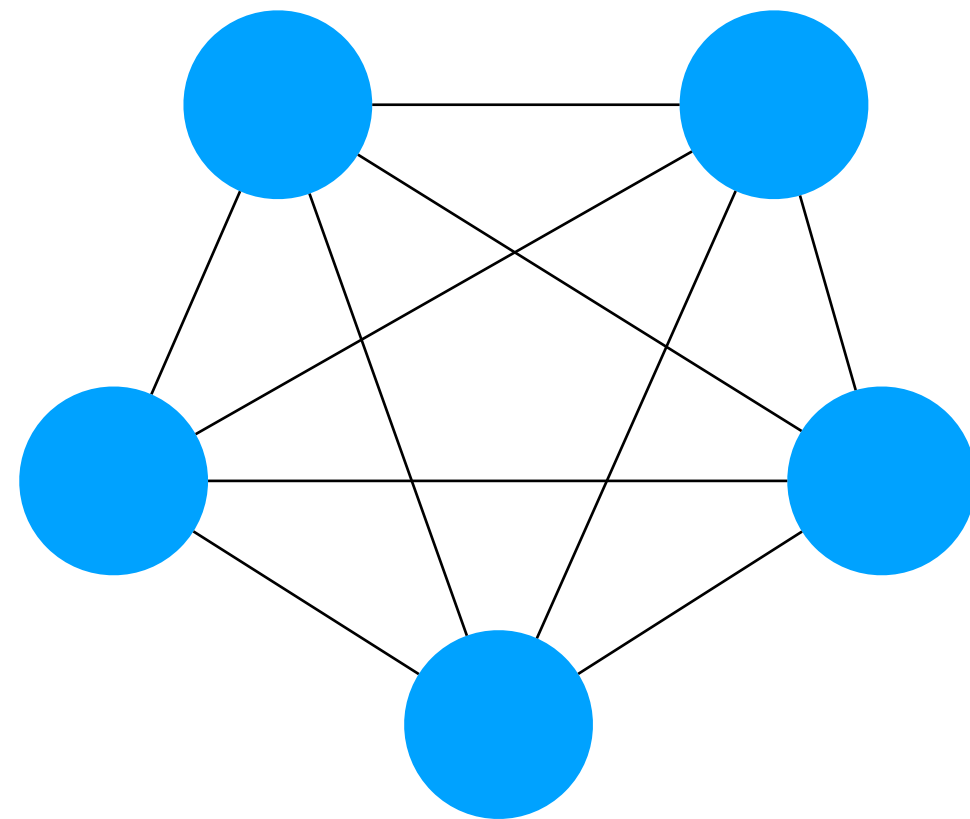
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A graph is **planar** if it can be embedded in the plane without edge crossings.



## Theorem (Kuratowski's Theorem)

A graph is planar if, and only if, it does not contain  $K_5$  or  $K_{3,3}$  as a minor.

# Treewidth and Minors

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## Observation

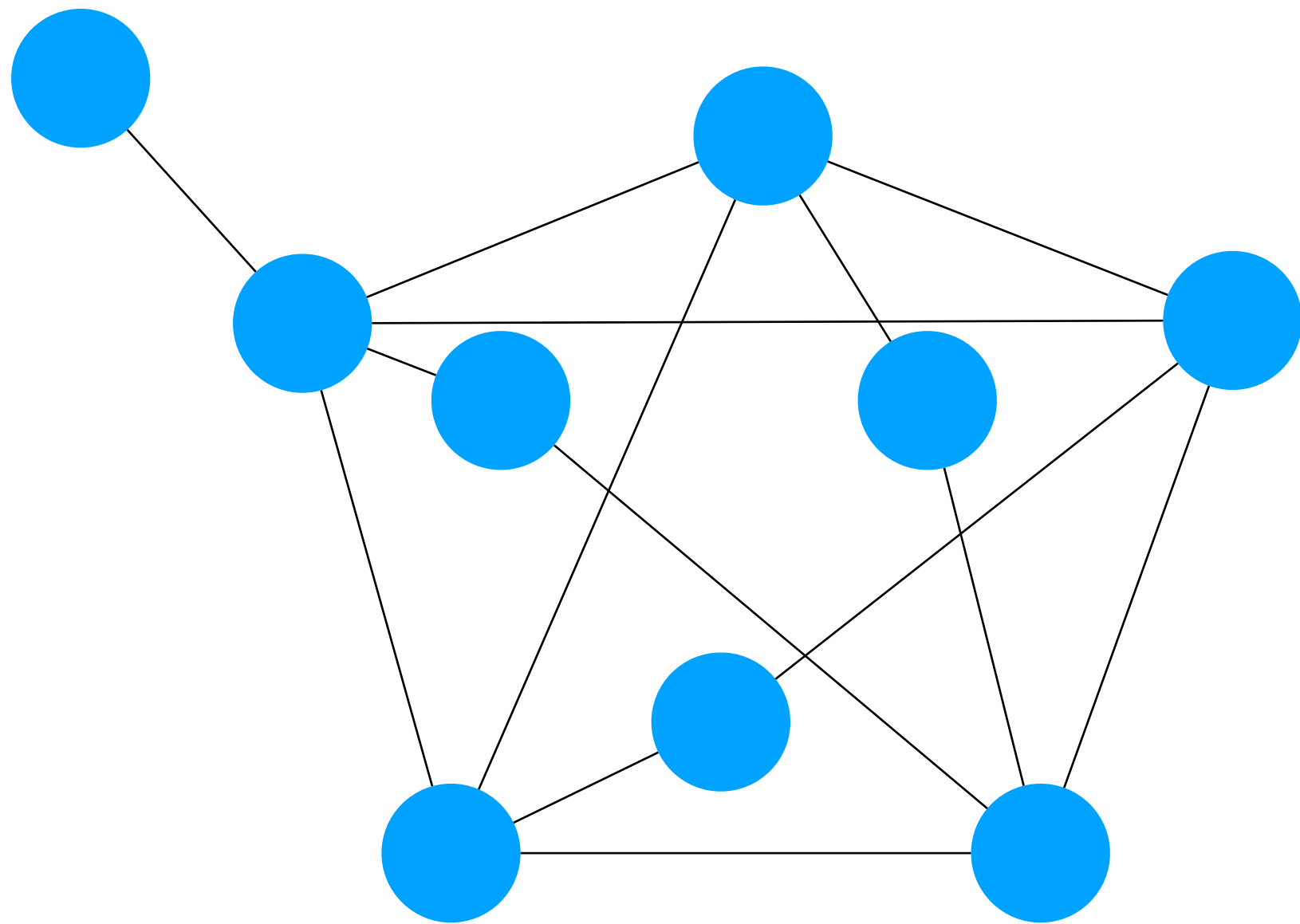
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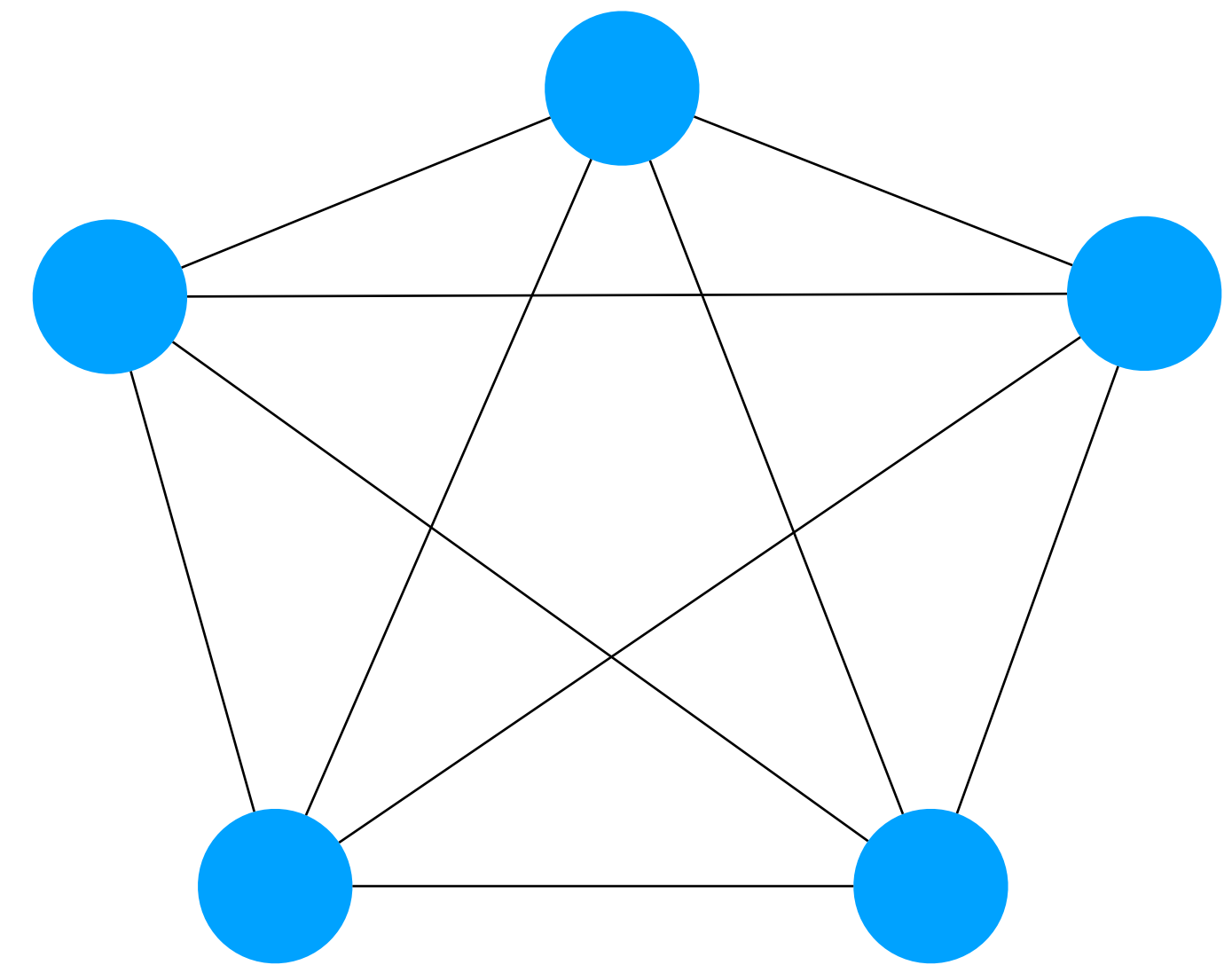
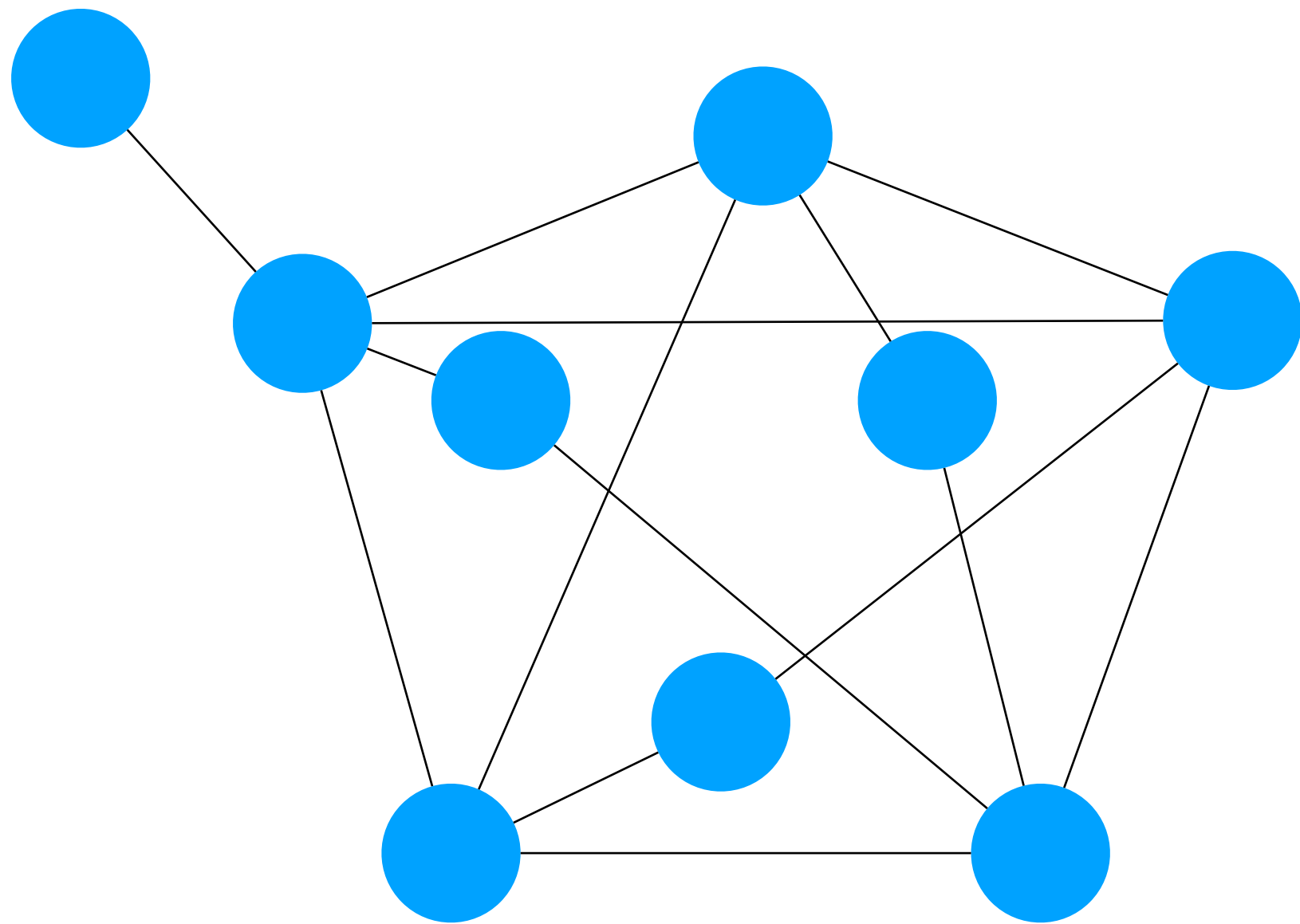
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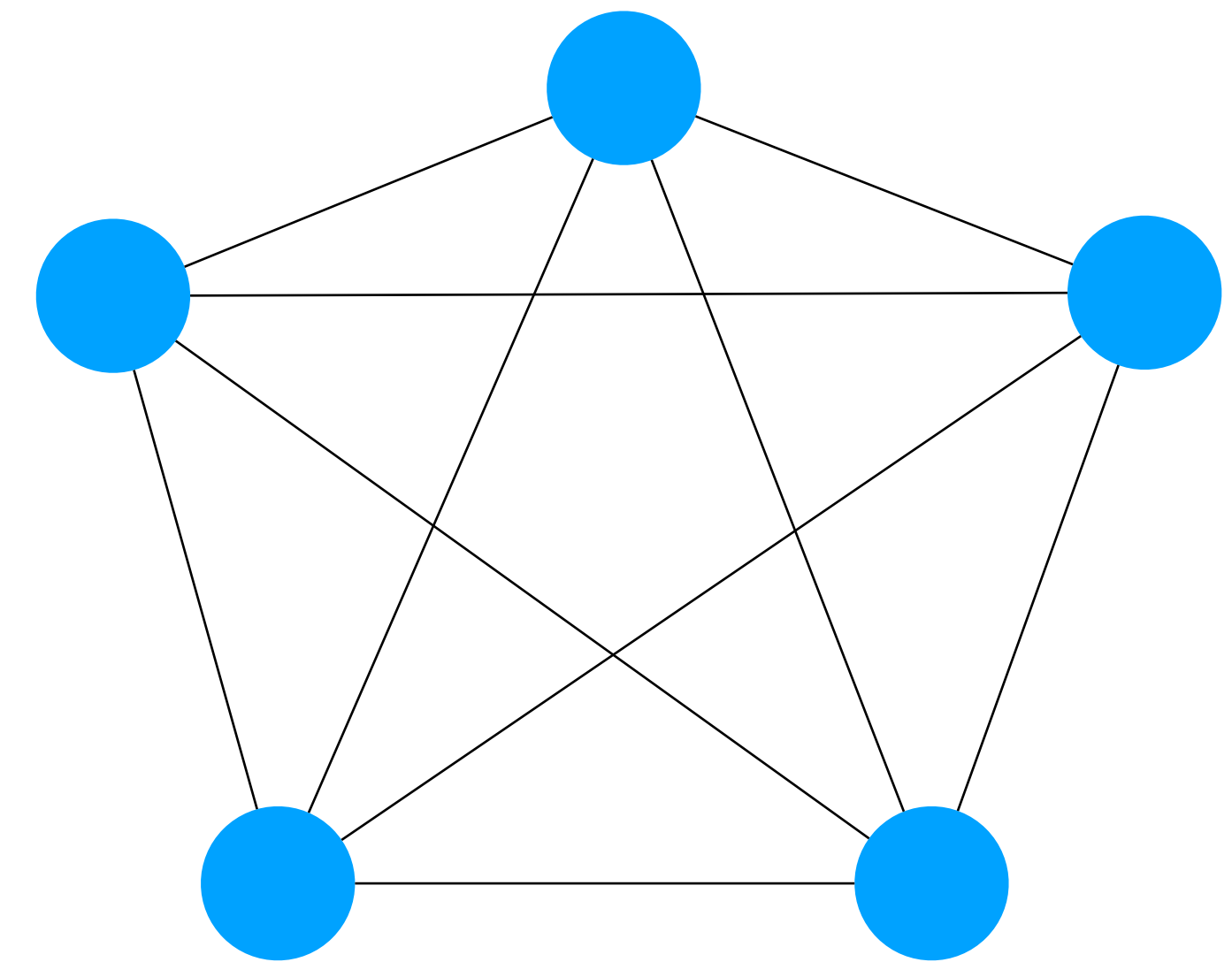
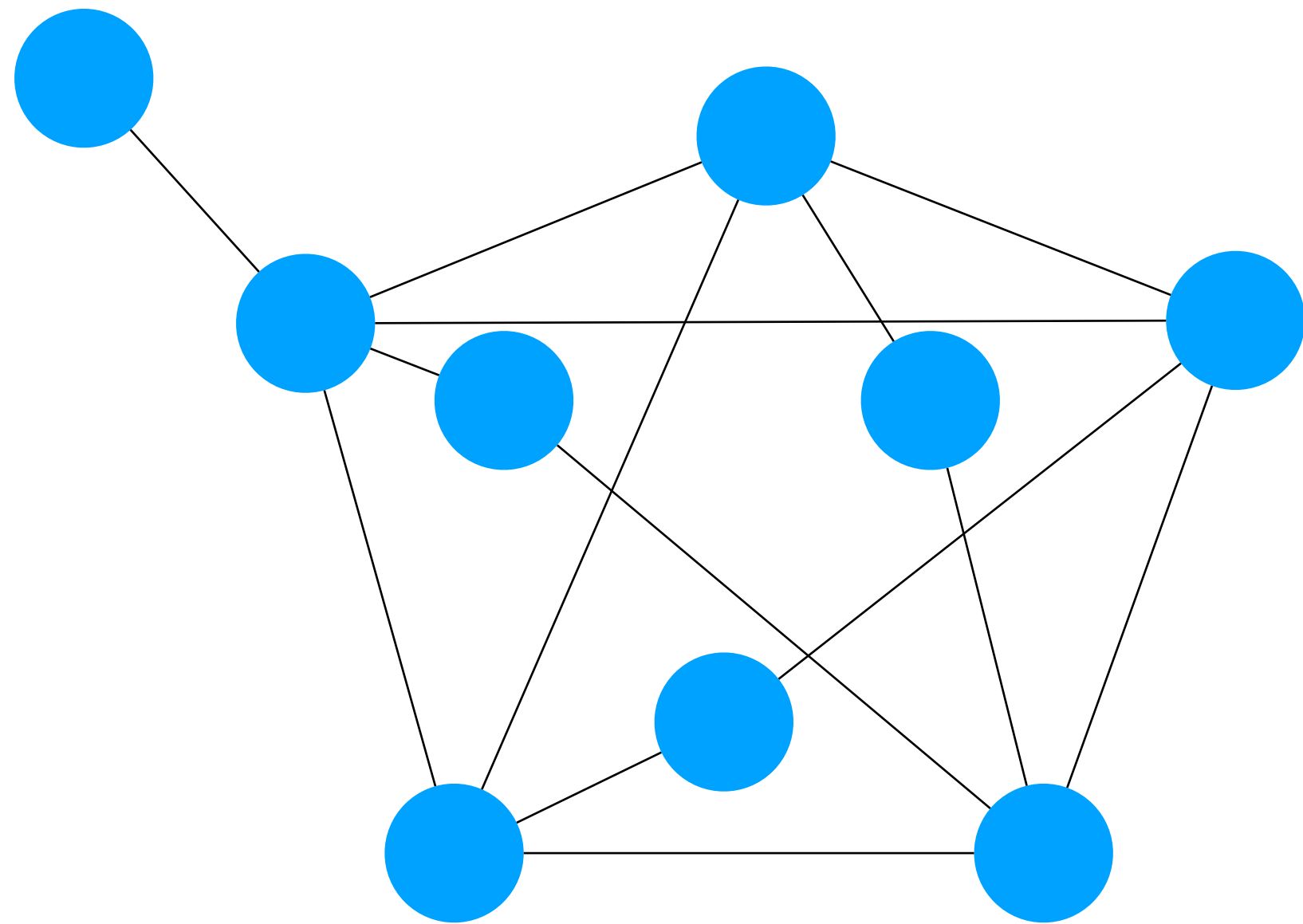
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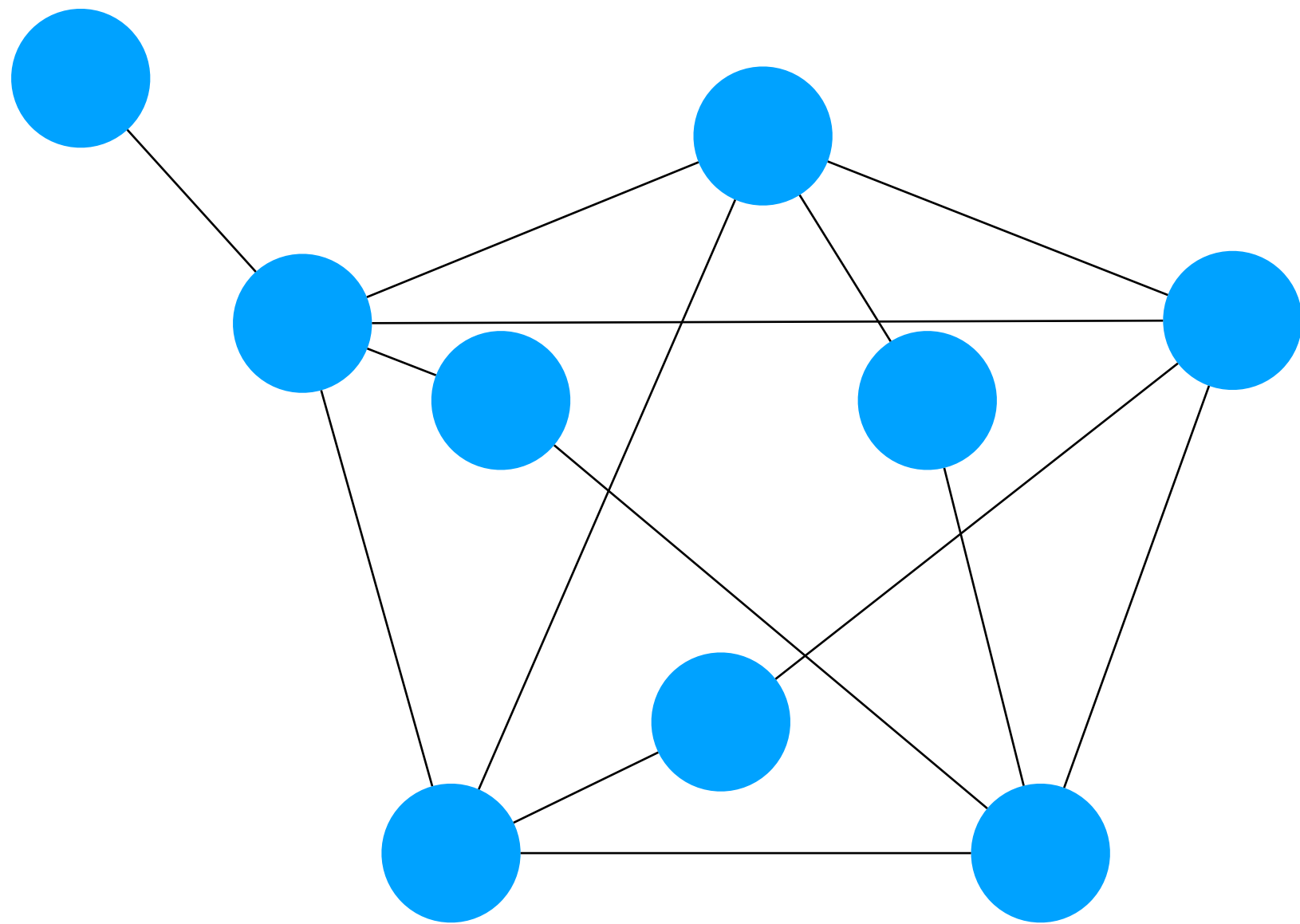


$K_5$  treewidth 4

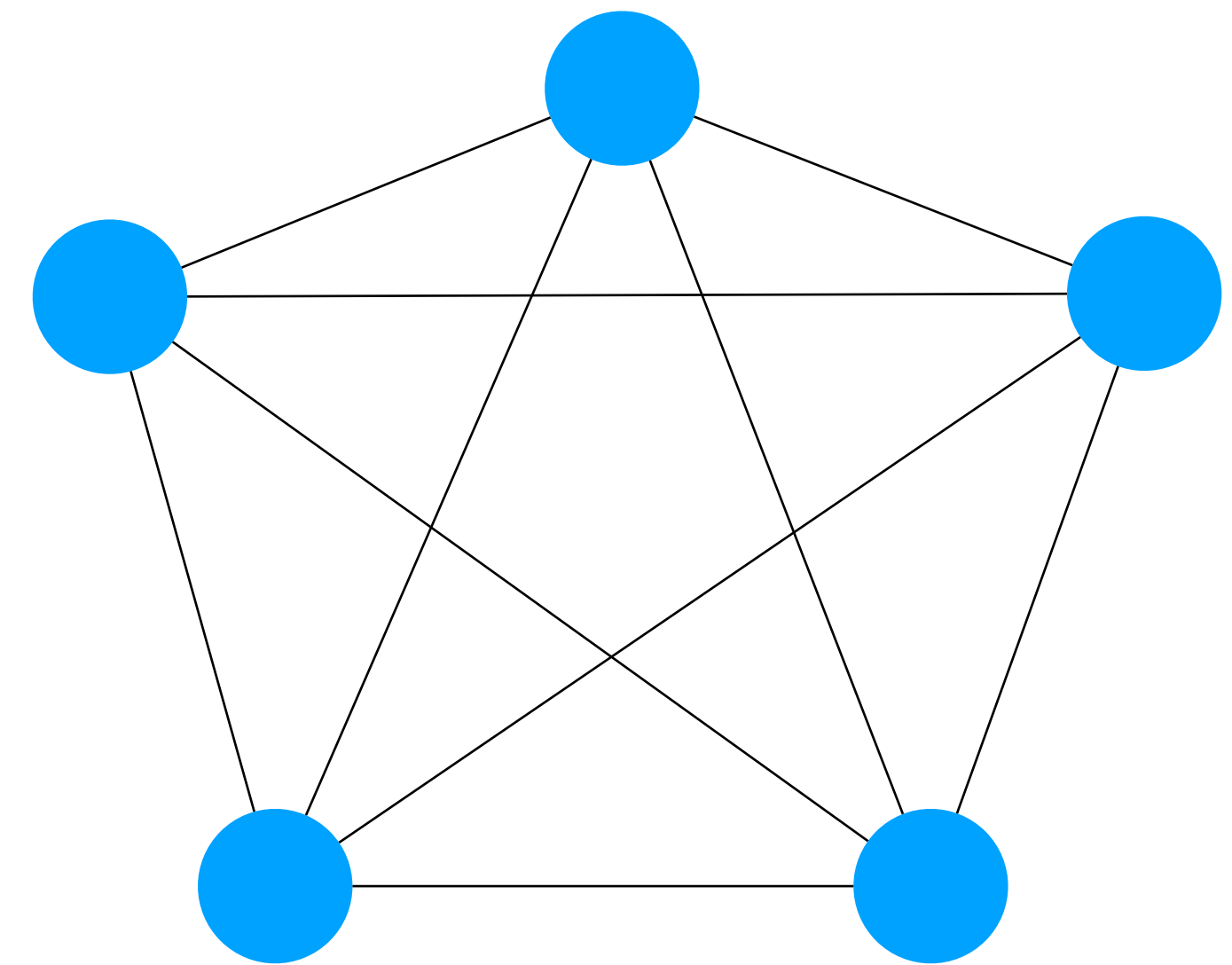
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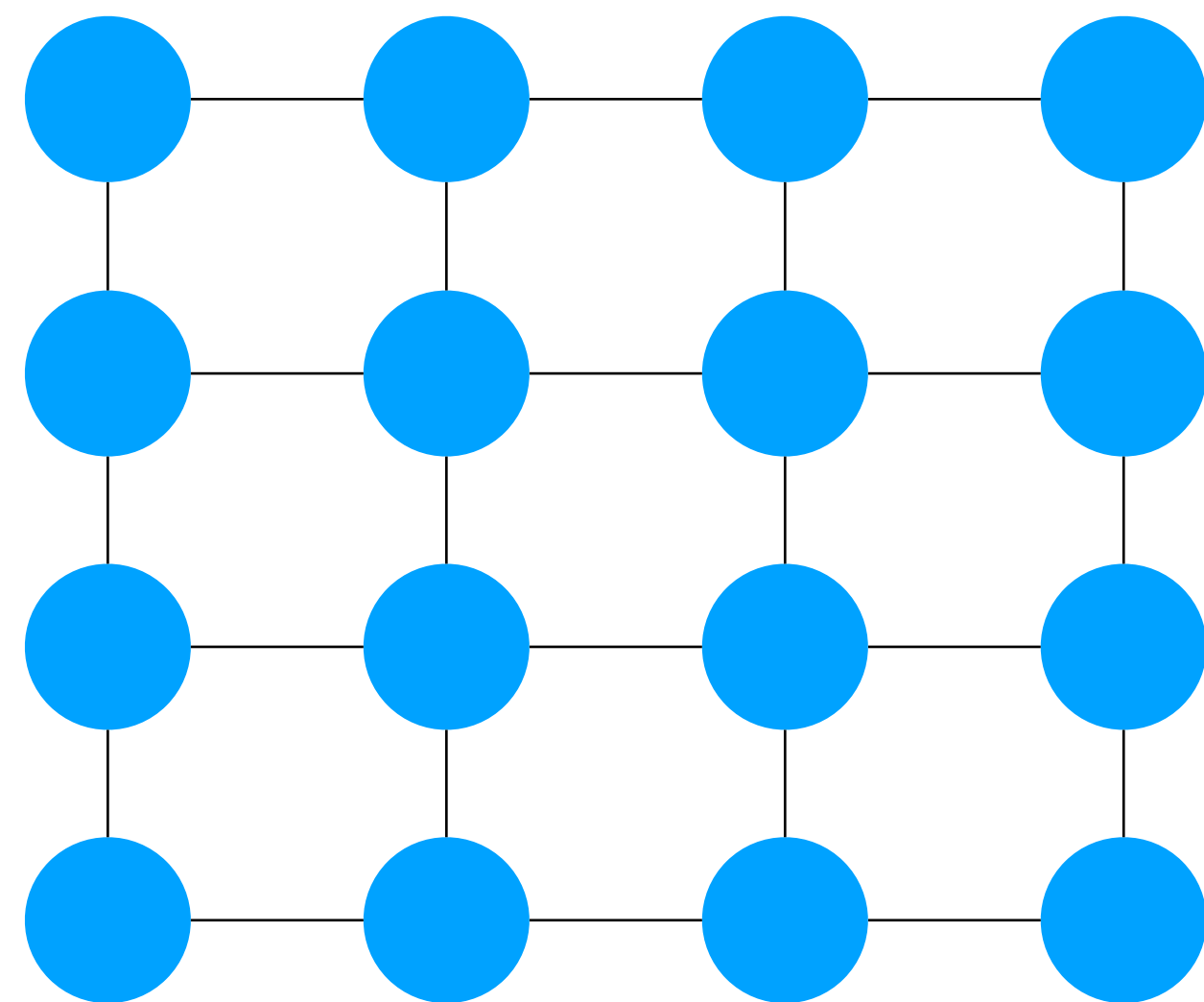


treewidth  $\geq 4$



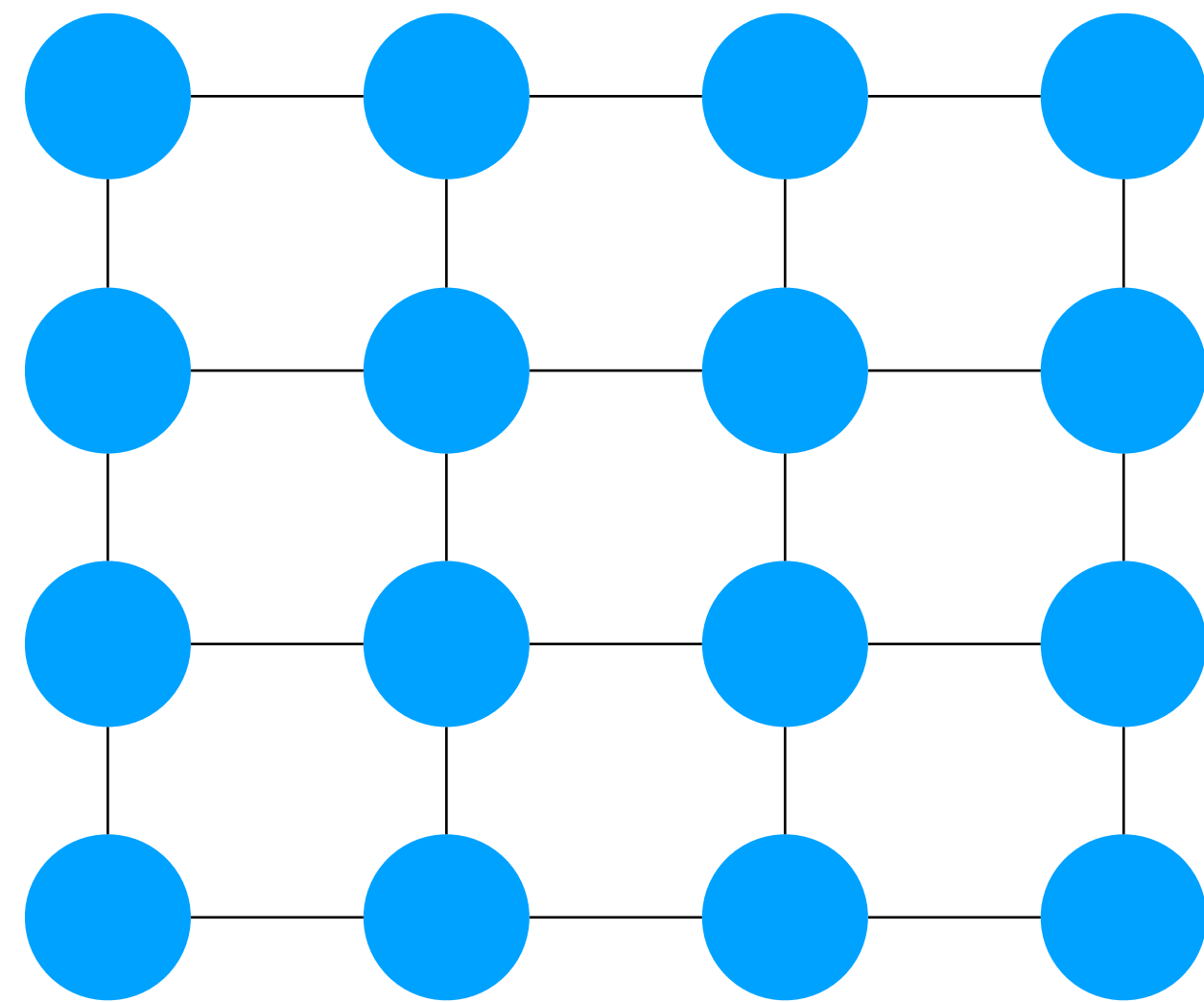
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# Treewidth and Grid Minors



$Q_4$

# Treewidth and Grid Minors

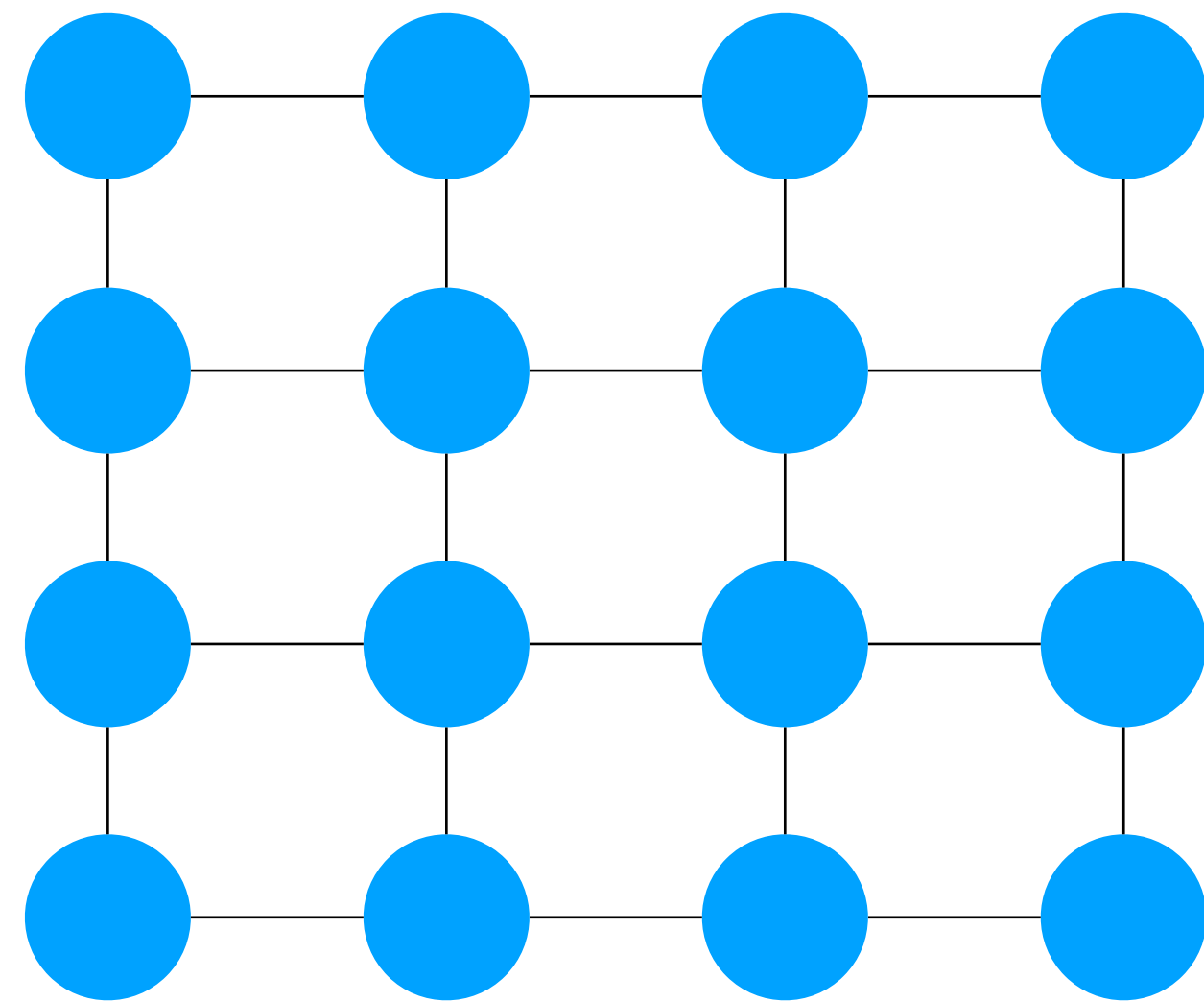


$Q_4$

## Theorem (Robertson and Seymour)

A graph class  $\mathcal{C}$  has bounded treewidth if, and only if, there is a  $k$  such that the grid  $Q_k$  is not a minor of any graph in  $\mathcal{C}$ .

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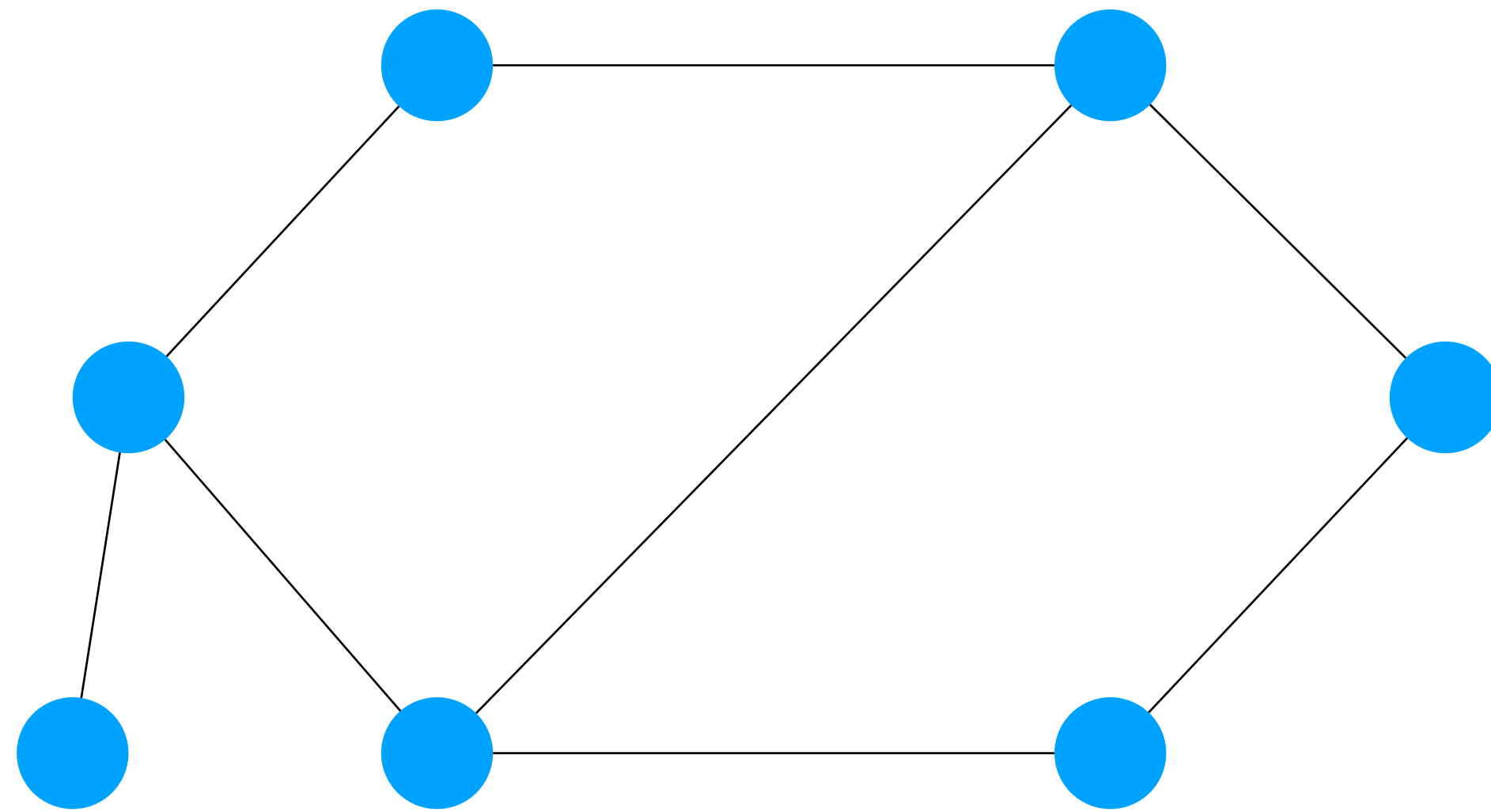
## Theorem (Chekuri and Chuzhoy)

There is a polynomial  $p$  such that every graph of treewidth larger than  $p(k)$  contains  $Q_k$  as a minor.

# Cops and Robbers

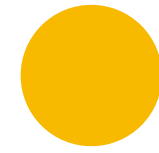
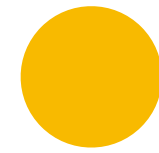


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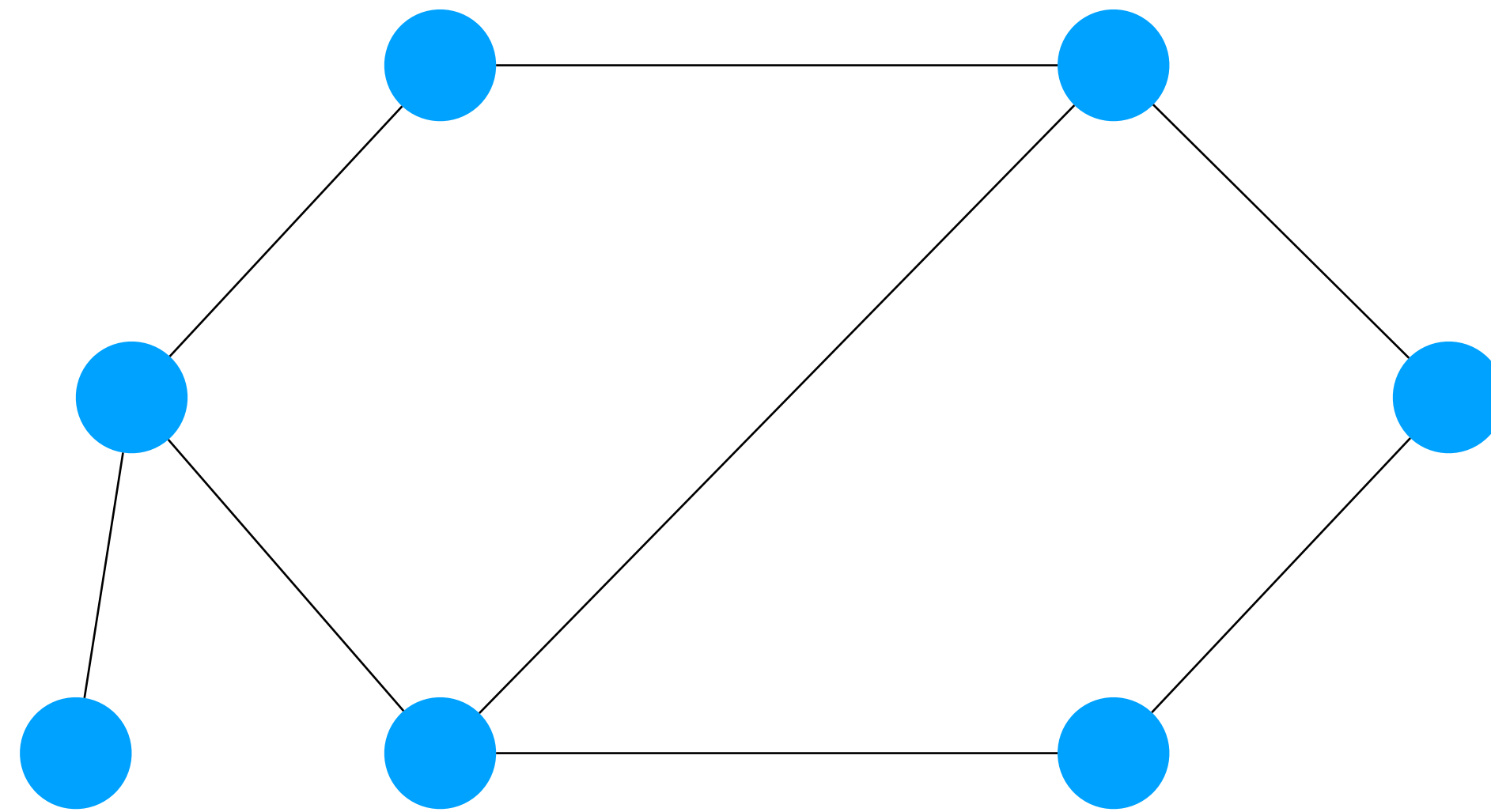
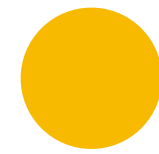


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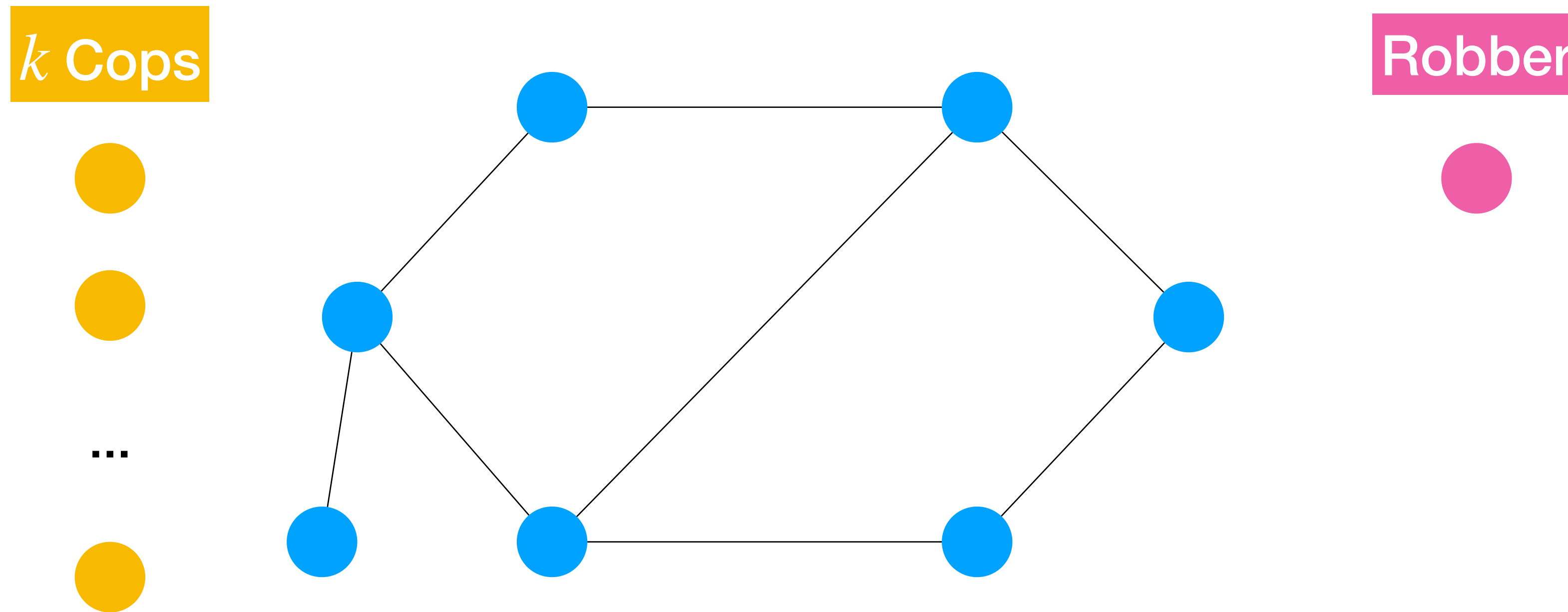
$k$  Cops



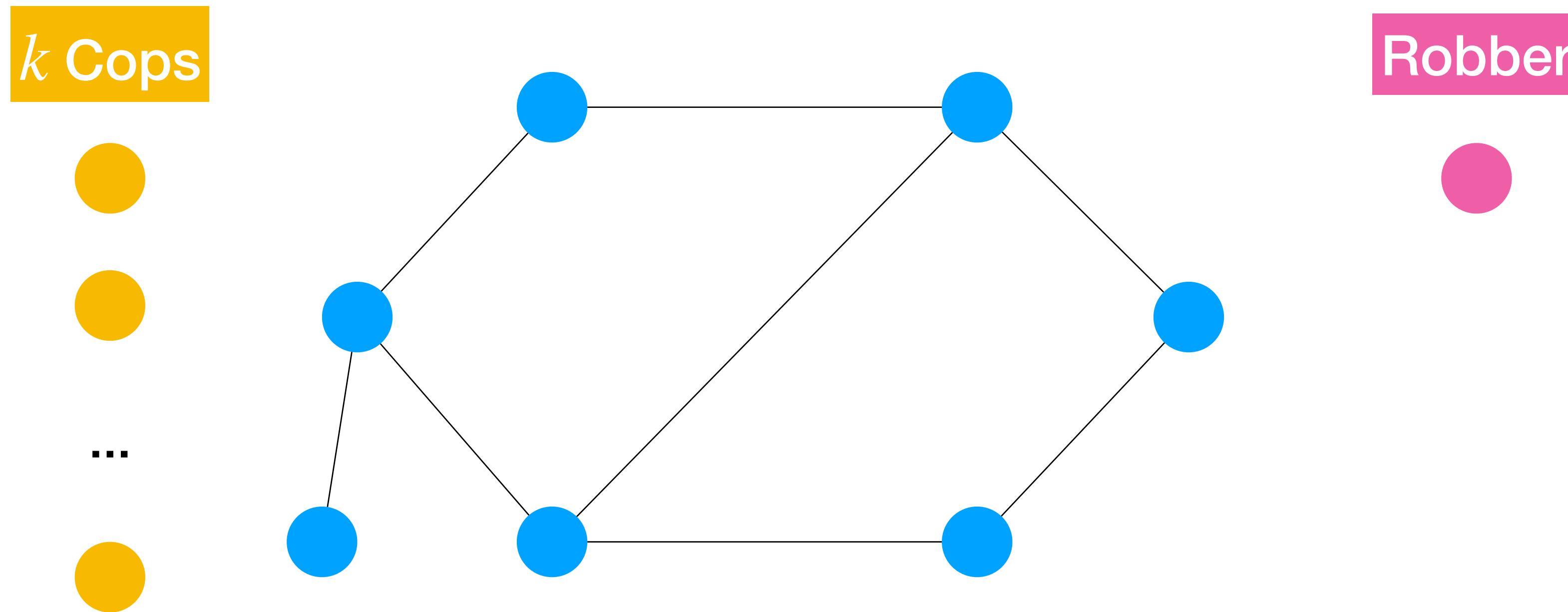
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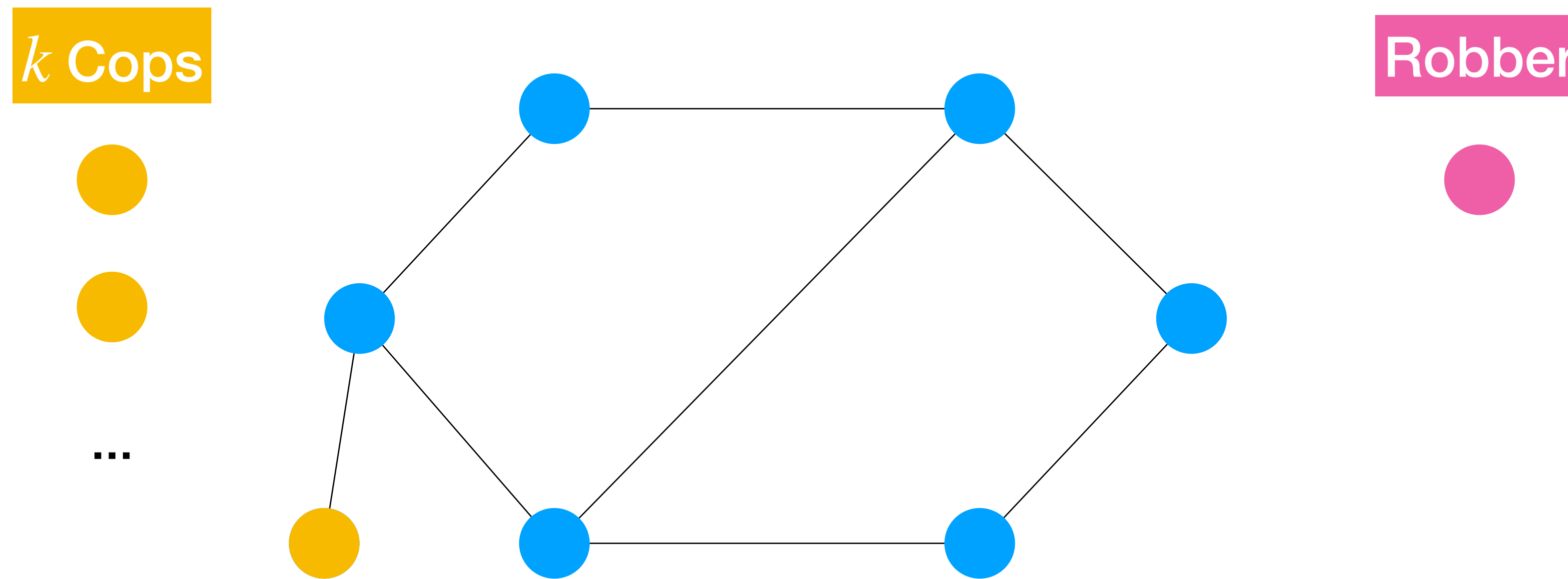


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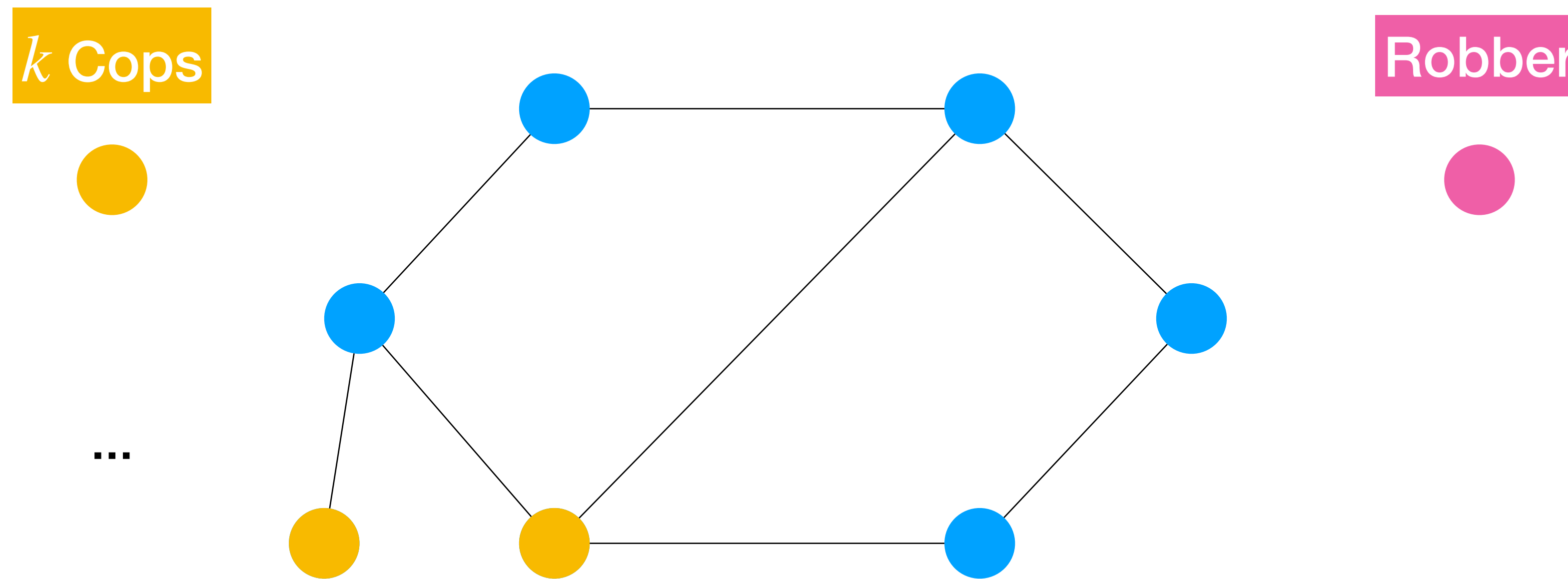
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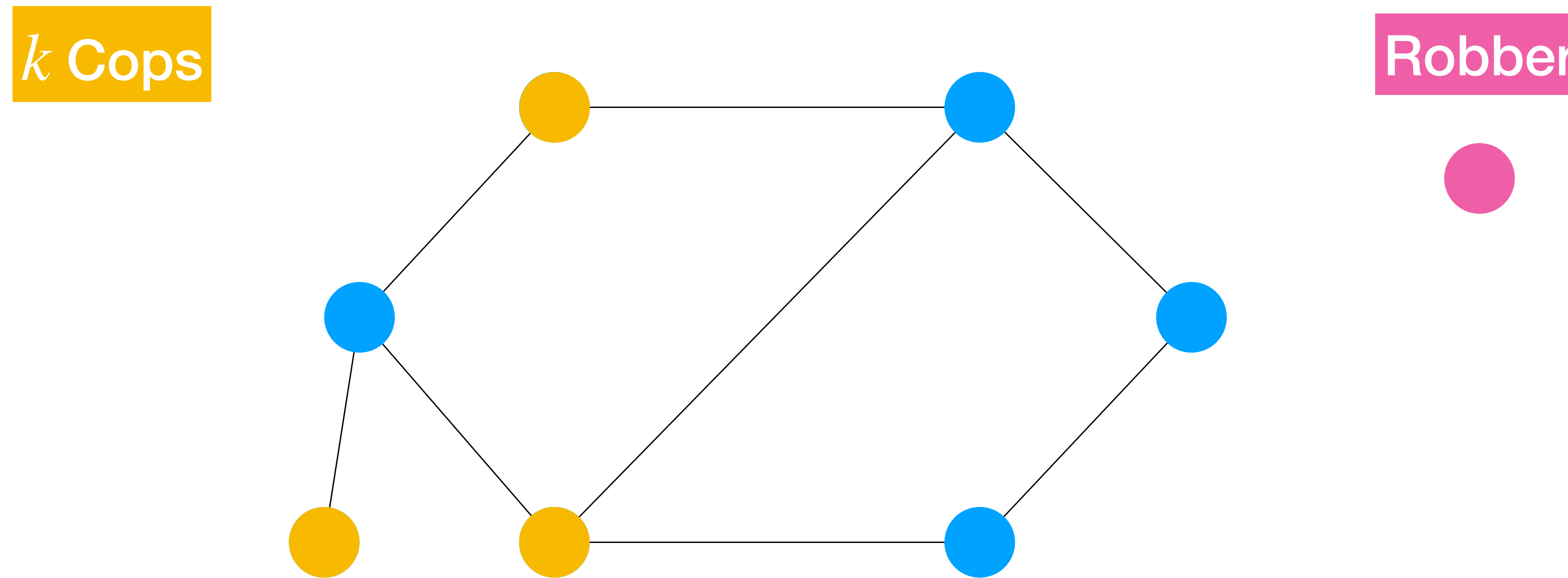
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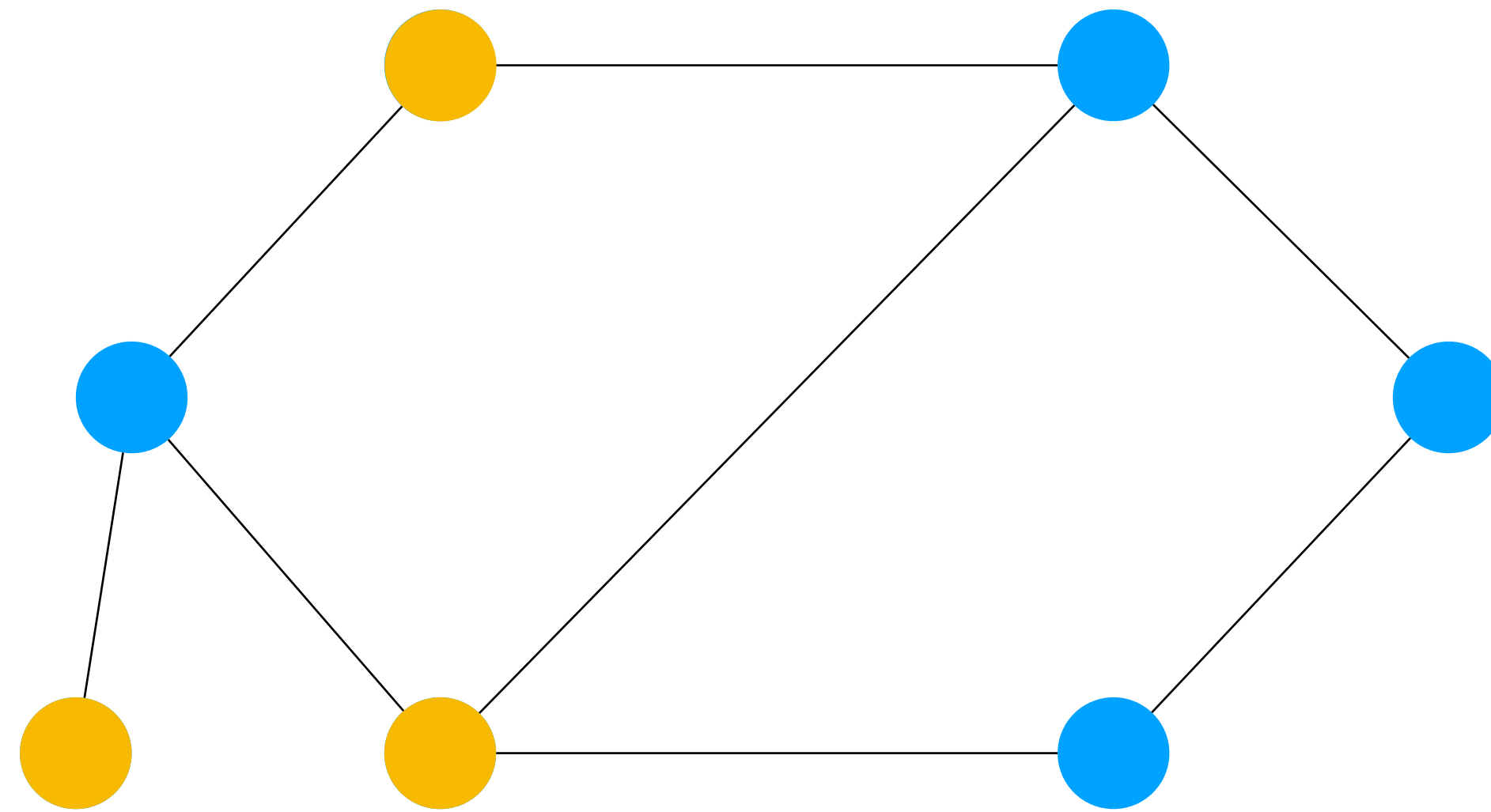
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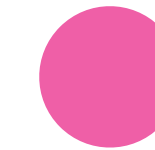
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Robber



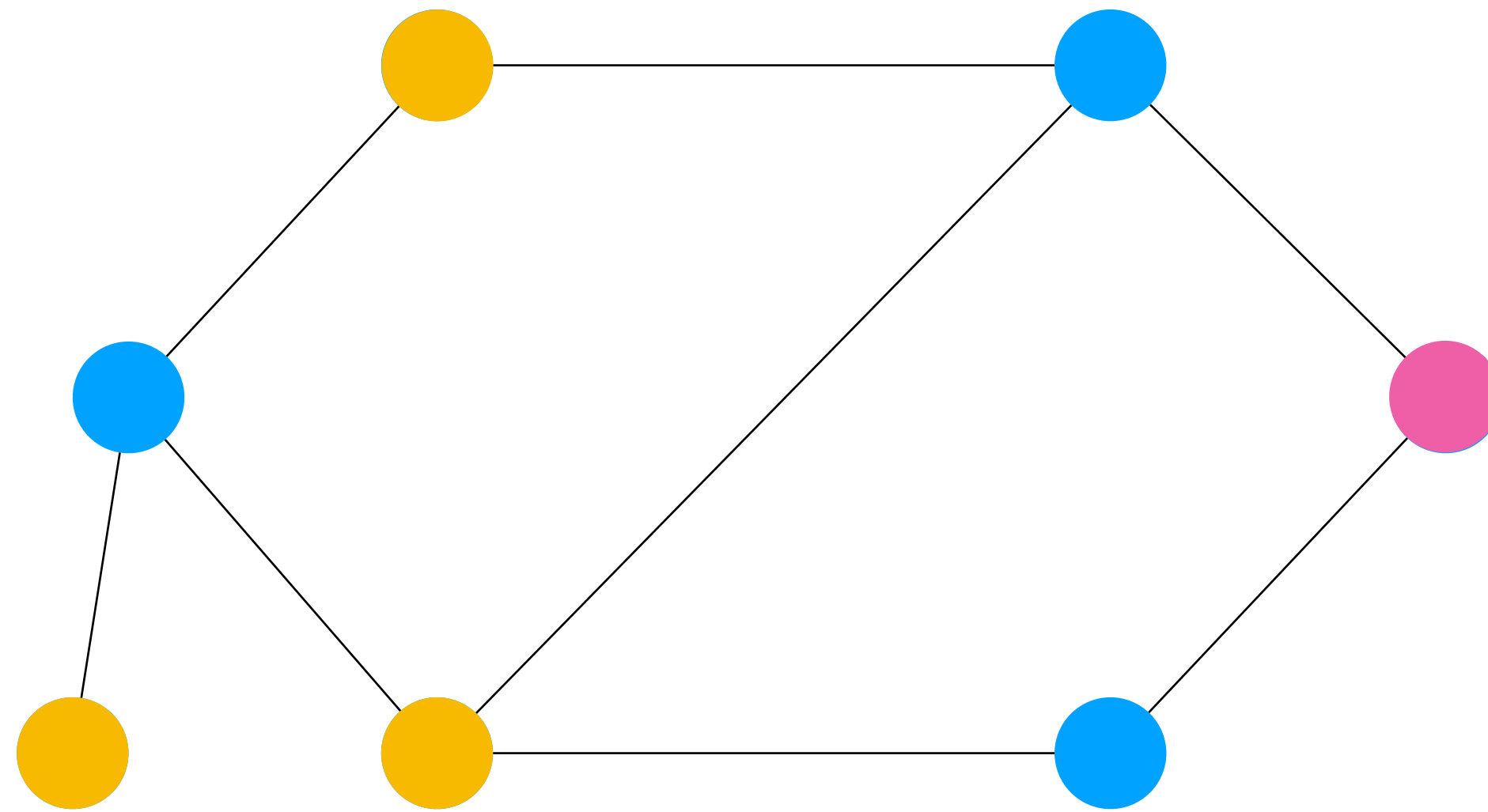
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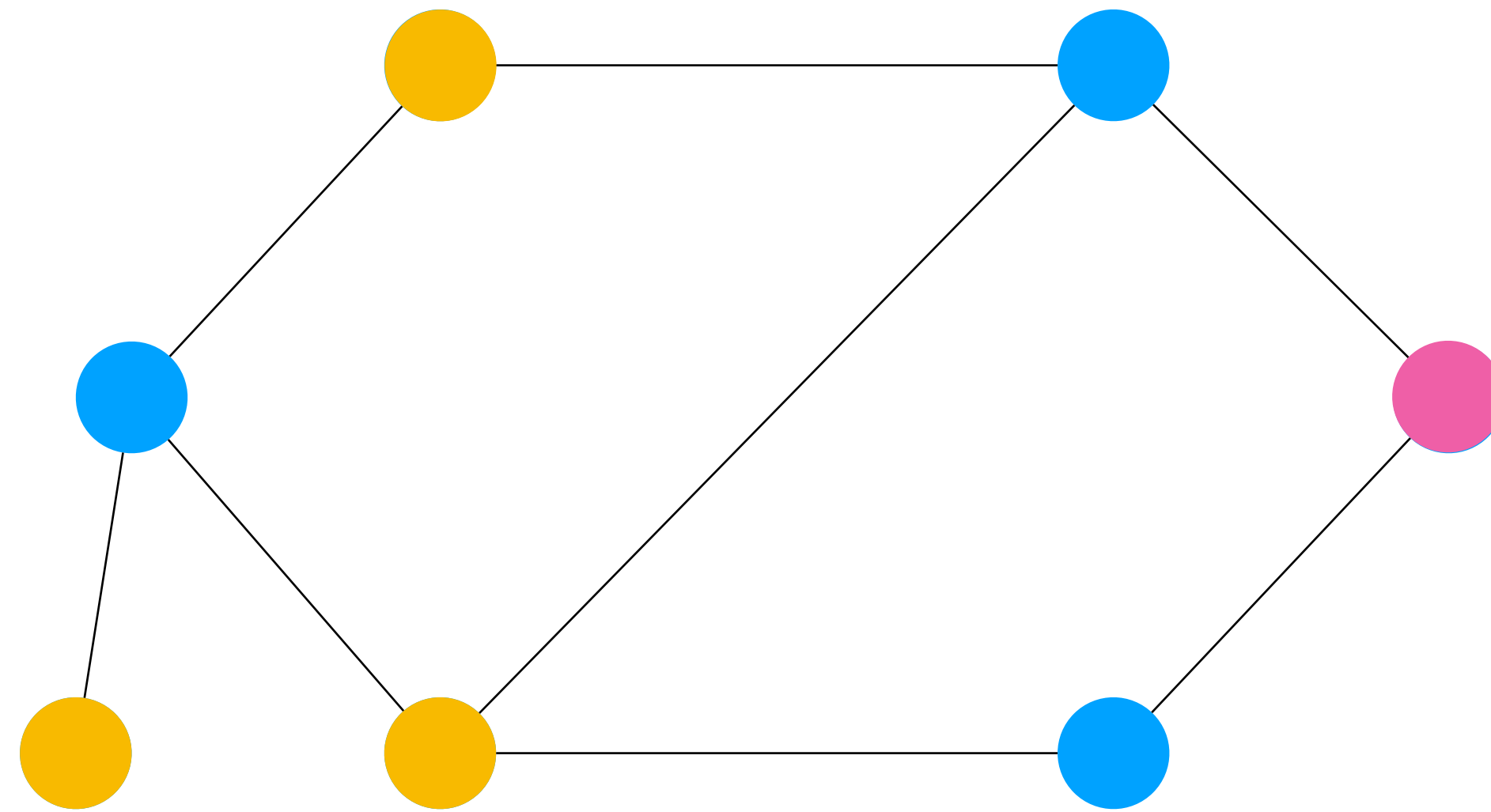


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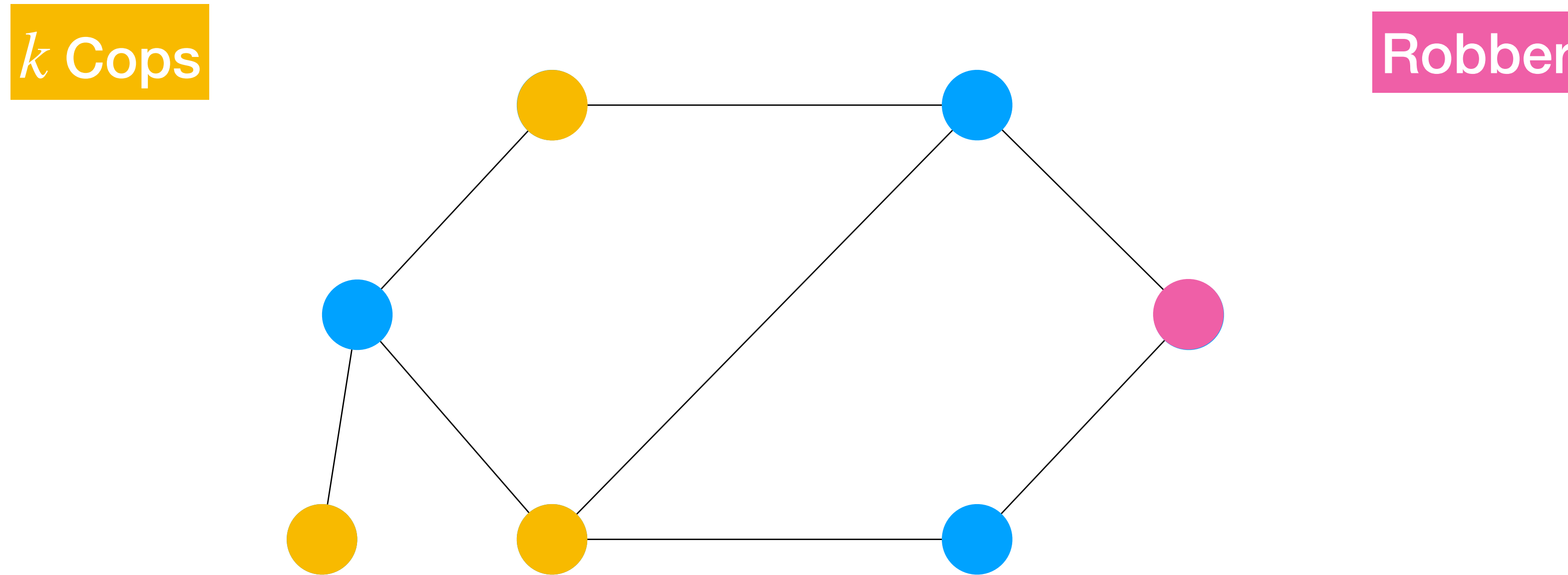
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Robber

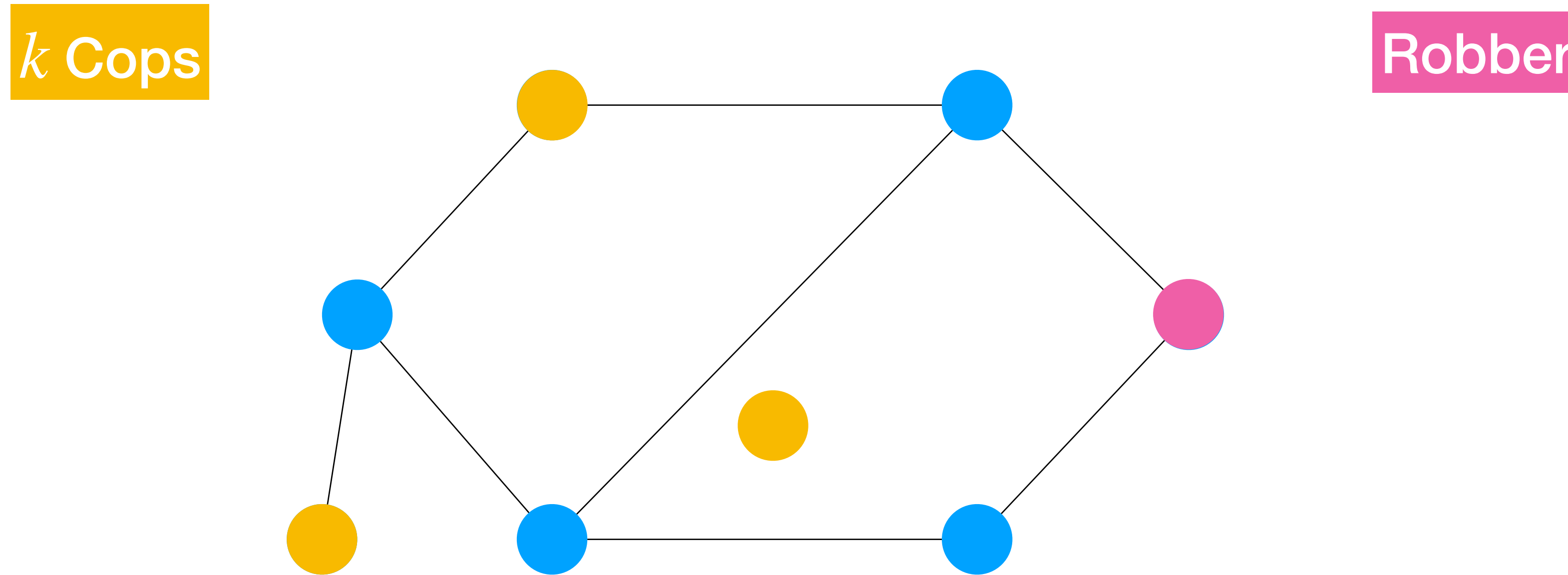


# Cops and Robbers



- Cops may move anywhere, but have to “**leave the graph**” to do so.

# Cops and Robbers

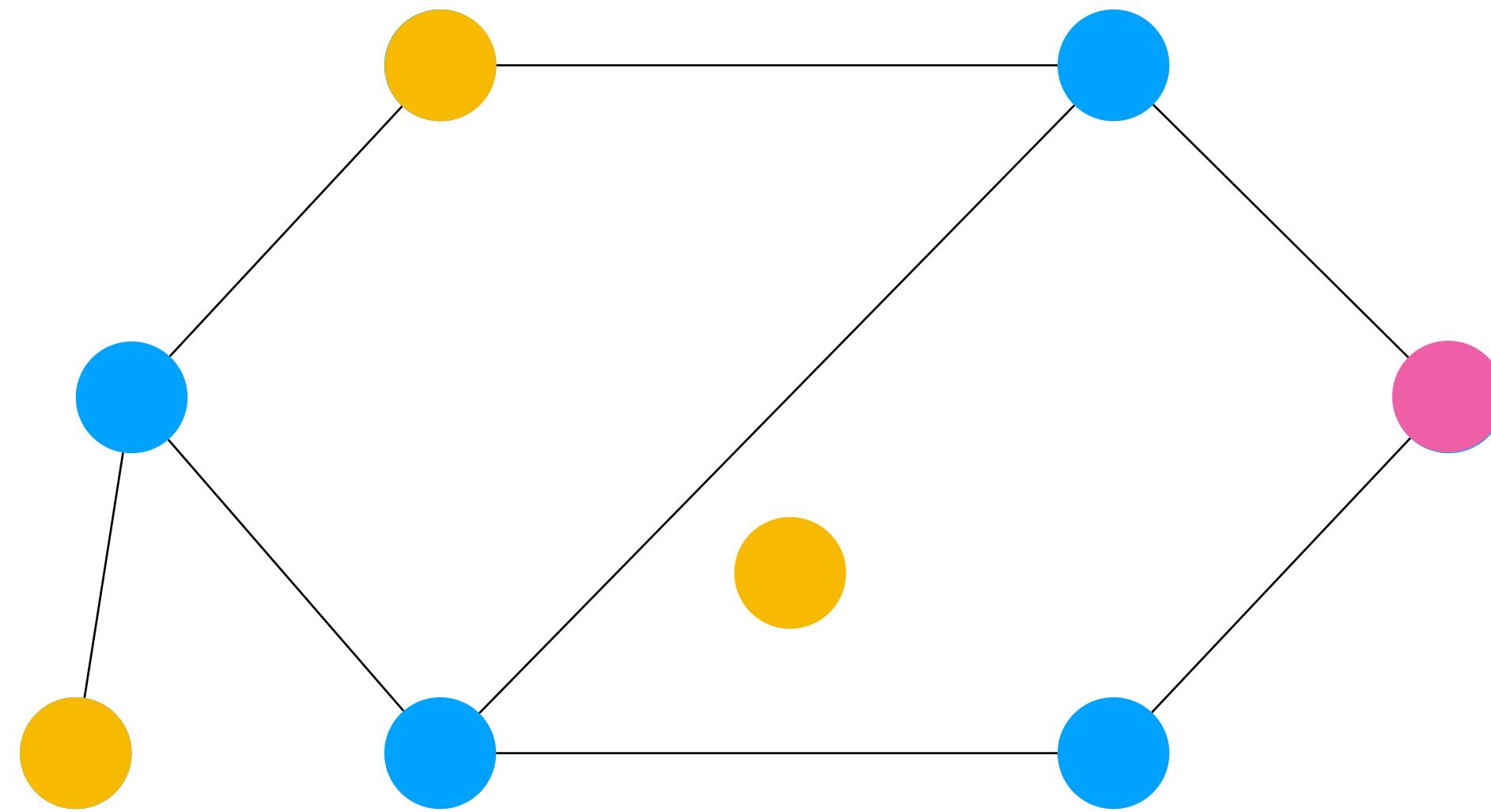


- Cops may move anywhere, but have to “**leave the graph**” to do so.

# Cops and Robbers

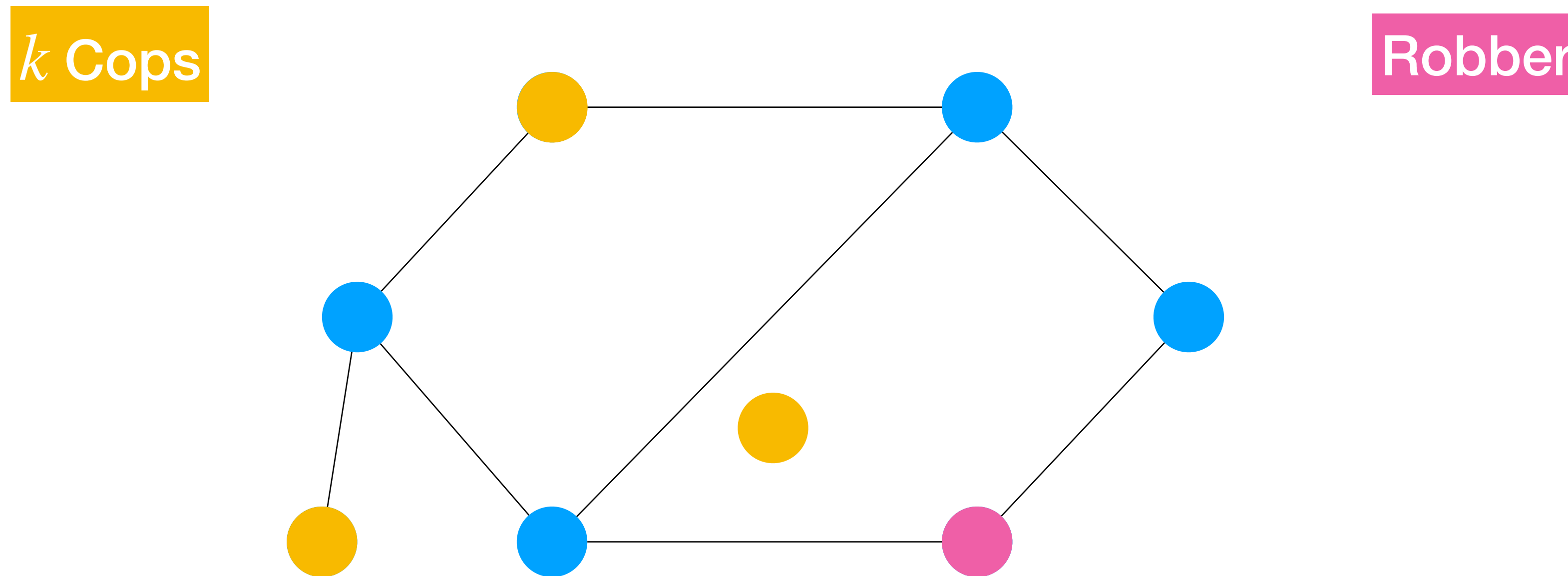
$k$  Cops

Robber



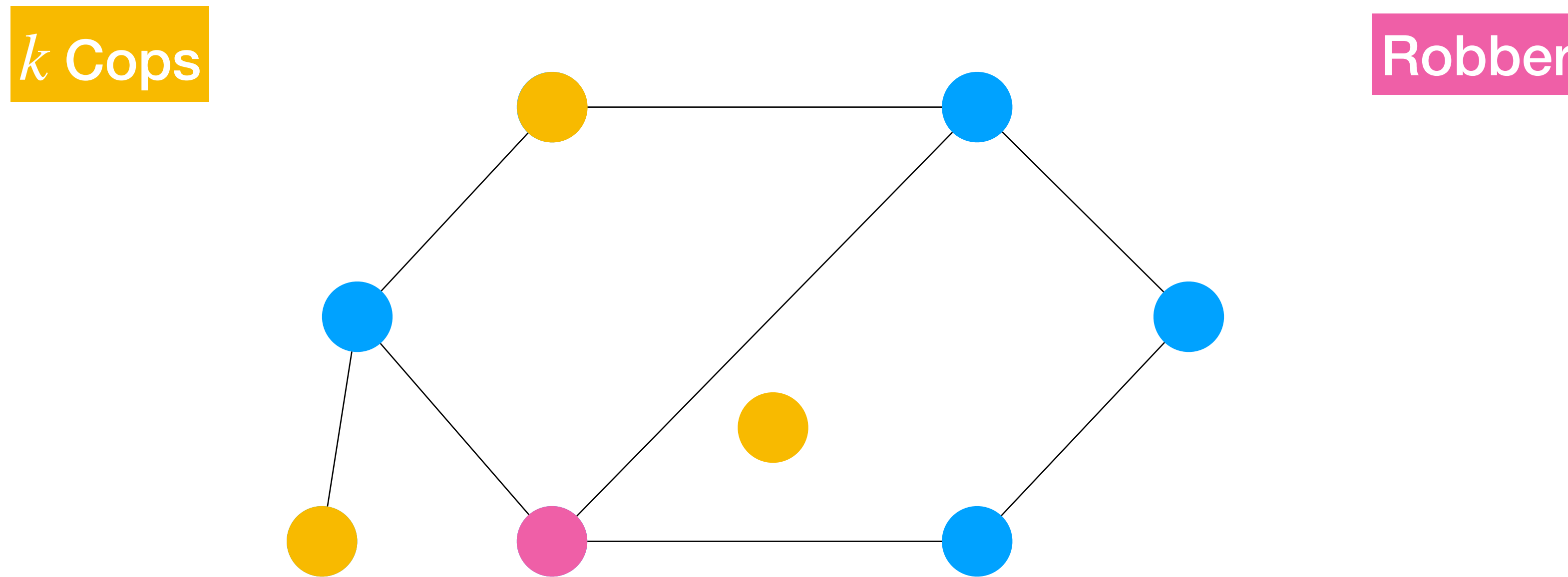
- Cops may move anywhere, but have to “leave the graph” to do so.
- Robber moves along the edges at “infinite speed”, must avoid cops.

# Cops and Robbers



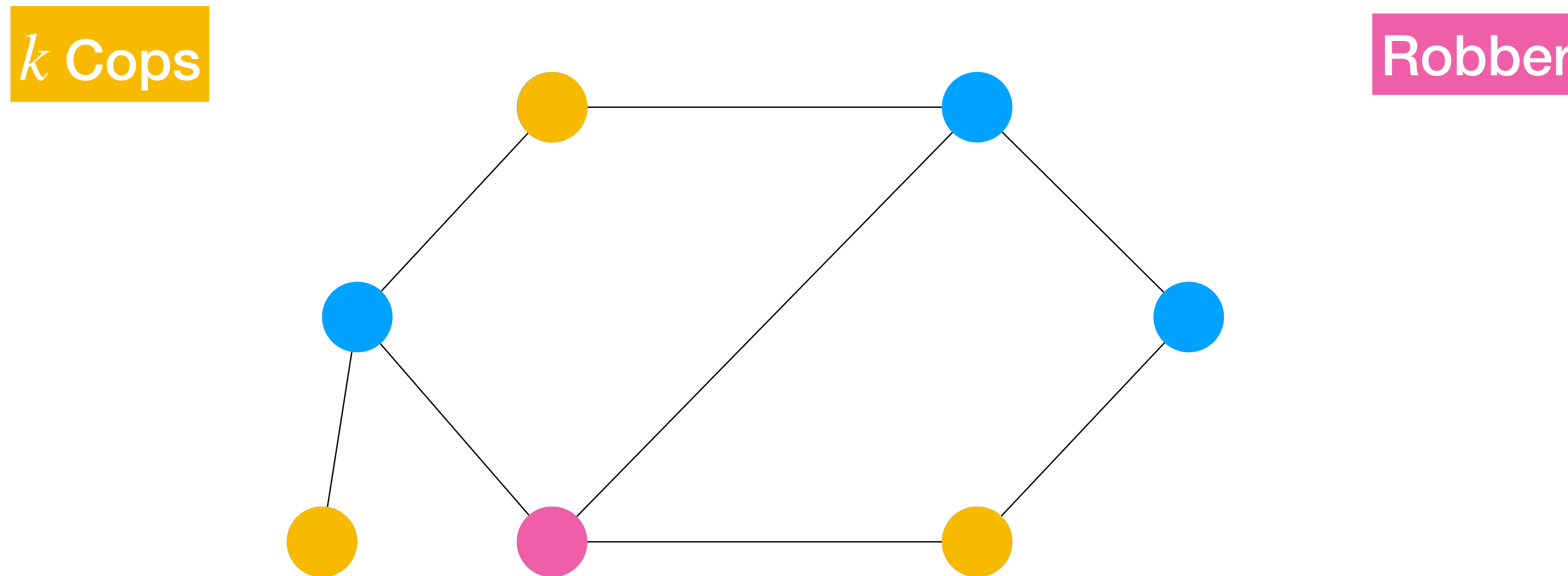
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# Cops and Robbers



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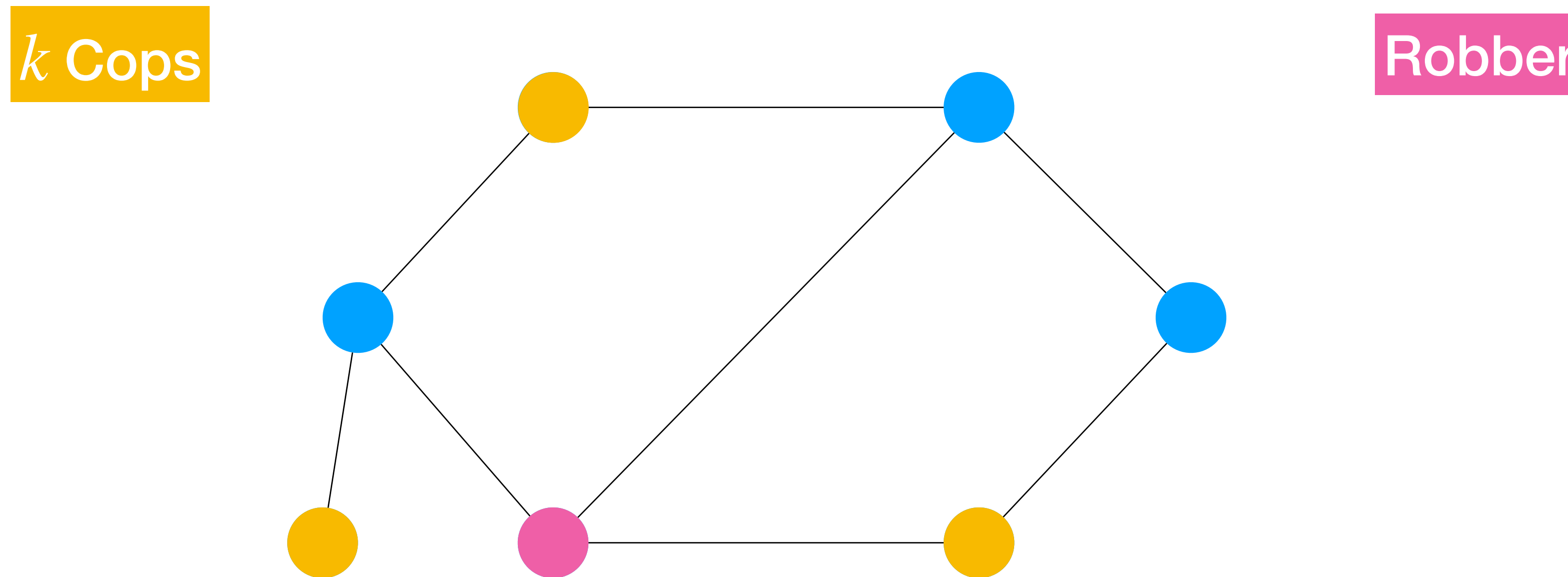
# Cops and Robbers



- Cops may move anywhere, but have to “leave the graph” to do so.
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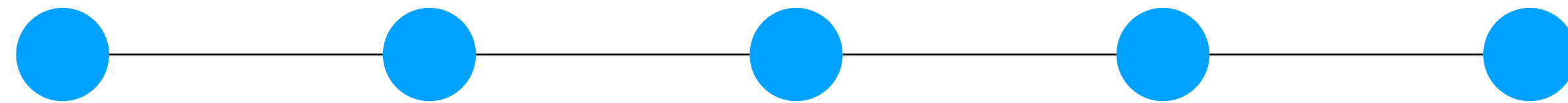
# Cops and Robbers



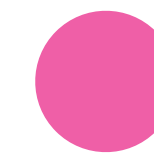
**Cops win the game if the robber is caught.**

- Cops may move anywhere, but have to “**leave the graph**” to do so.
- Robber moves along the edges at “**infinite speed**”, must avoid cops.

# Cops and Robbers

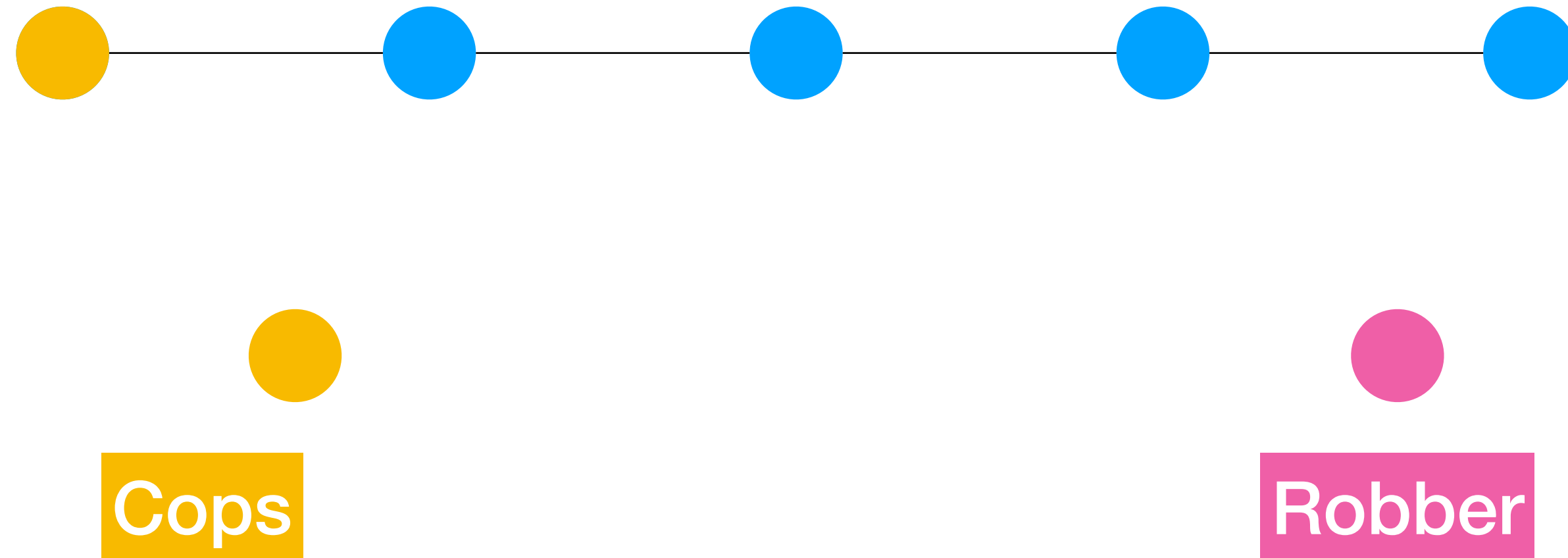


Cops

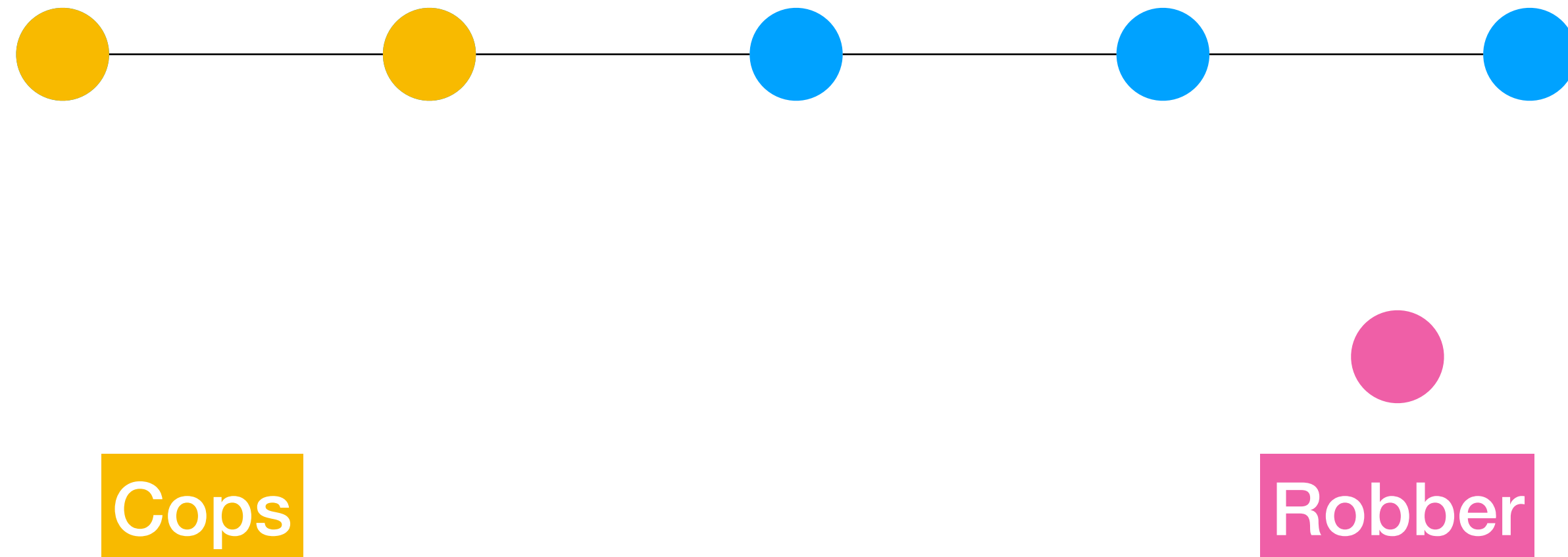


Robber

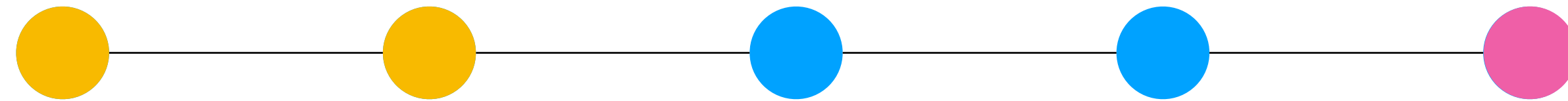
# Cops and Robbers



# Cops and Robbers



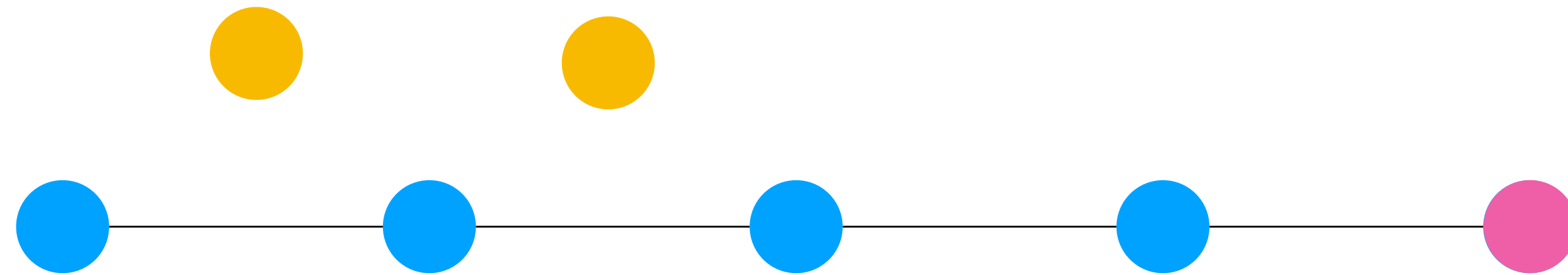
# Cops and Robbers



Cops

Robber

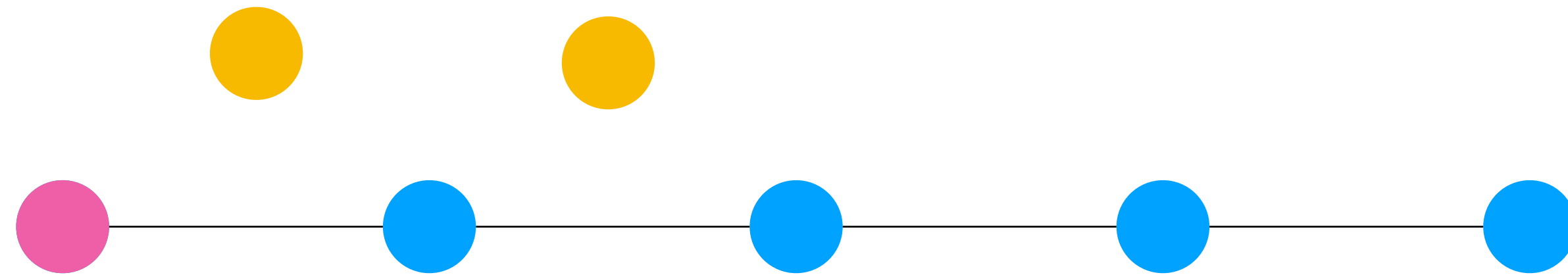
# Cops and Robbers



Cops

Robber

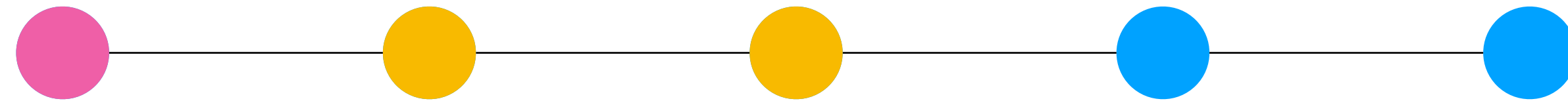
# Cops and Robbers



Cops

Robber

# Cops and Robbers

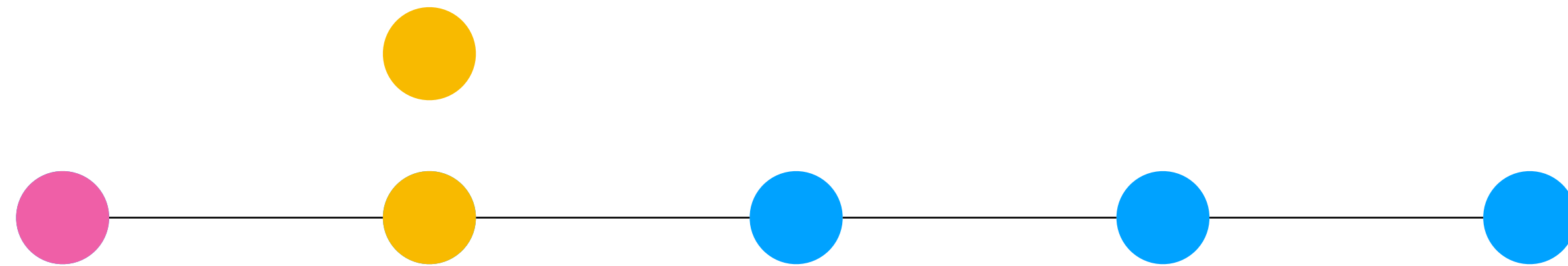


Cops

Robber



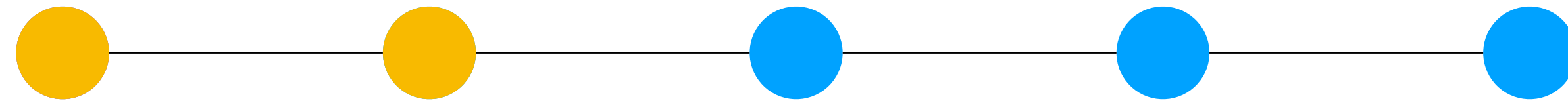
# Cops and Robbers



Cops

Robber

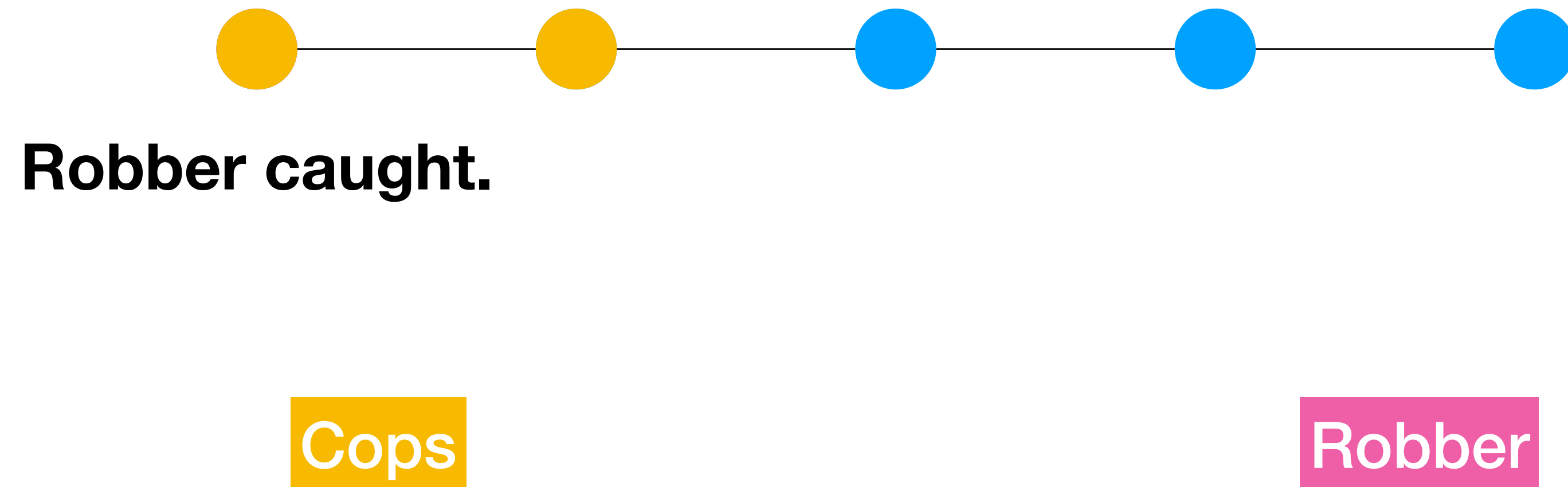
# Cops and Robbers



Cops

Robber

# Cops and Robbers

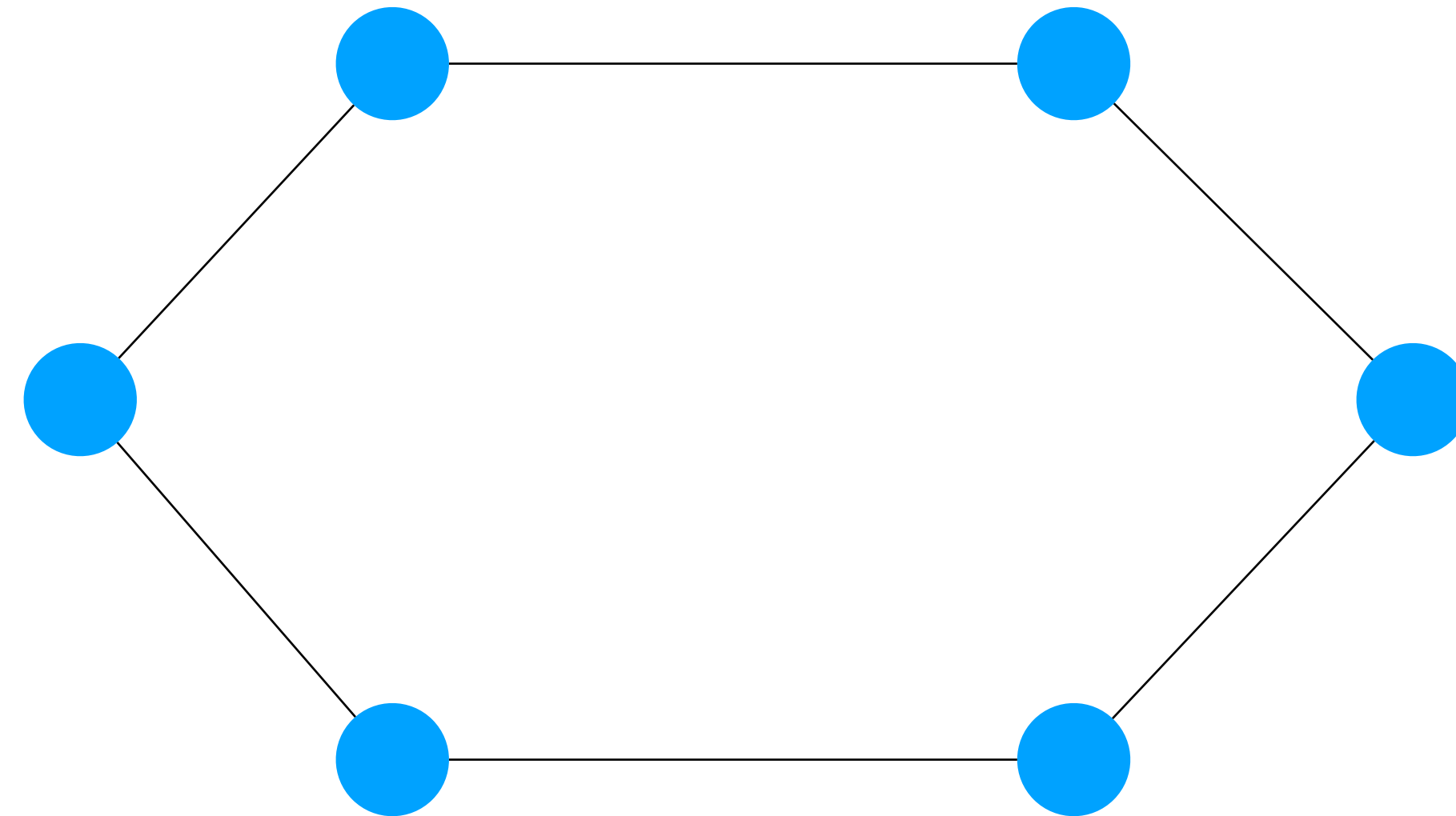


# Another Definition of Treewidth

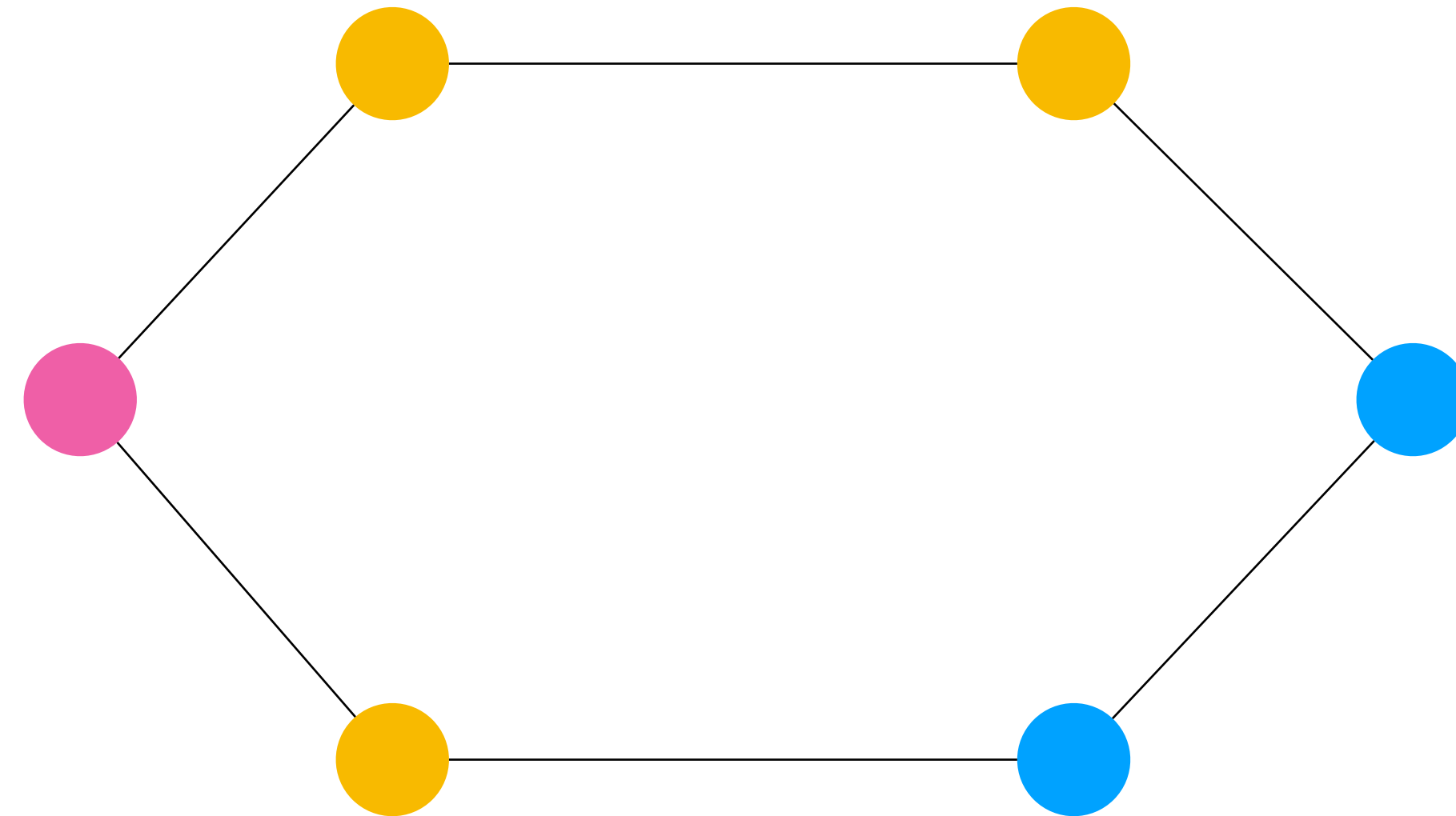
## Fact

The **treewidth** of a graph  $G$  is the minimum  $k$  such that  $k + 1$  cops have a strategy to catch a robber in  $G$ .

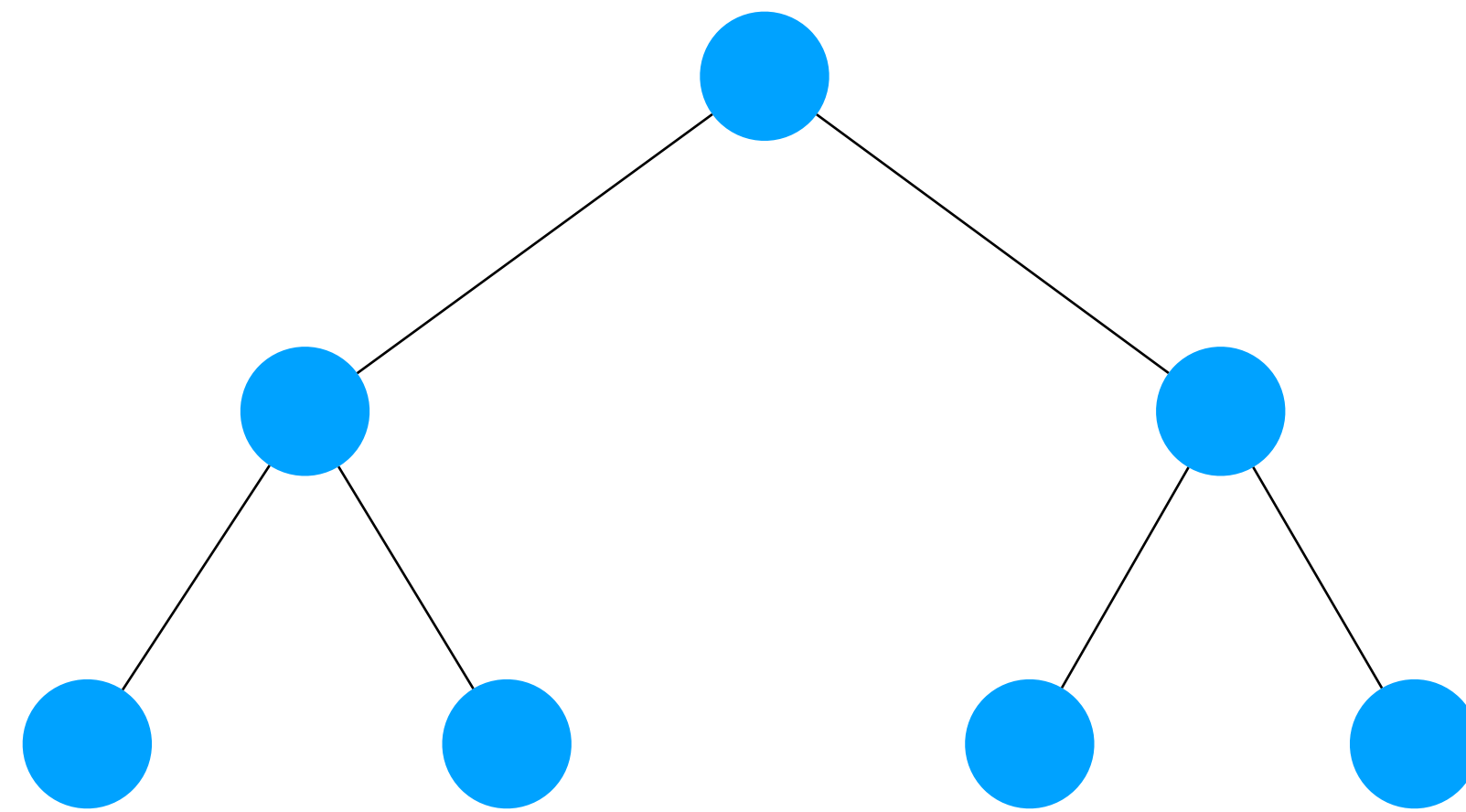
# Example



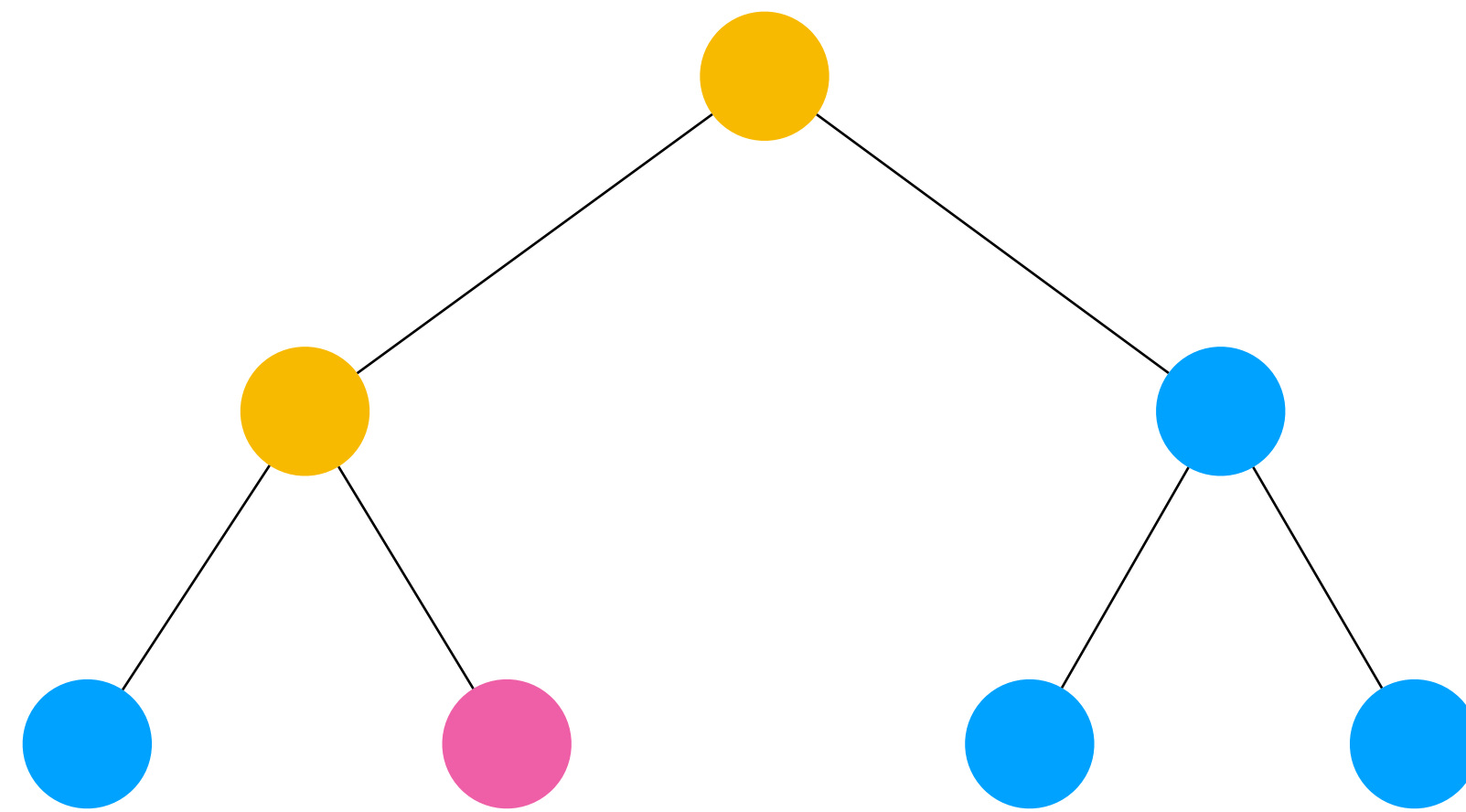
# Example



# Example

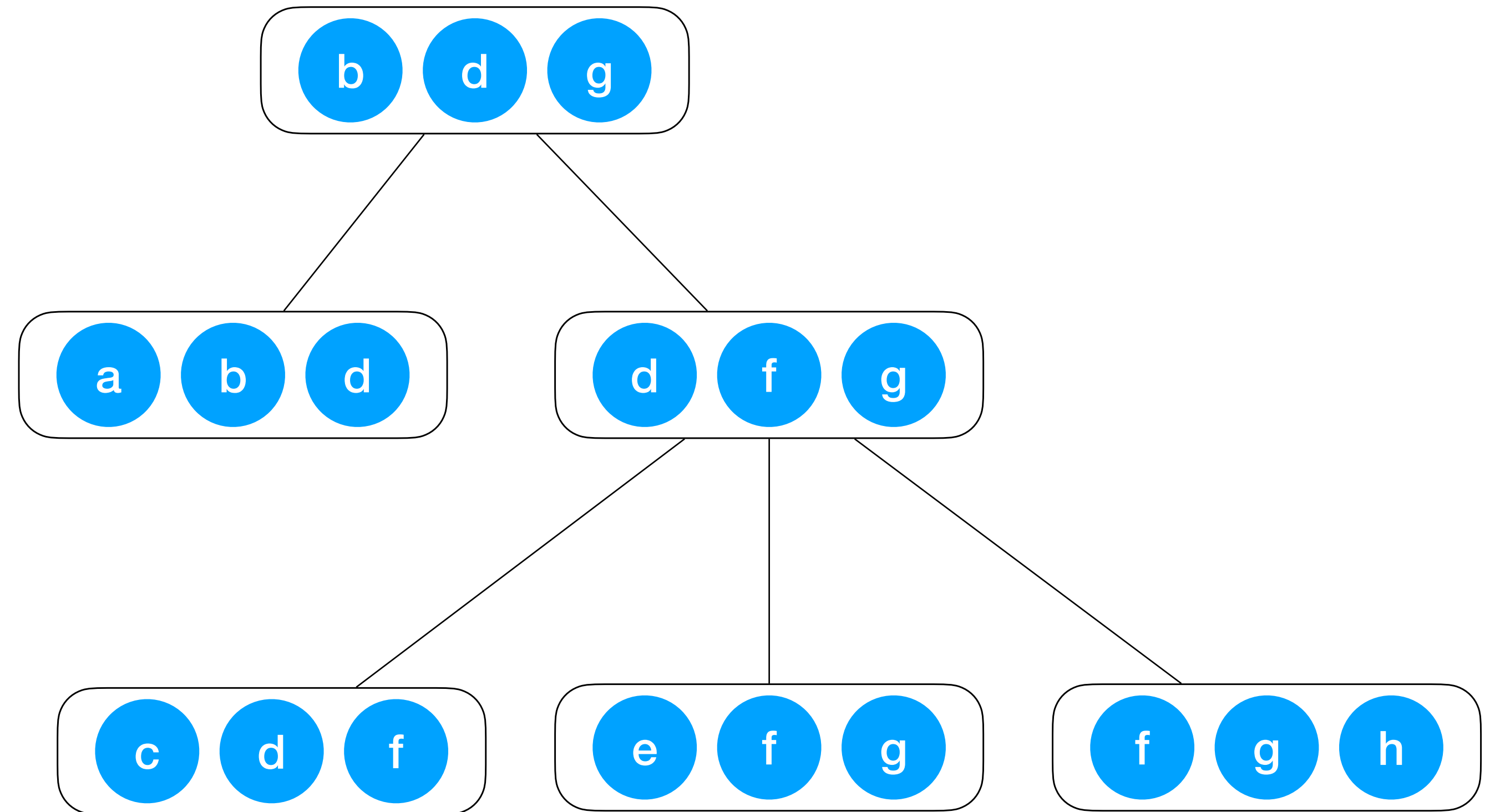
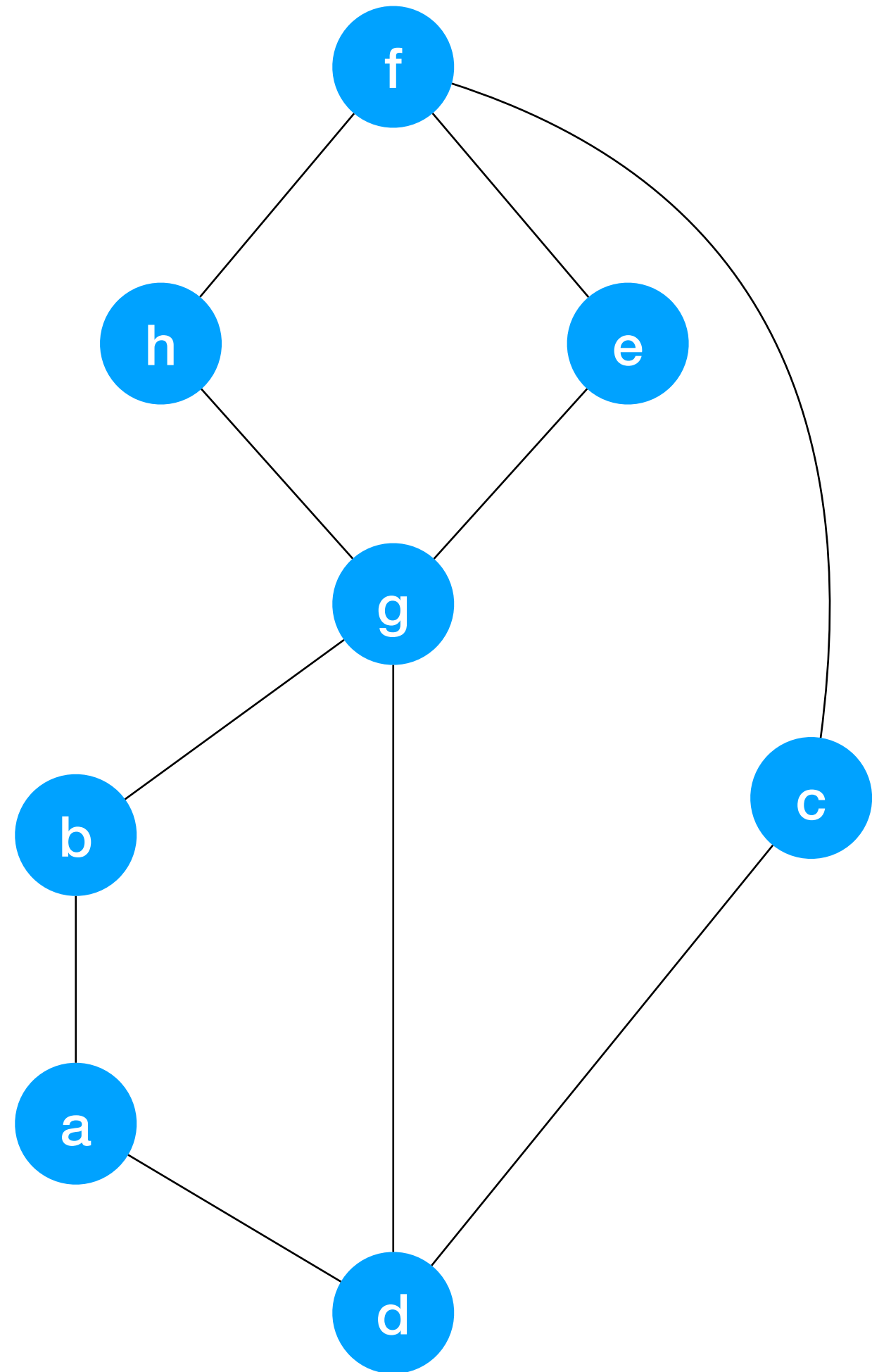


# Example

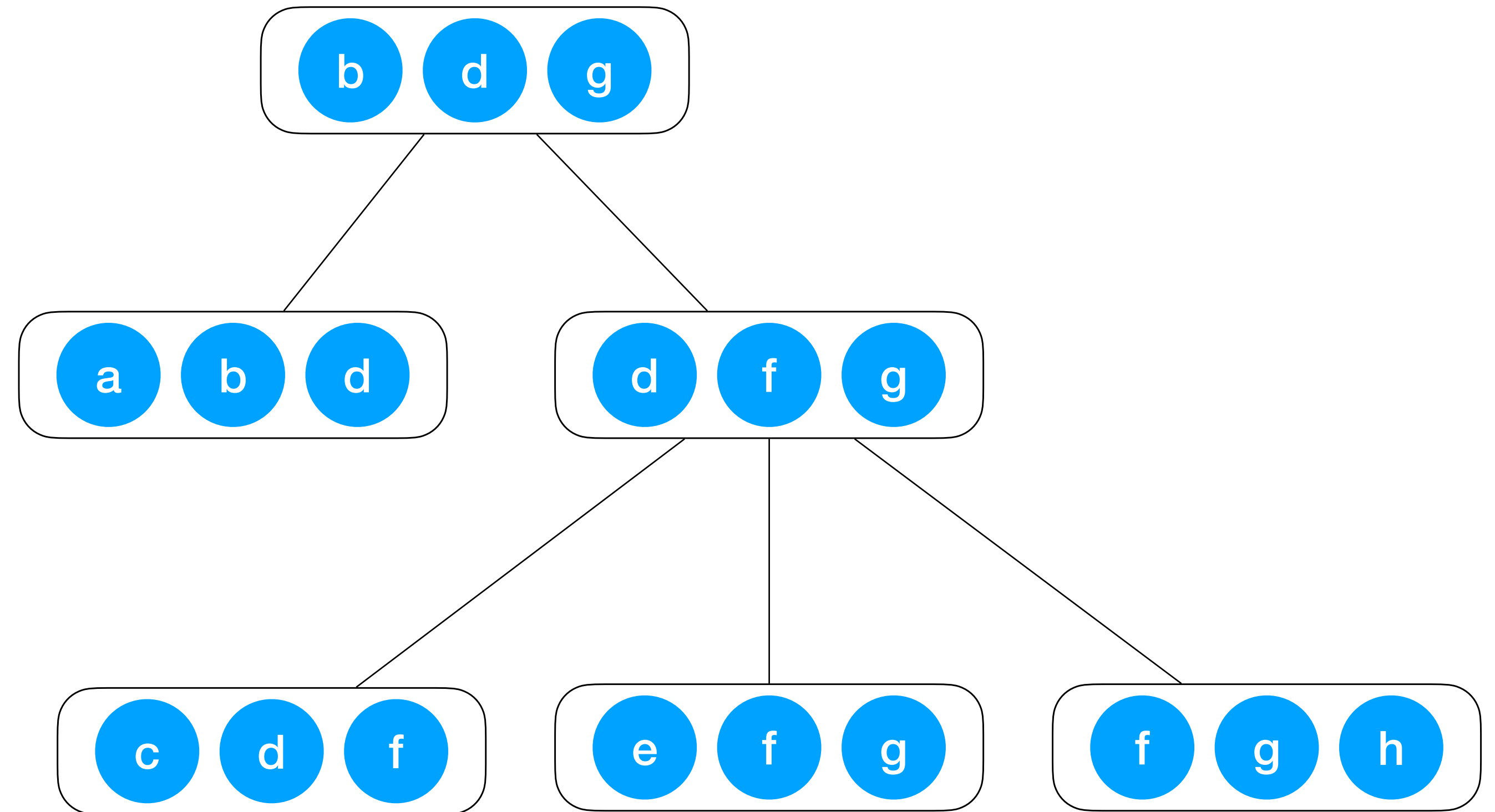
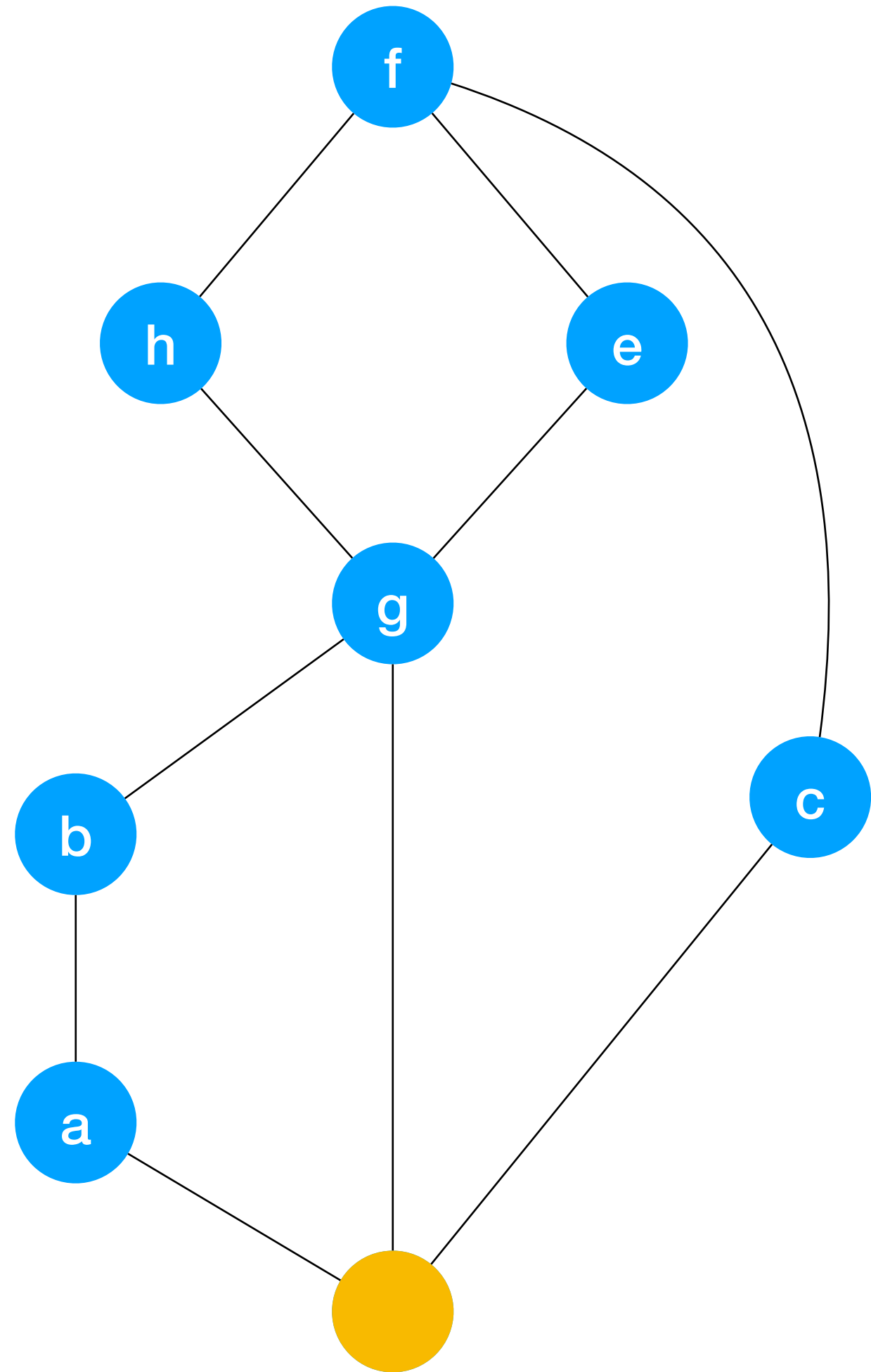




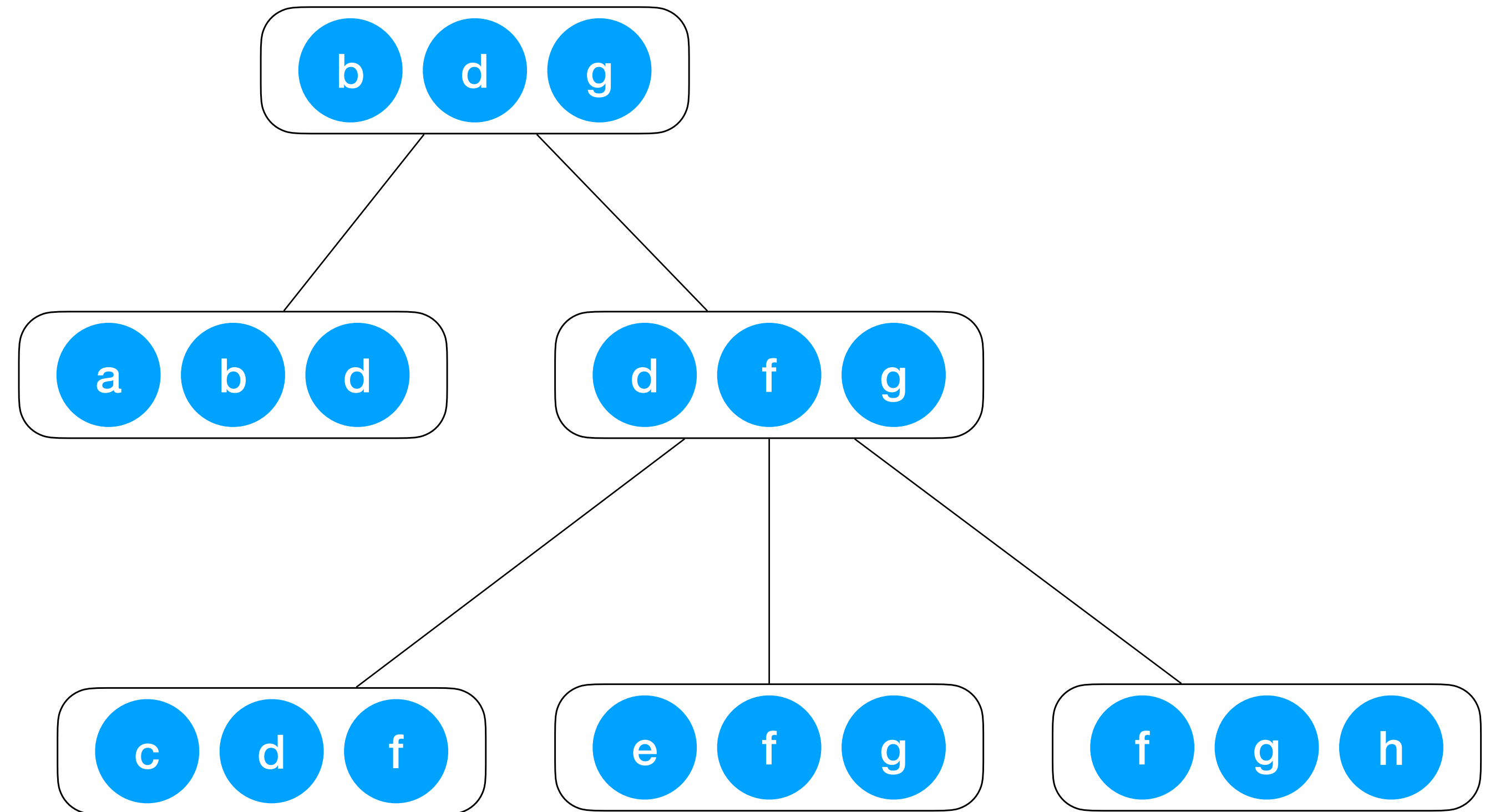
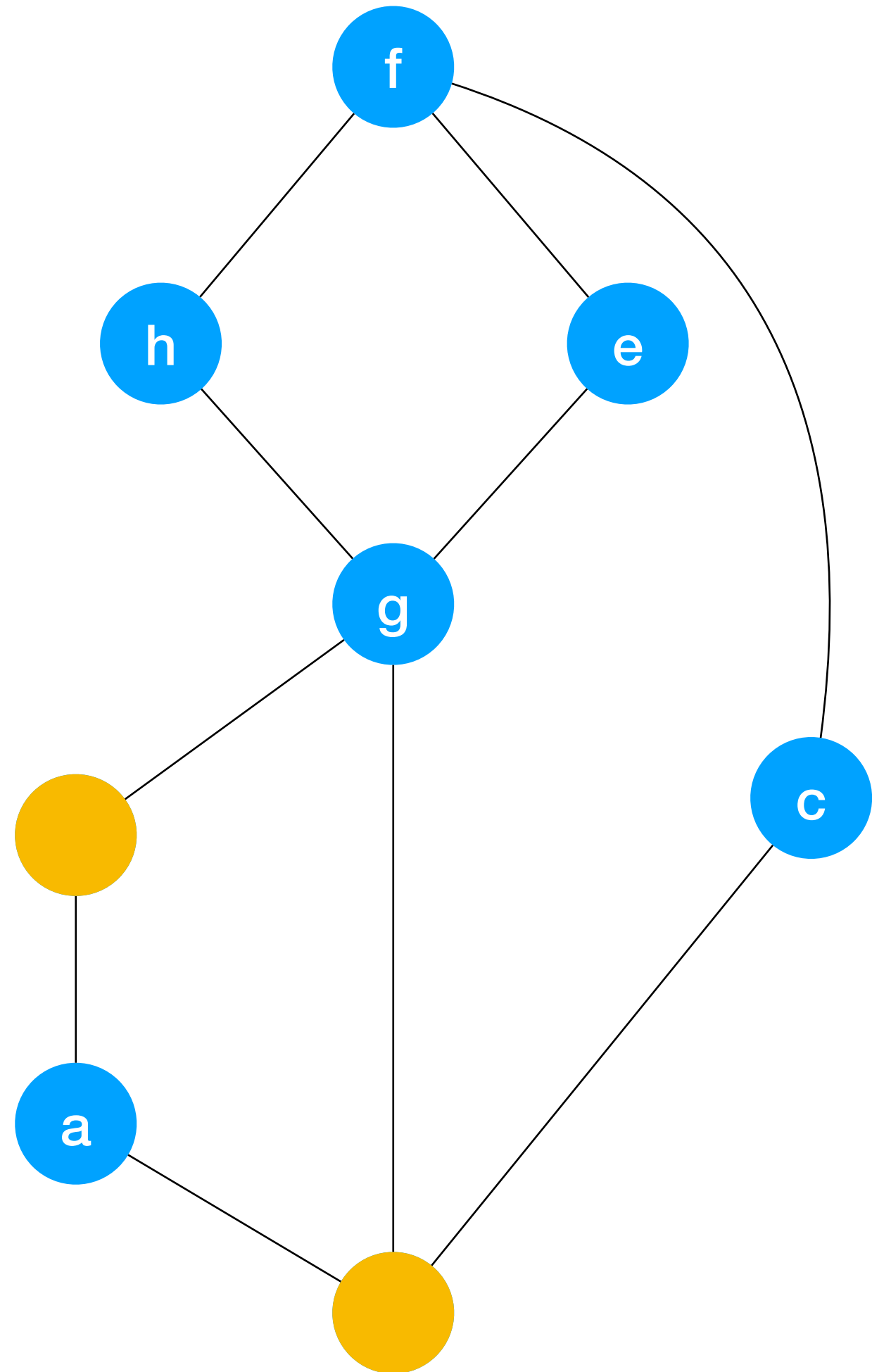
# Example



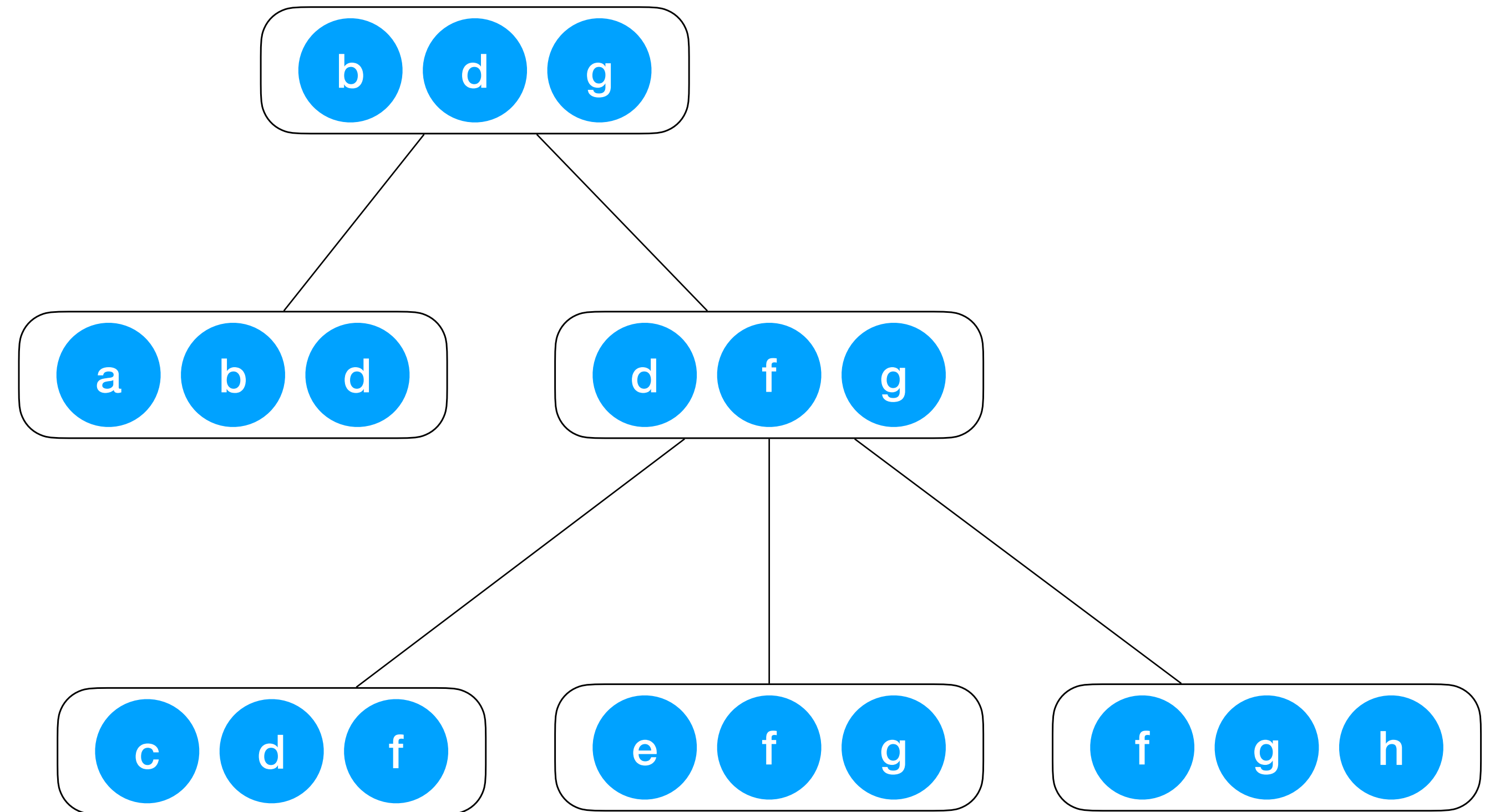
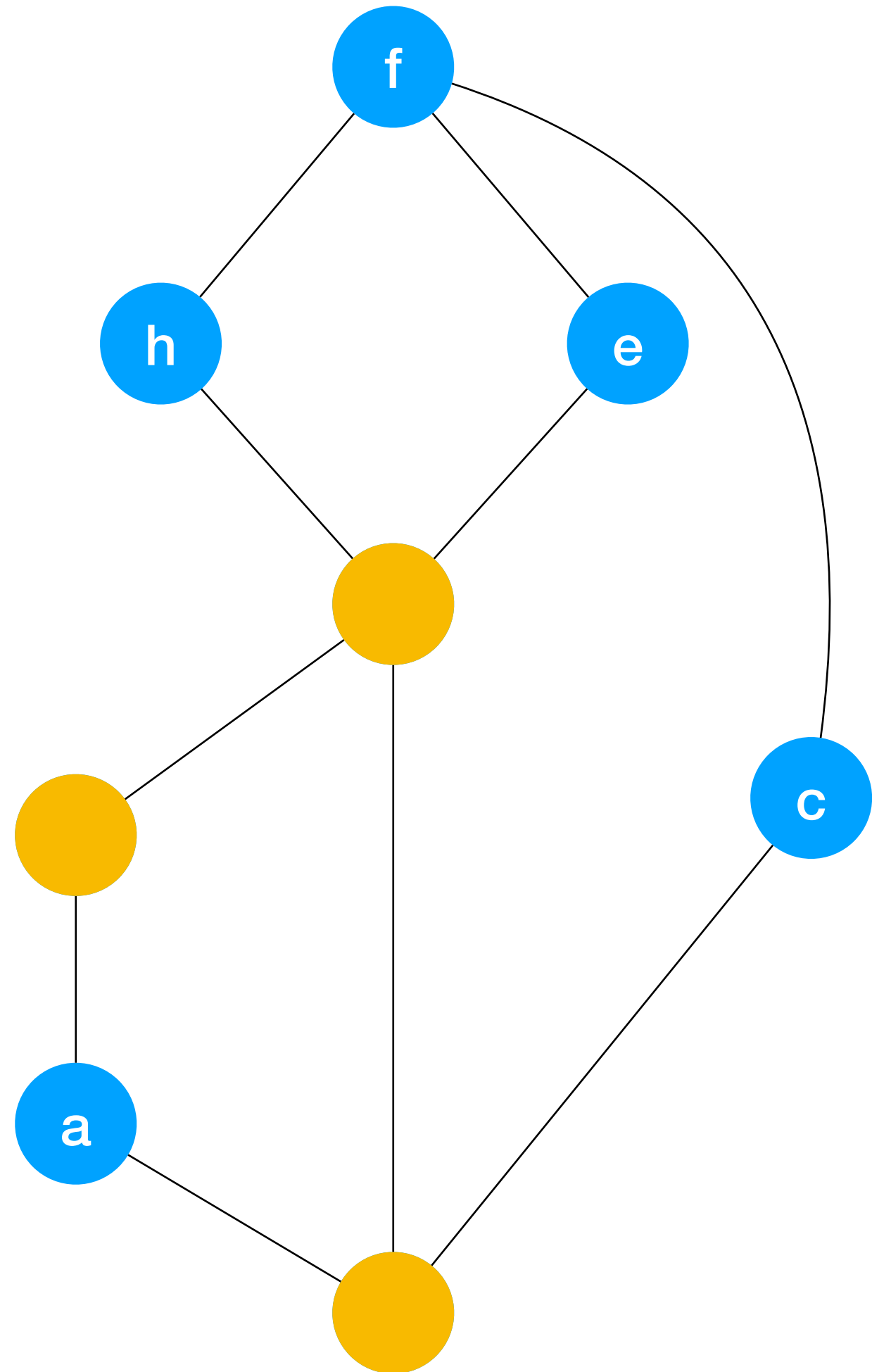
# Example



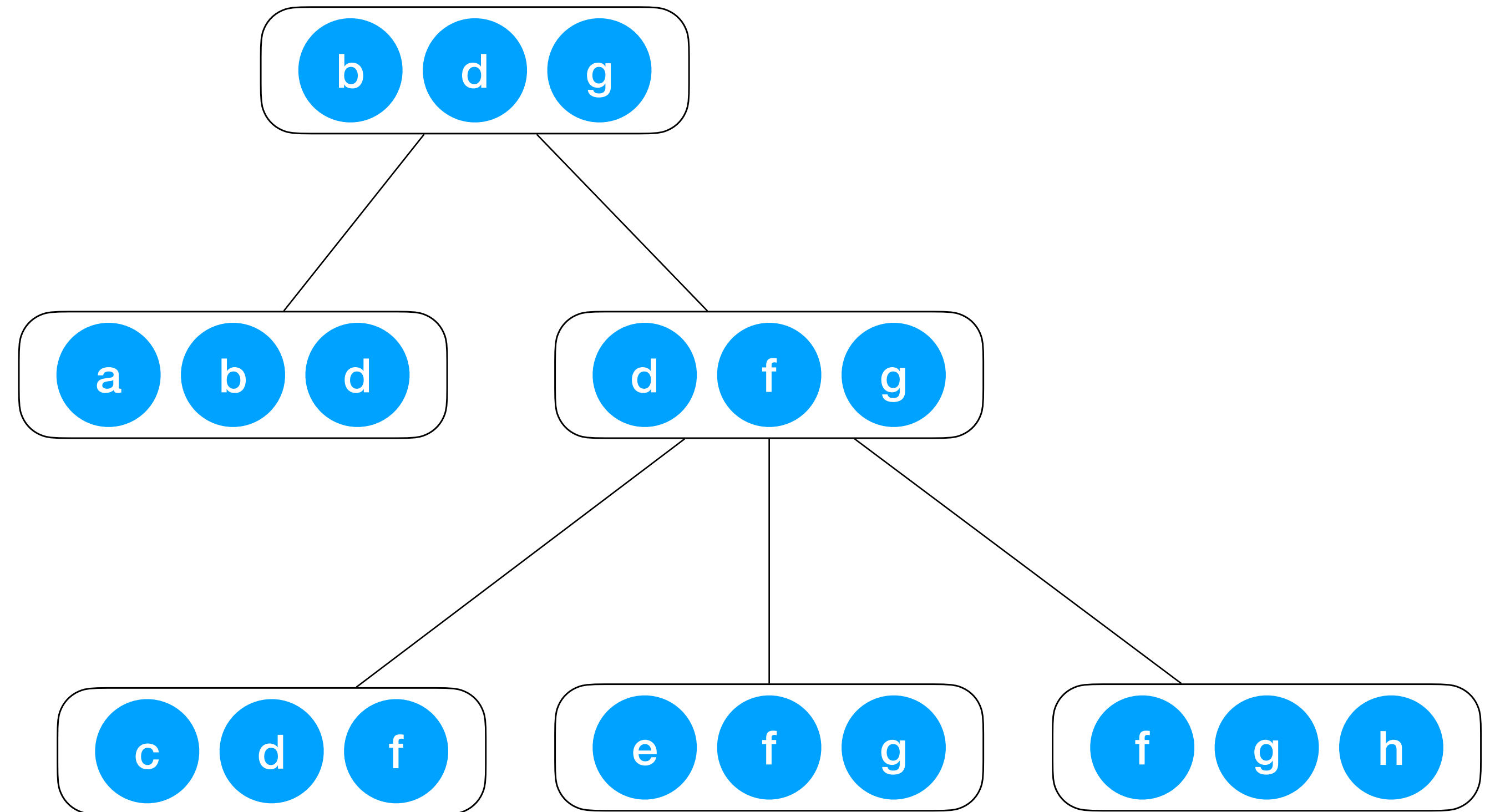
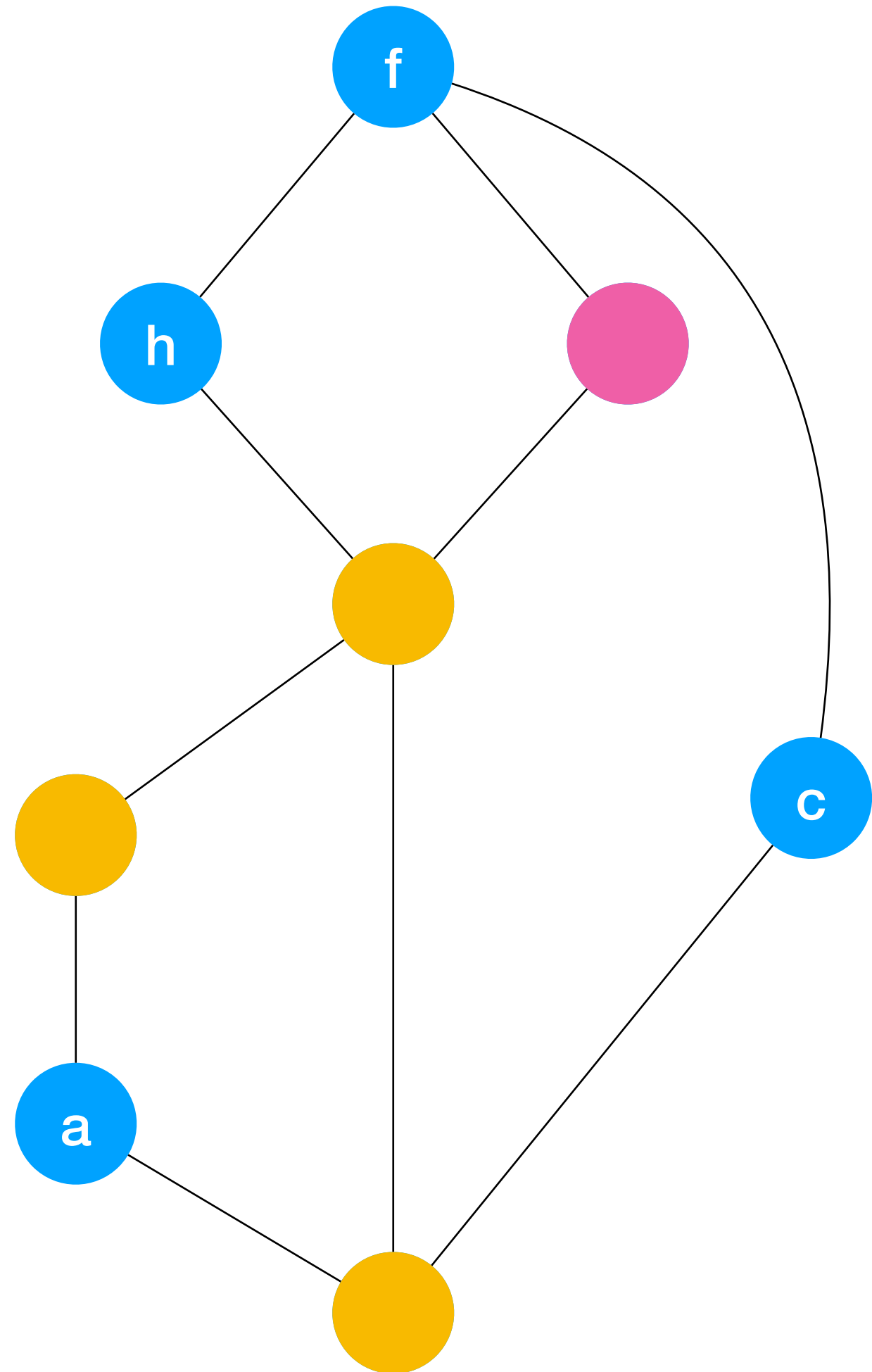
# Example



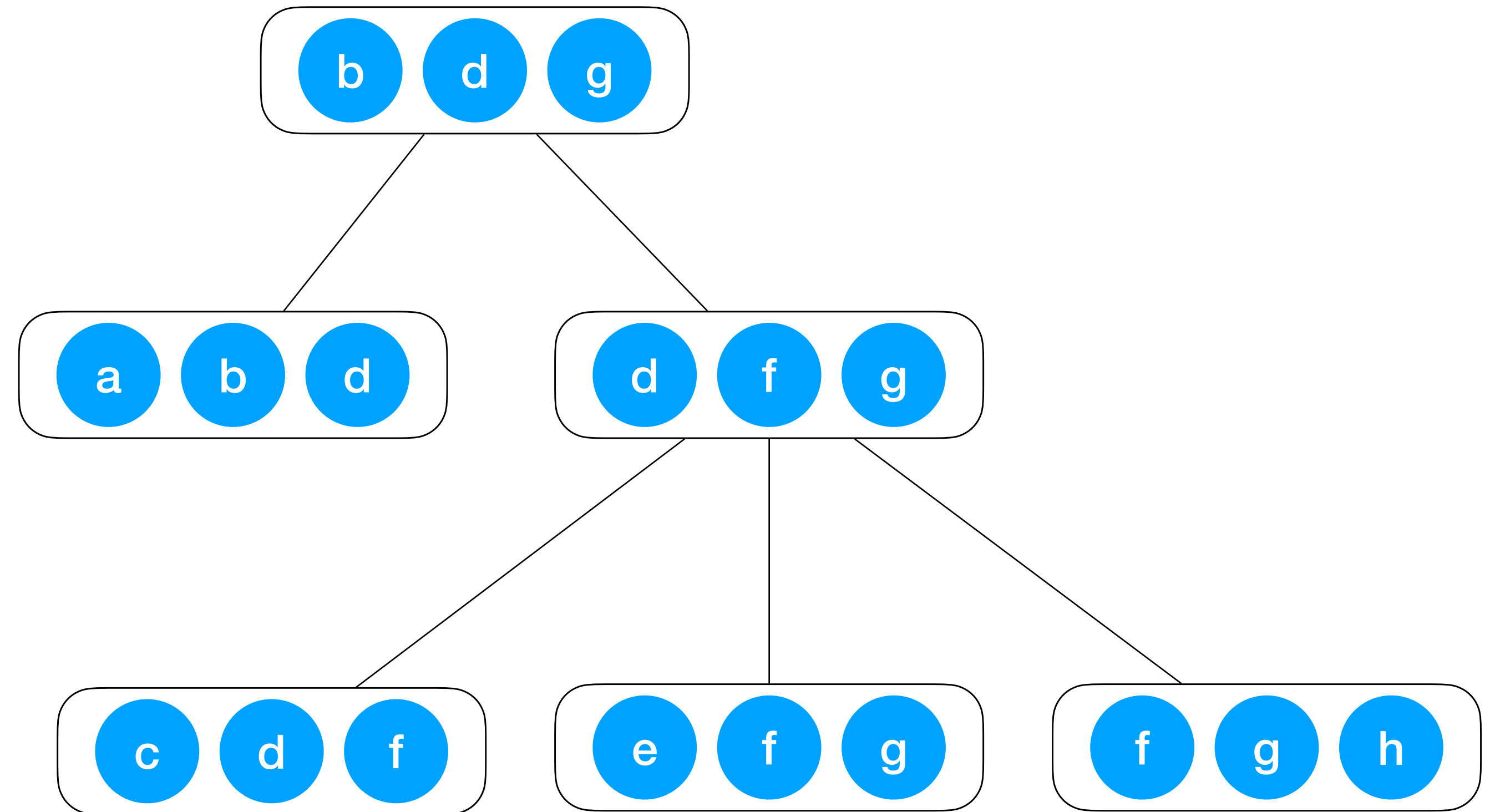
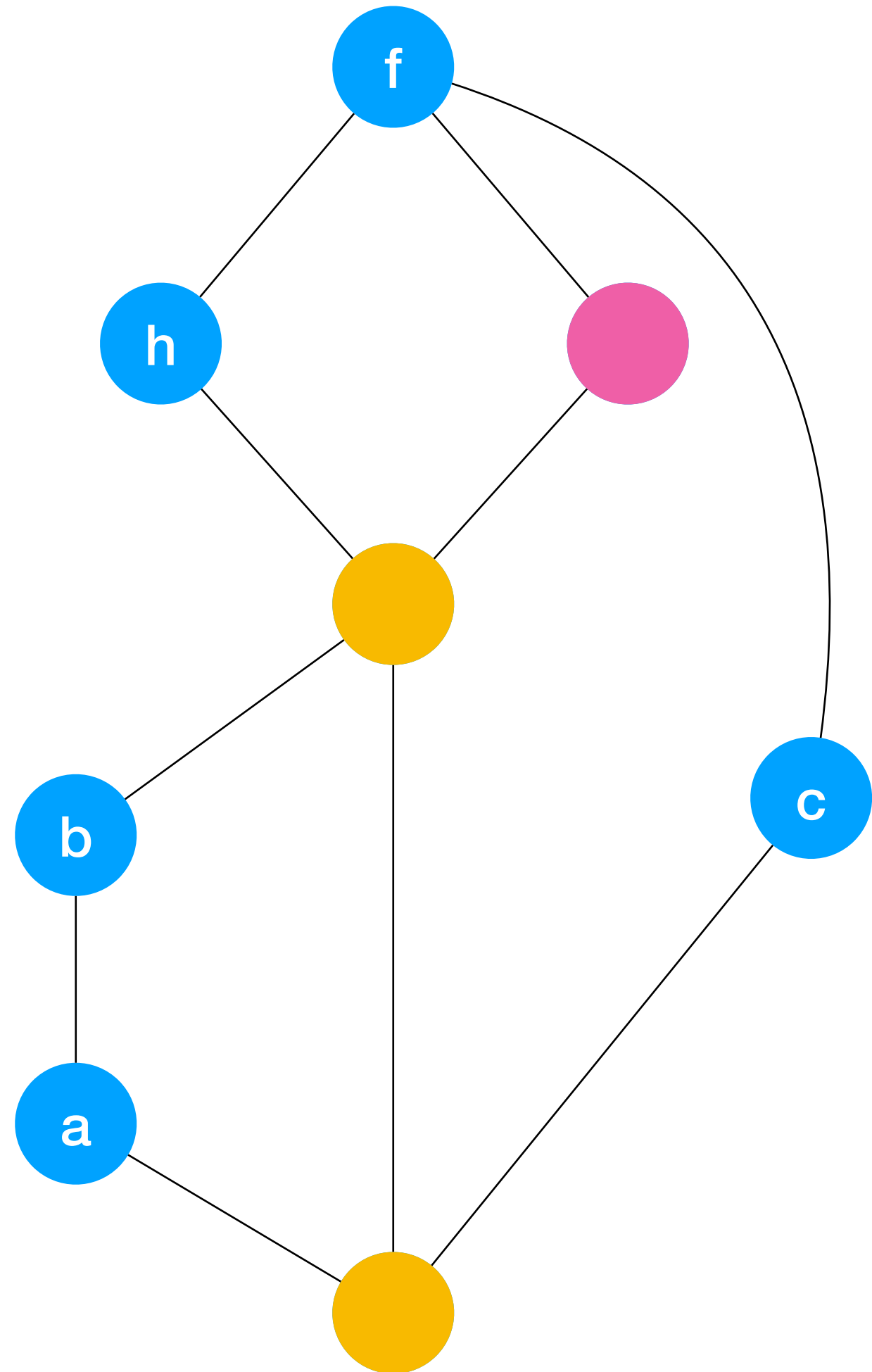
# Example



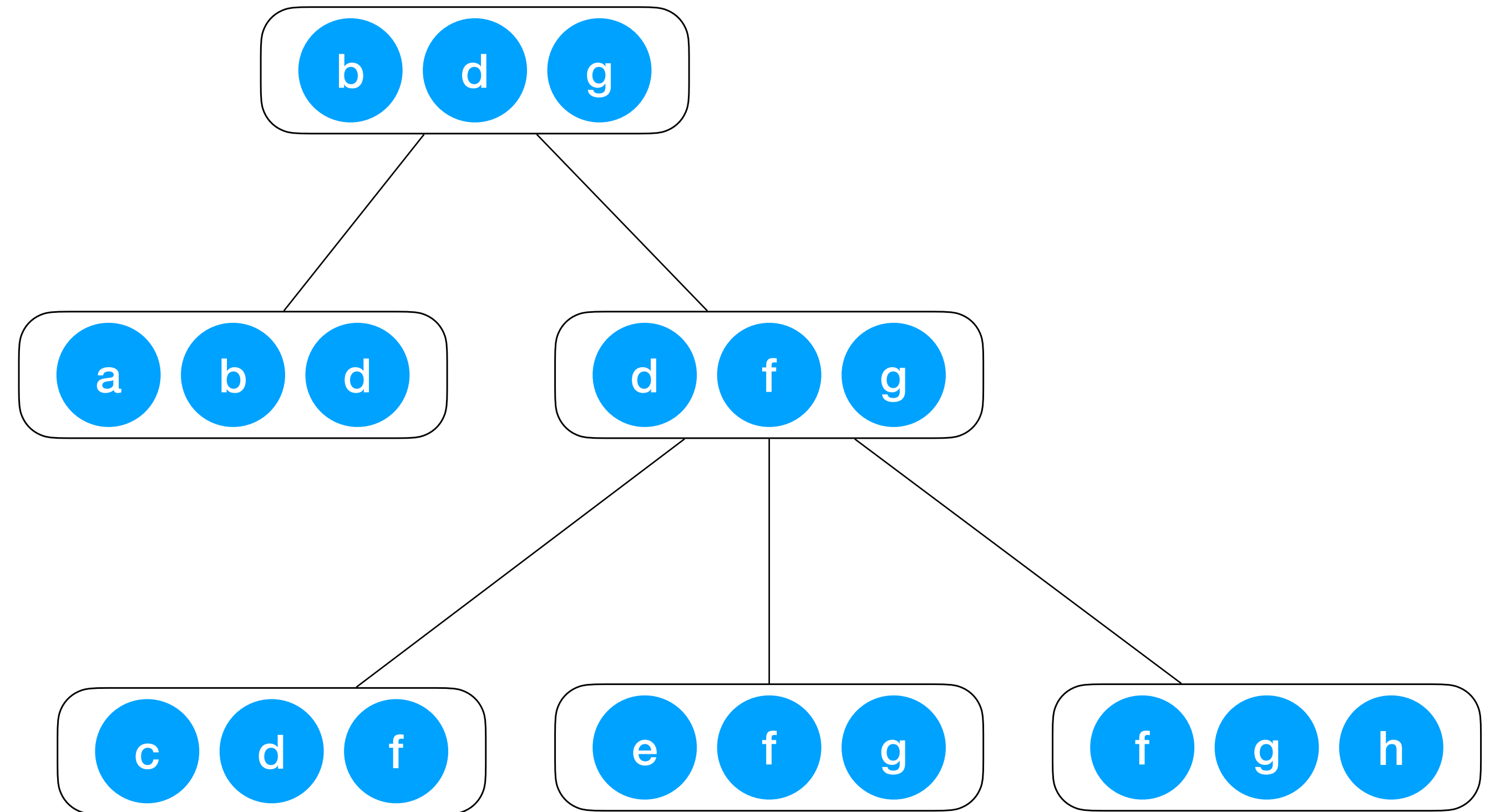
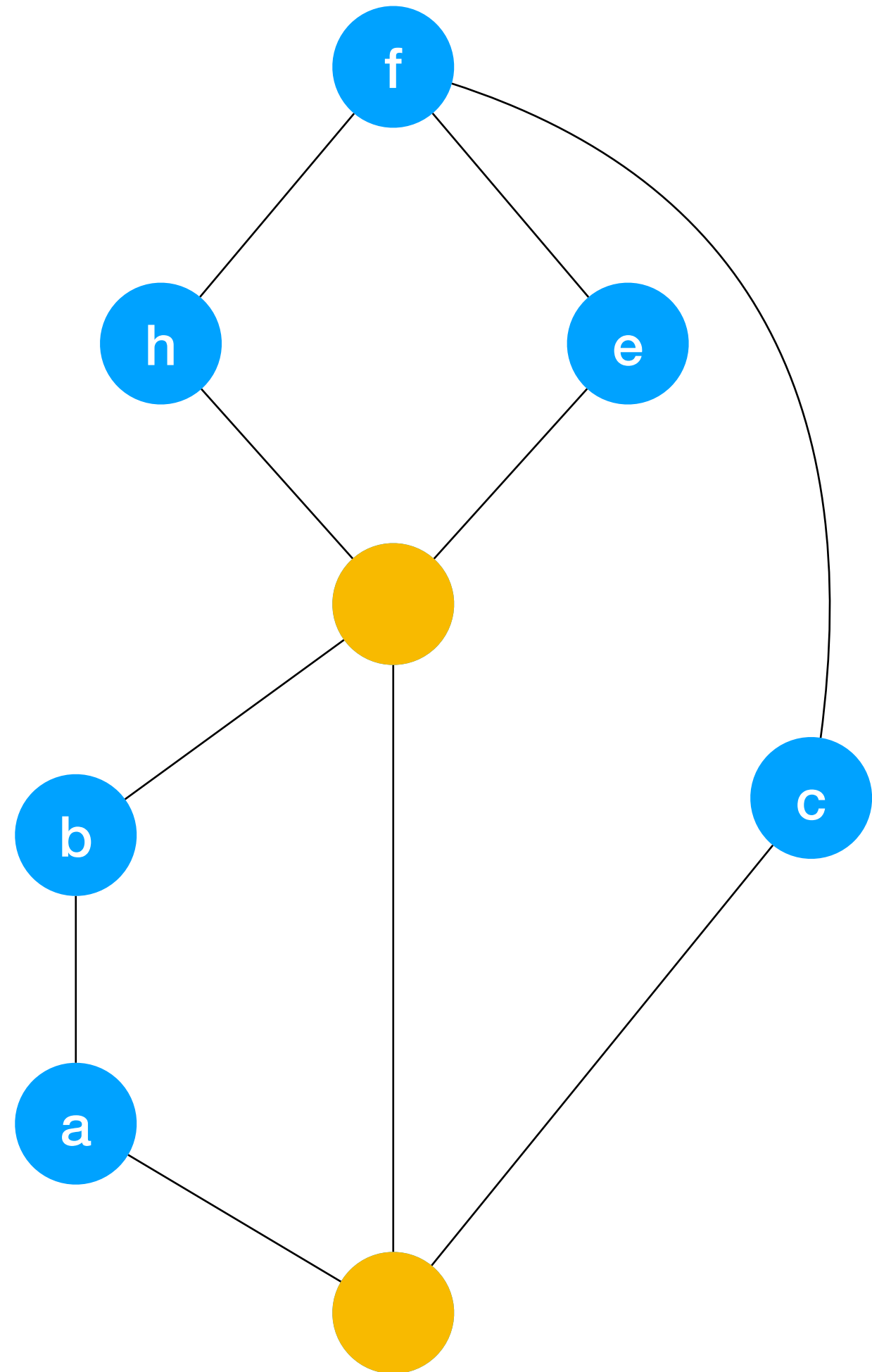
# Example



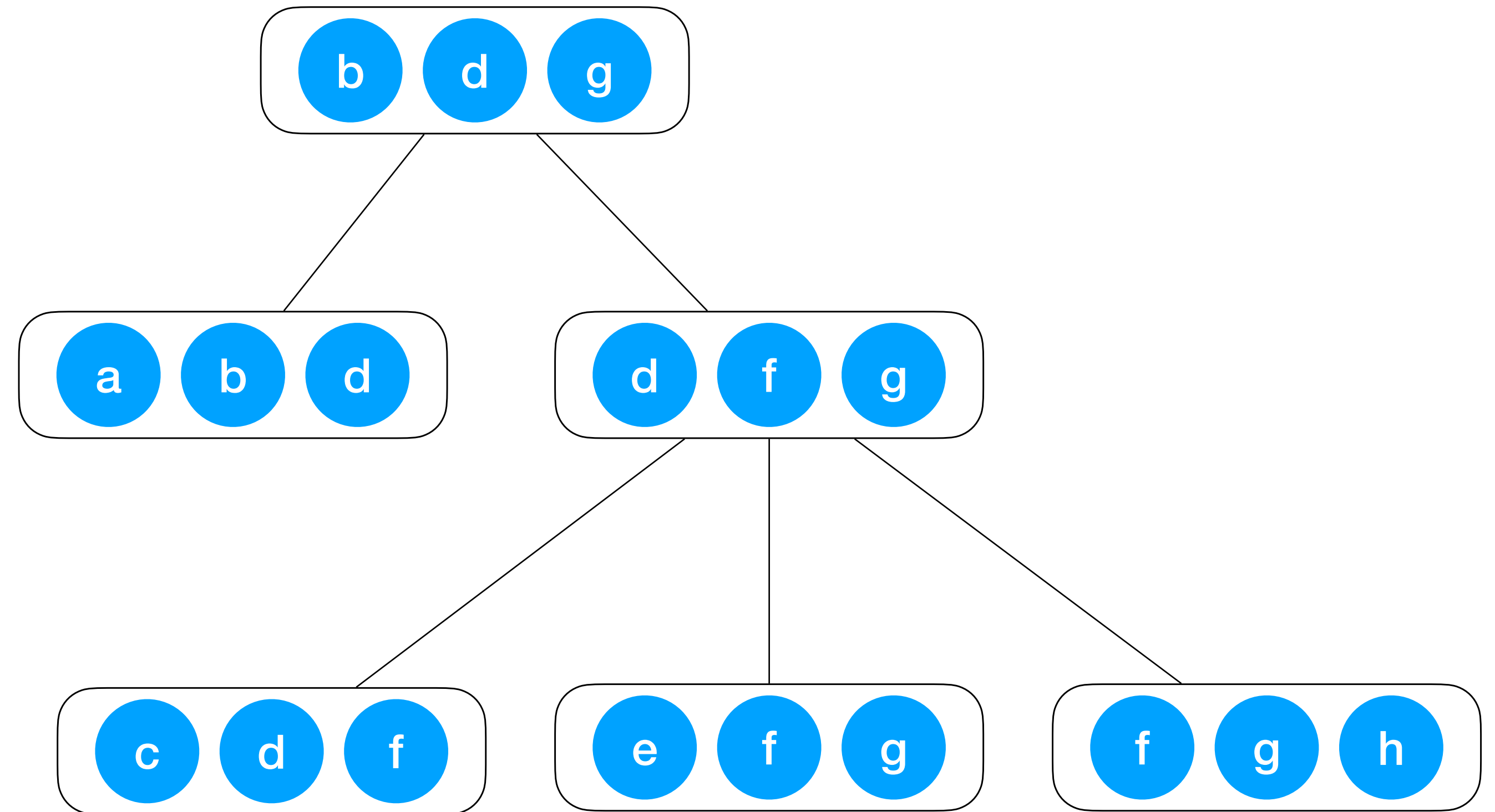
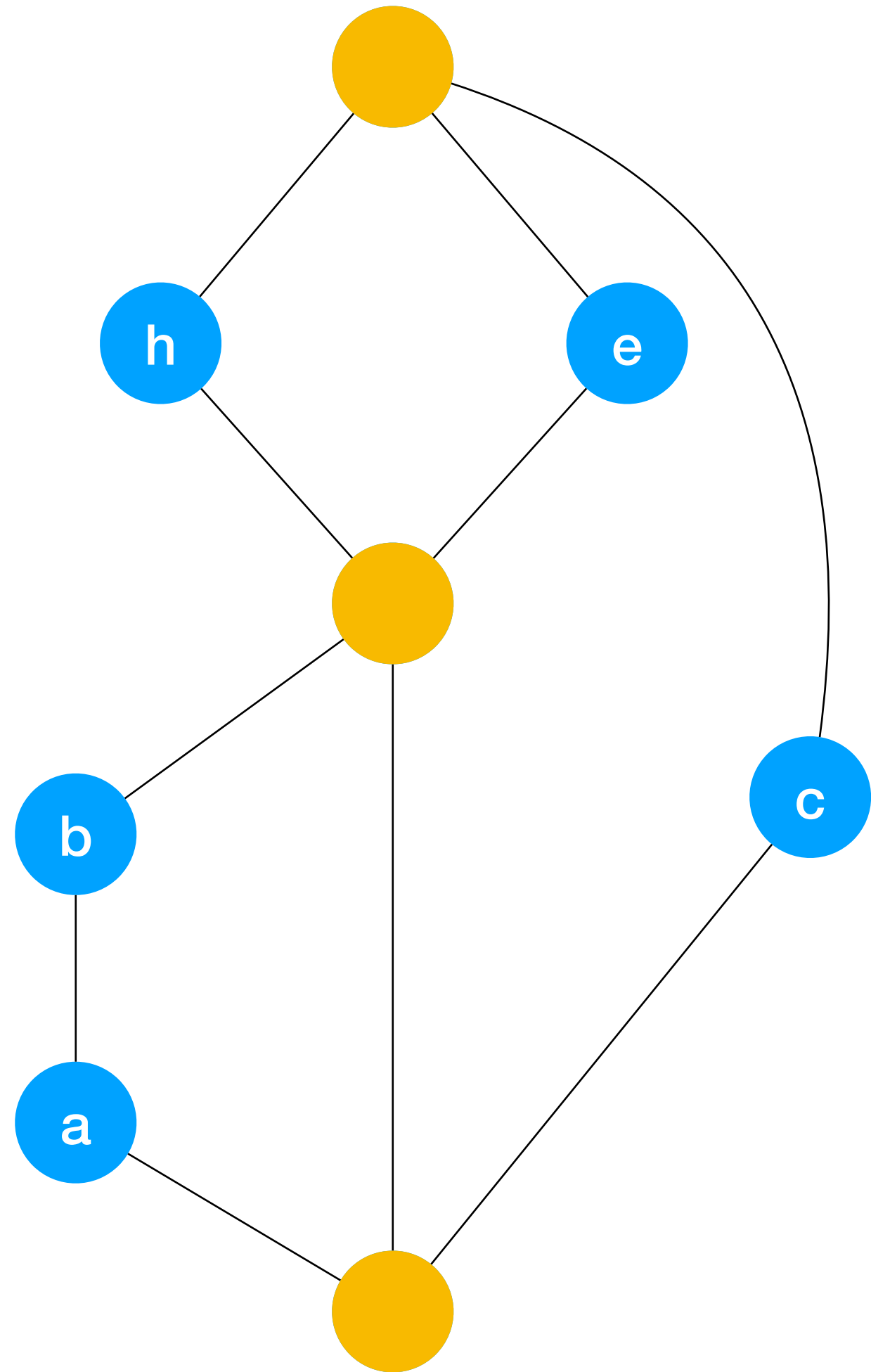
# Example



# Example

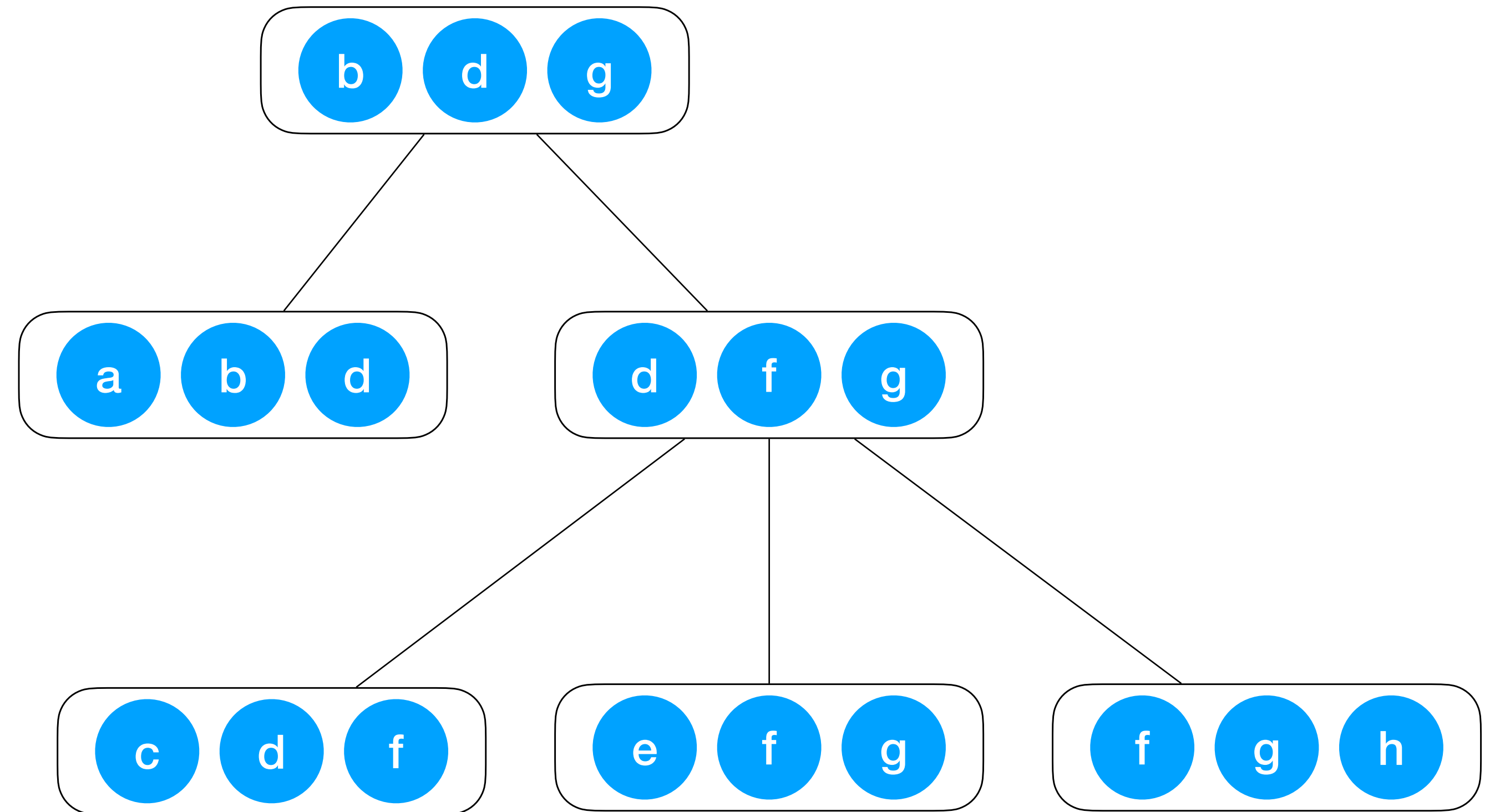
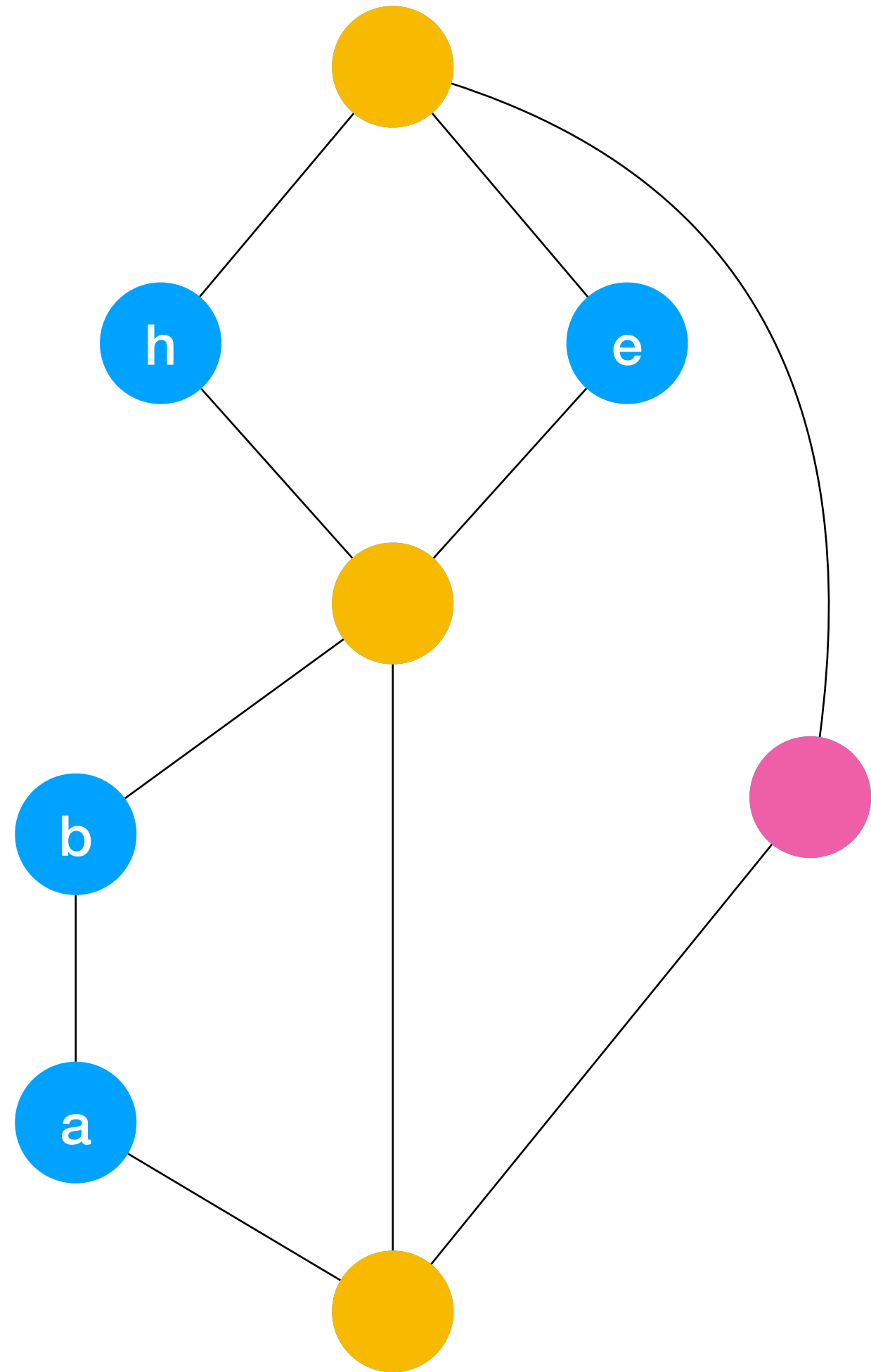


# Example

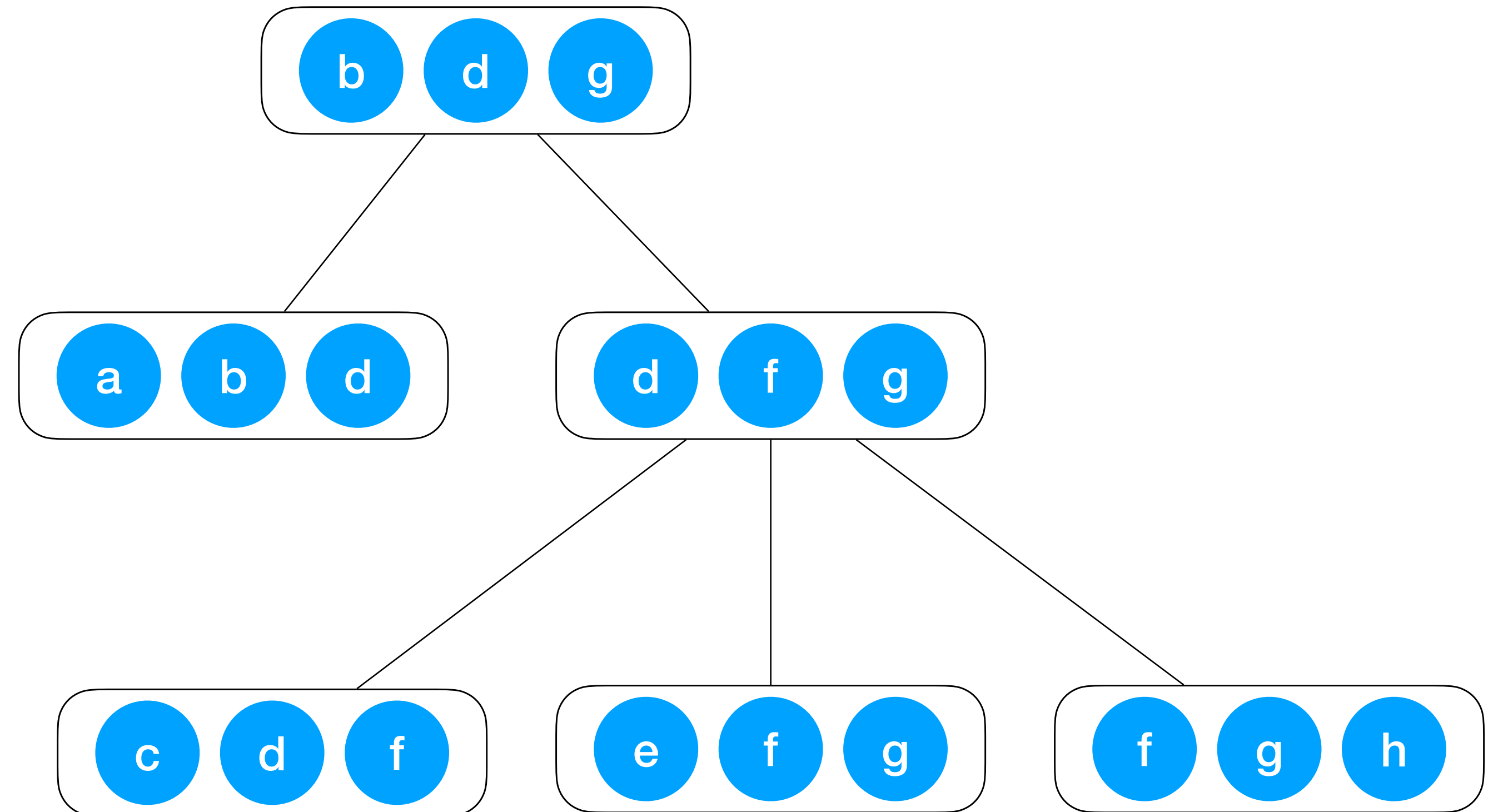
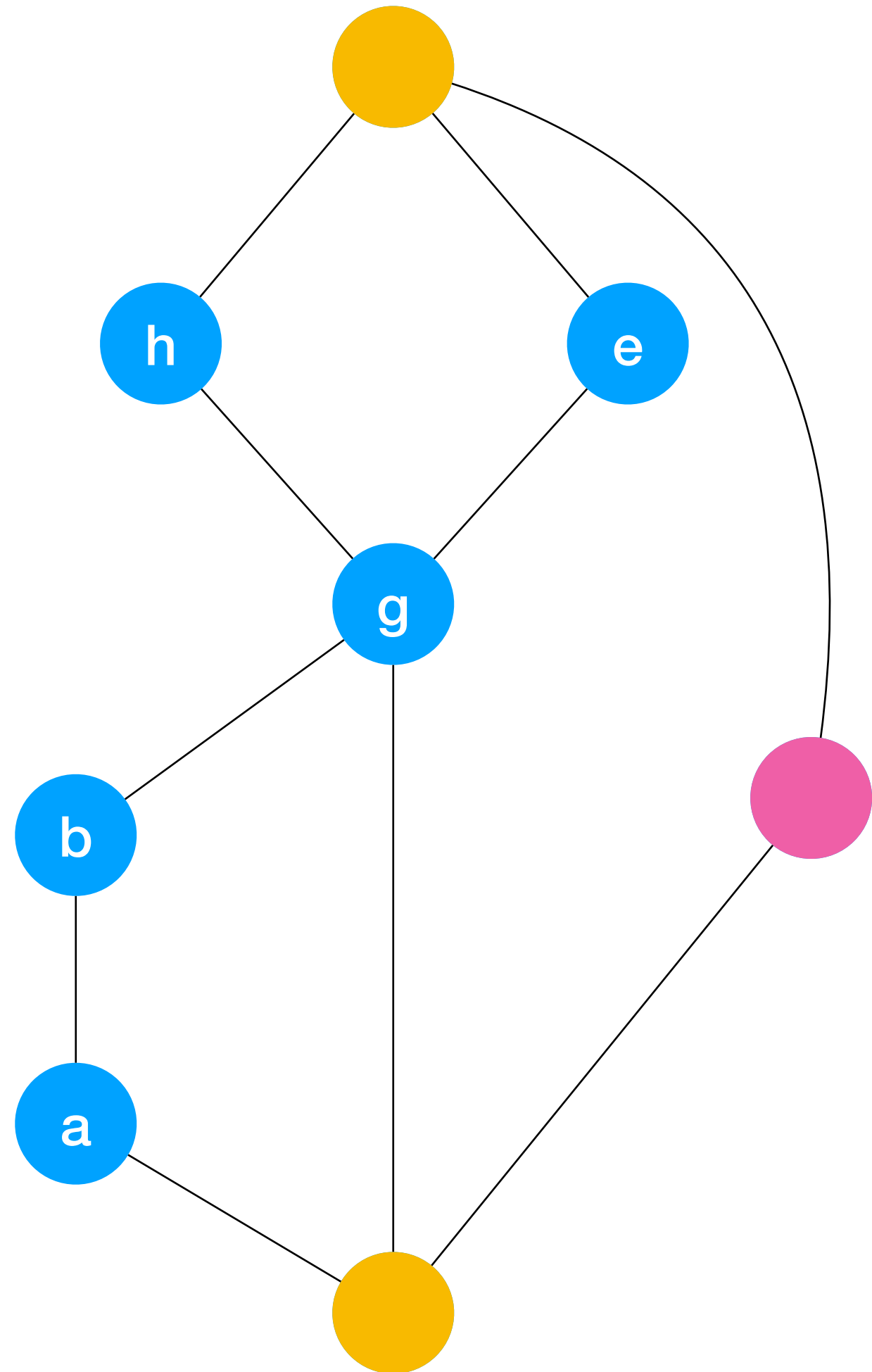




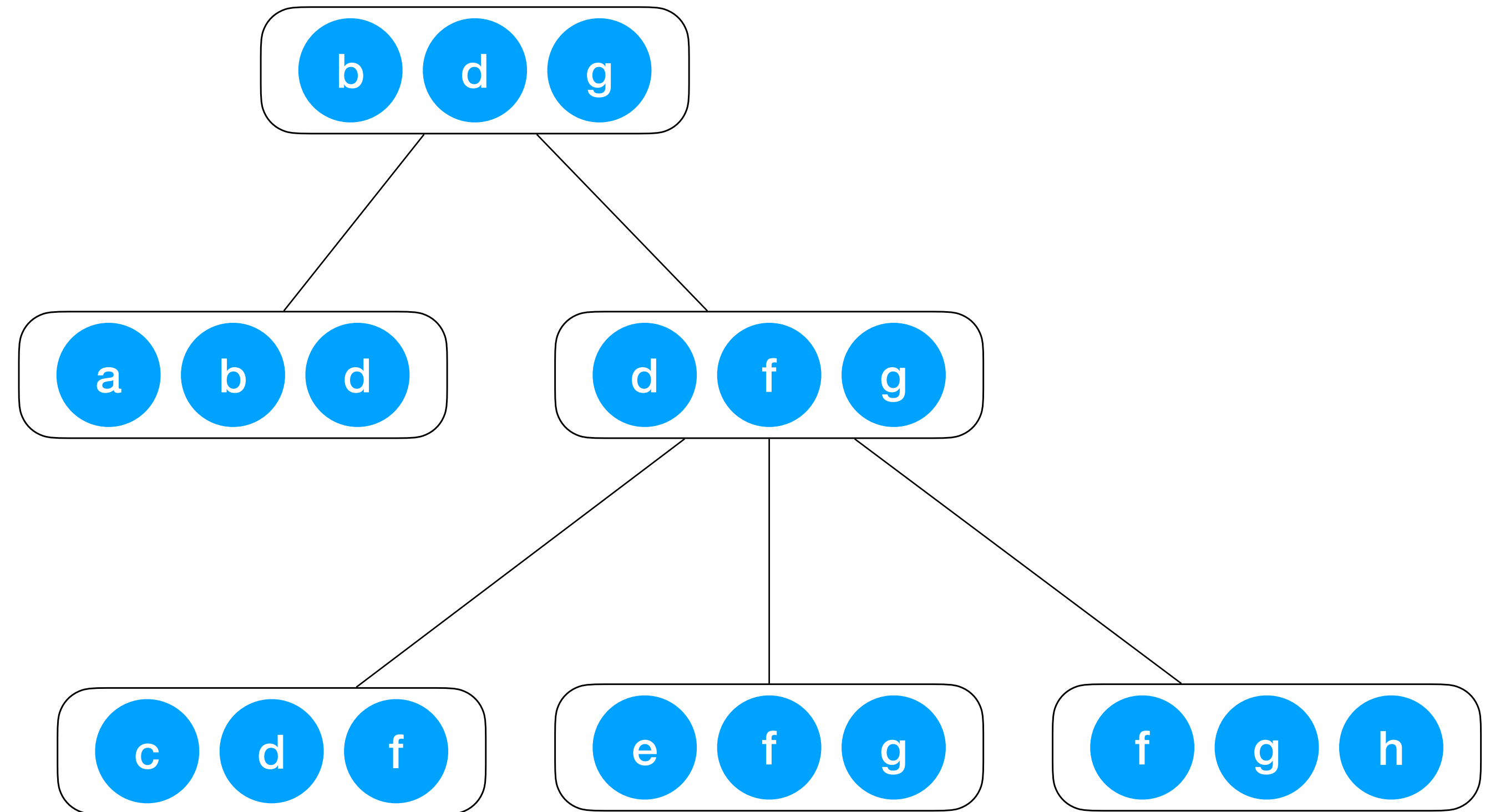
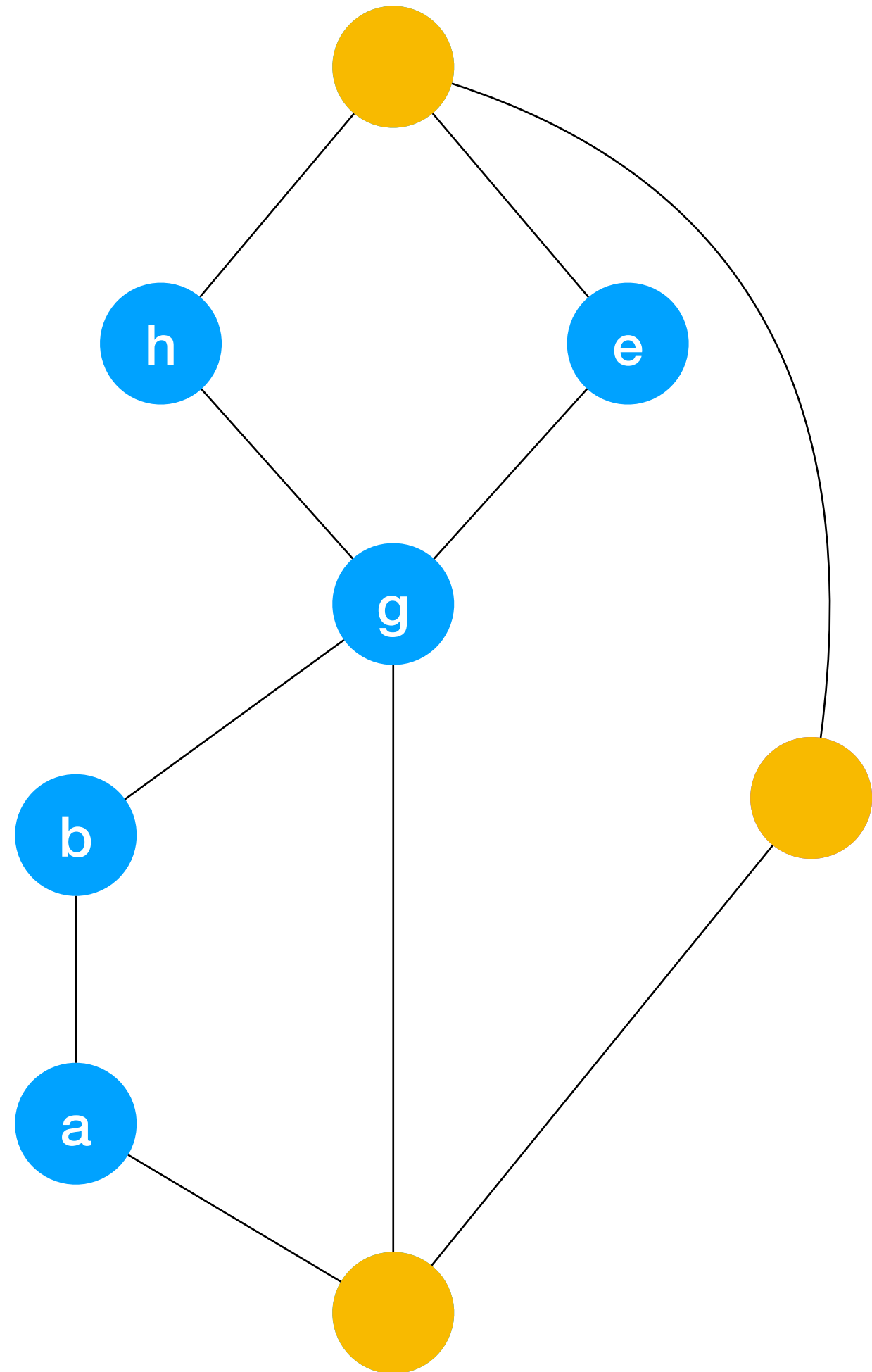
# Example



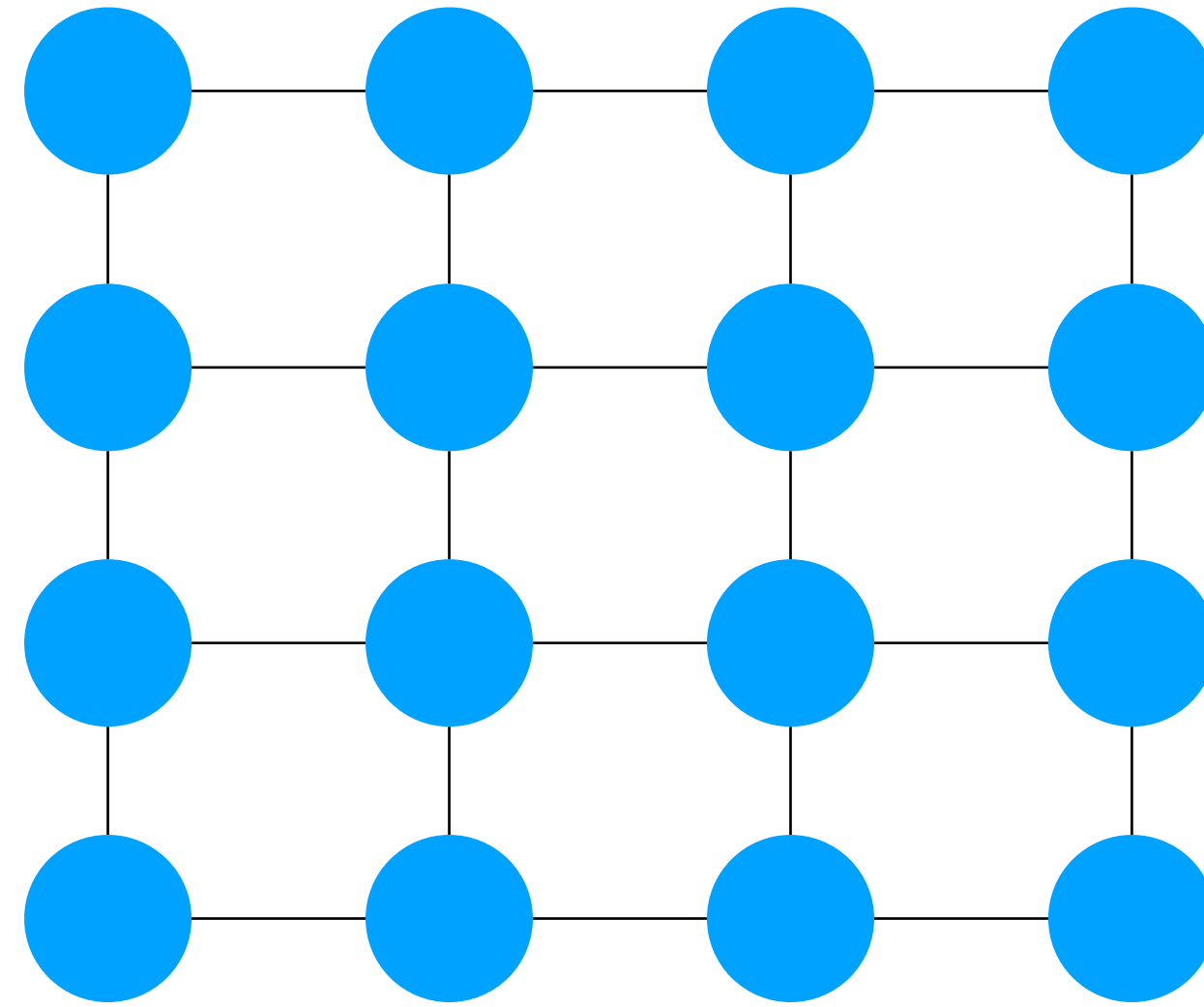
# Example



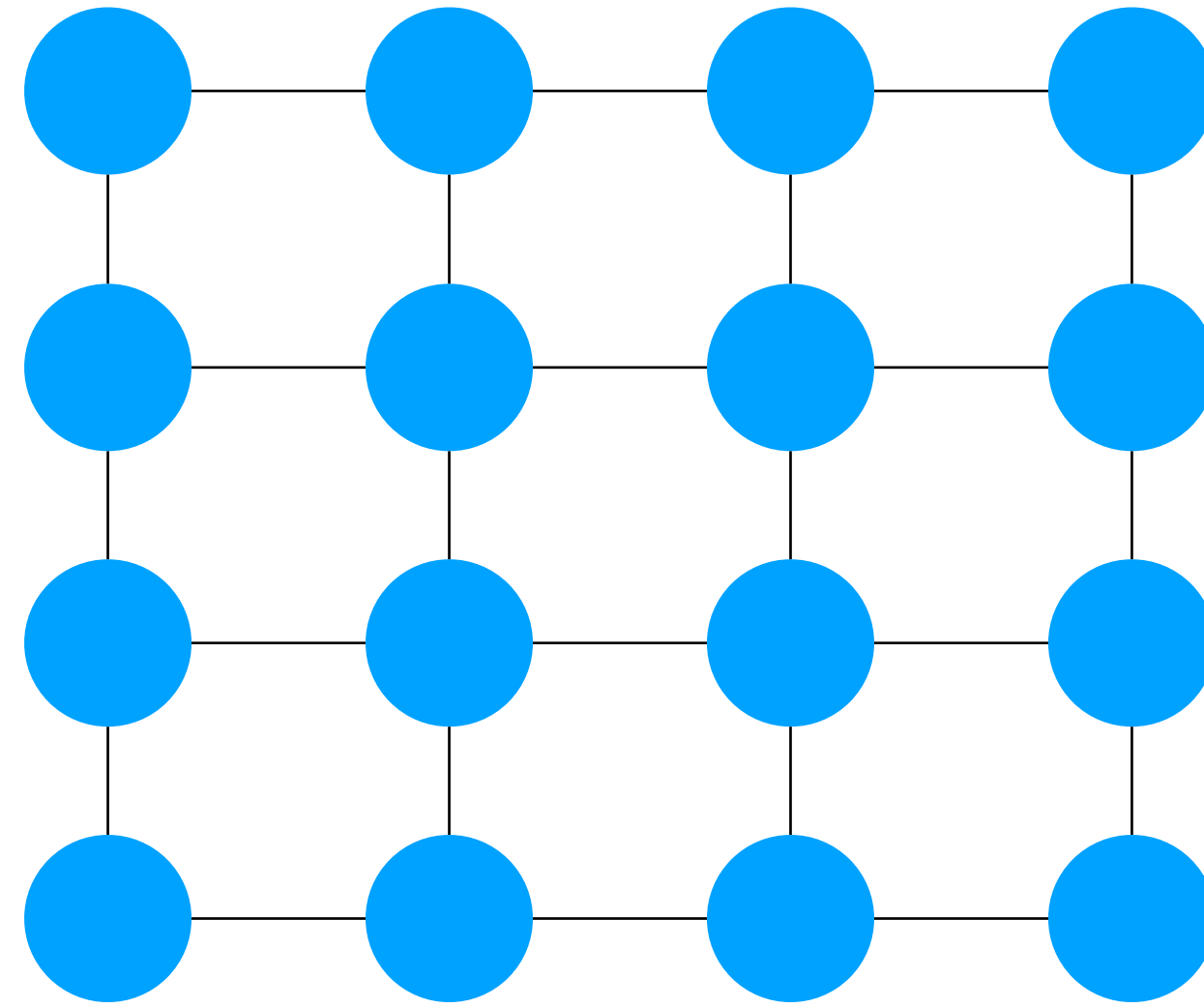
# Example



# Grids have large Treewidth

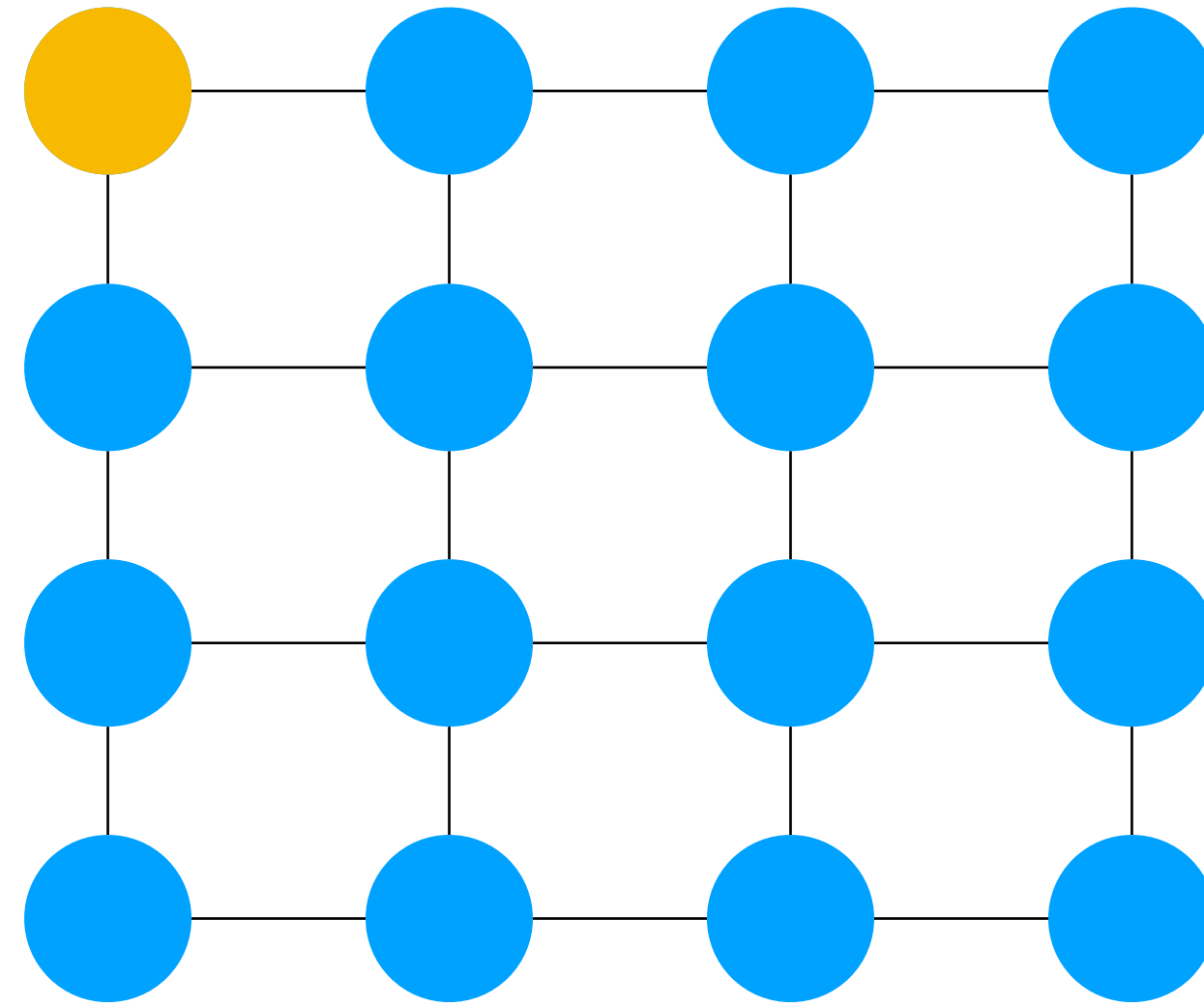


# Grids have large Treewidth



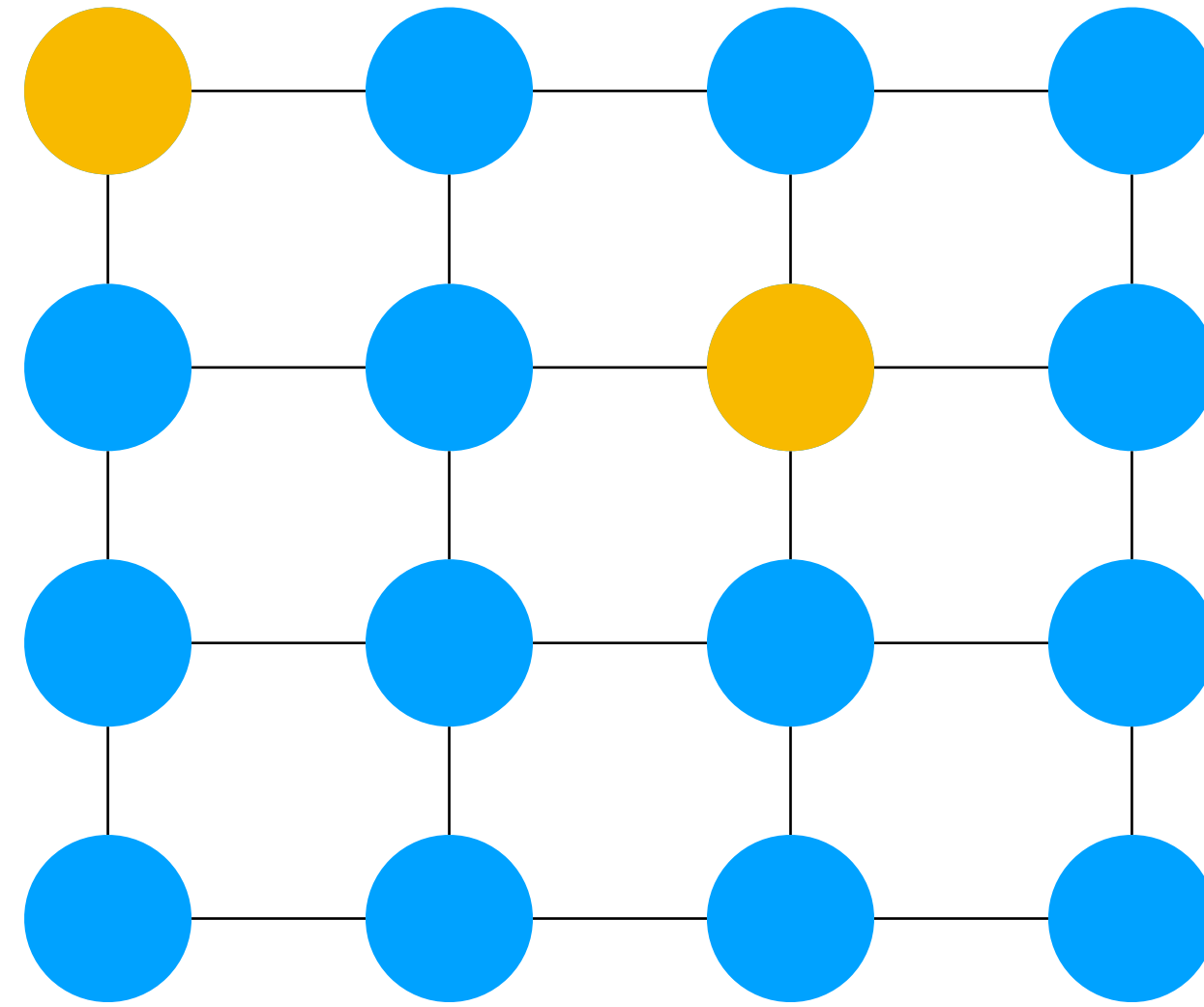
$n - 1$  cops cannot win on  $Q_n$

# Grids have large Treewidth



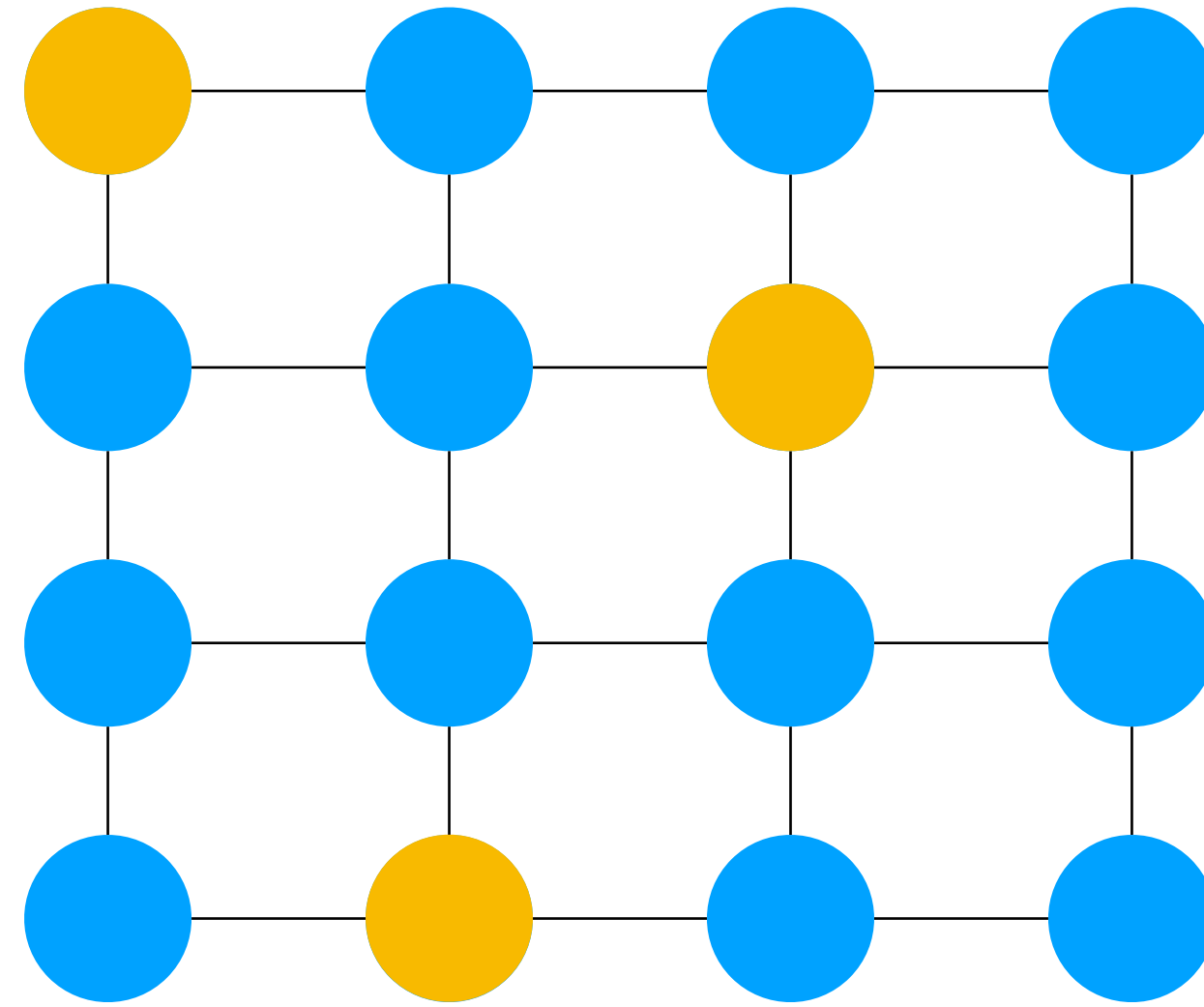
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# Grids have large Treewidth



$n - 1$  cops cannot win on  $Q_n$

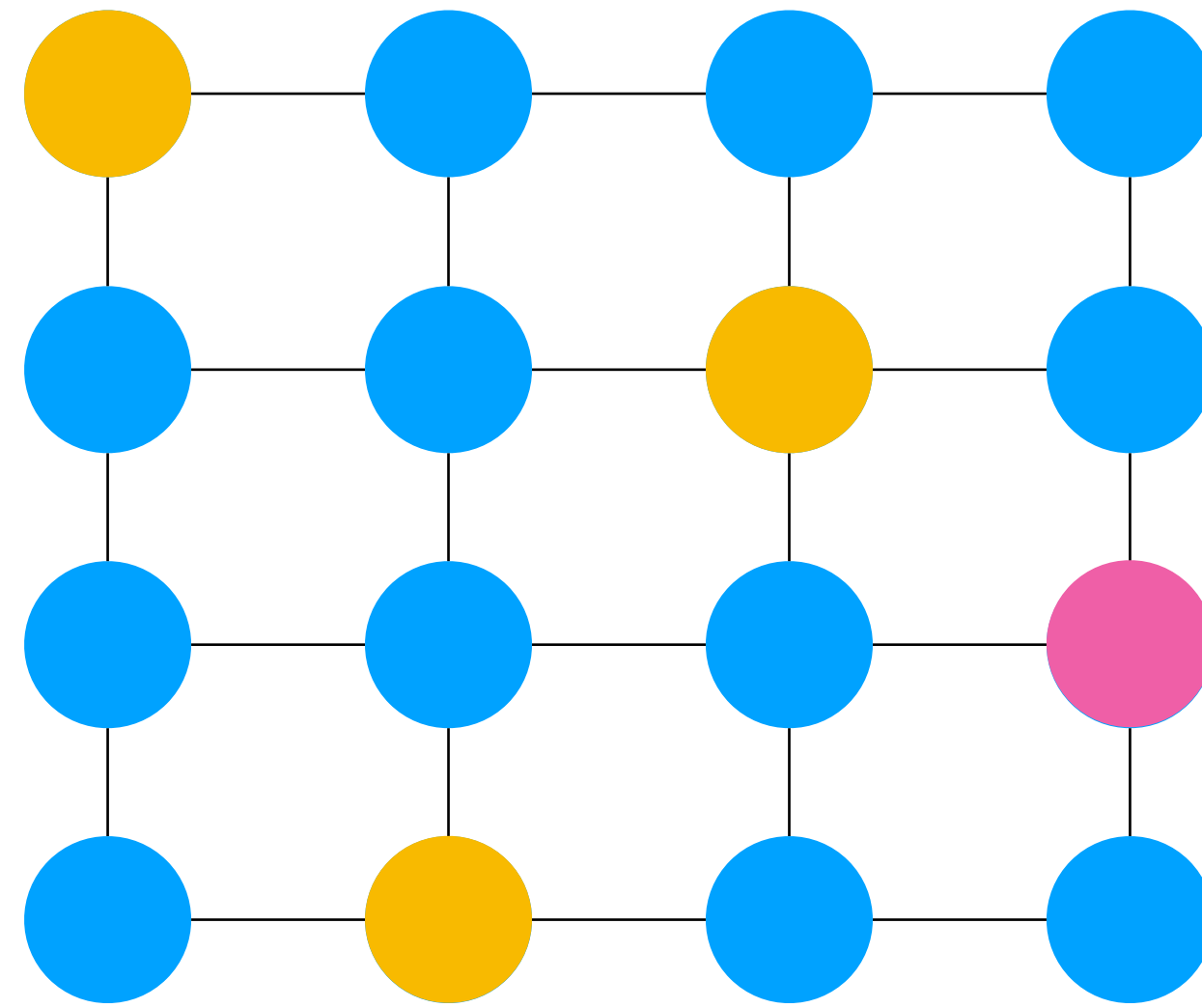
# Grids have large Treewidth



$n - 1$  cops cannot win on  $Q_n$



# Grids have large Treewidth



$n - 1$  cops cannot win on  $Q_n$

In fact, the treewidth of  $Q_n$  is  $n$ .

# **Application: Variable Elimination for SAT**

# Propositional Satisfiability

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$$(x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2) \wedge (x_3 \vee \neg x_2)$$

# Propositional Satisfiability

$$(x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2) \wedge (x_3 \vee \neg x_2)$$

## SAT

**Input:** A CNF formula  $F$ .

**Question:** Does  $F$  have a satisfying assignment?

# Propositional Satisfiability

$$(x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2) \wedge (x_3 \vee \neg x_2)$$

$x_1$

$x_2$

$x_3$

**SAT**

**Input:** A CNF formula  $F$ .

**Question:** Does  $F$  have a satisfying assignment?

# Propositional Satisfiability

$$(x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2) \wedge (x_3 \vee \neg x_2)$$

$x_1$

$x_2$

$x_3$

**SAT**

**Input:** A CNF formula  $F$ .

**Question:** Does  $F$  have a satisfying assignment?

# The Resolution Rule

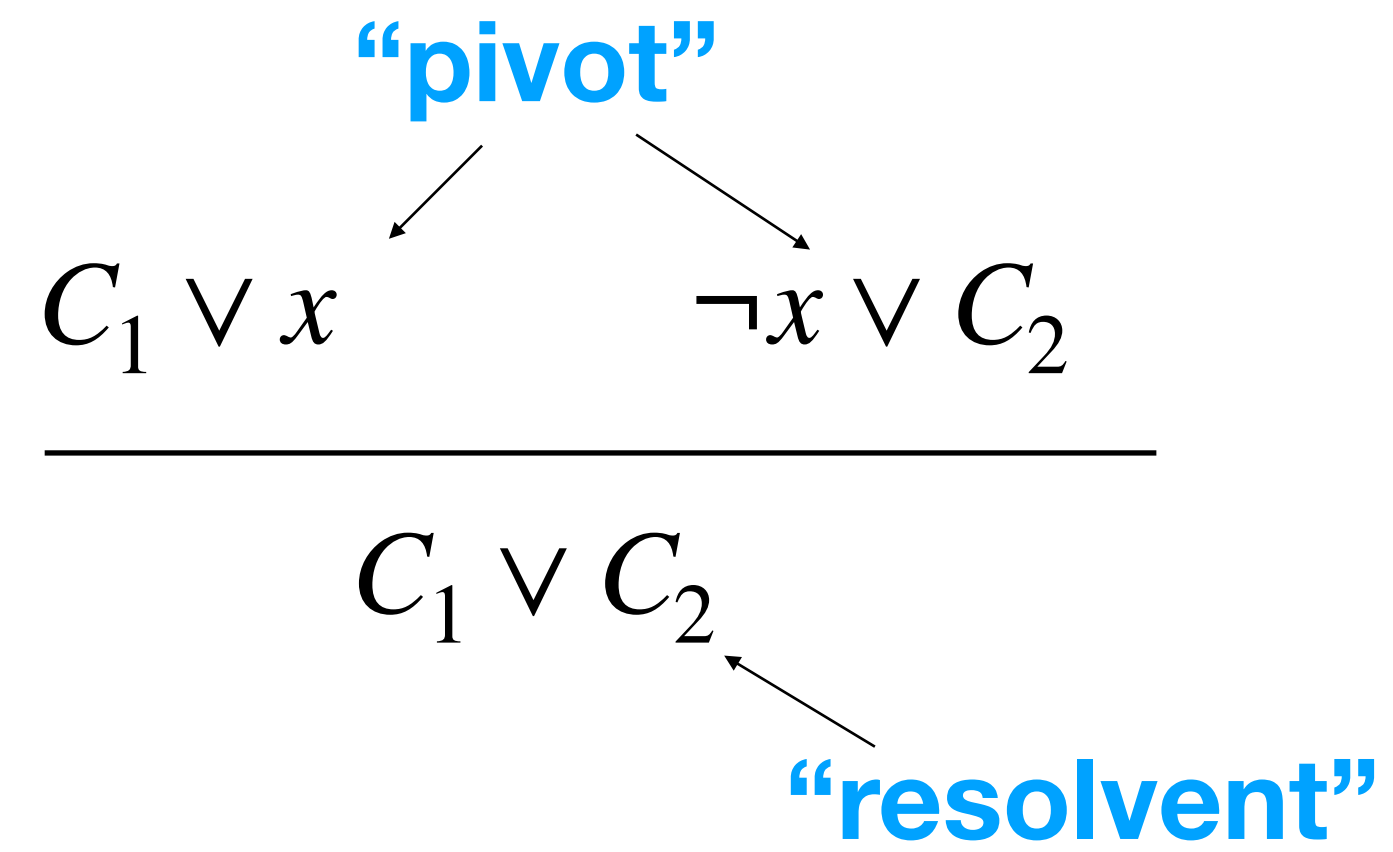
$$\frac{C_1 \vee x \quad \neg x \vee C_2}{C_1 \vee C_2}$$



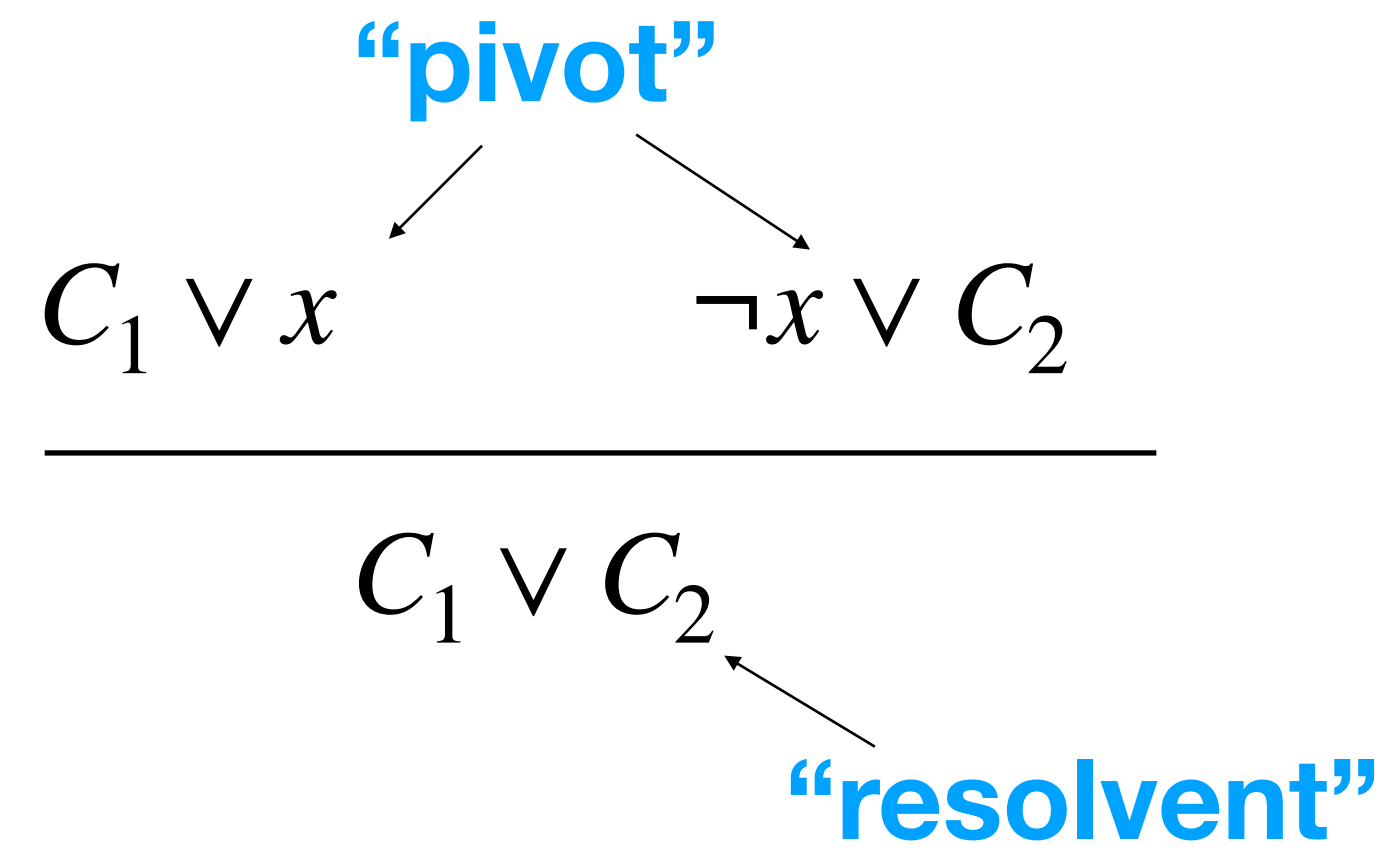
# The Resolution Rule

$$\begin{array}{c} \text{"pivot"} \\ \swarrow \quad \searrow \\ C_1 \vee x \quad \neg x \vee C_2 \\ \hline C_1 \vee C_2 \end{array}$$

# The Resolution Rule

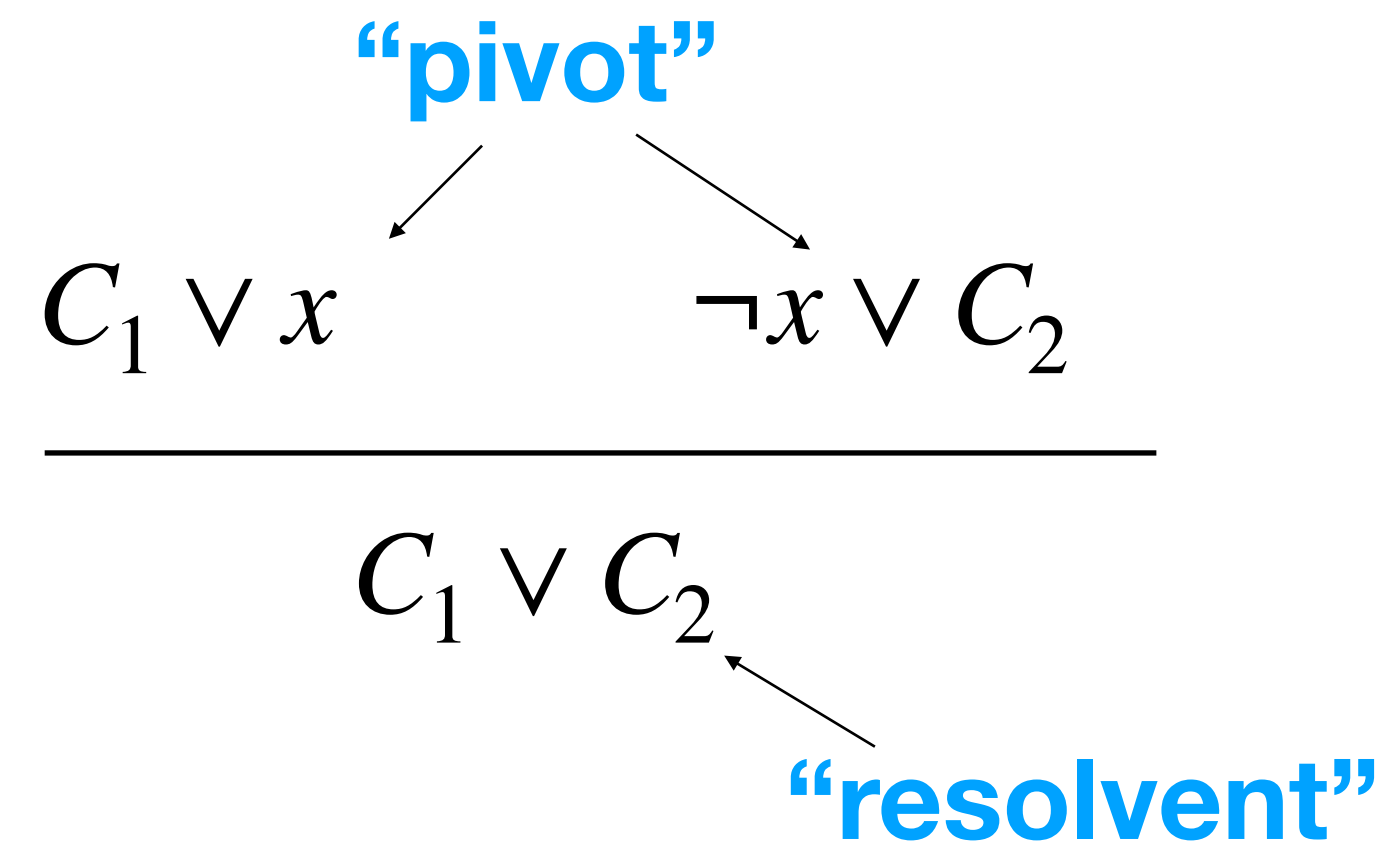


# The Resolution Rule



$C_1 \vee x$      $\neg x \vee C_2$     both satisfied

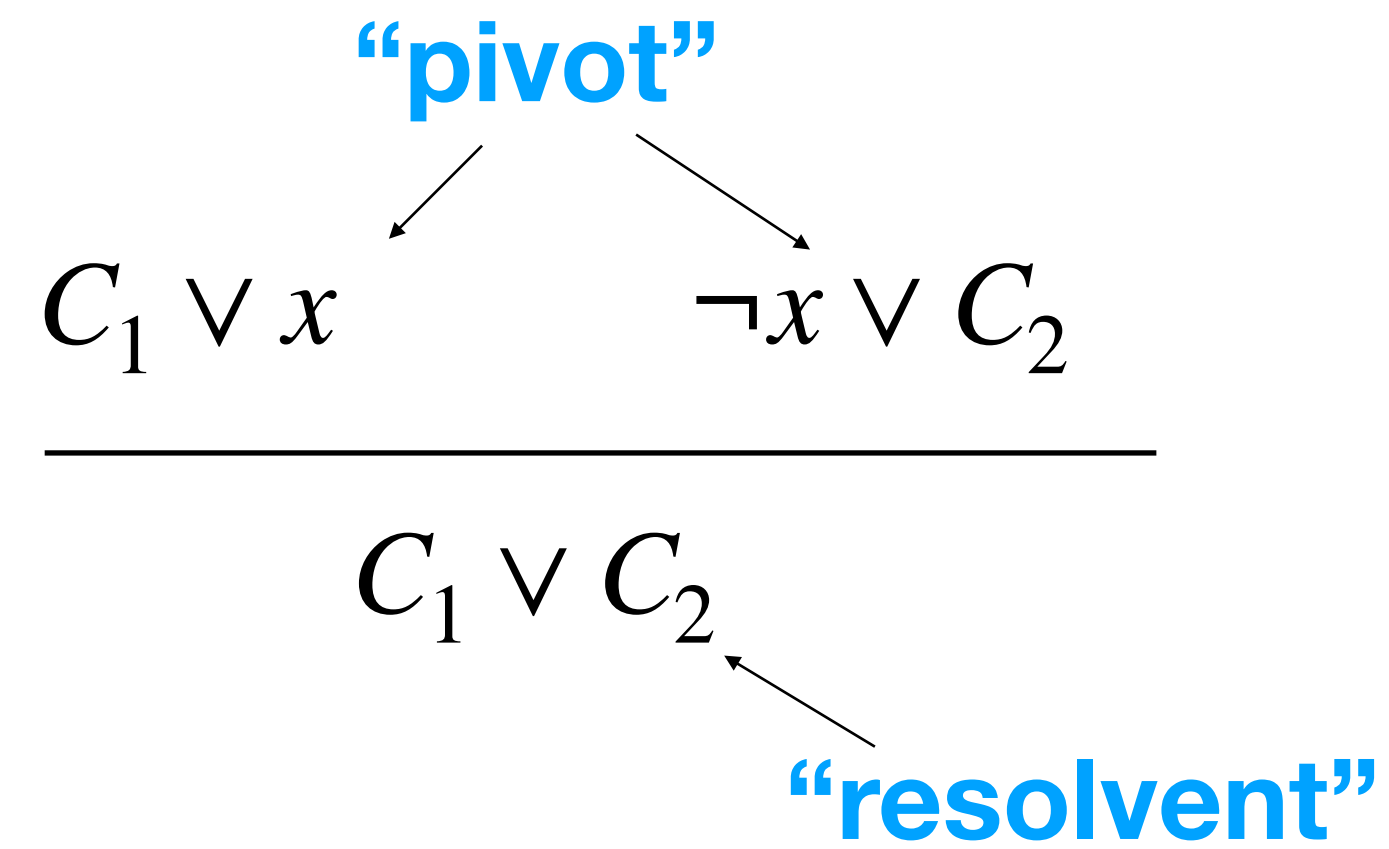
# The Resolution Rule

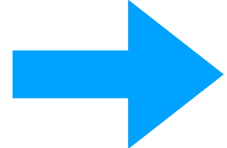


$C_1 \vee x$      $\neg x \vee C_2$     both satisfied

$x$  false

# The Resolution Rule



$x$  false   $C_1 \vee x$      $\neg x \vee C_2$     both satisfied

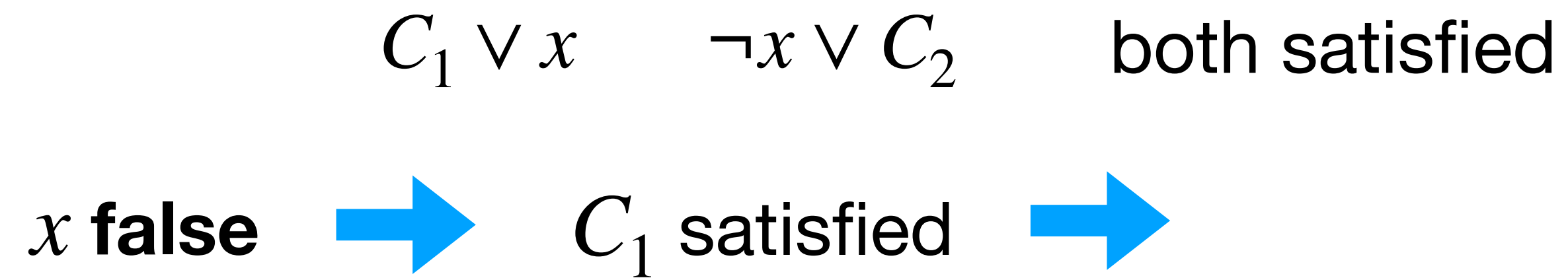
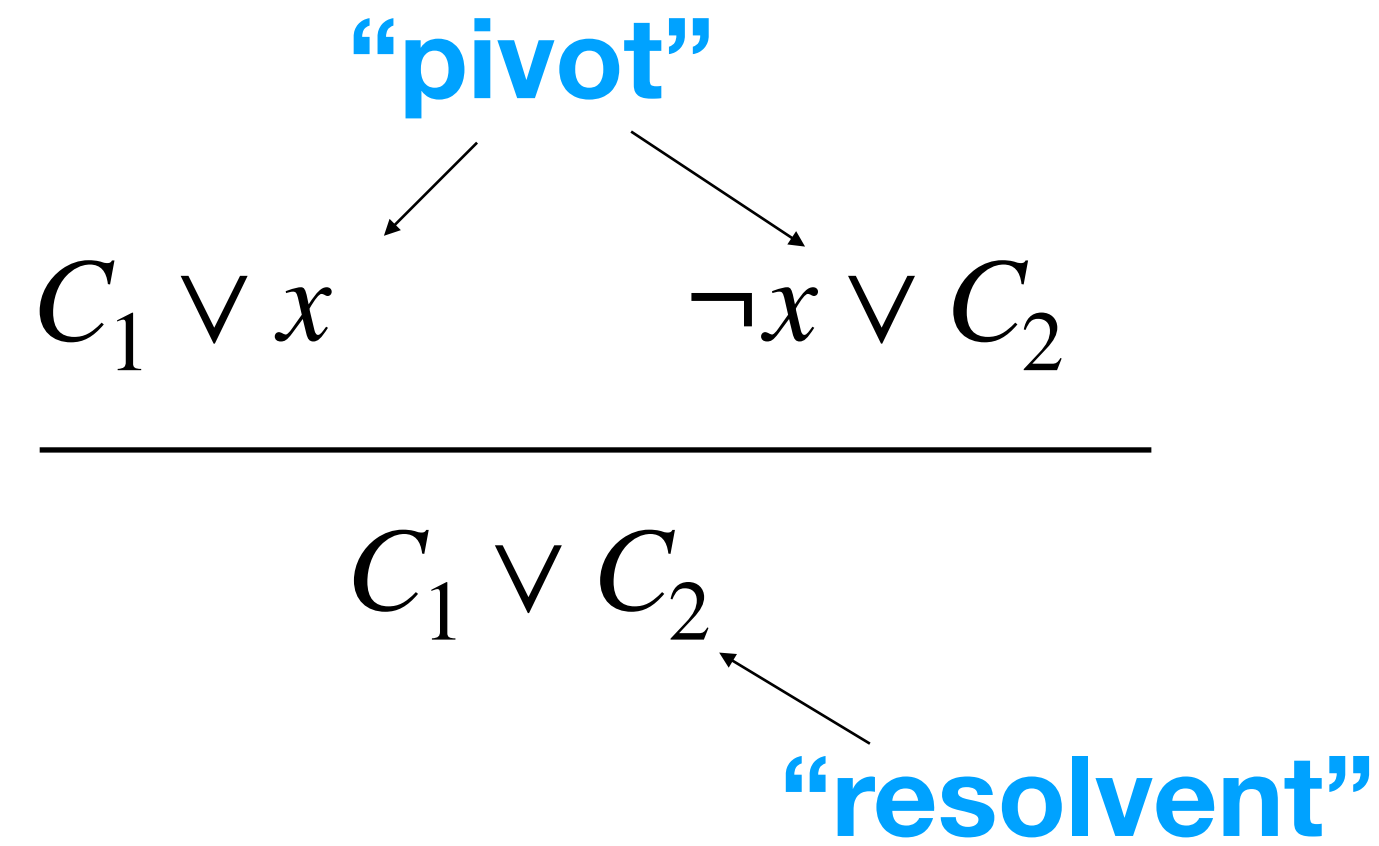
# The Resolution Rule

$$\begin{array}{c} \text{"pivot"} \\ \swarrow \quad \searrow \\ C_1 \vee x \quad \neg x \vee C_2 \\ \hline C_1 \vee C_2 \\ \swarrow \\ \text{"resolvent"} \end{array}$$

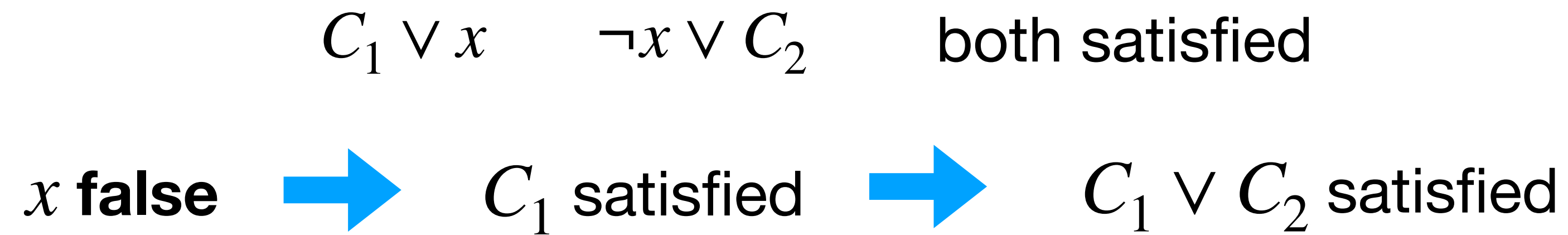
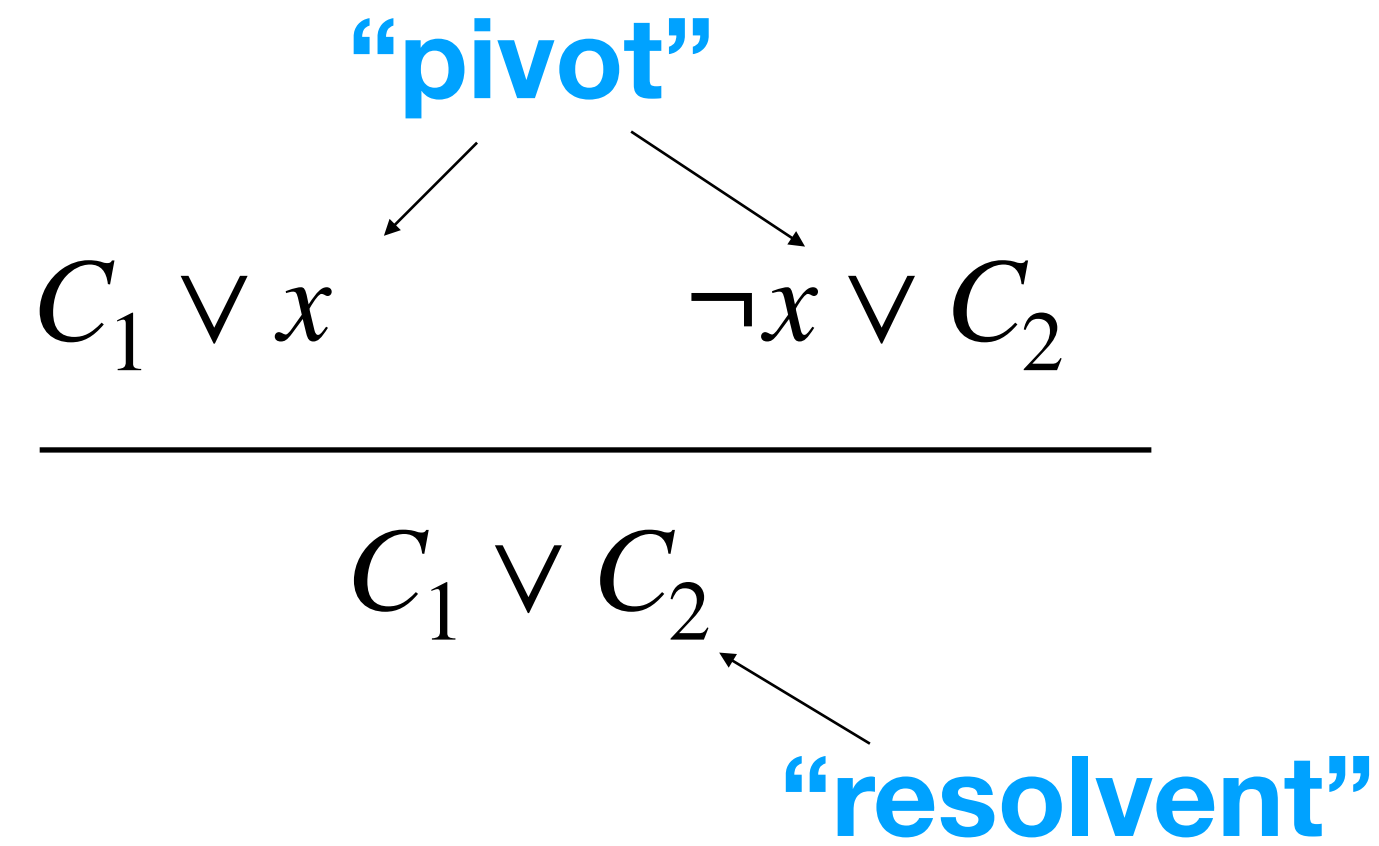
$C_1 \vee x$      $\neg x \vee C_2$     both satisfied

$x$  false  $\rightarrow$   $C_1$  satisfied

# The Resolution Rule



# The Resolution Rule





# The Resolution Rule

$$\frac{C_1 \vee x \quad \neg x \vee C_2}{C_1 \vee C_2}$$

“pivot”

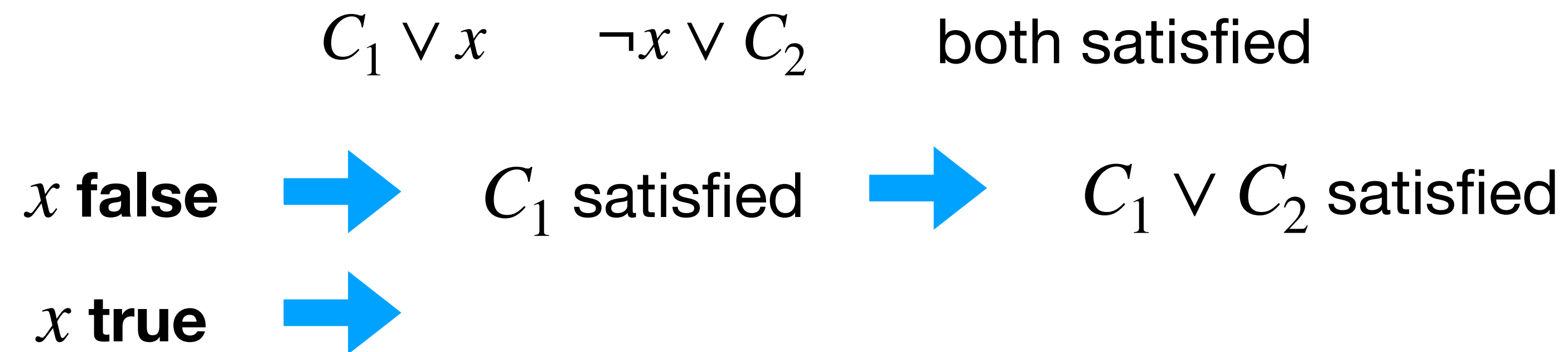
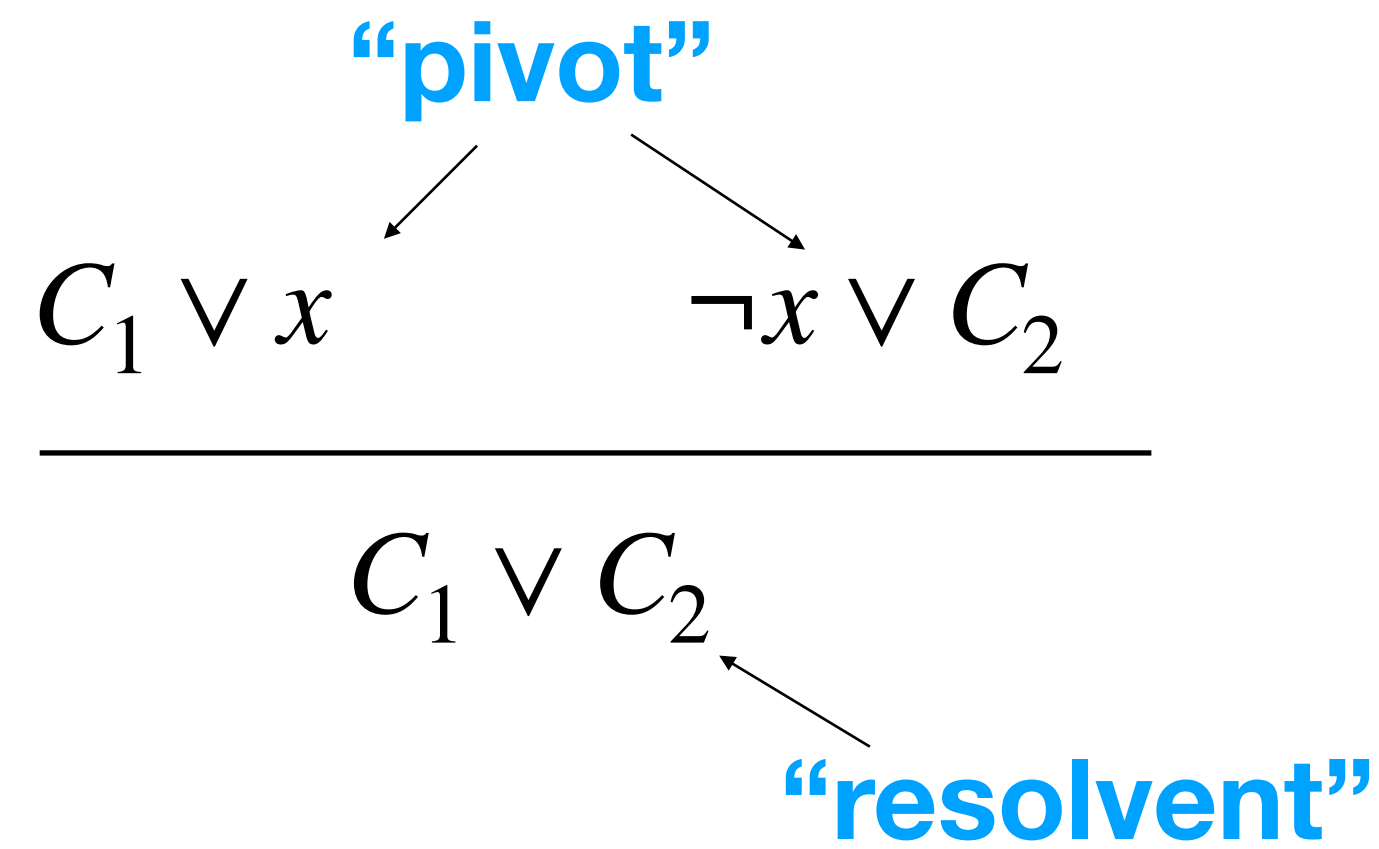
“resolvent”

$C_1 \vee x$      $\neg x \vee C_2$     both satisfied

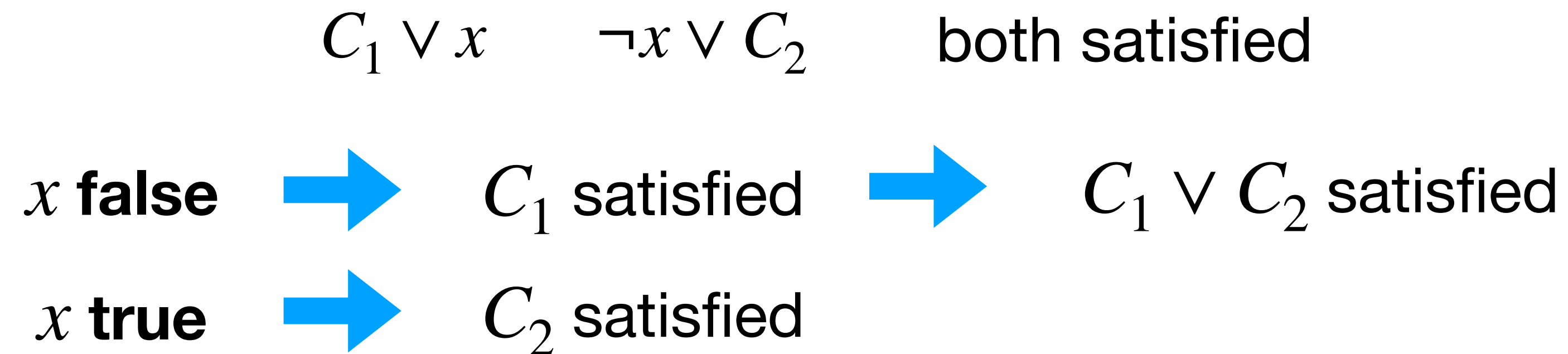
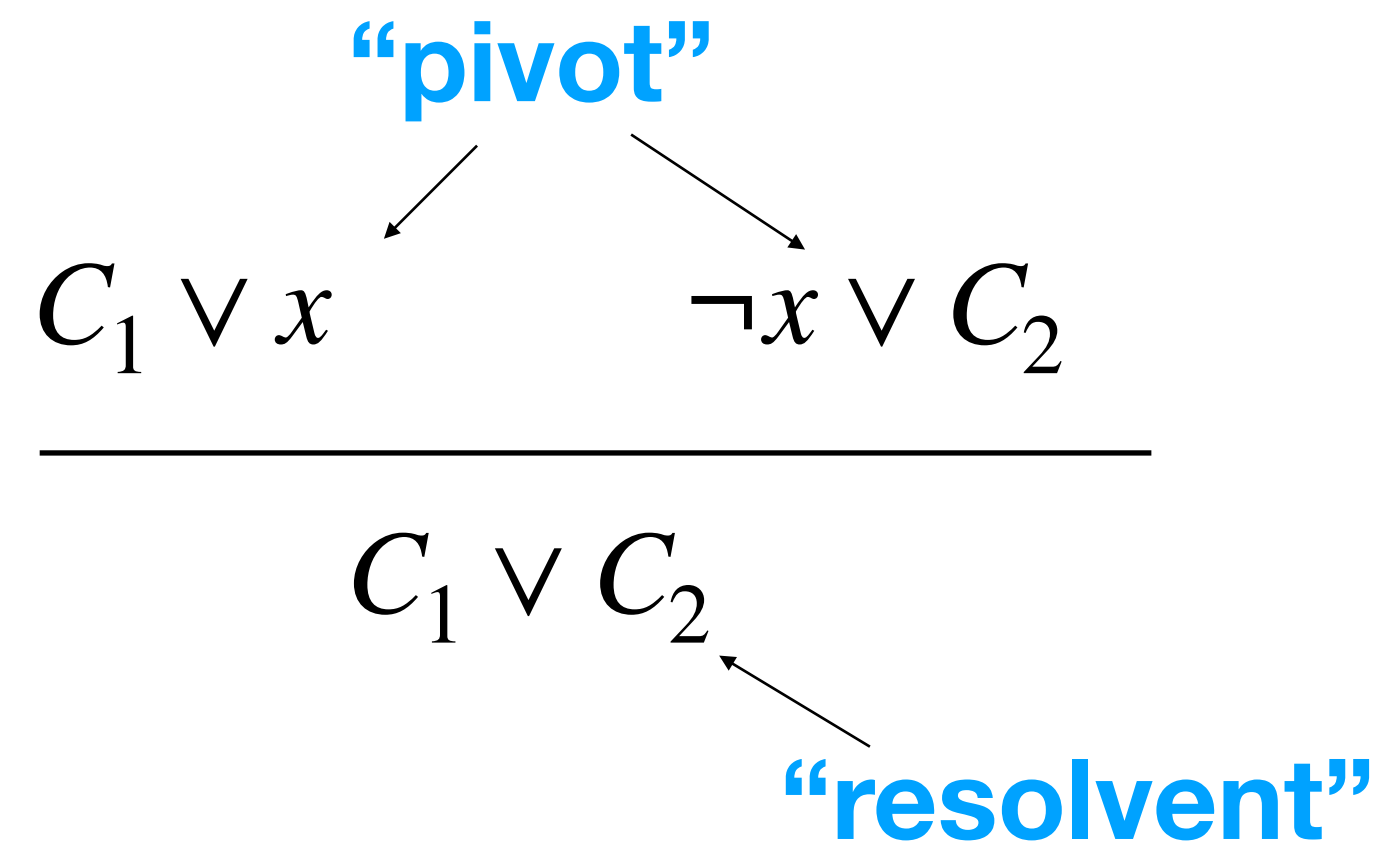
$x$  false  $\rightarrow$   $C_1$  satisfied  $\rightarrow$   $C_1 \vee C_2$  satisfied

$x$  true

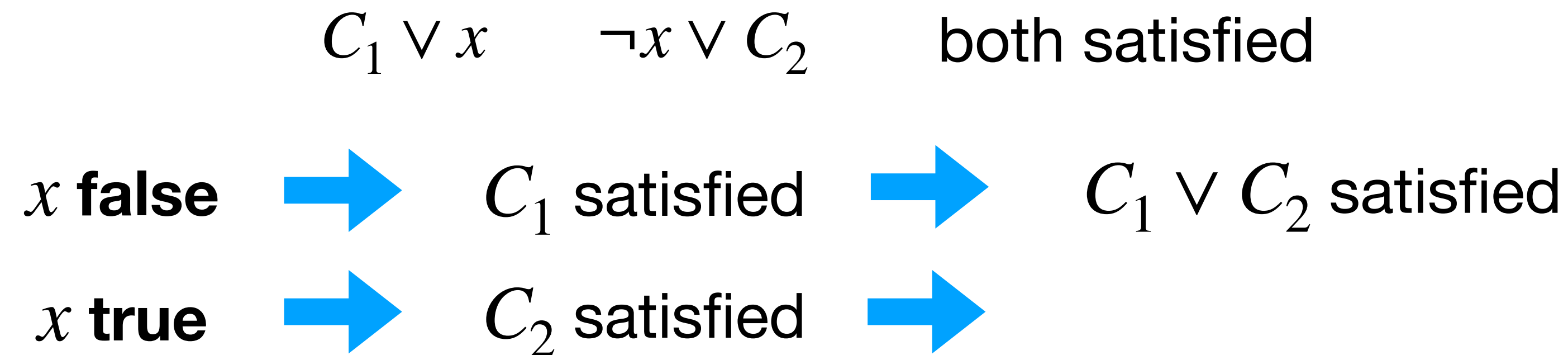
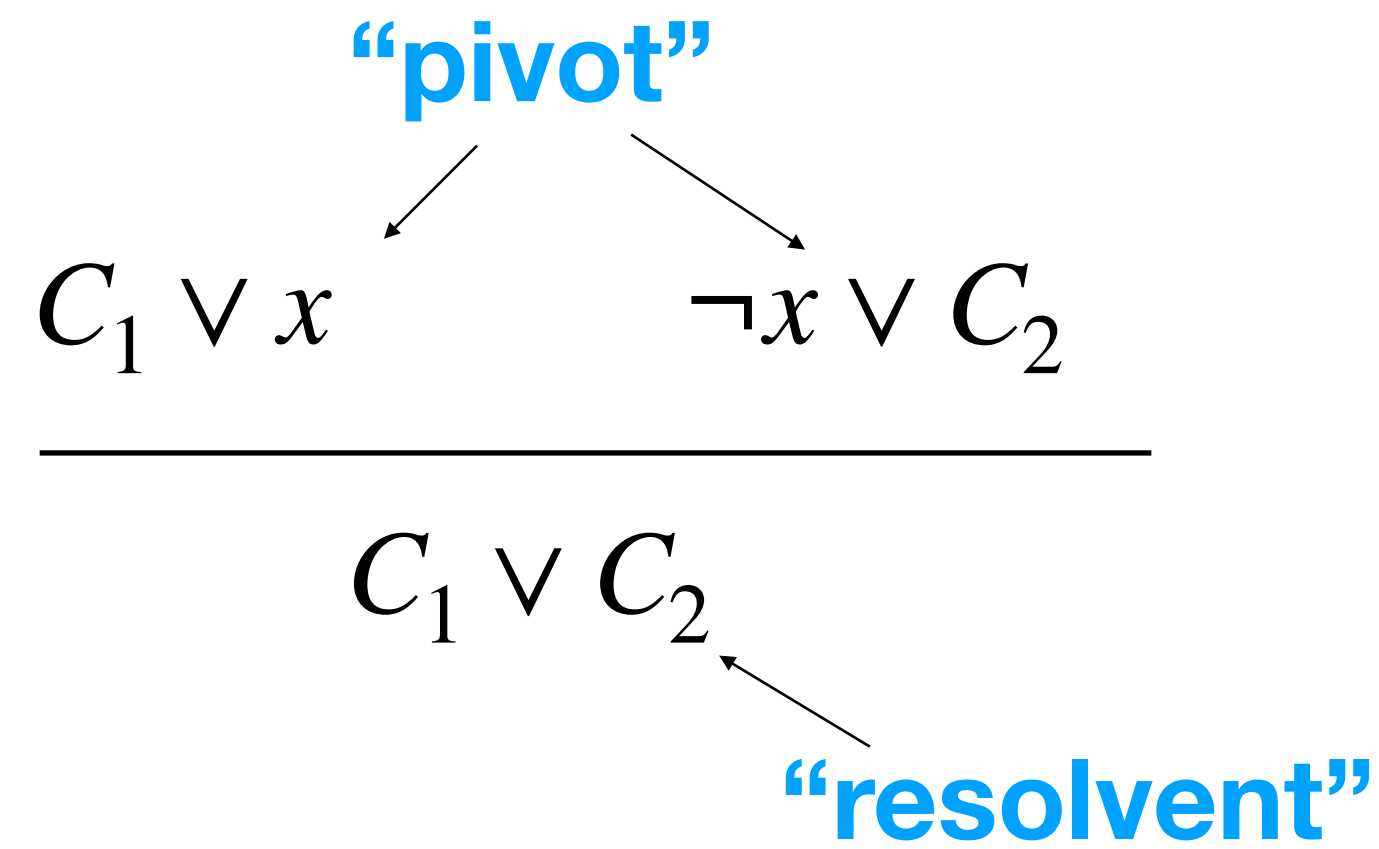
# The Resolution Rule



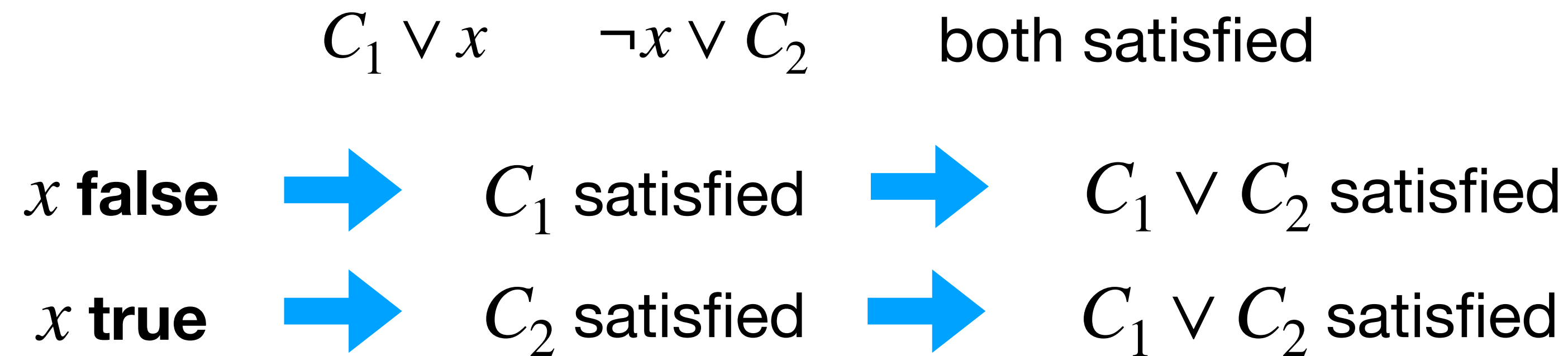
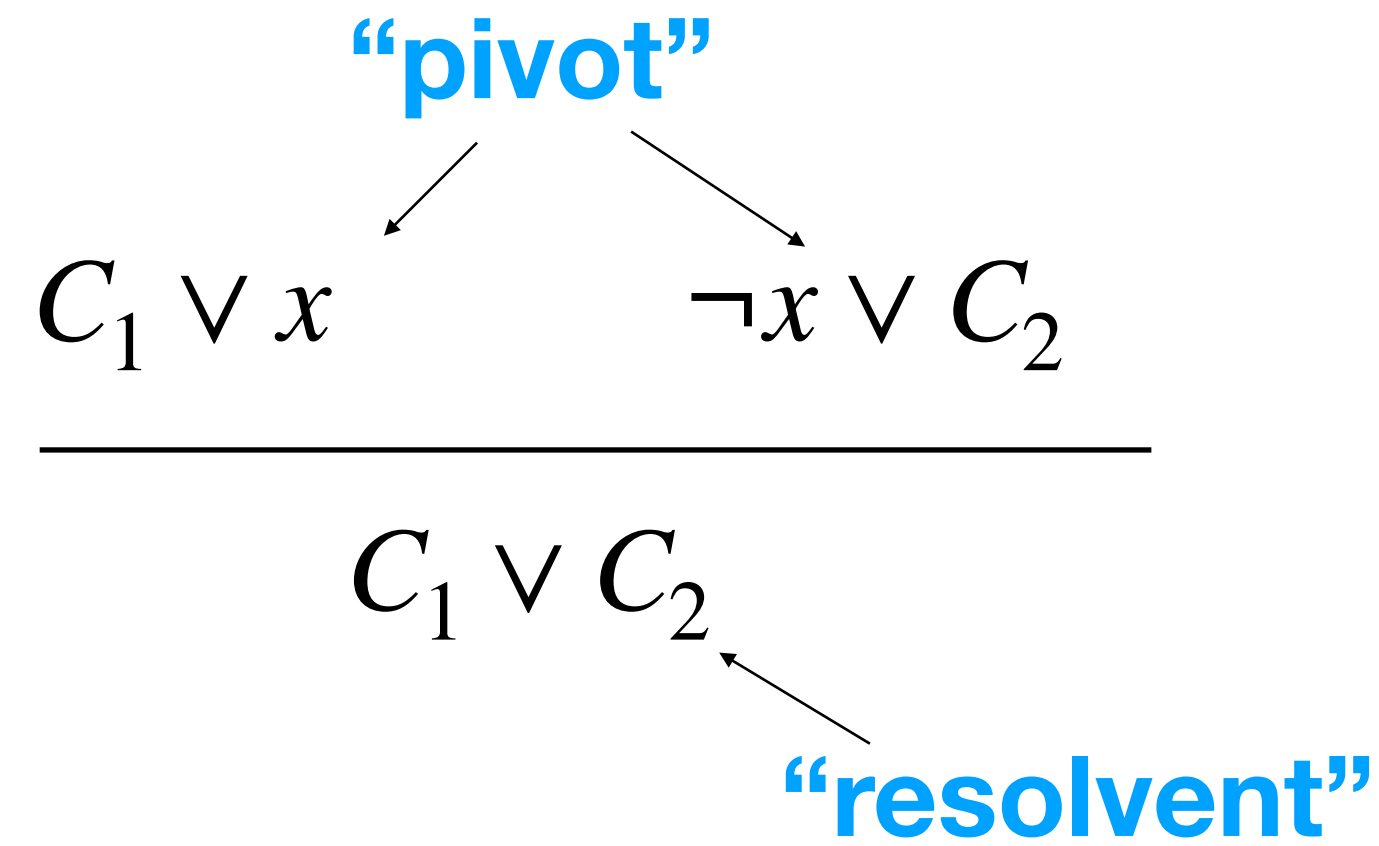
# The Resolution Rule



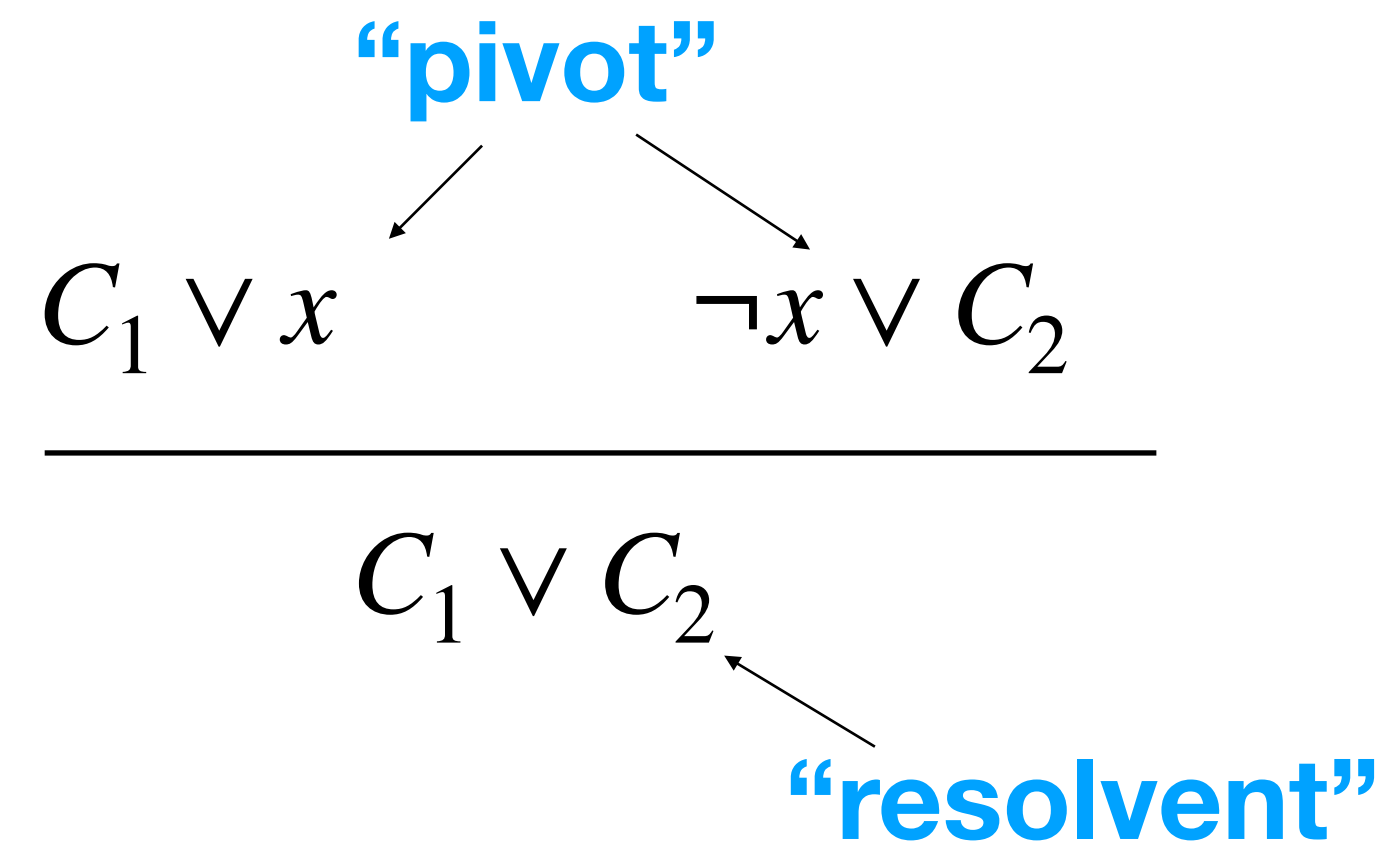
# The Resolution Rule



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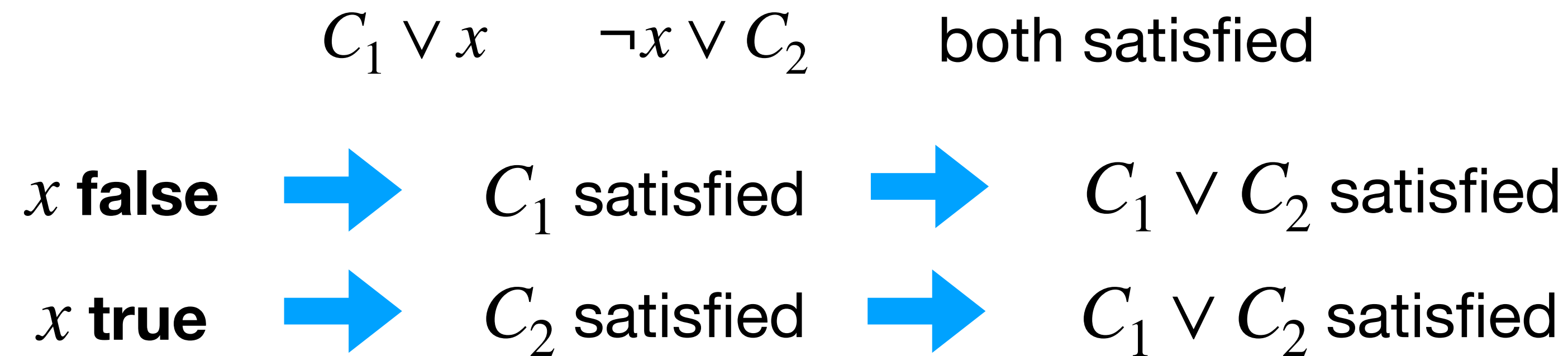


# The Resolution Rule

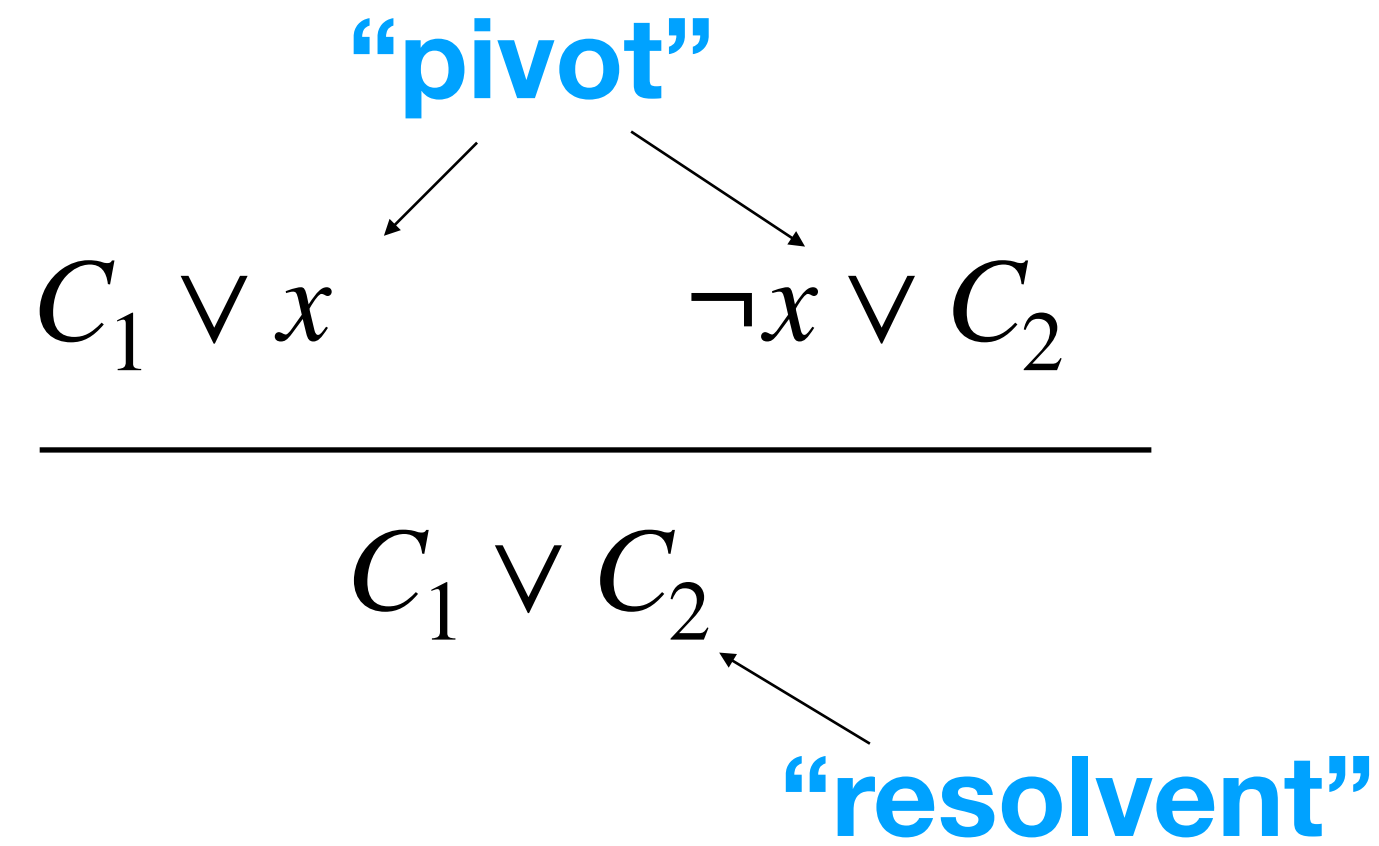


## Theorem

Resolution is sound.



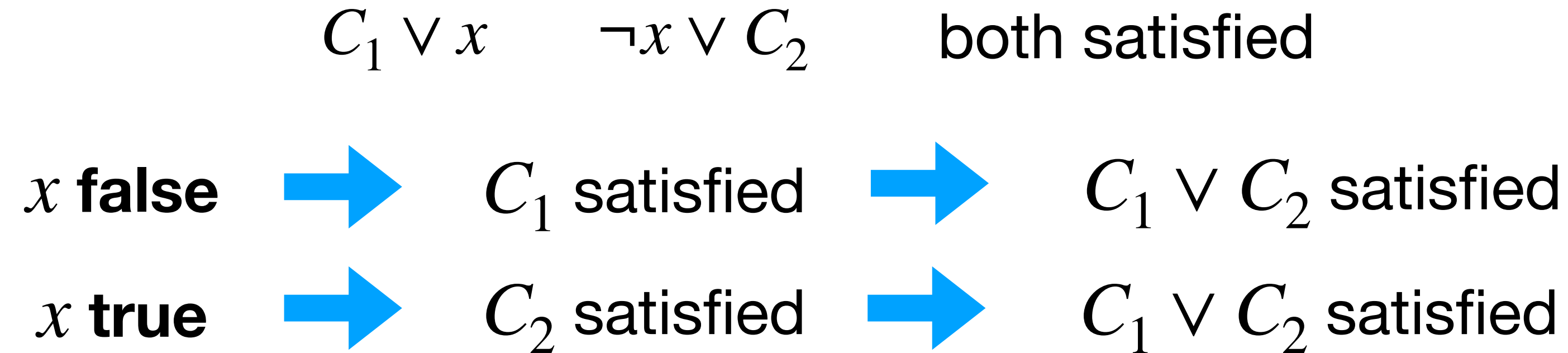
# The Resolution Rule



## Theorem

Resolution is sound.

Adding resolvents does not make a formula unsatisfiable.



# Examples



# Examples

$$(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)$$

# Examples

$$\underline{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}$$

# Examples

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}{(x_2 \vee \neg x_3)}$$

# Examples

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}{(x_2 \vee \neg x_3)}$$

$$(x_1 \vee x_2 \vee \neg x_3) \quad (x_3 \vee \neg x_2)$$

# Examples

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}{(x_2 \vee \neg x_3)}$$

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# Examples

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}{(x_2 \vee \neg x_3)}$$

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# Examples

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (\neg x_1 \vee x_2)}{(x_2 \vee \neg x_3)}$$

$$\frac{(x_1 \vee x_2 \vee \neg x_3) \quad (x_3 \vee \neg x_2)}{(x_1 \vee x_2 \vee \neg x_2)}$$

**tautology** (always satisfied)

# Examples

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$(x)$      $(\neg x)$

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“empty clause”

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“empty clause”

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The empty clause cannot be satisfied.

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“refutation”

↙

## Corollary

If a formula has a refutation it is unsatisfiable.

# Completeness

## Theorem

Every unsatisfiable formula has a refutation.

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**Algorithm for SAT: decide if there is a refutation.**

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Davis & Putnam 1960

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Worst case:  $m^2$  resolvents  
in each iteration.

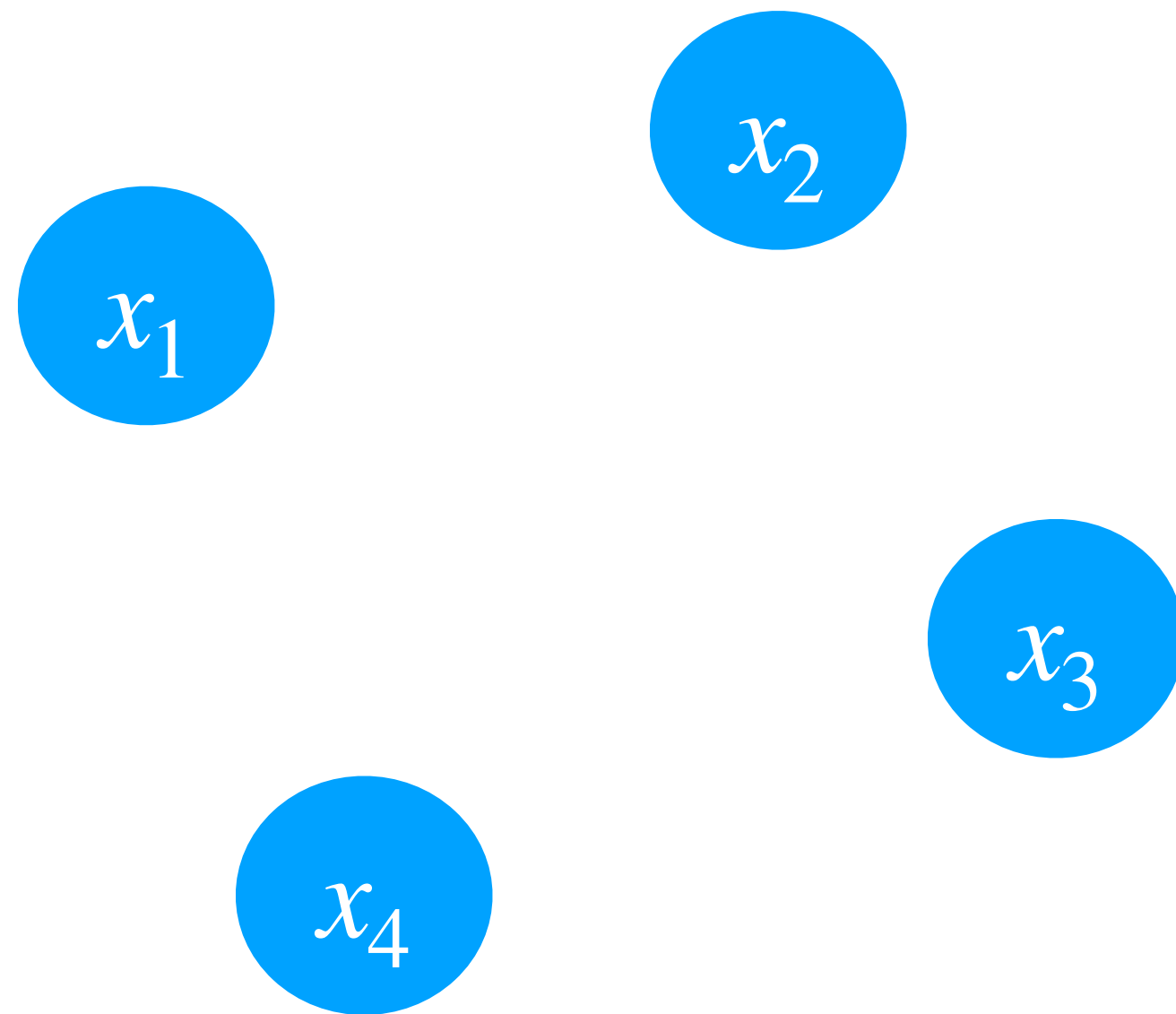
“variable elimination”

# Primal Graph

$$(x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3) \wedge (x_3 \vee \neg x_2) \wedge (x_1 \vee x_2 \vee x_4)$$

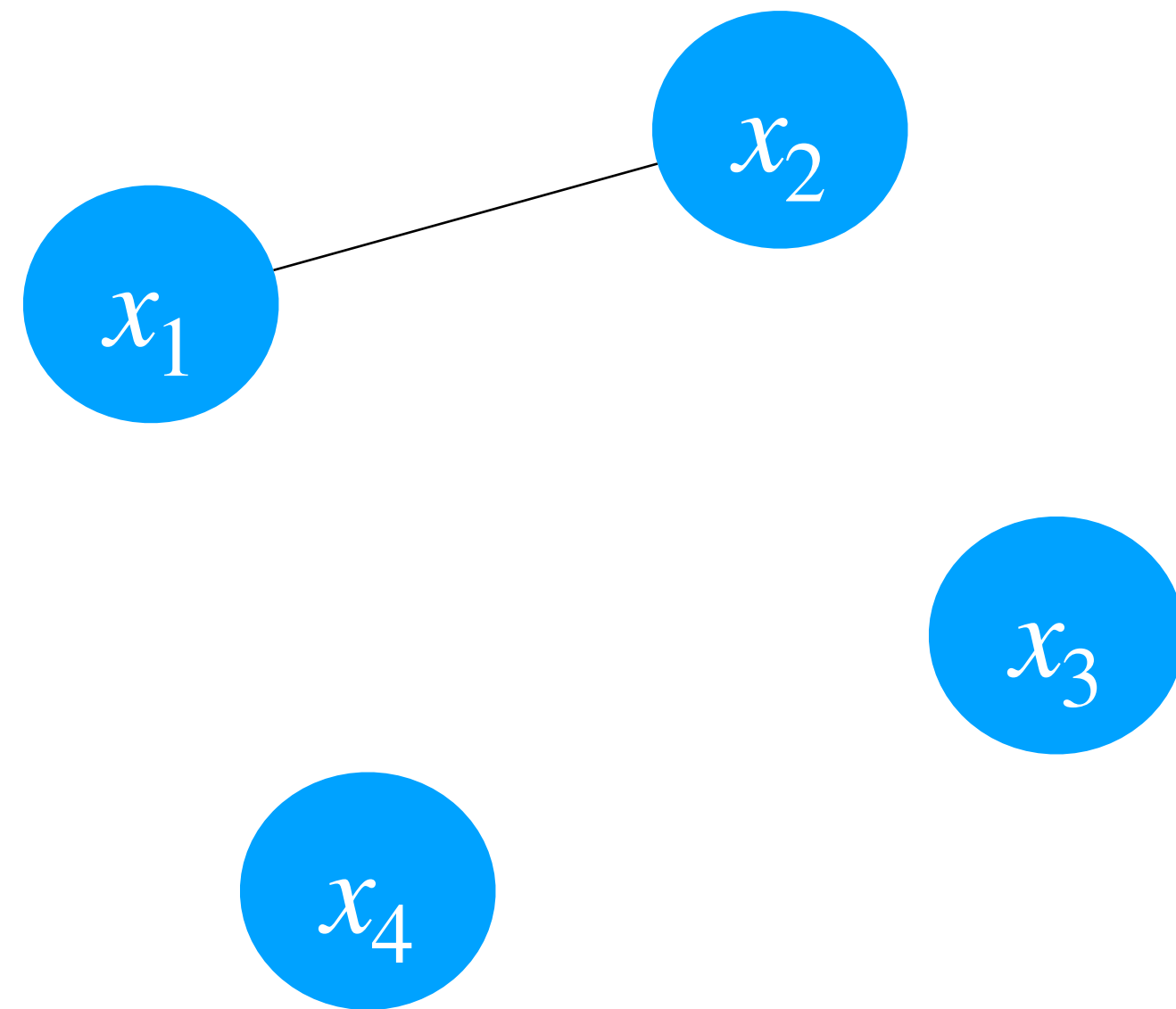
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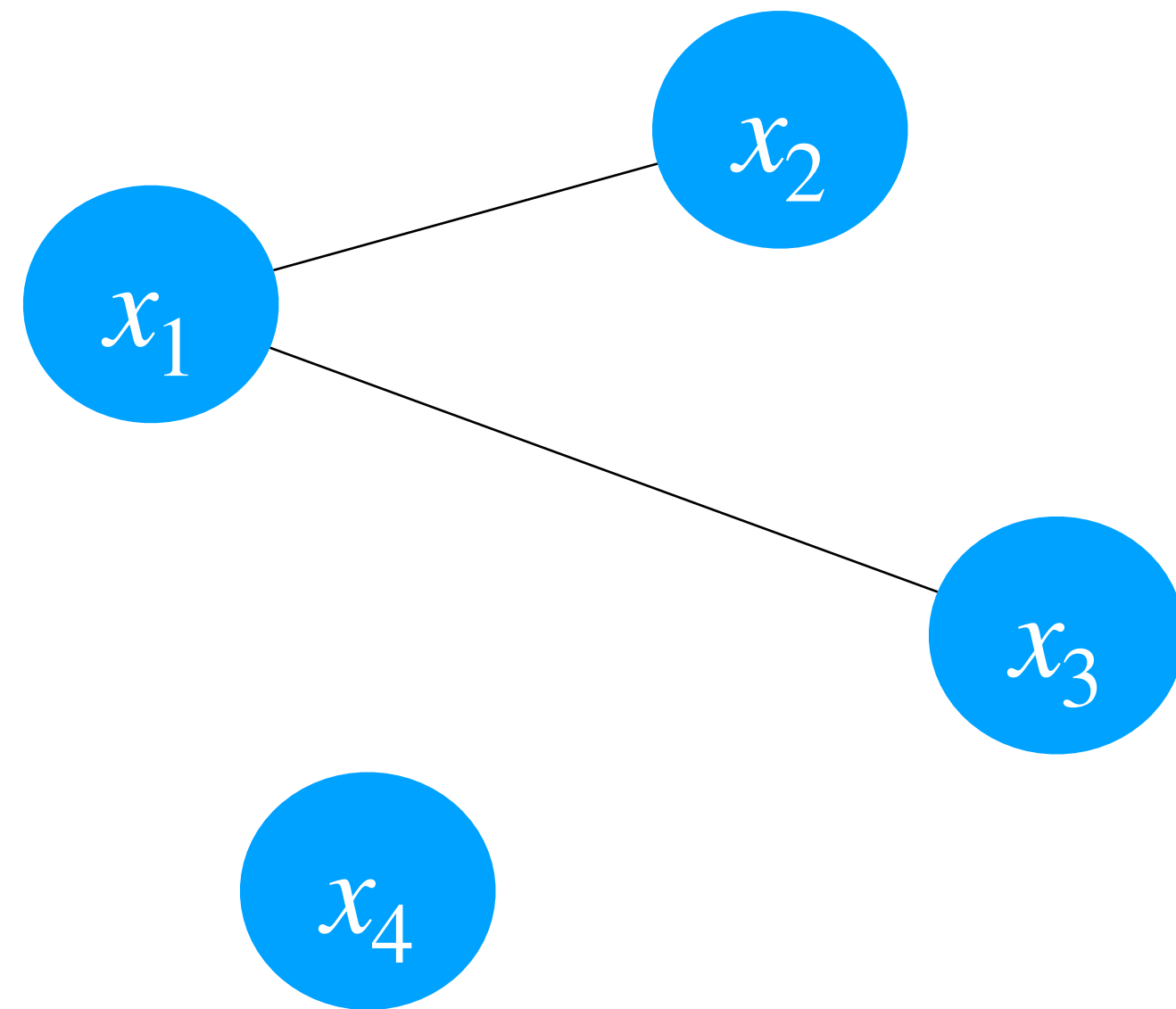
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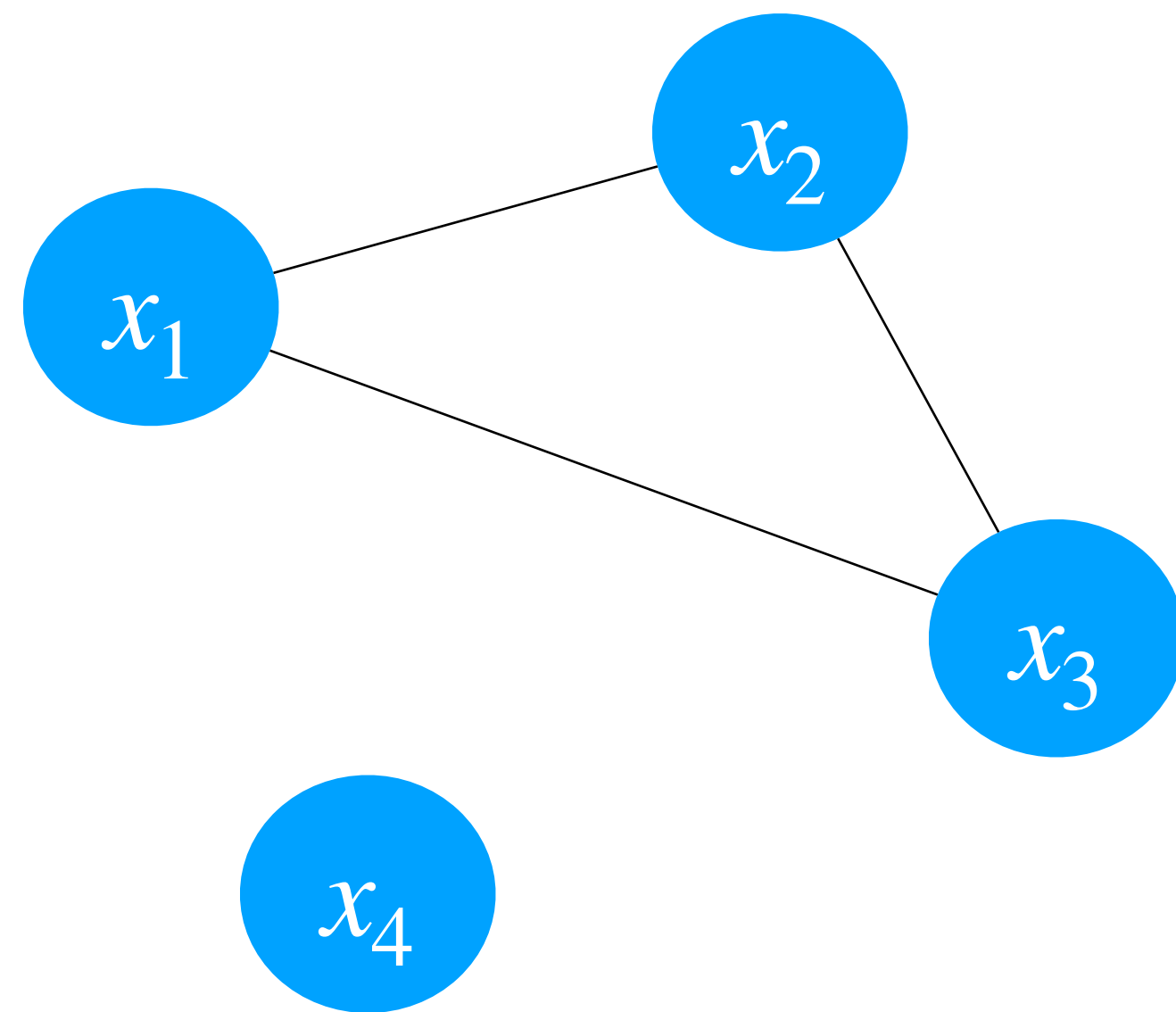
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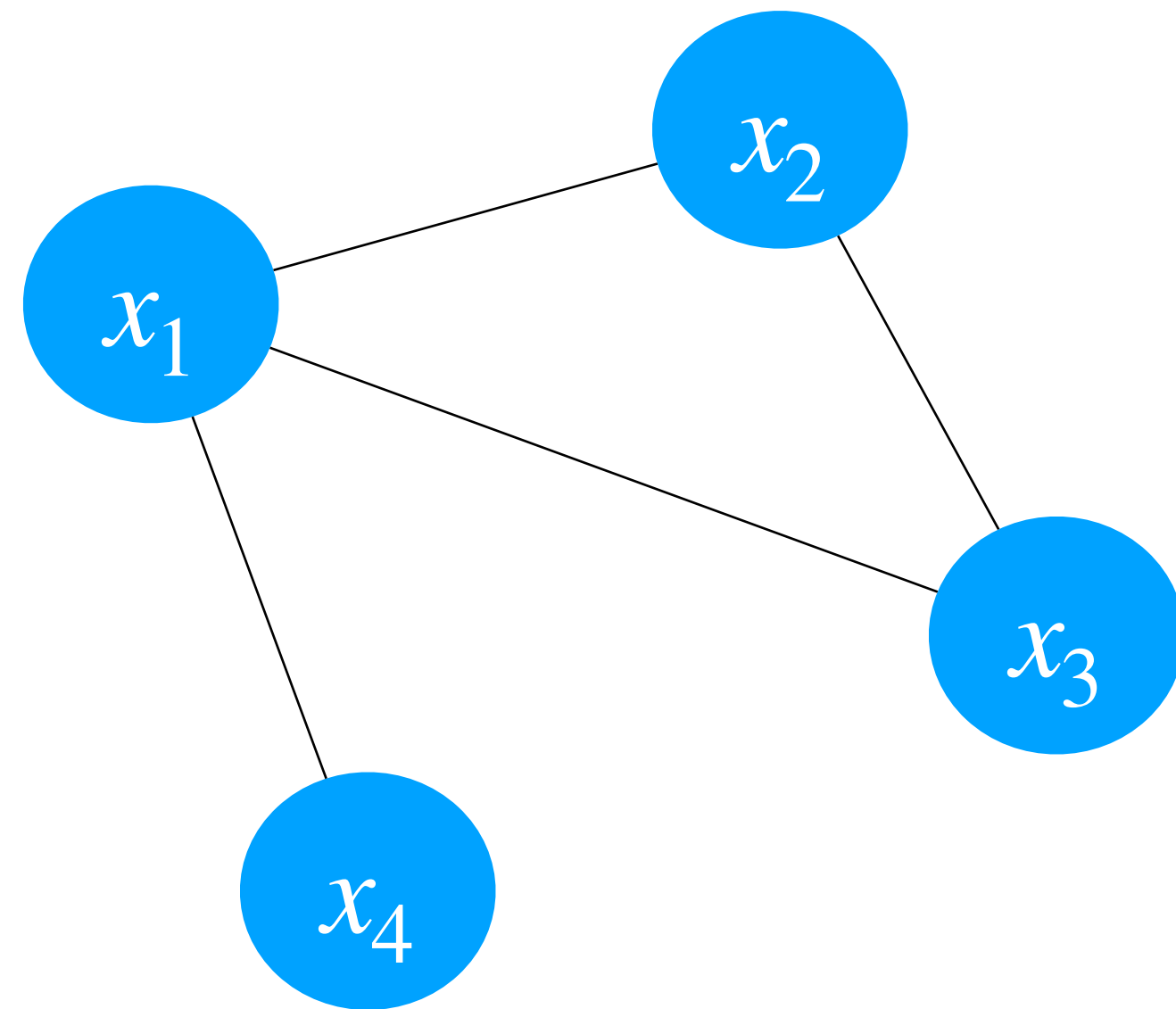
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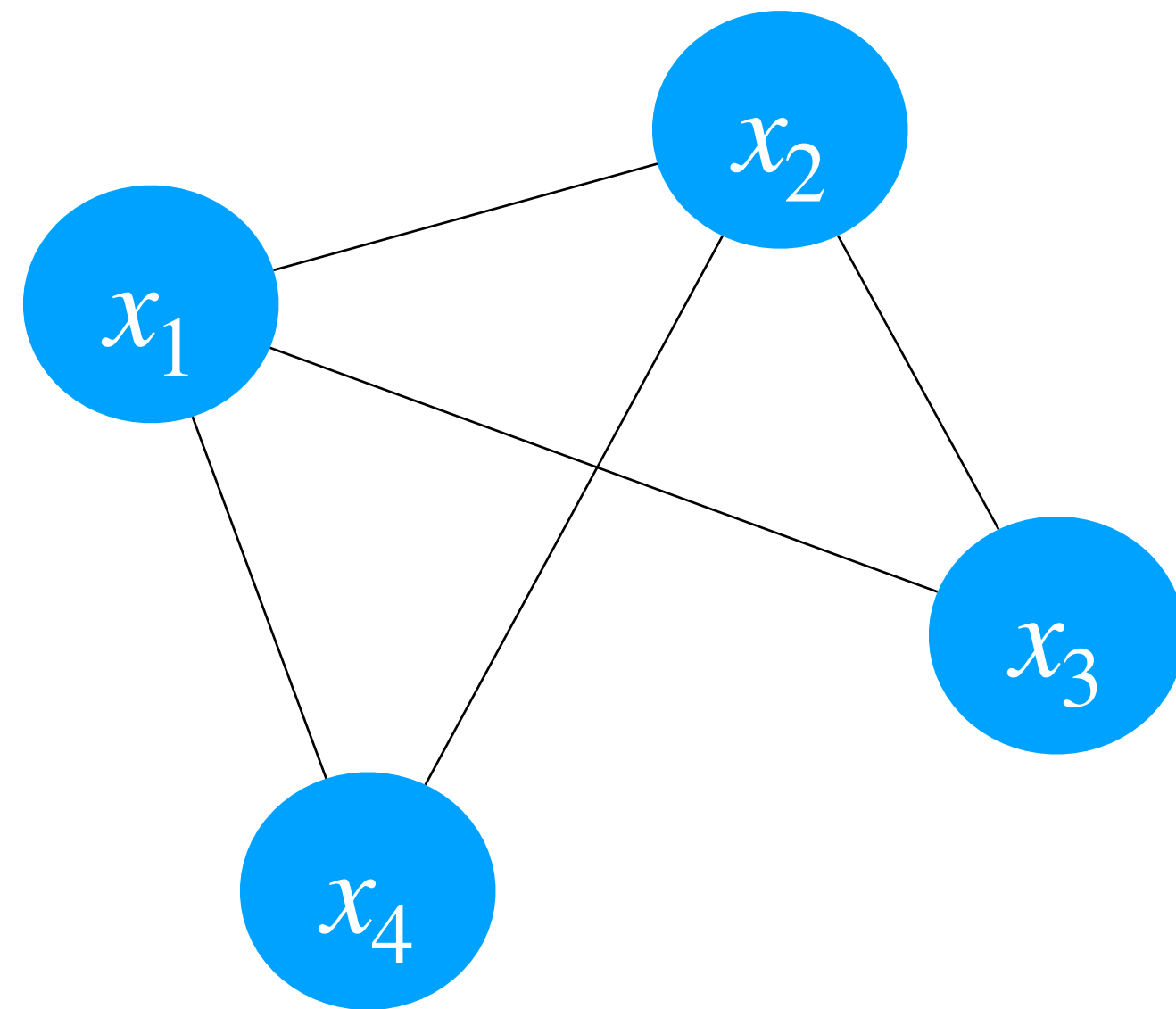
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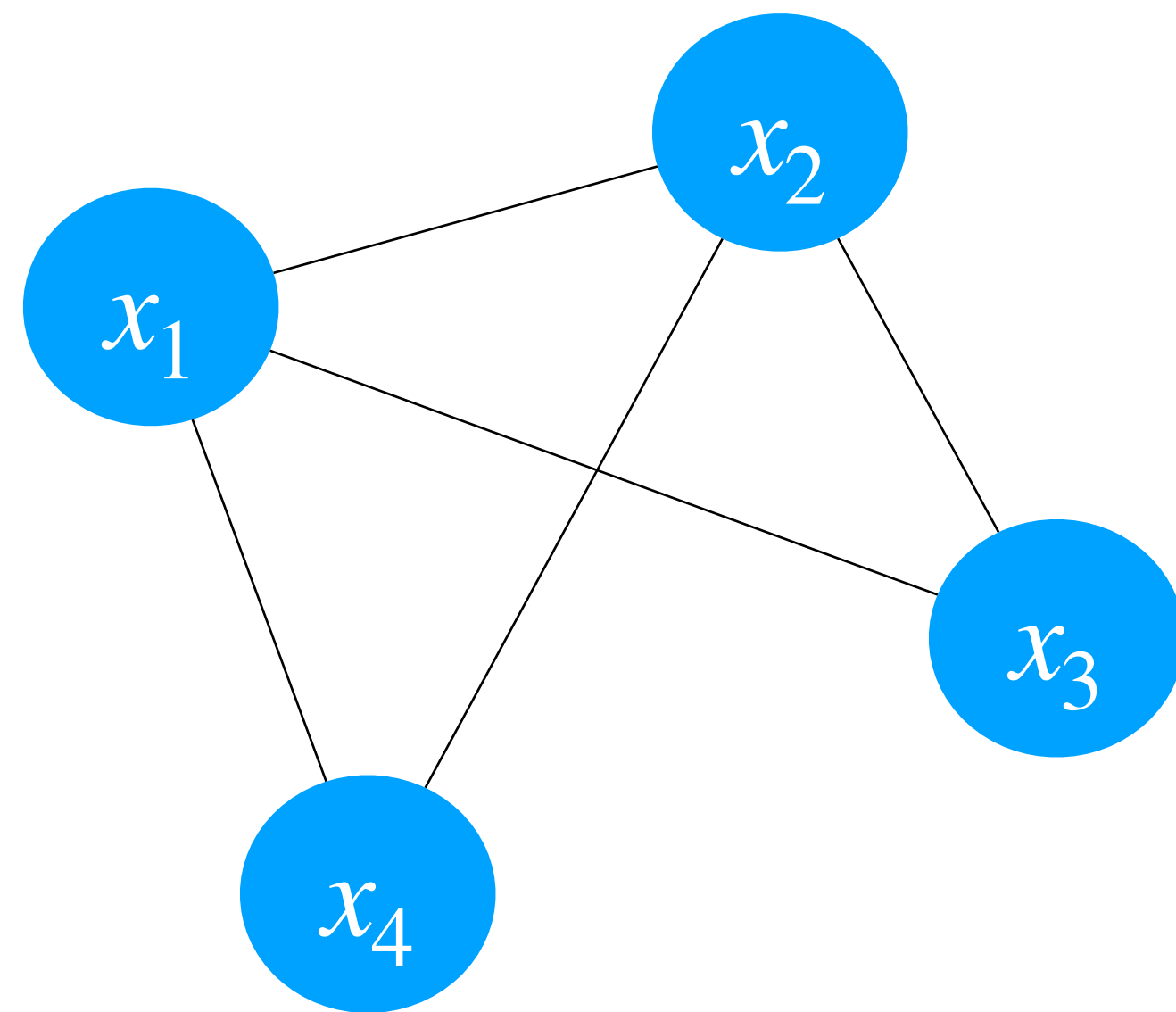
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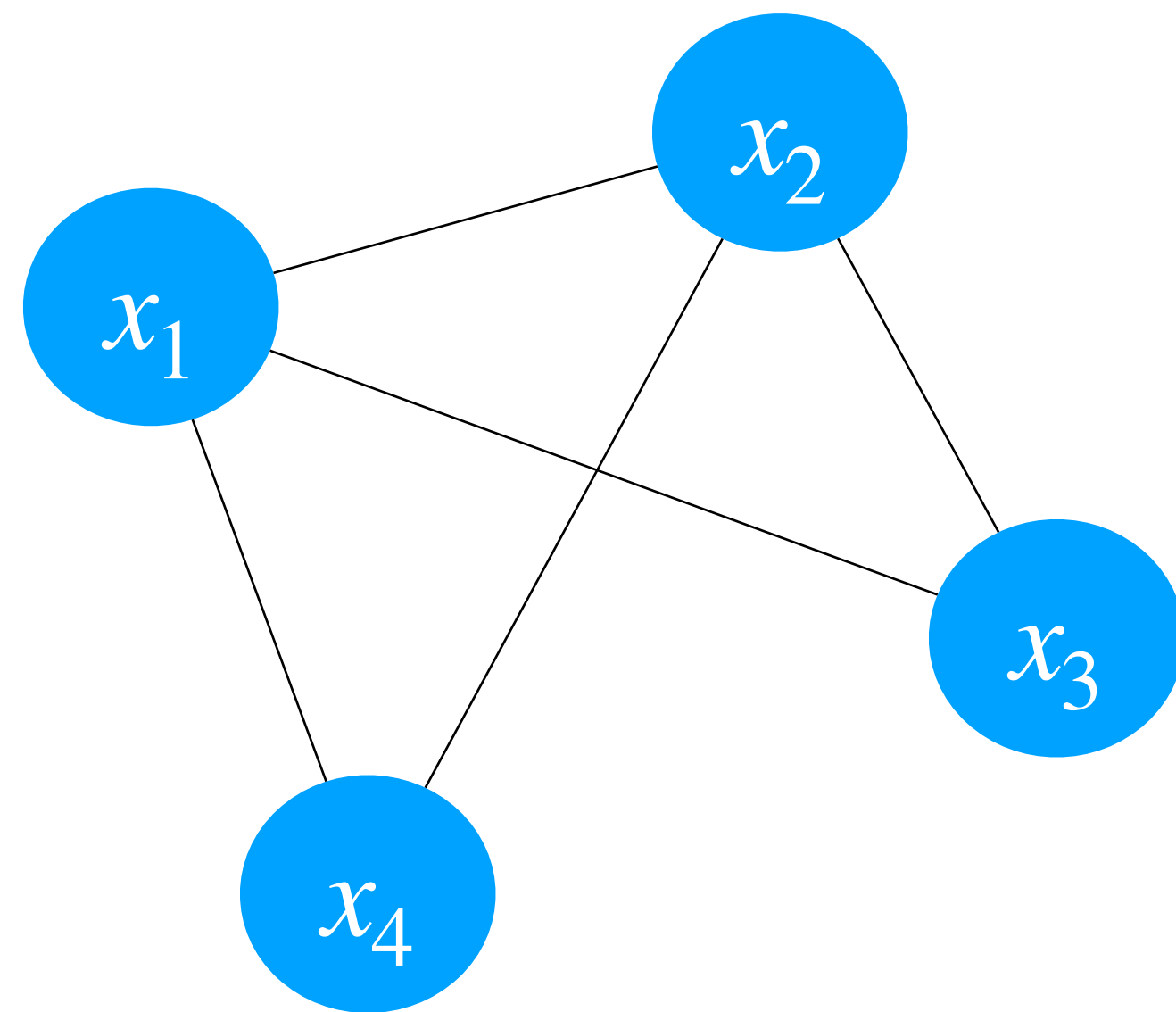
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$$(x_3 \vee \neg x_2)$$

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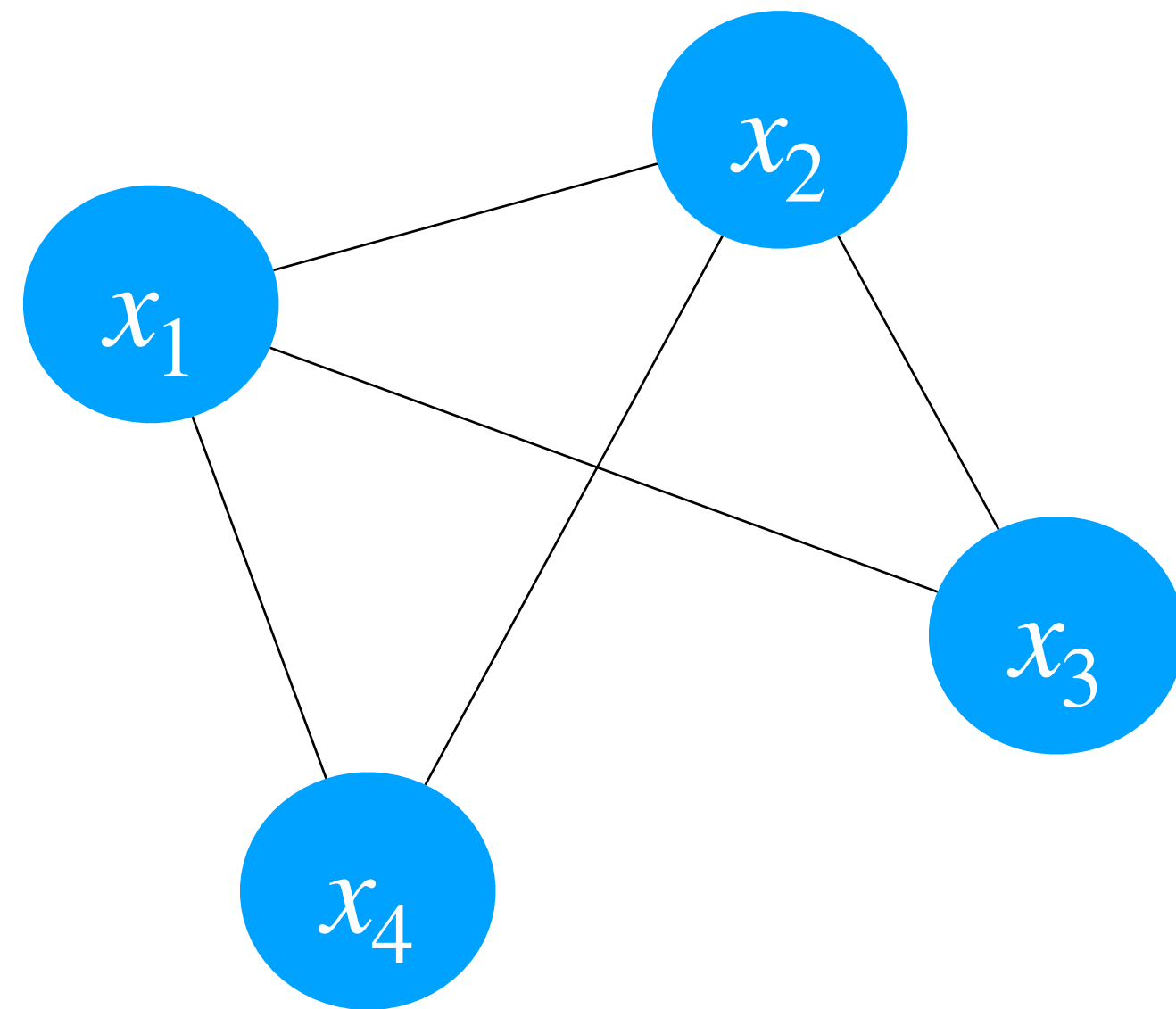


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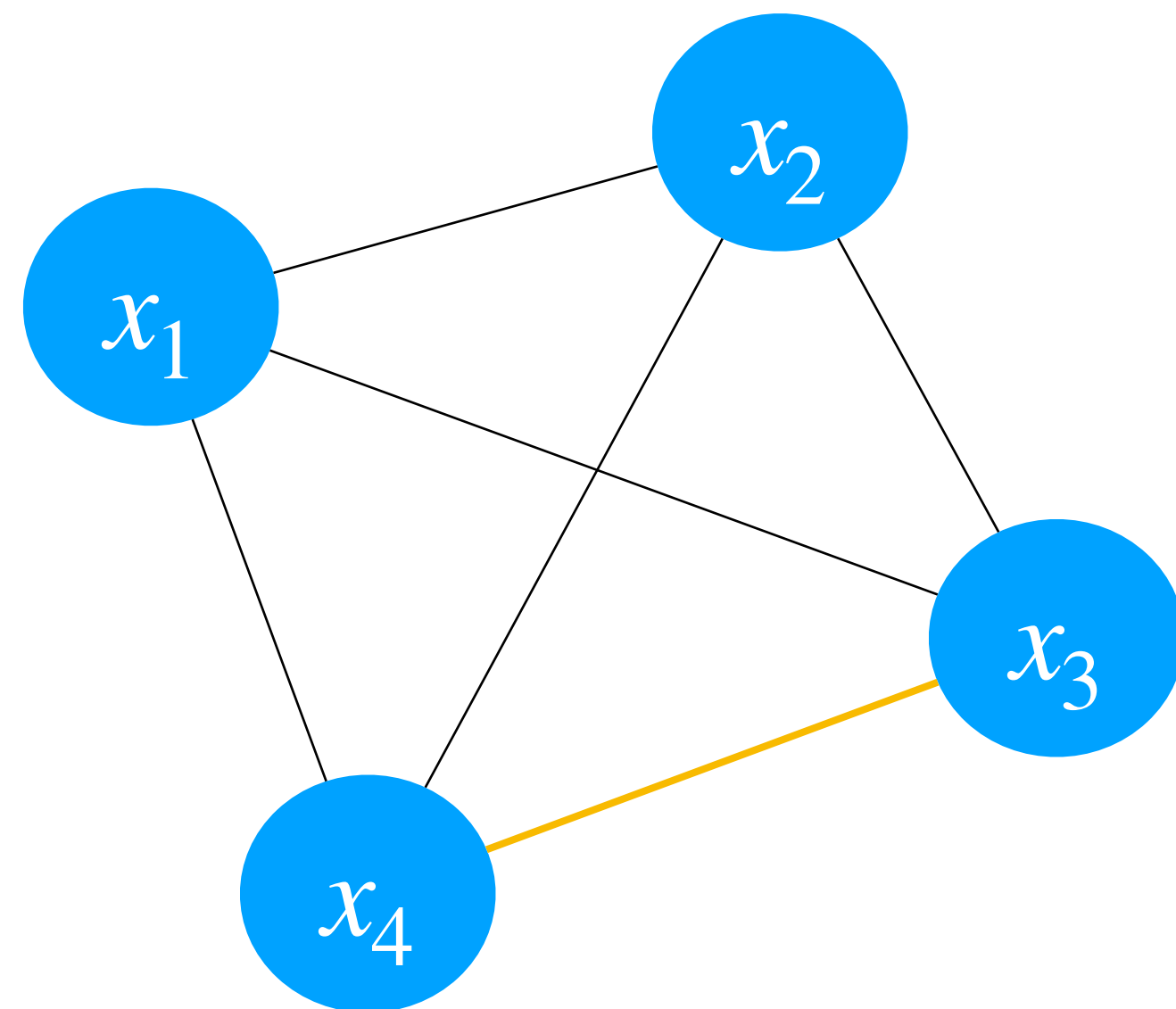
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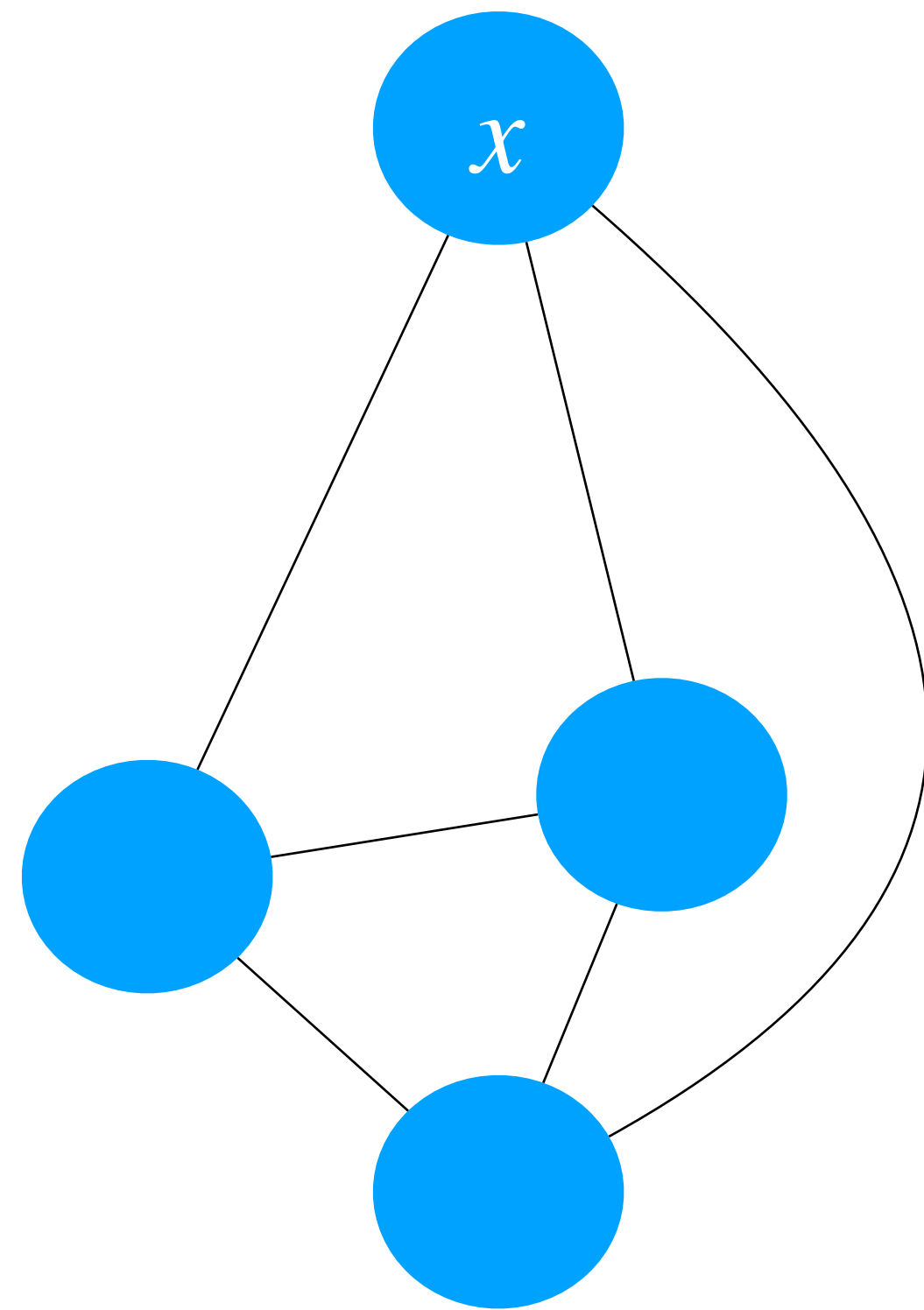
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# Resolution in the Primal Graph

$$\frac{C_1 \vee x \quad \neg x \vee C_2}{C_1 \vee C_2}$$

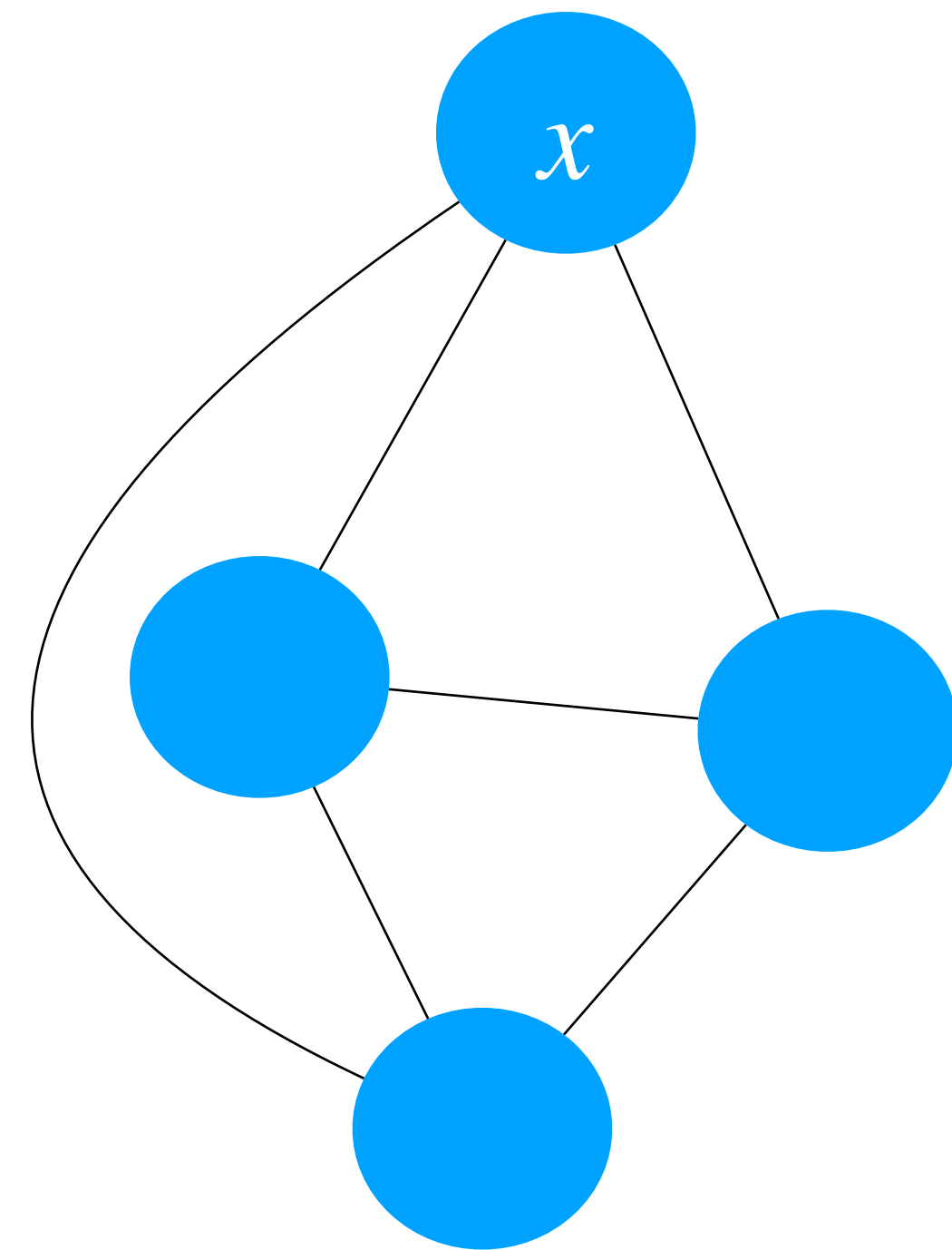
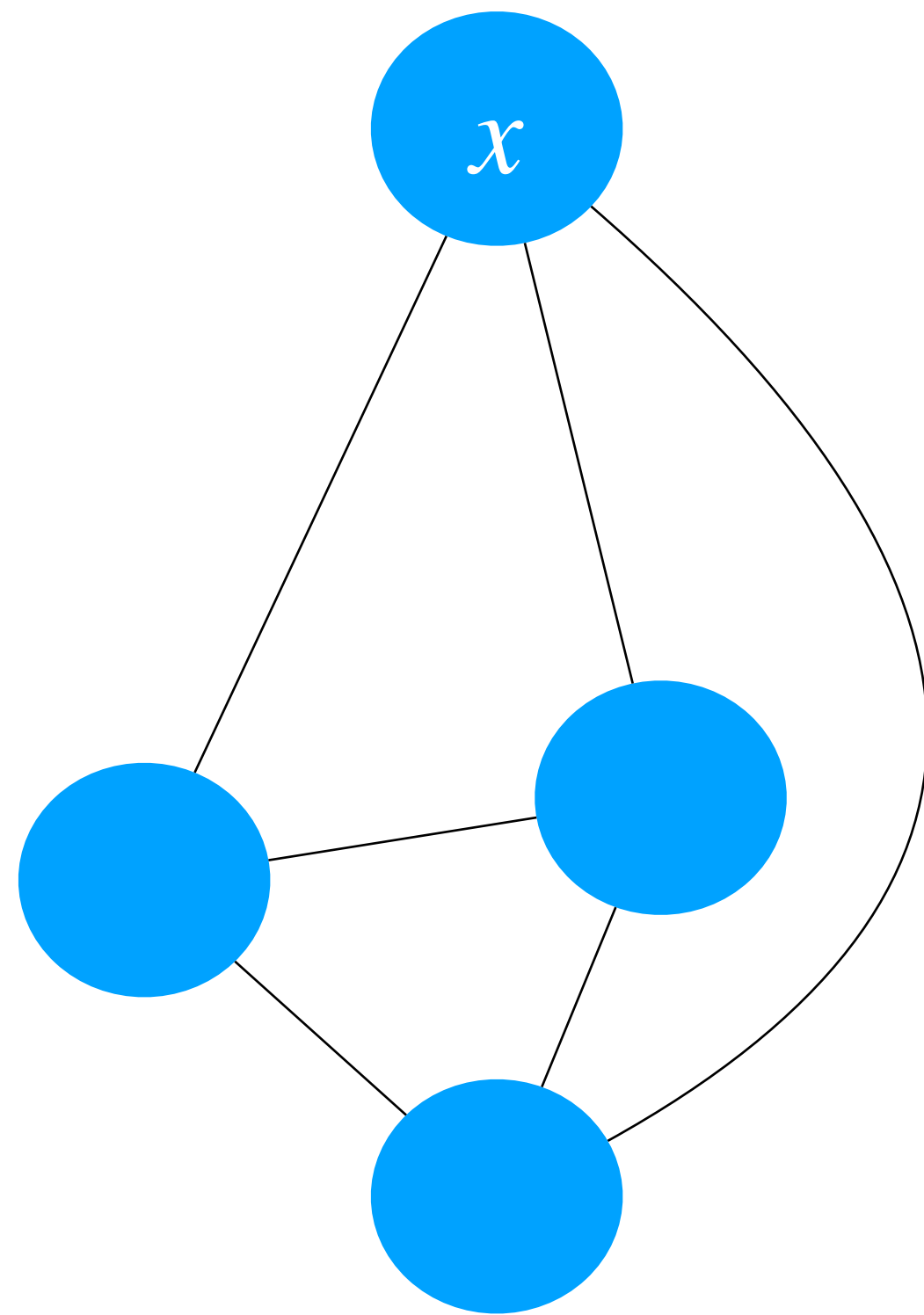
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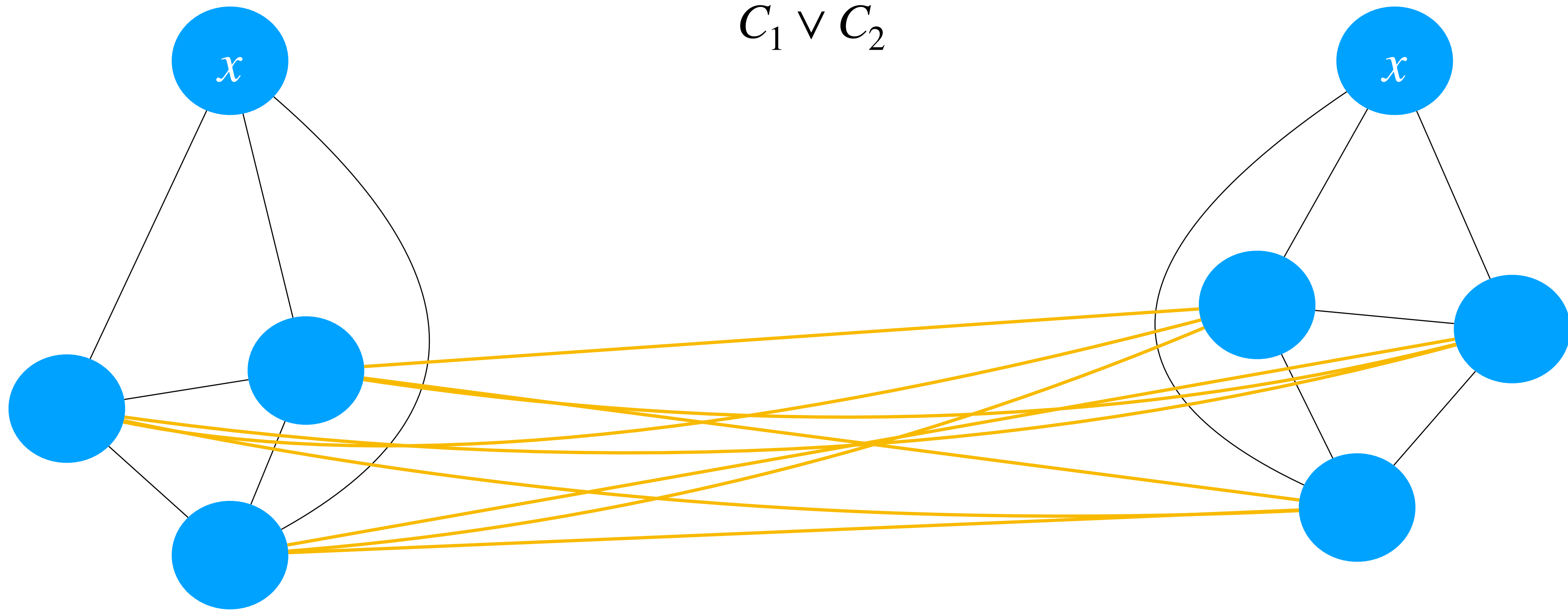
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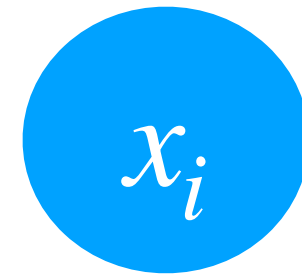
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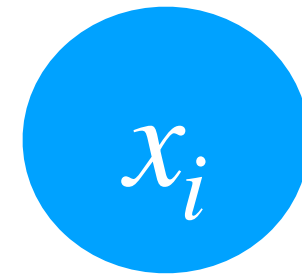


# Variable Elimination in the Primal Graph

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$$C_1 \vee x_i$$

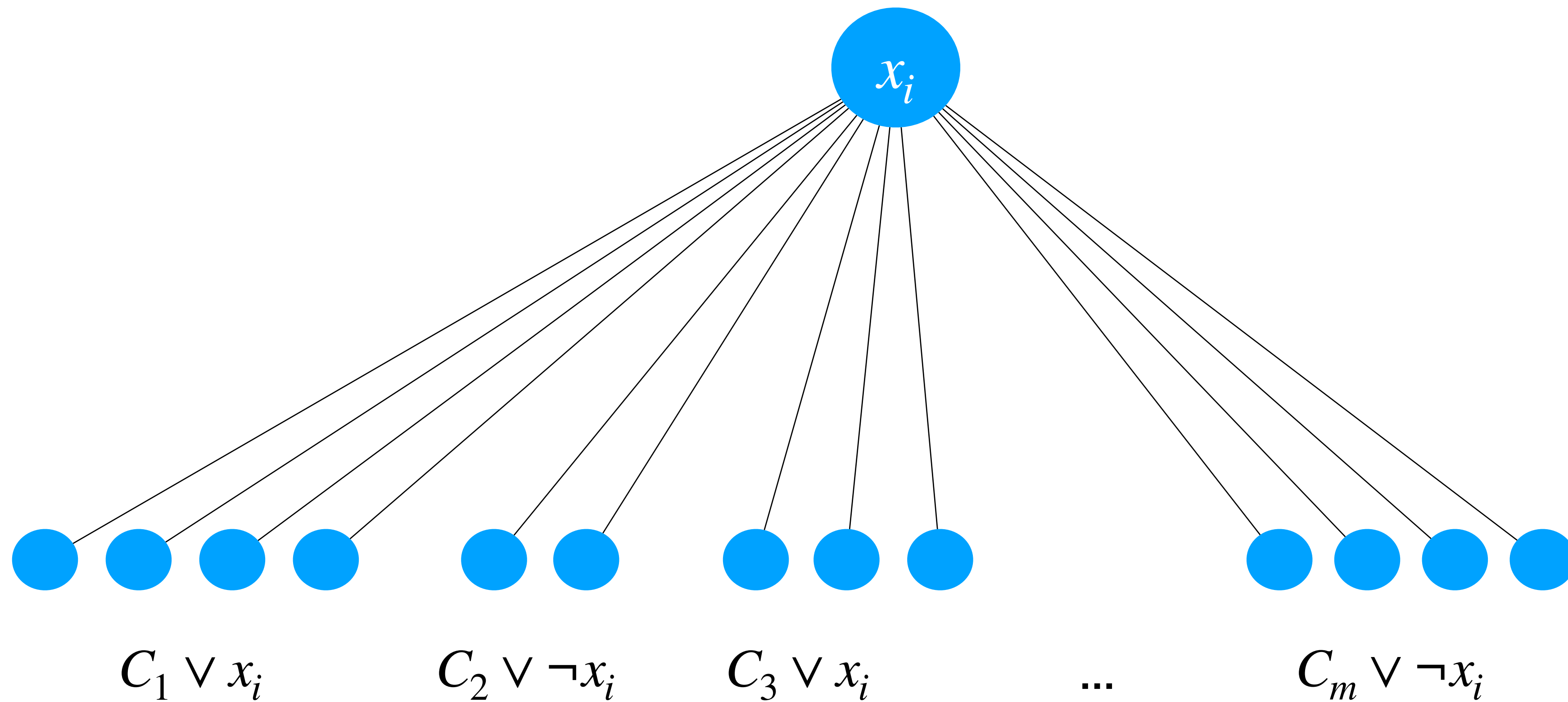
$$C_2 \vee \neg x_i$$

$$C_3 \vee x_i$$

...

$$C_m \vee \neg x_i$$

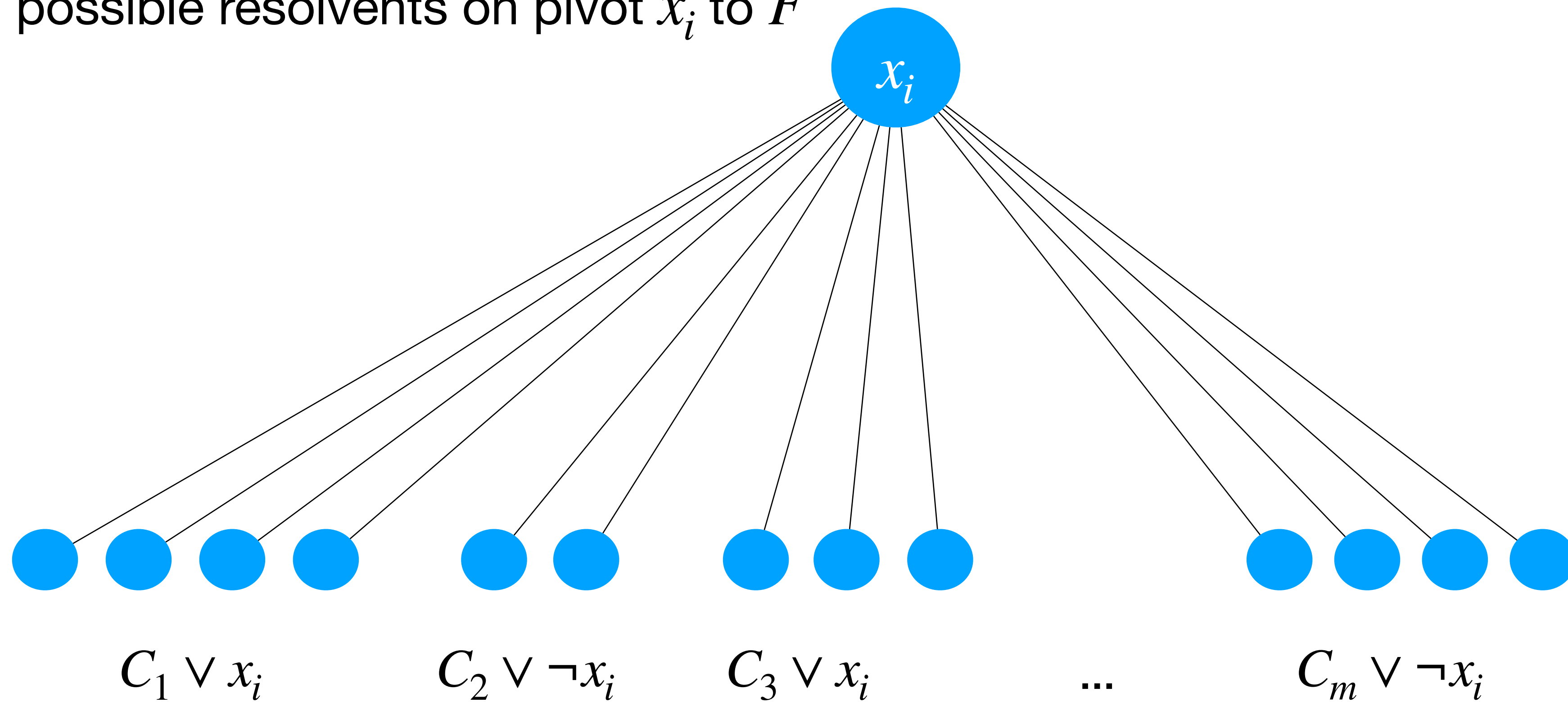
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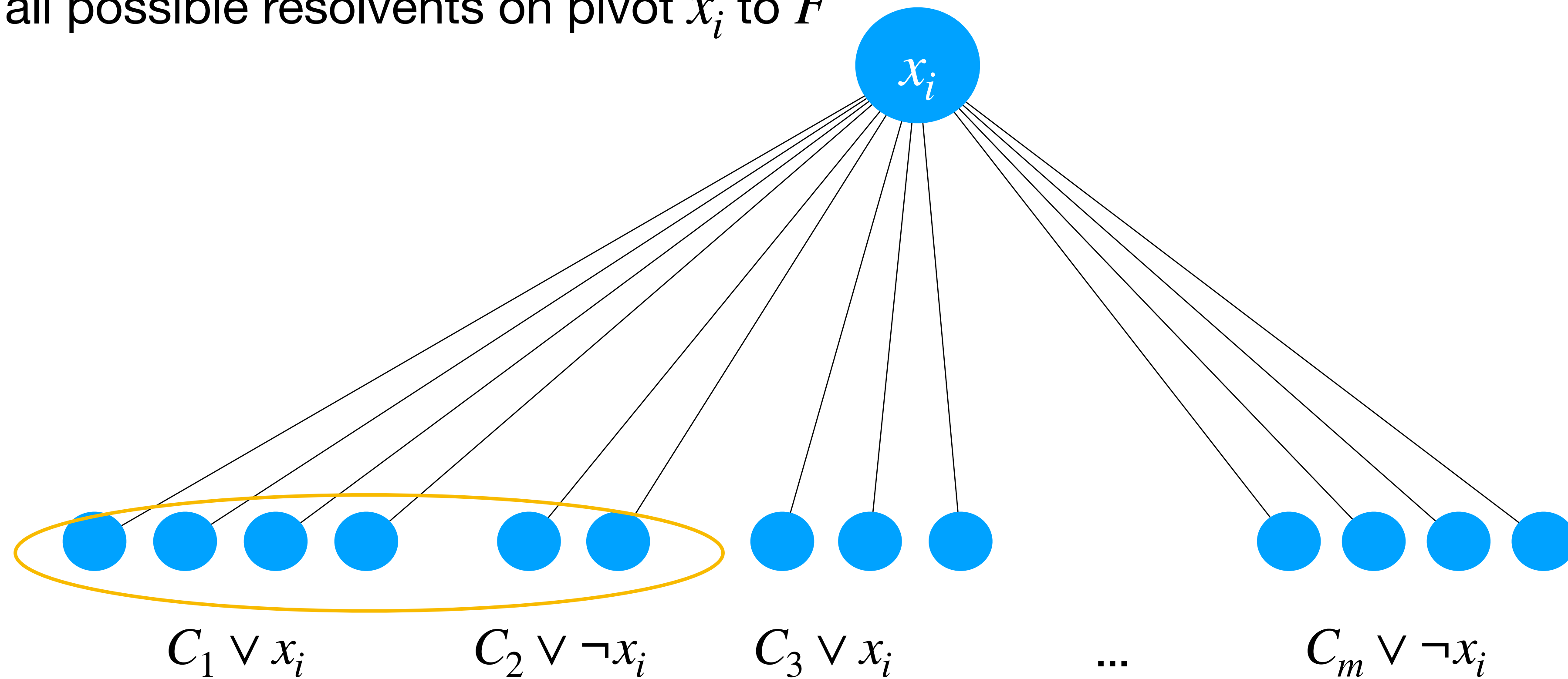
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add all possible resolvents on pivot  $x_i$  to  $F$



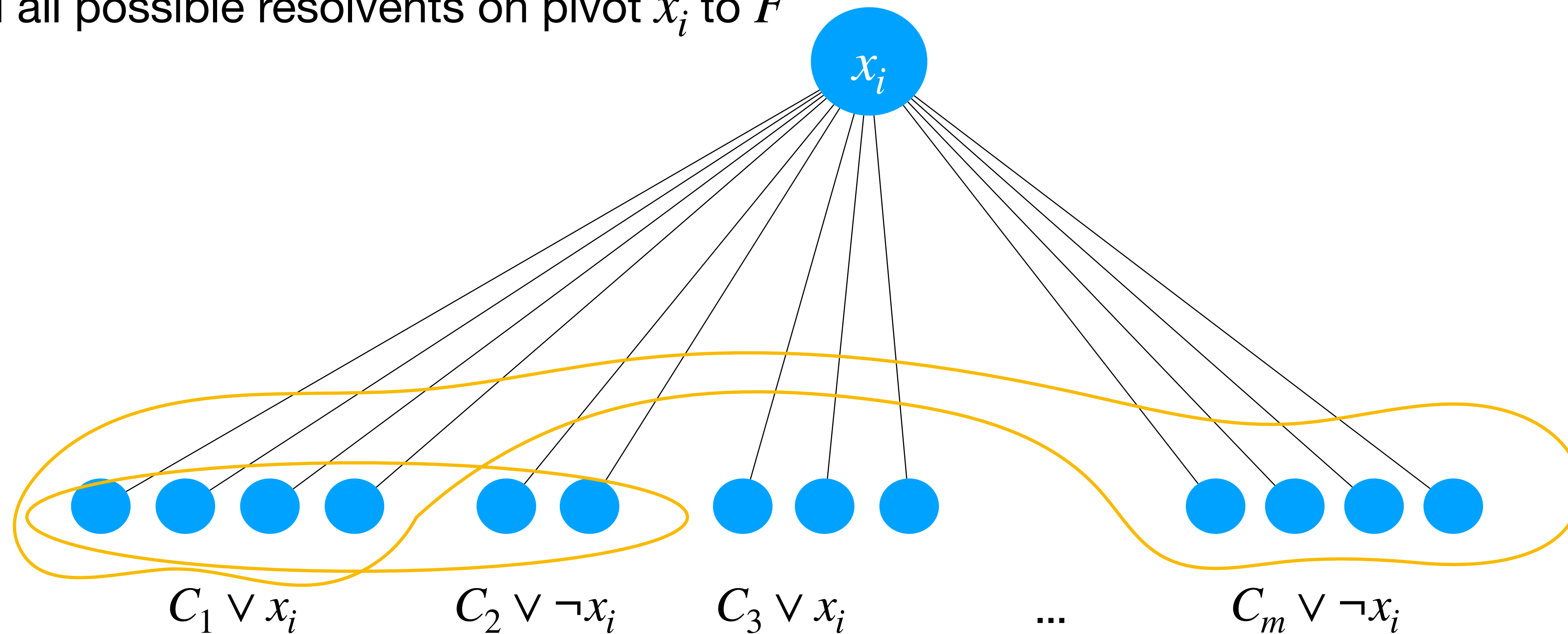
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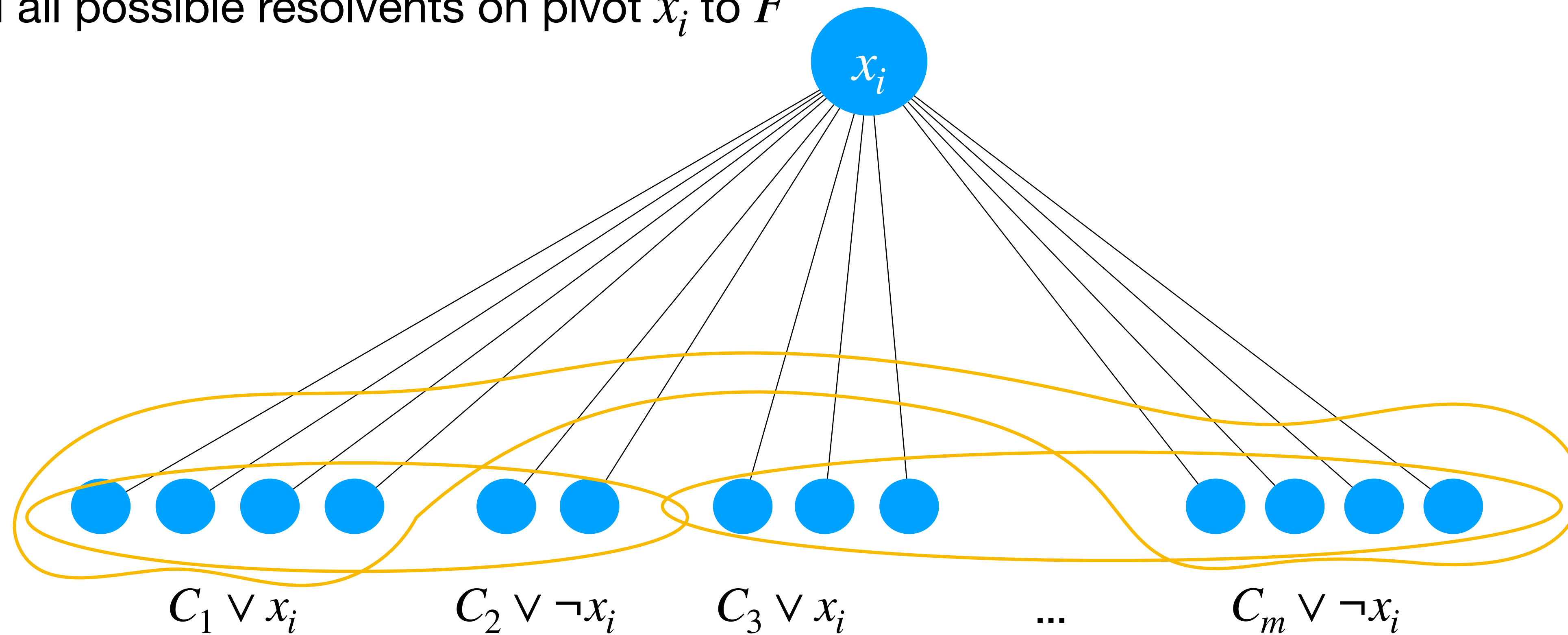
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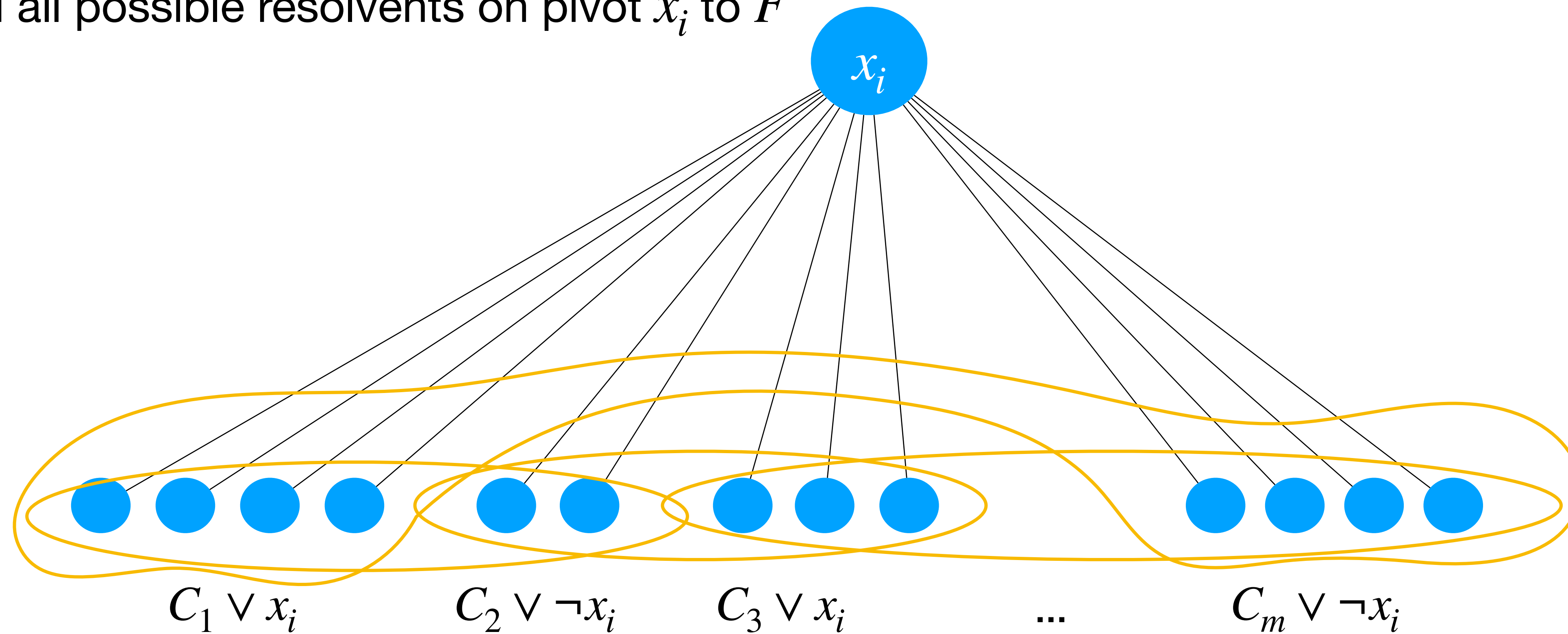
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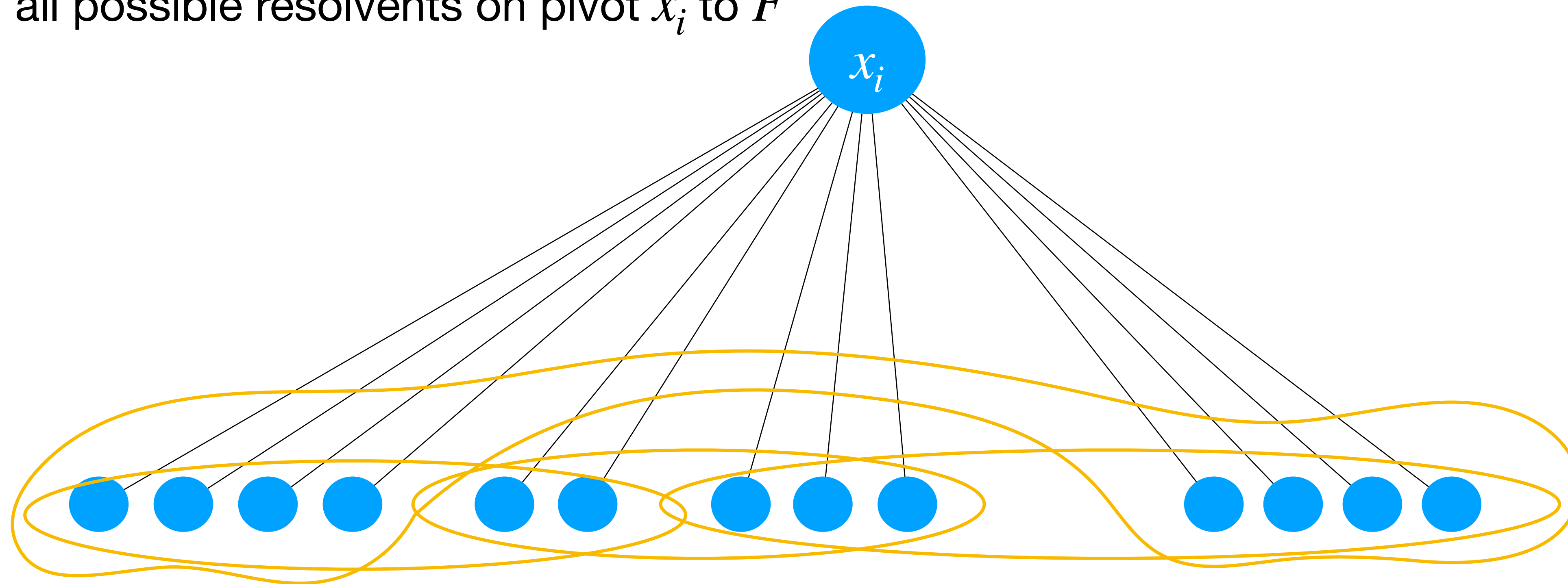
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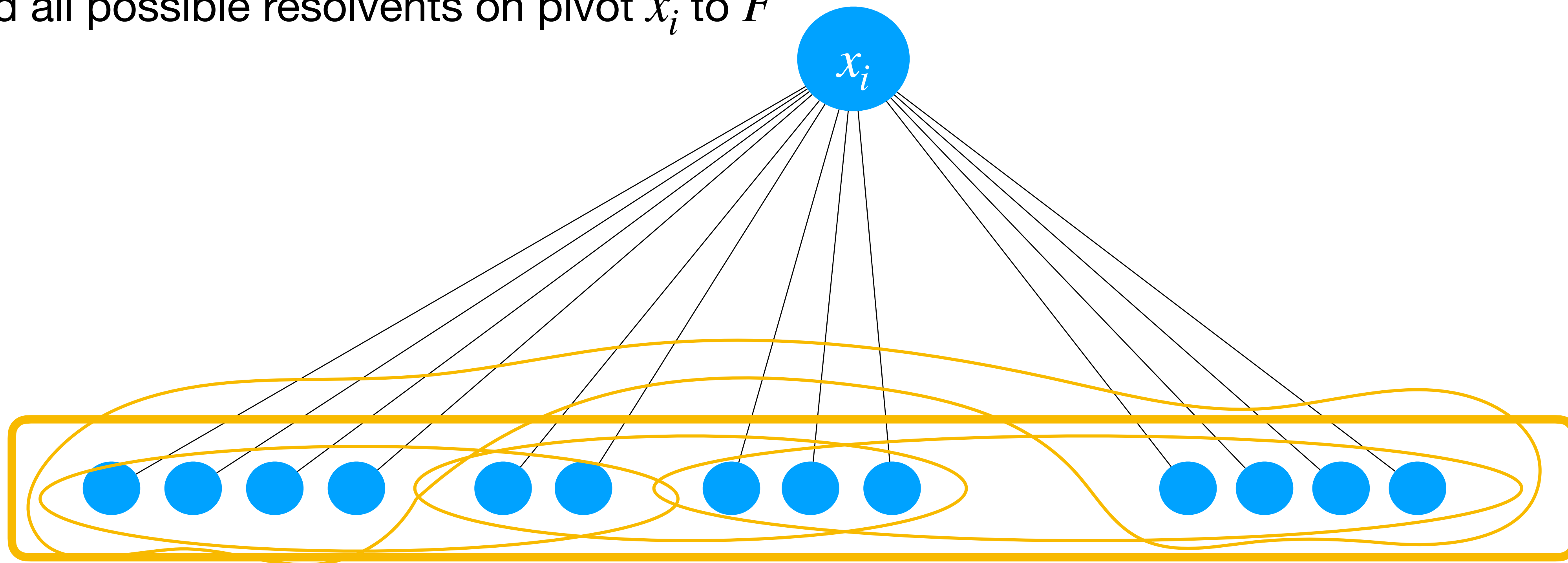
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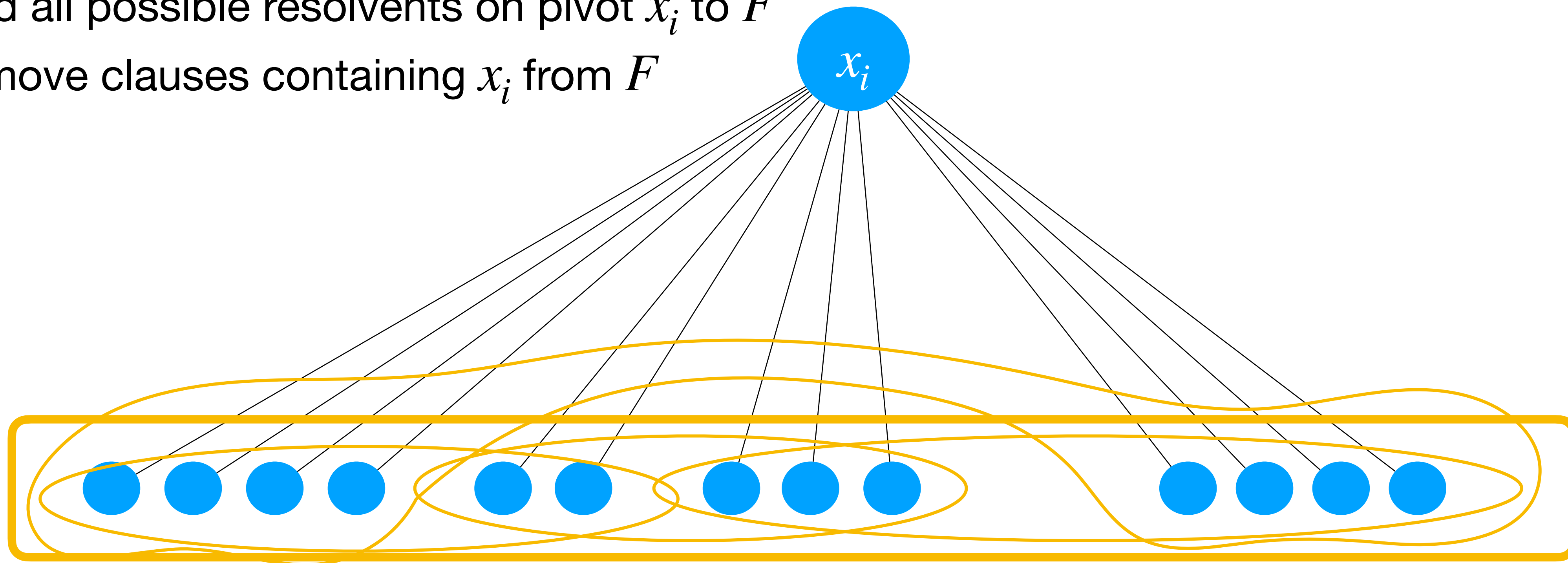
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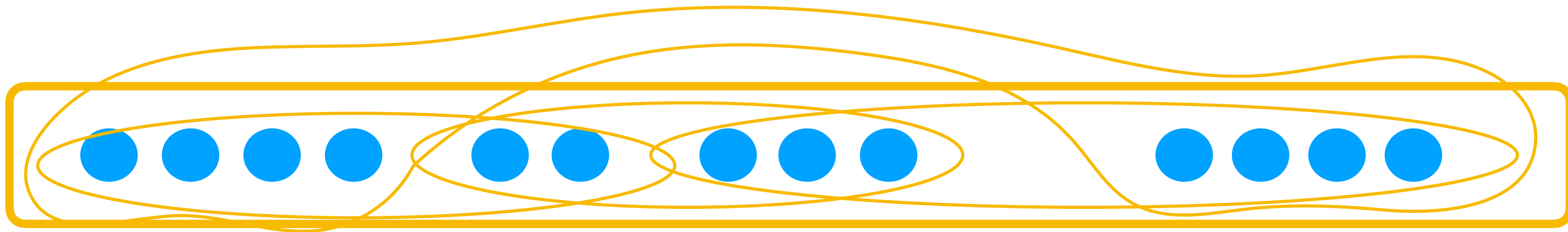
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remove clauses containing  $x_i$  from  $F$





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# DP-Resolution and Elimination Orderings

**Input:** A CNF formula  $F$  with  $m$  clauses

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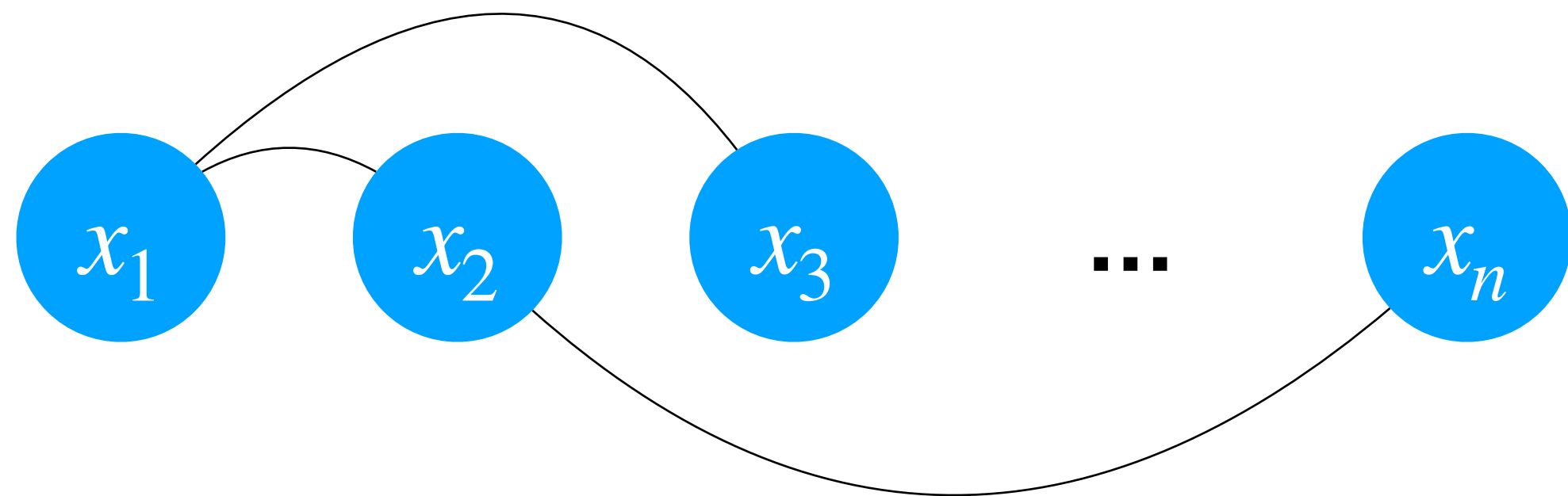
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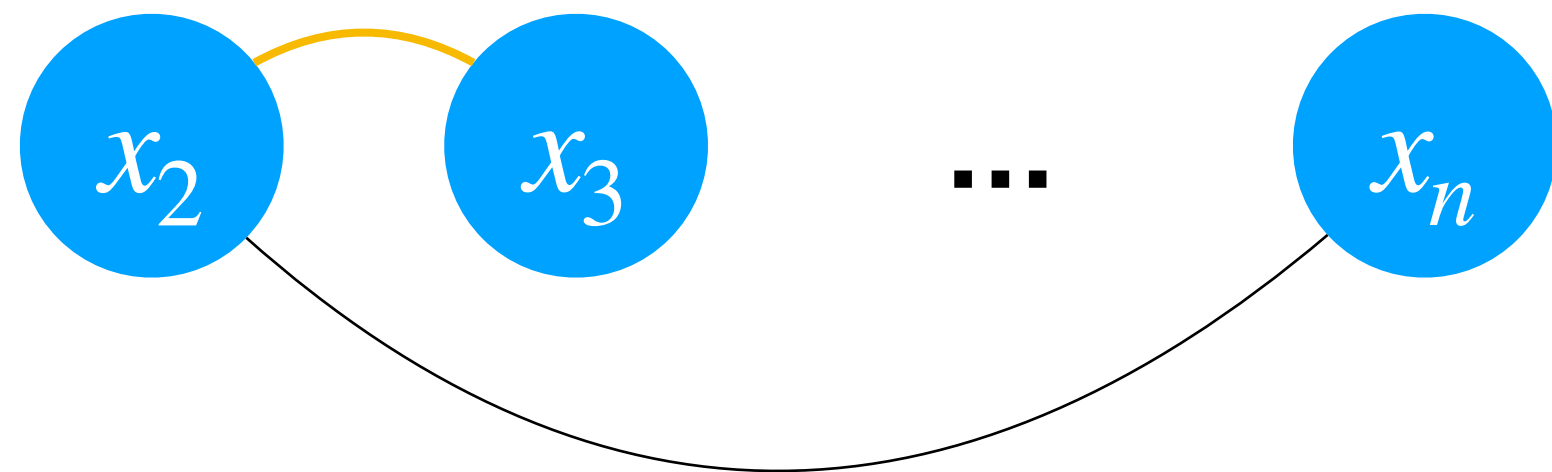
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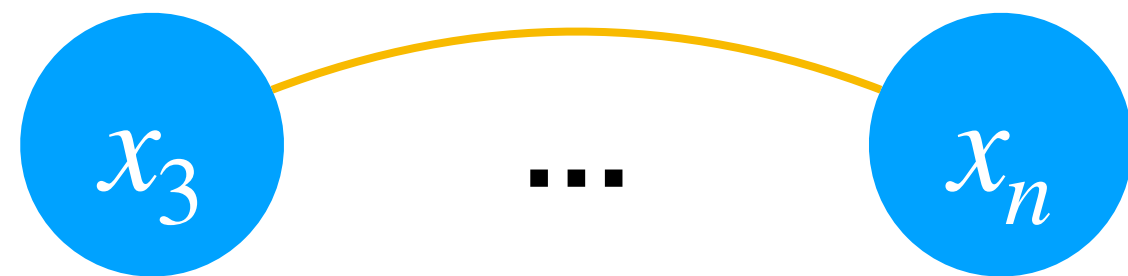
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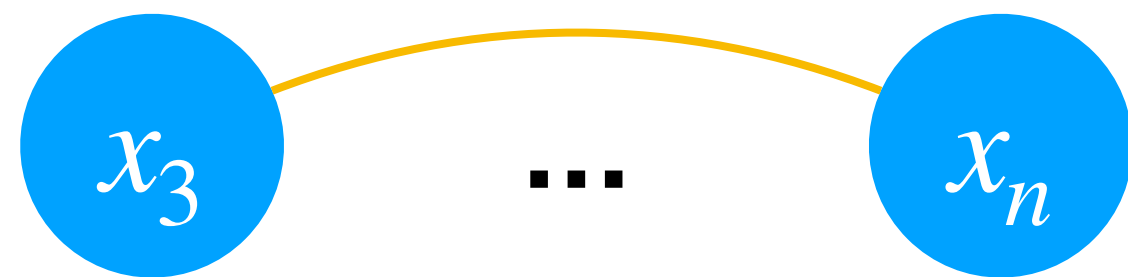
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**Observation**

The size of any clause generated by Davis-Putnam Resolution is at most the width  $k$  of the elimination ordering in the primal graph.



# DP-Resolution and Elimination Orderings

**Input:** A CNF formula  $F$  with  $m$  clauses

Pick an ordering  $\sigma := x_1, \dots, x_n$  of variables

**for**  $x_i$  **in**  $\sigma$  :

add all possible resolvents on pivot  $x_i$  to  $F$

remove clauses containing  $x_i$  from  $F$

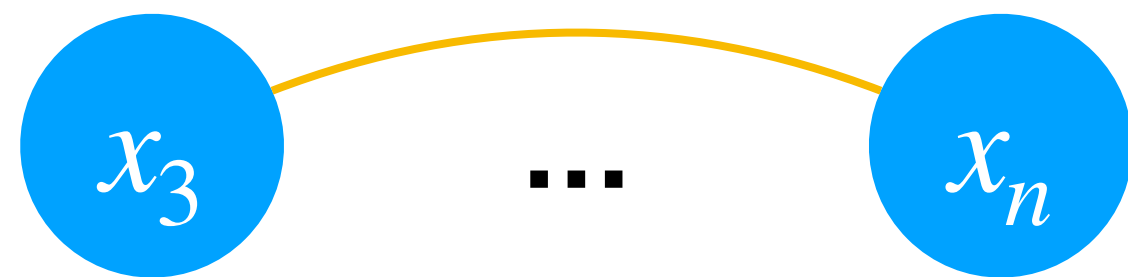
remove tautologies from  $F$

**return false**  $\Leftrightarrow F$  contains the empty clause

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Worst case:  $3^k$  resolvents in each iteration.



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## Theorem

SAT is FPT parameterized by the treewidth of the primal graph.

