

# Fundamental Graph Algorithms

## Part One

# Announcements

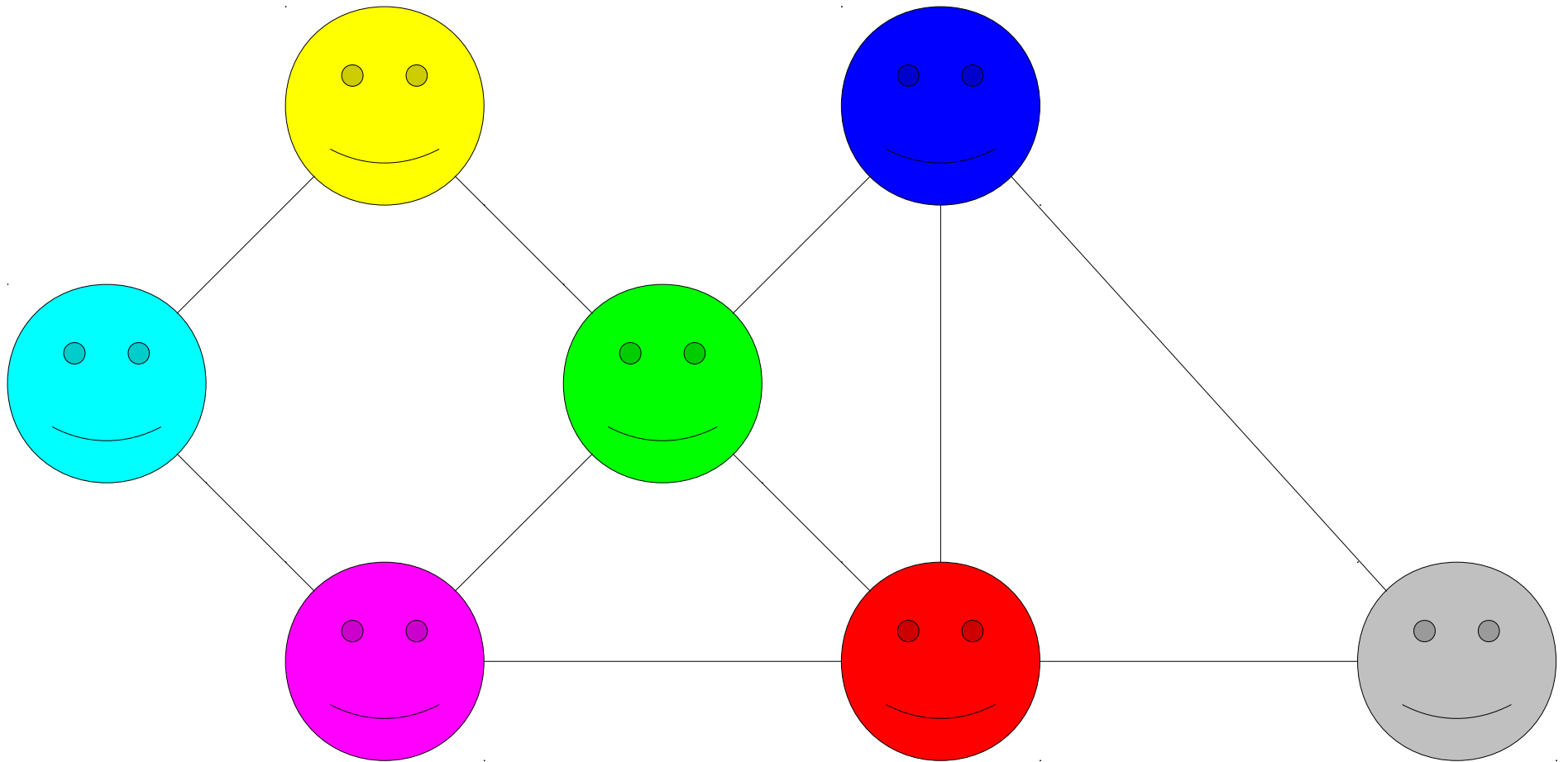
- Problem Set One out, due **Wednesday, July 3**.
  - Play around with  $O$ ,  $\Omega$ , and  $\Theta$  notations!
  - Get your feet wet designing and analyzing algorithms.
  - Explore today's material on graphs.
- Can be completed using just material from the first two lectures.
- We suggest reading through the handout on how to approach the problem sets. There's a lot of useful information there!
- Office hours schedule will be announced tomorrow.

# Announcements

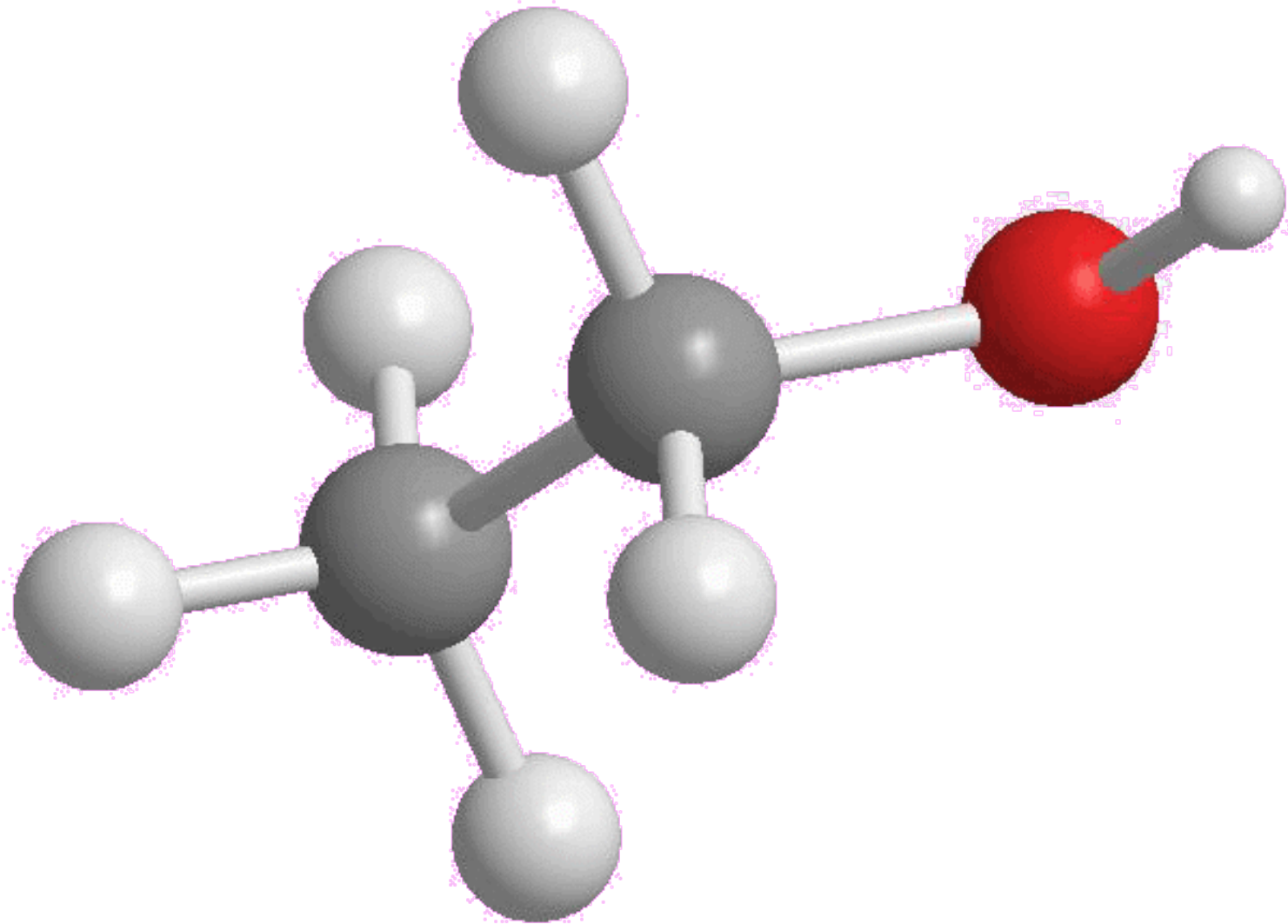
- We will not be writing any code in CS161; we'll focus more on the design and analysis techniques.
- Each week, we will have an optional programming section where you can practice coding up these algorithms.
- Run by TA Andy Nguyen, who coaches Stanford's ACM programming team.
- Meets **Thursdays, 4:15PM - 5:05PM** in **Gates B08**.

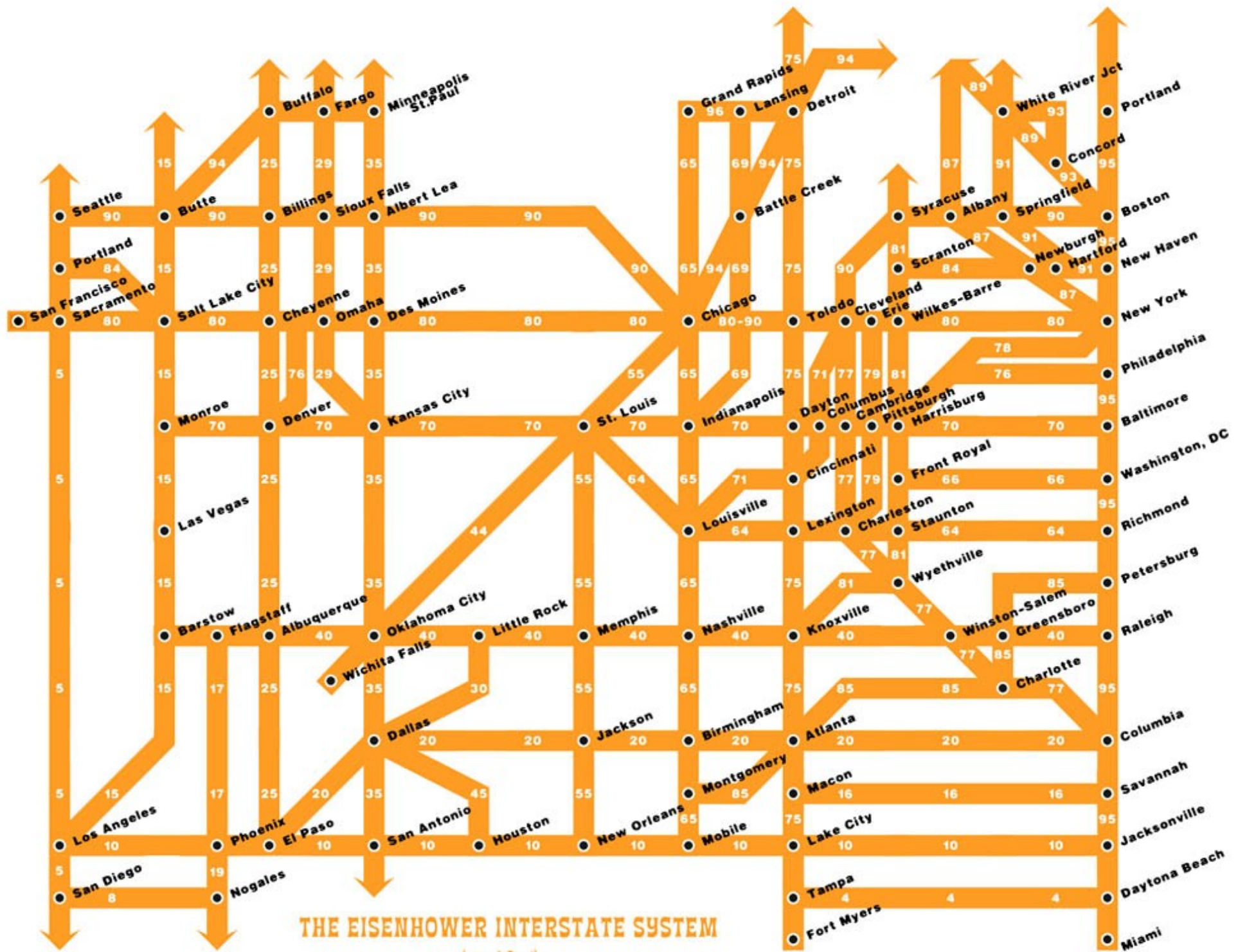
# Graphs

# A Social Network

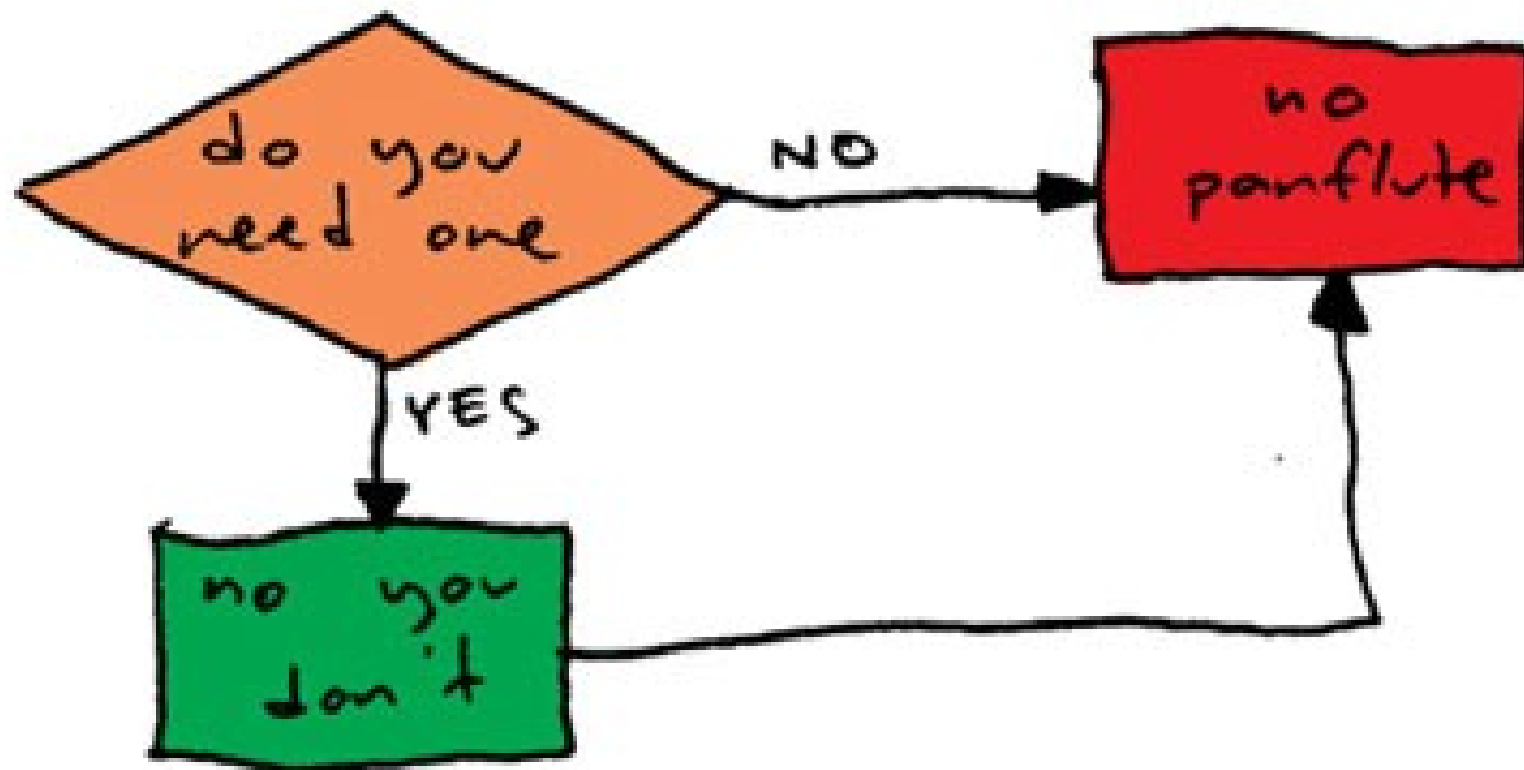


# Chemical Bonds

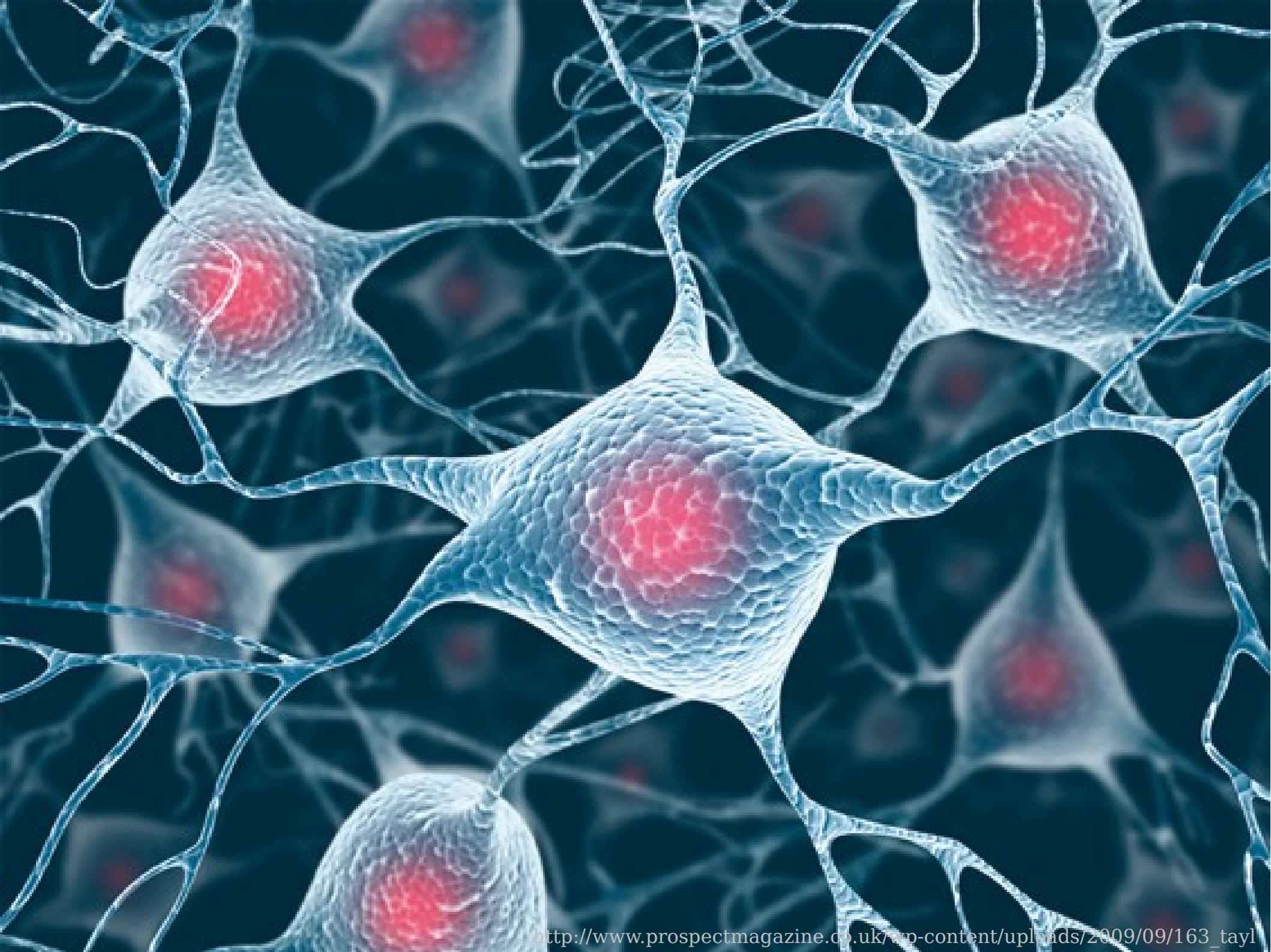


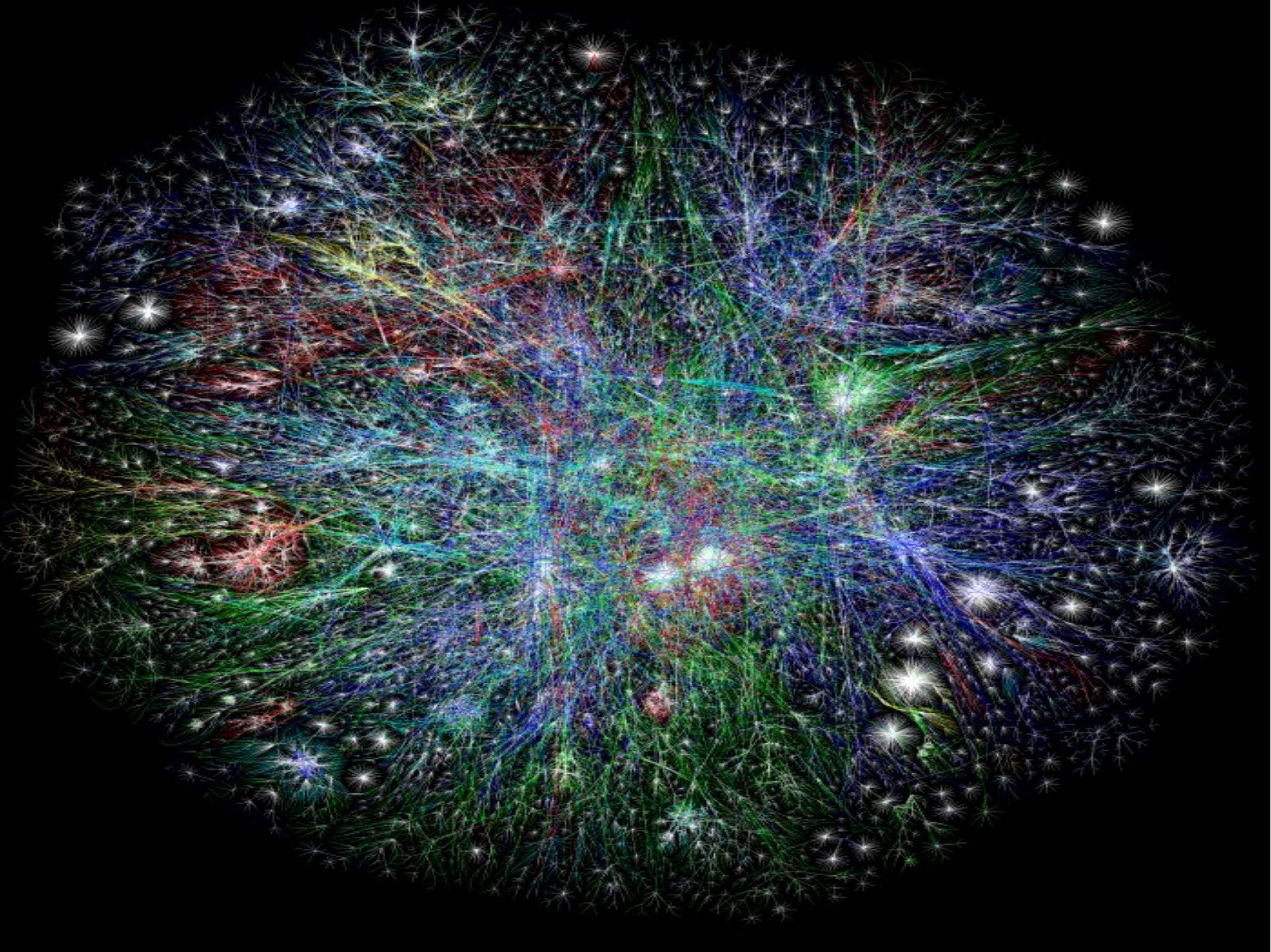


# PANFLUTE FLOWCHART





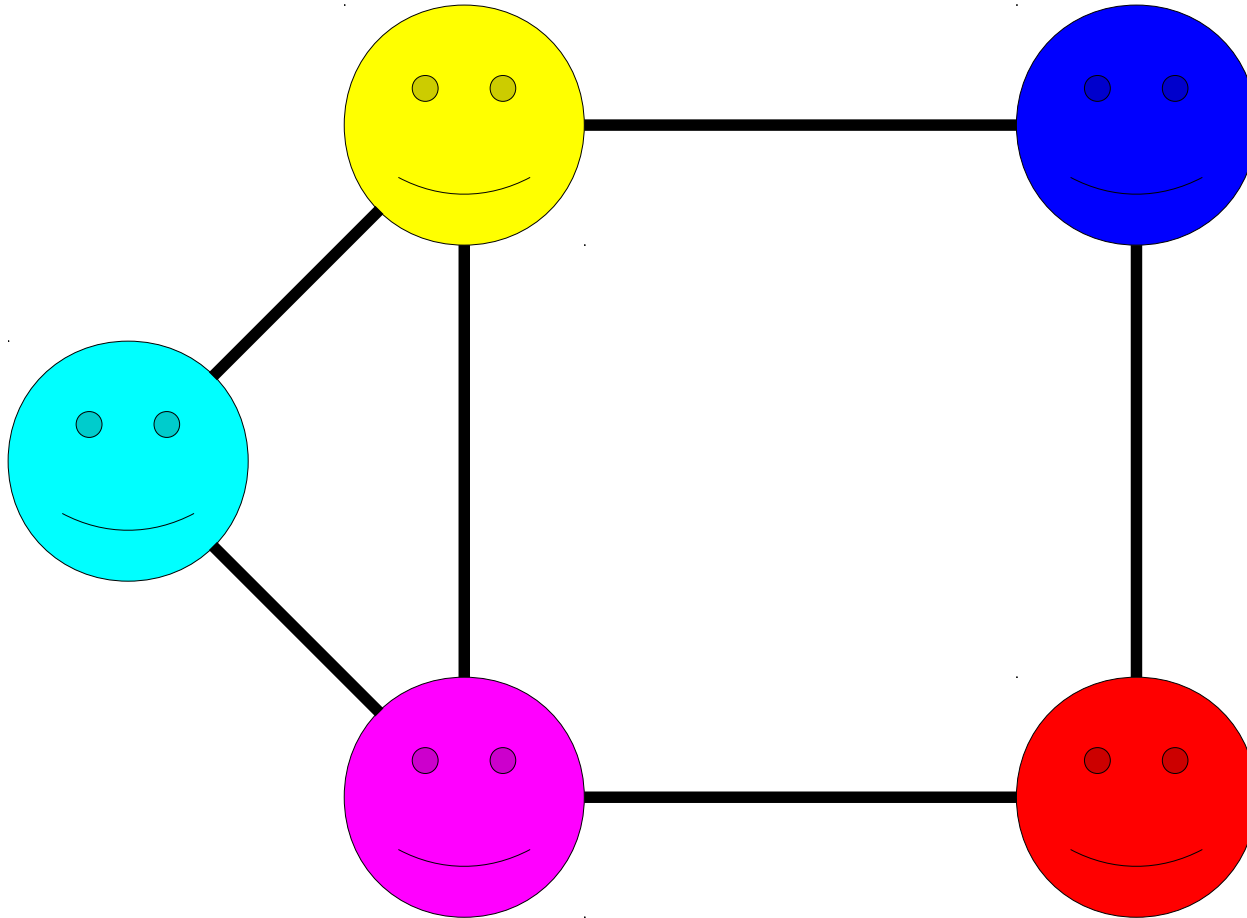




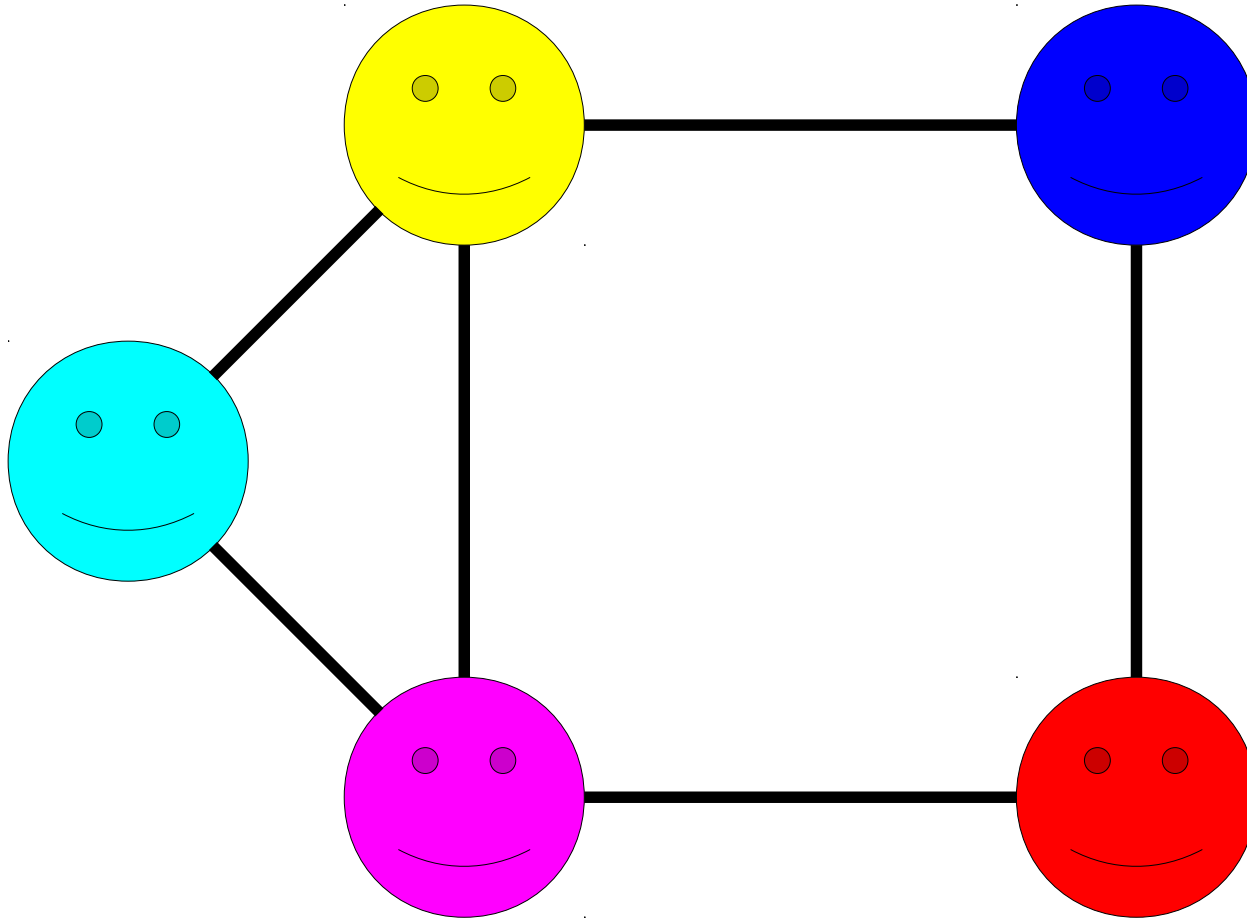
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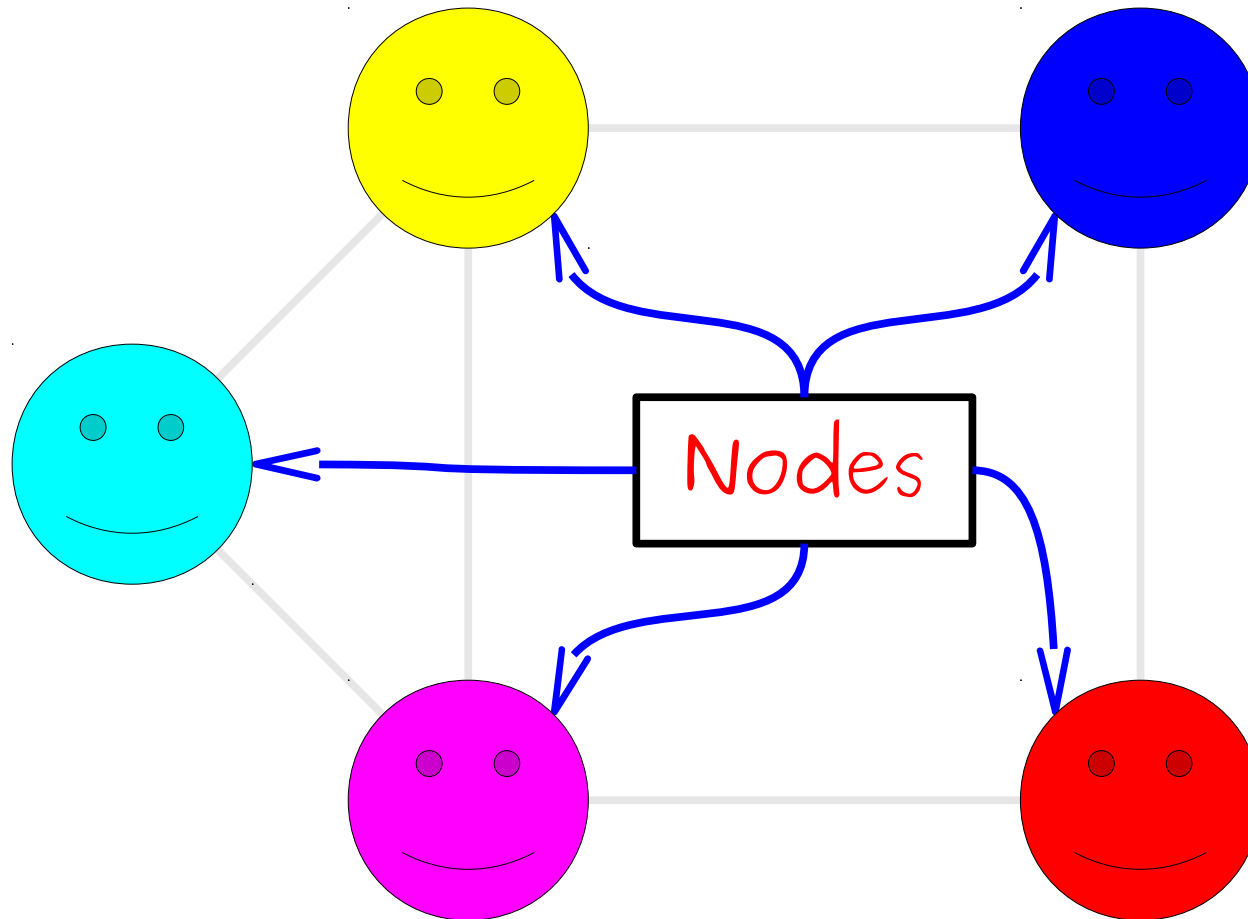


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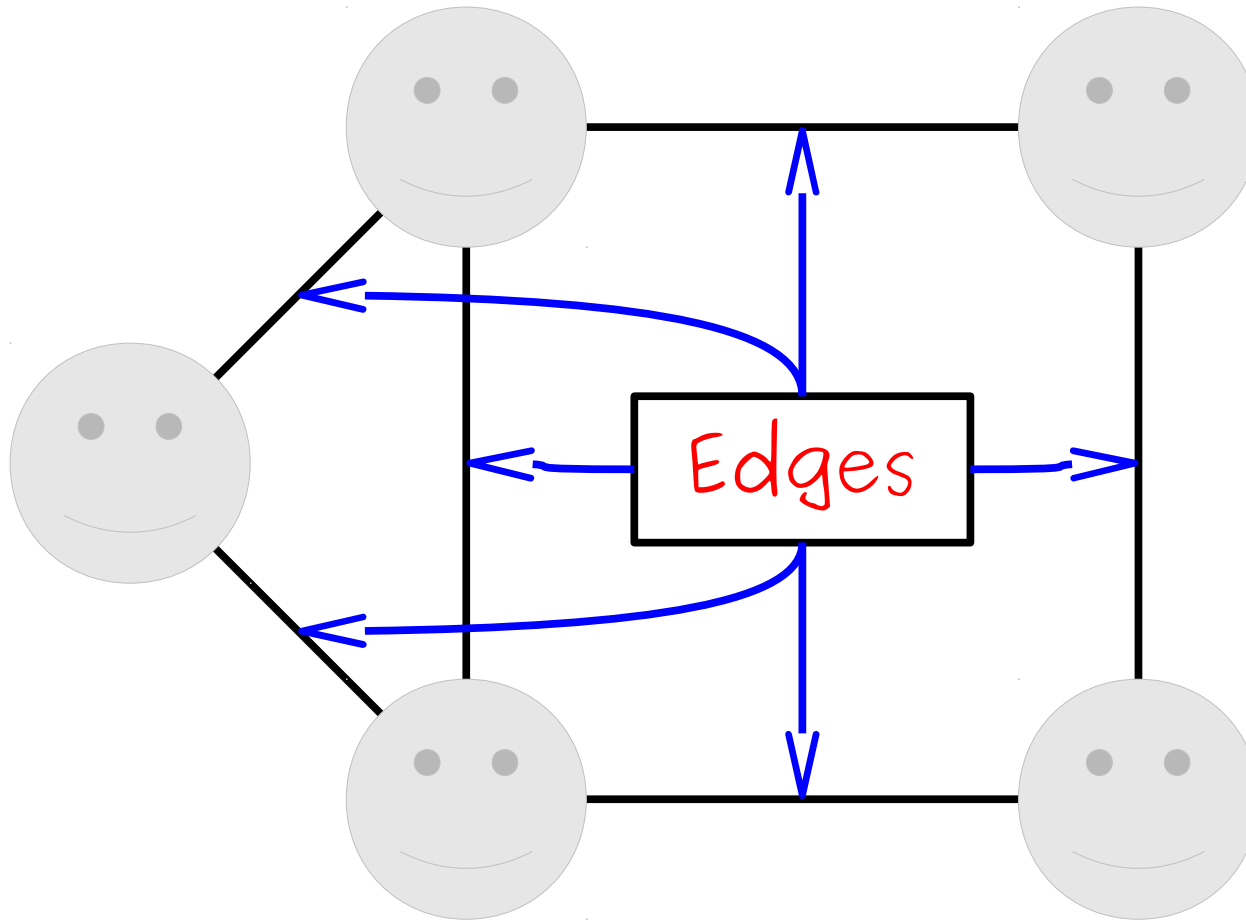
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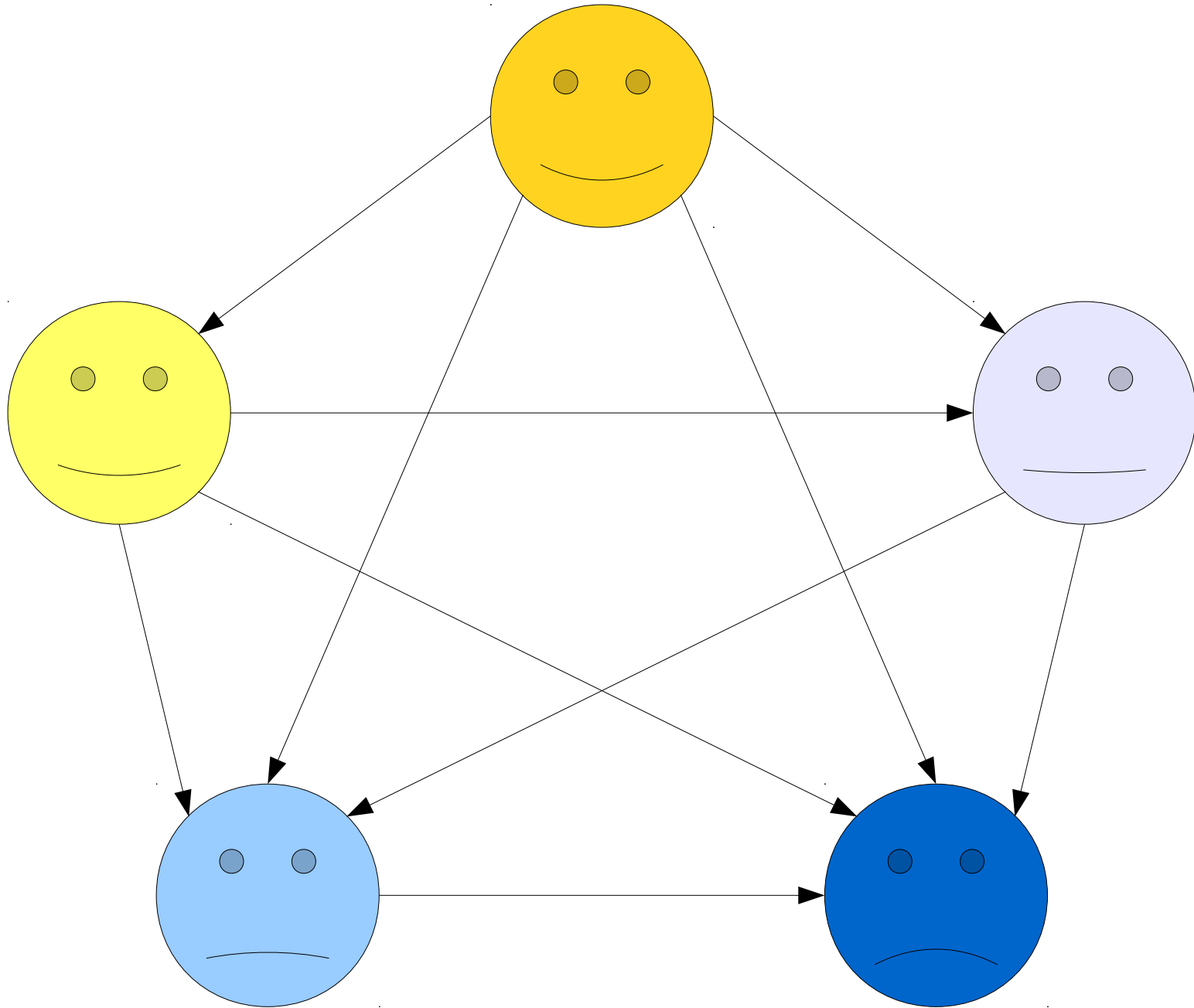
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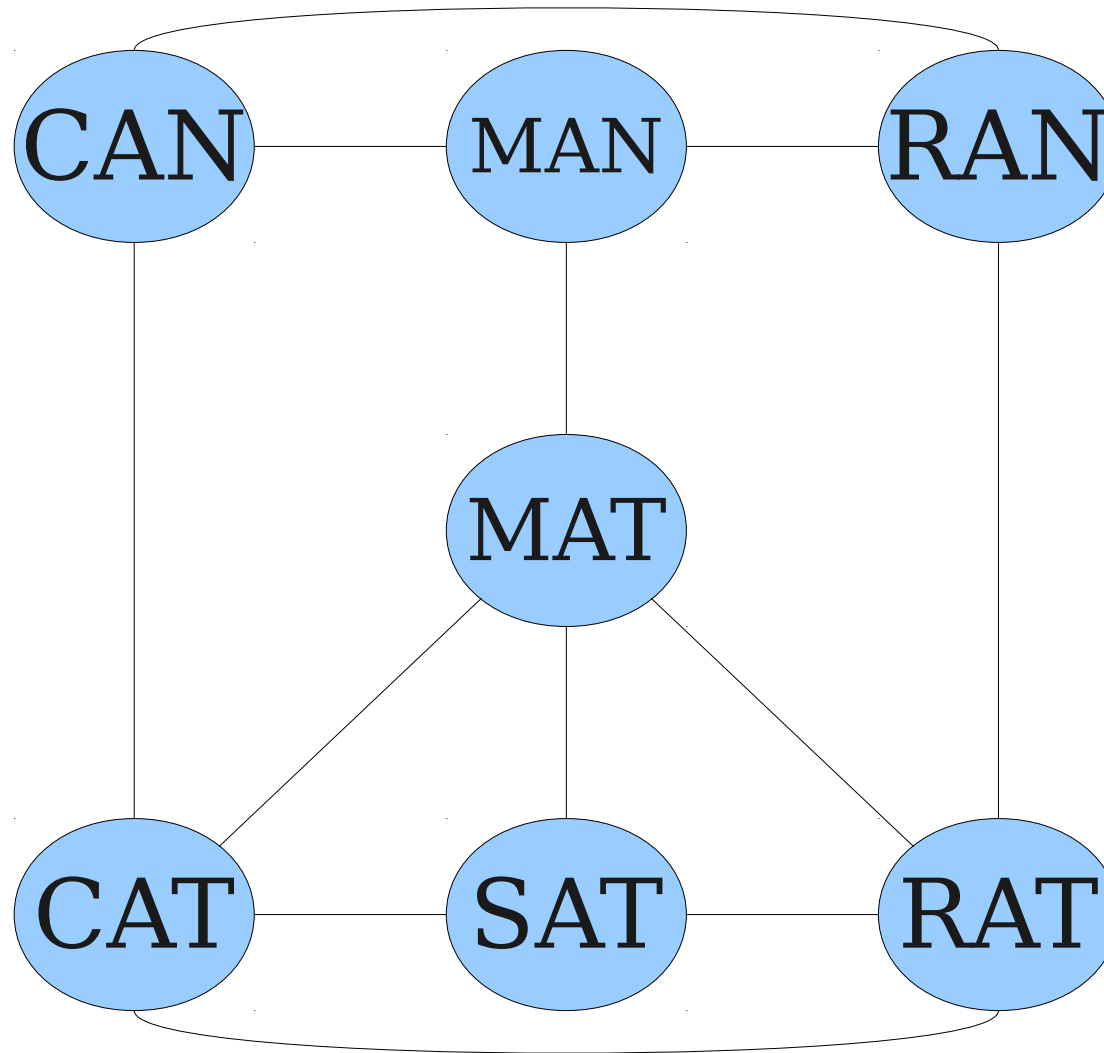
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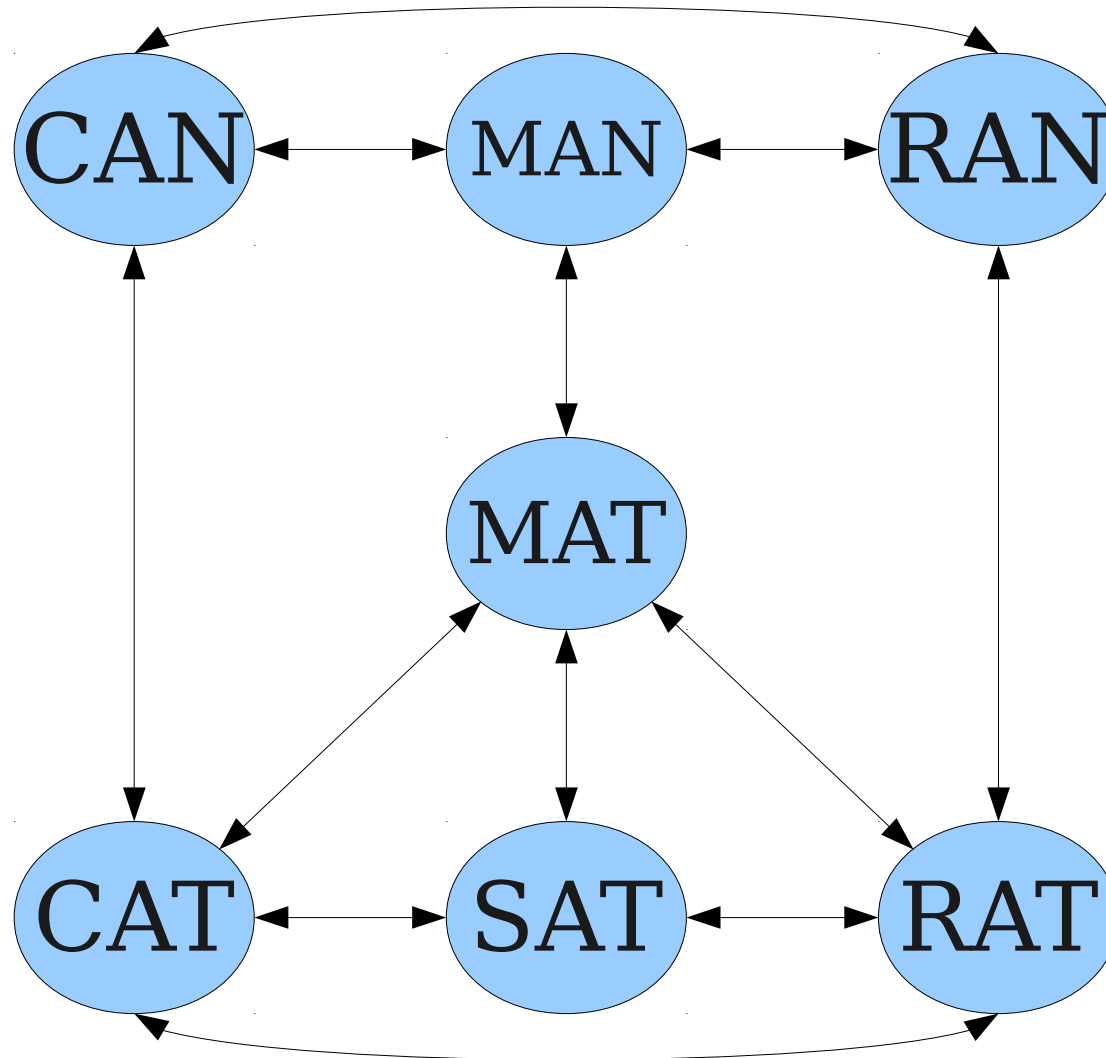
Some graphs are **directed**.



Some graphs are **undirected**.



Some graphs are **undirected**.



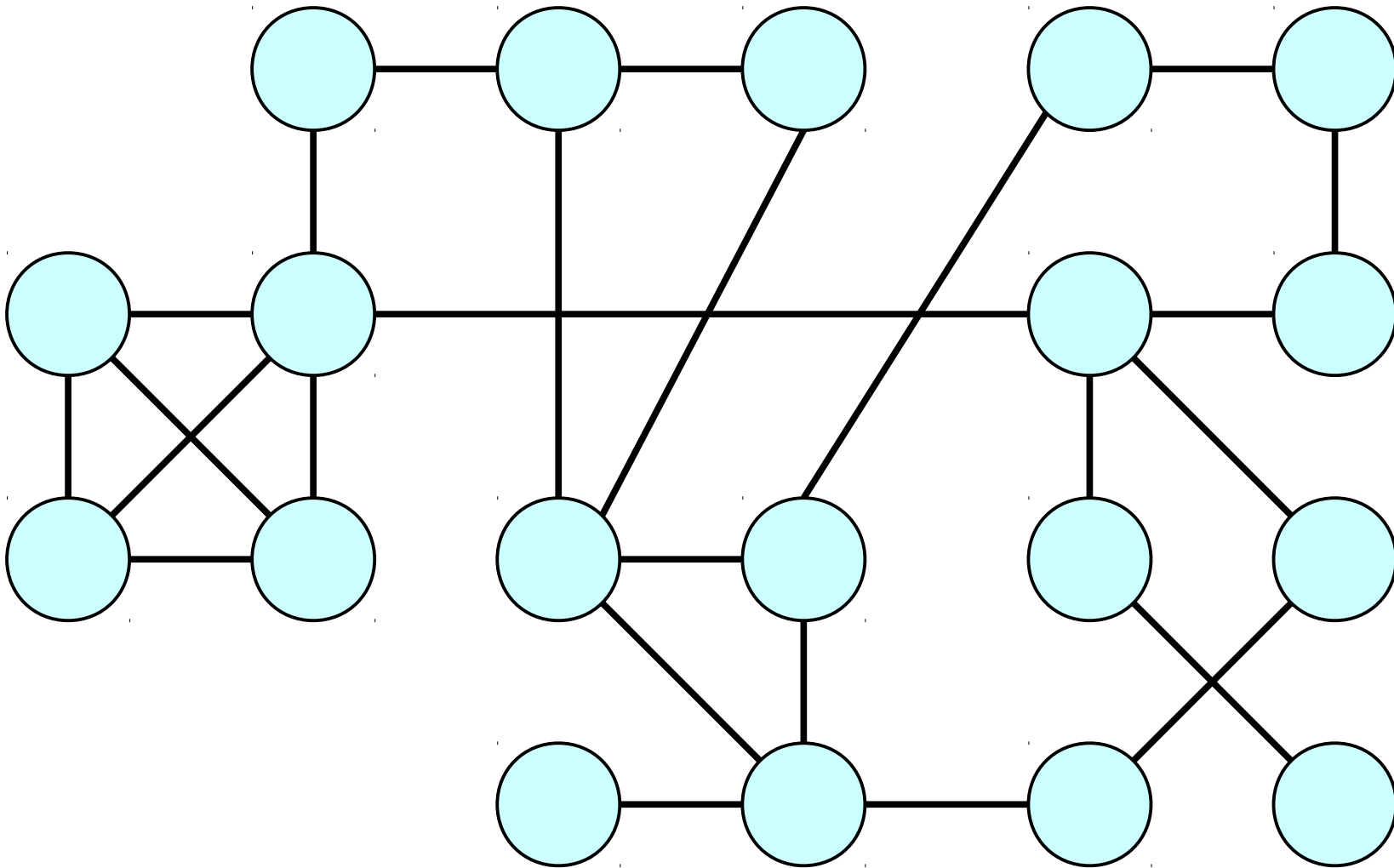
You can think of them as directed graphs with edges both ways.

# Formalisms

- A **graph** is an ordered pair  $G = (V, E)$  where
  - $V$  is a set of the **vertices** (nodes) of the graph.
  - $E$  is a set of the **edges** (arcs) of the graph.
- $E$  can be a set of ordered pairs or unordered pairs.
  - If  $E$  consists of ordered pairs,  $G$  is **directed**
  - If  $E$  consists of unordered pairs,  $G$  is **undirected**.
- In an *undirected* graph, the **degree** of node  $v$  (denoted  **$\deg(v)$** ) is the number of edges incident to  $v$ .
- In a *directed* graph, the **indegree** of a node  $v$  (denoted  **$\deg^-(v)$** ) is the number of edges entering  $v$  and the **outdegree** of a node  $v$  (denoted  **$\deg^+(v)$** ) is the number of edges leaving  $v$ .

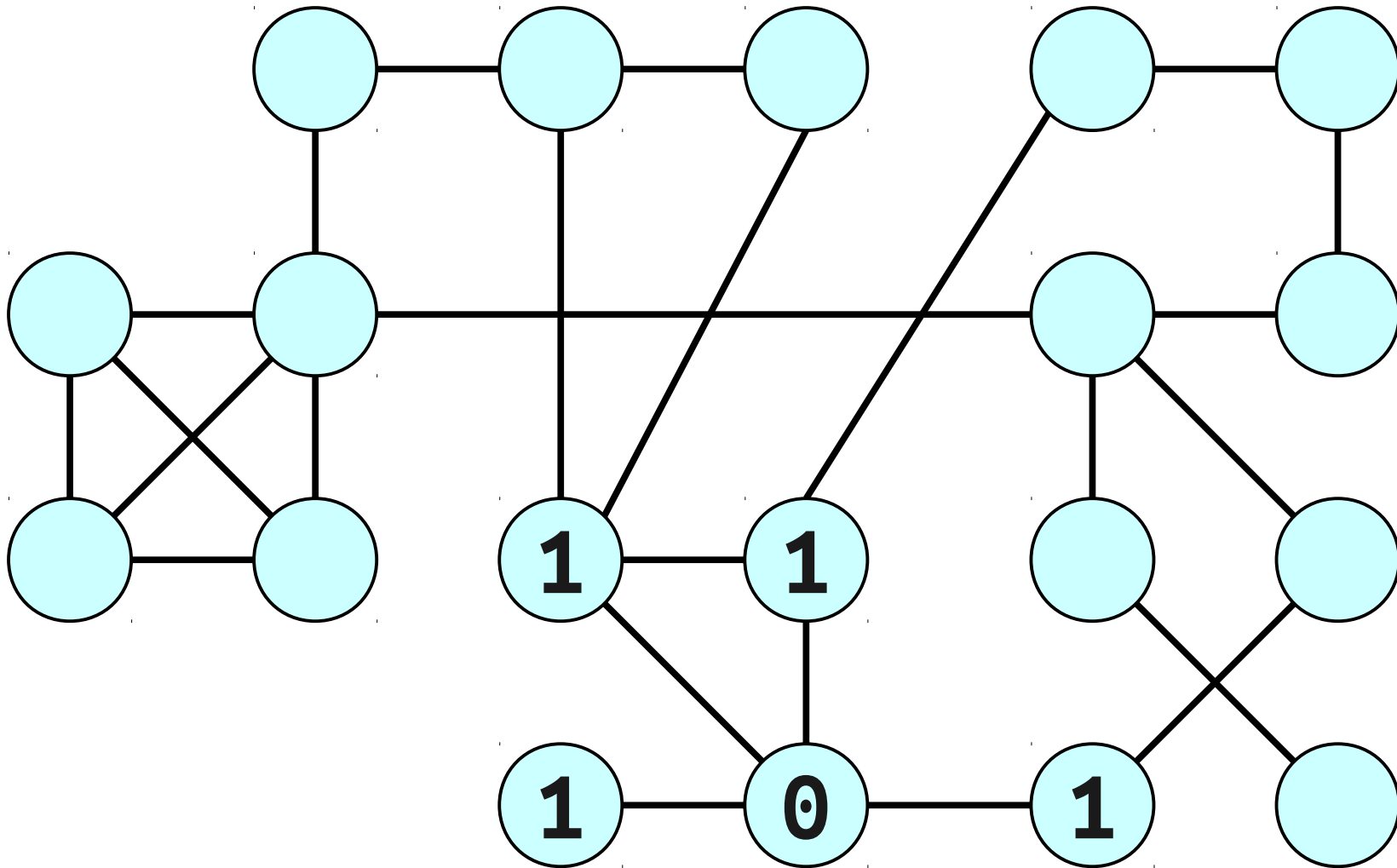
# An Application: Six Degrees of Separation

# A Social Network



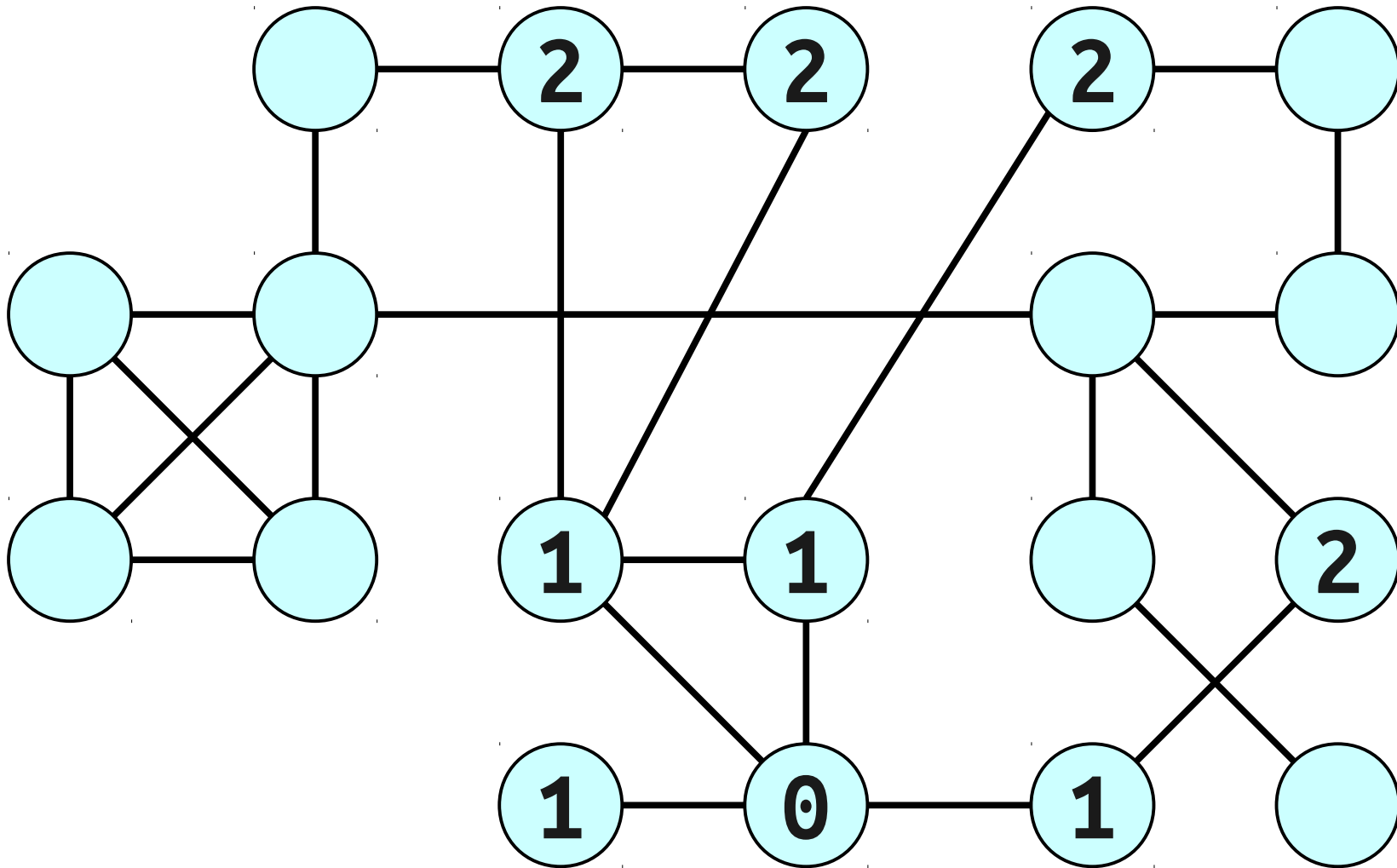


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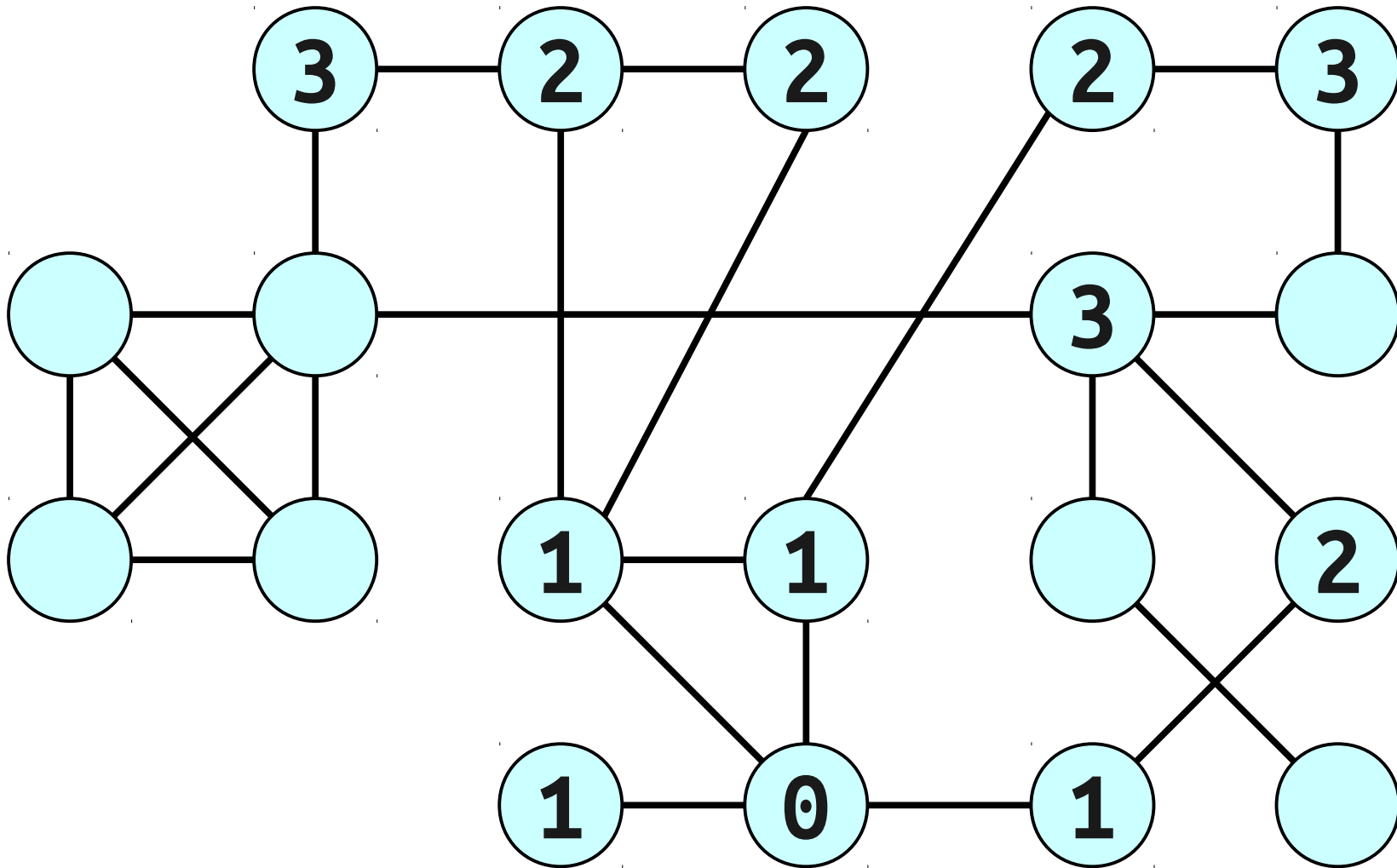




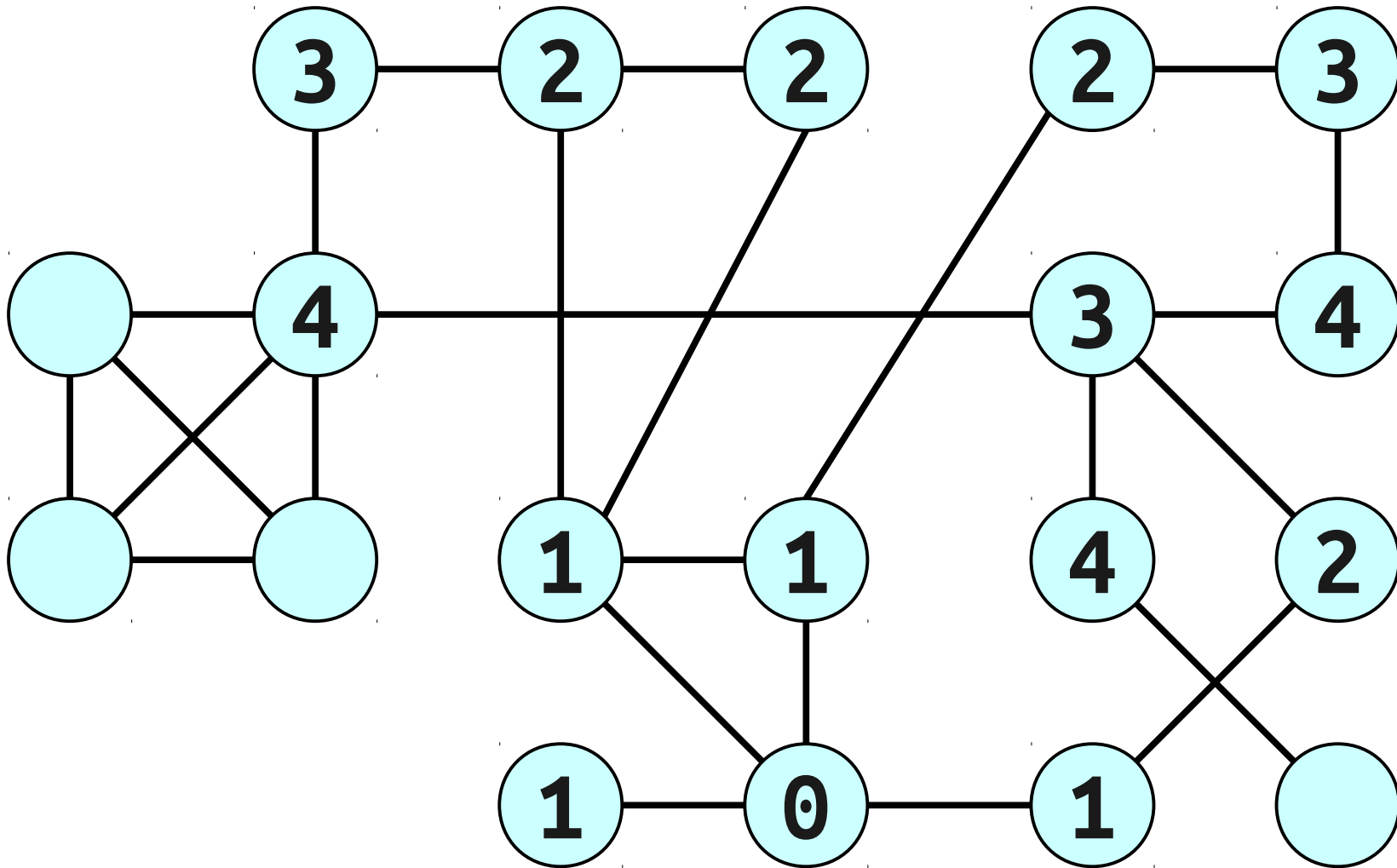
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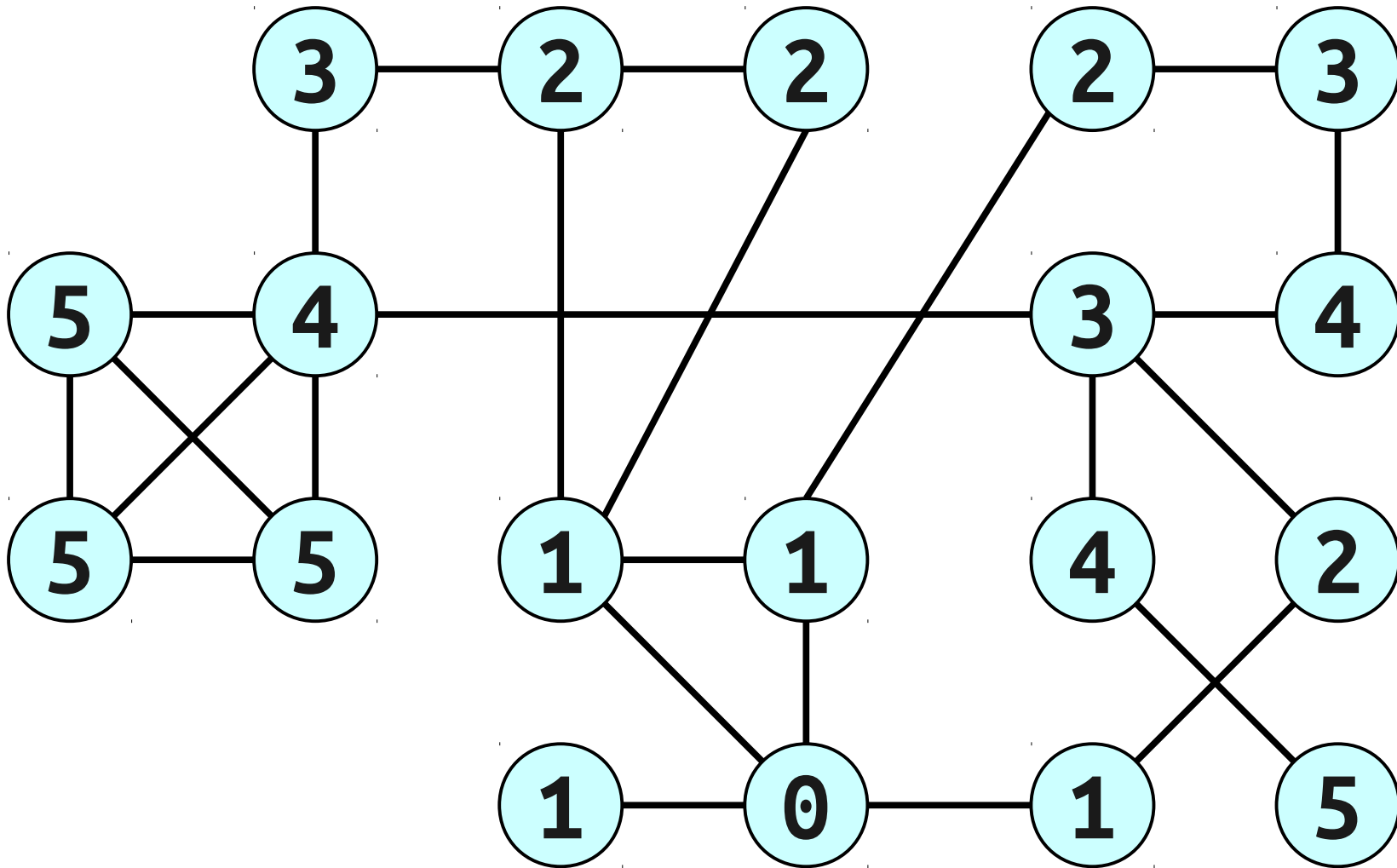
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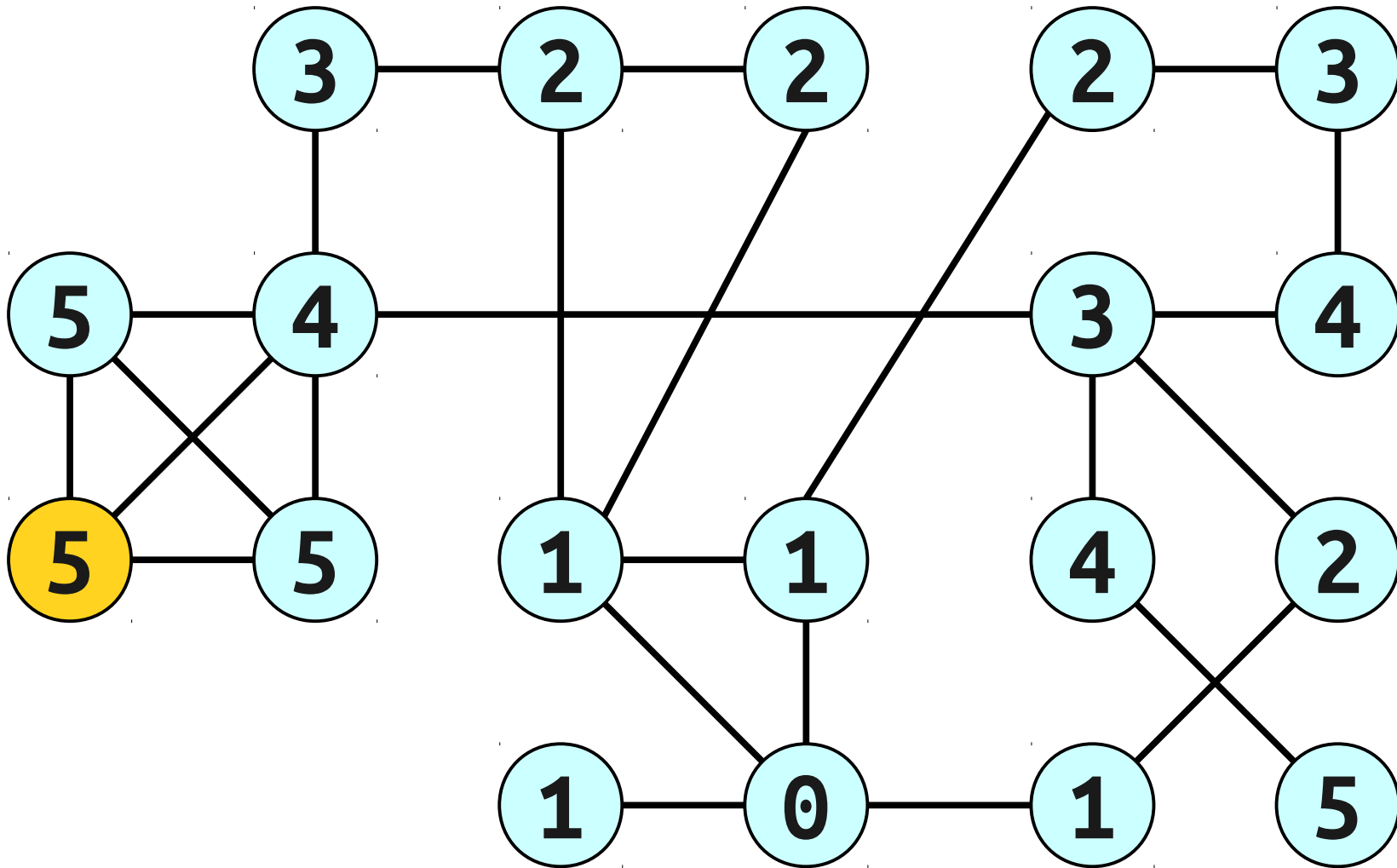
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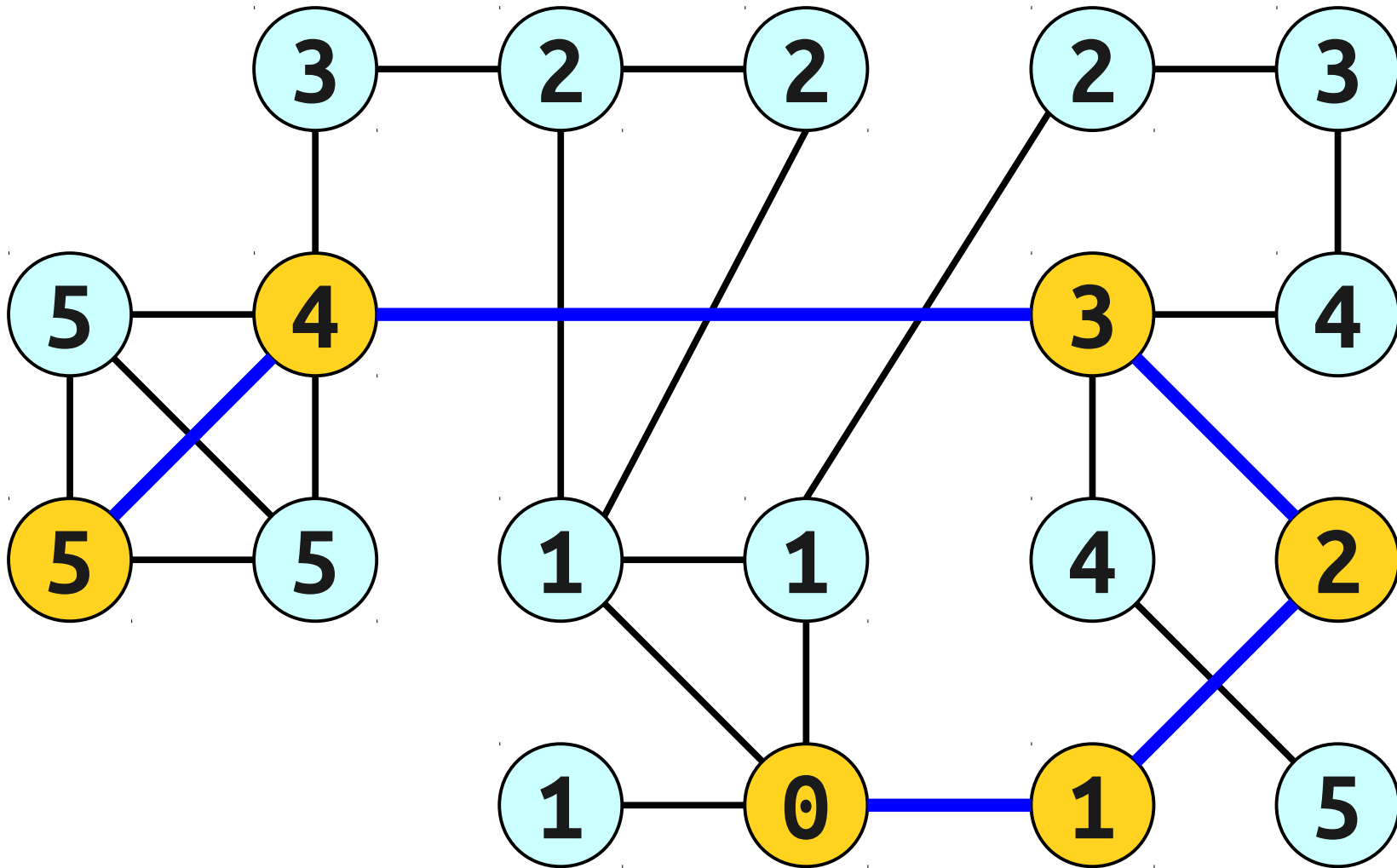
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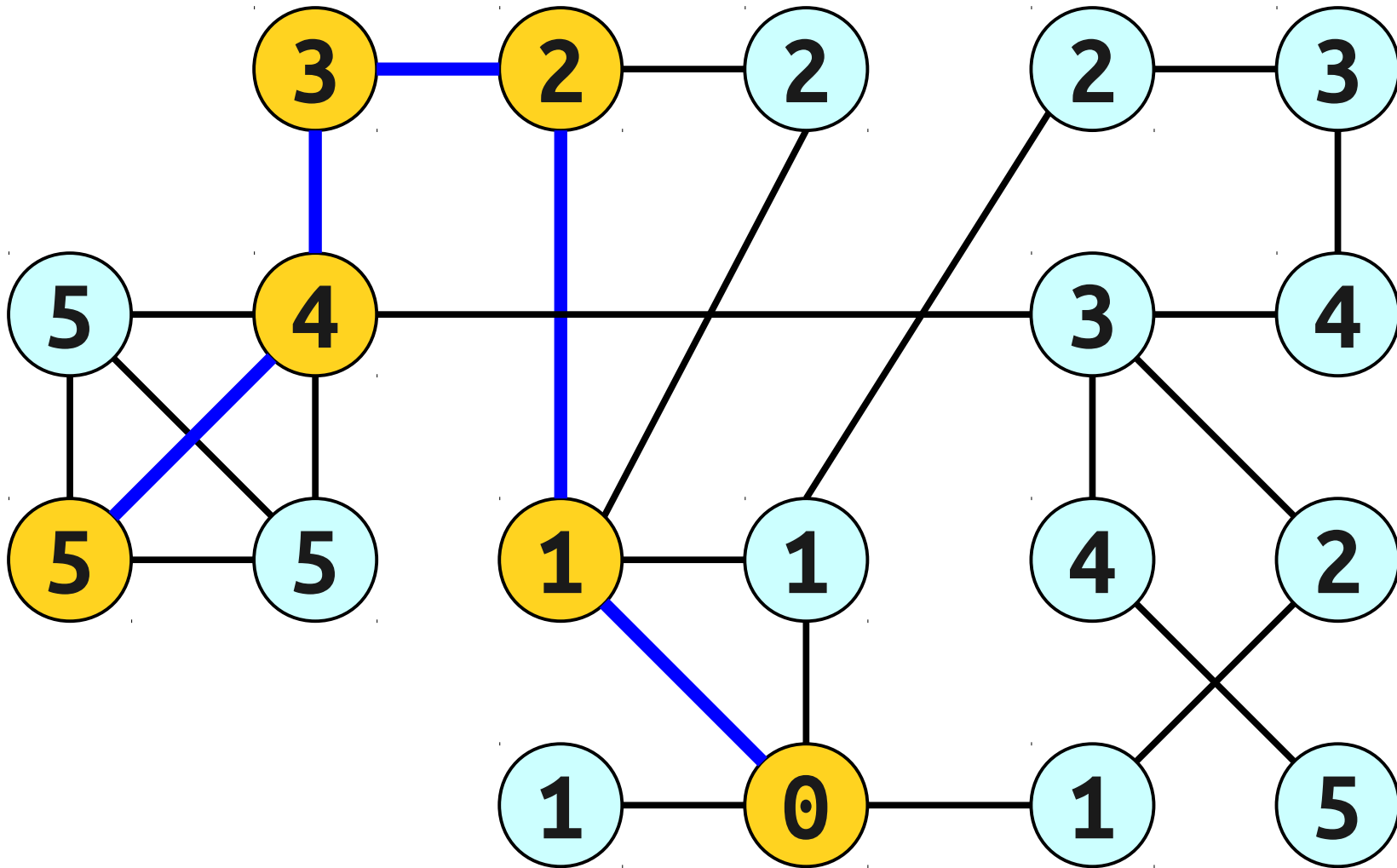
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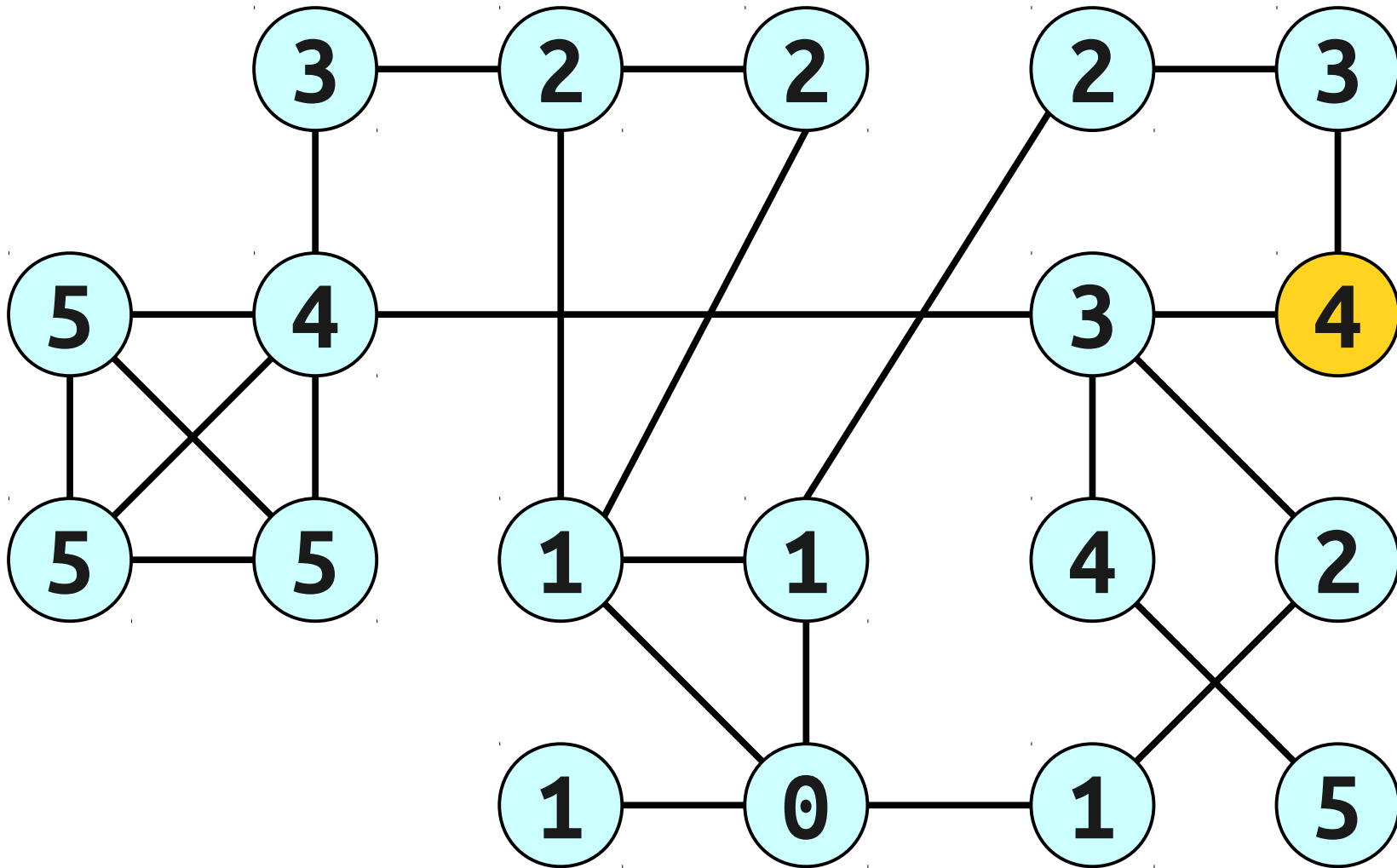
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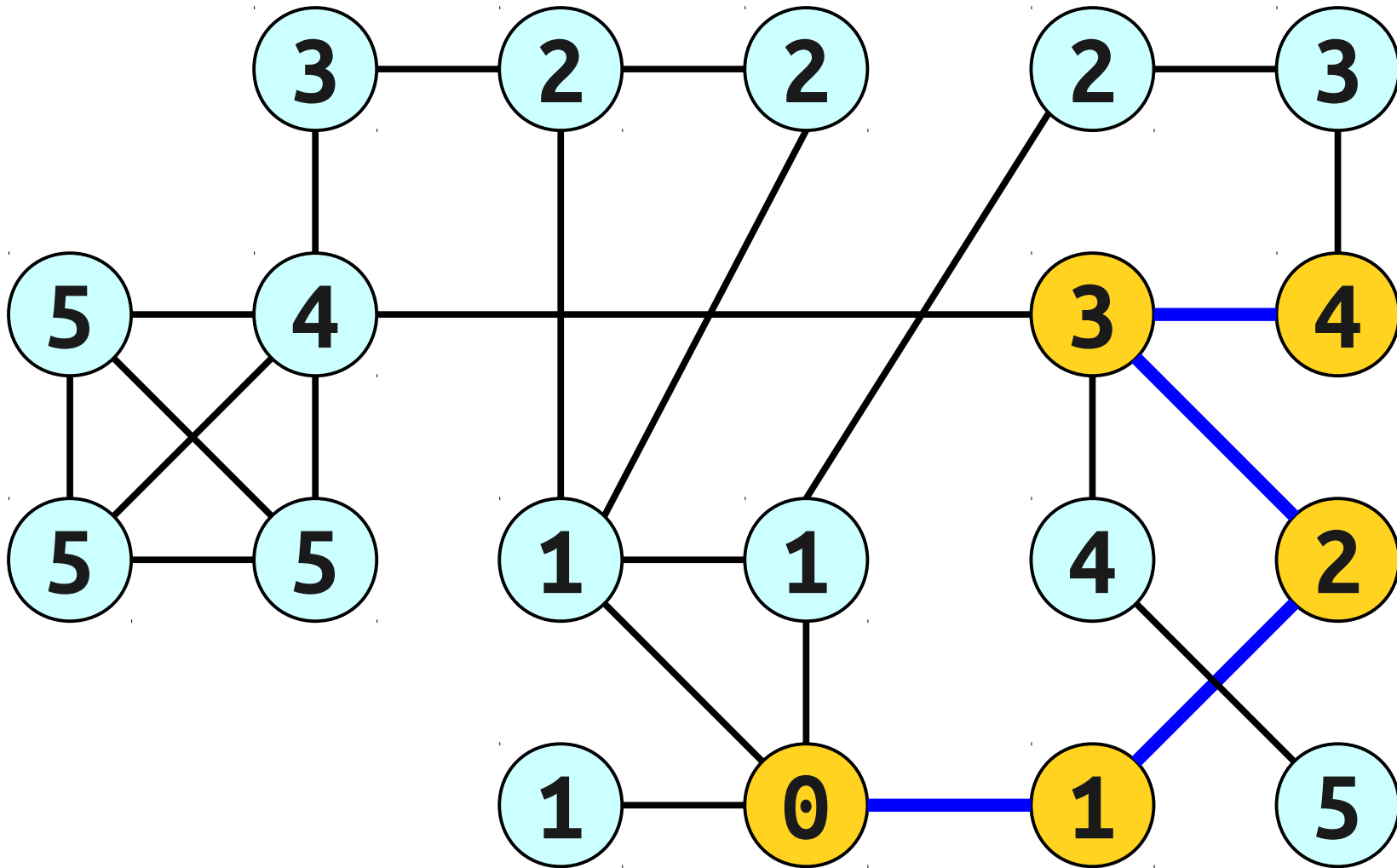


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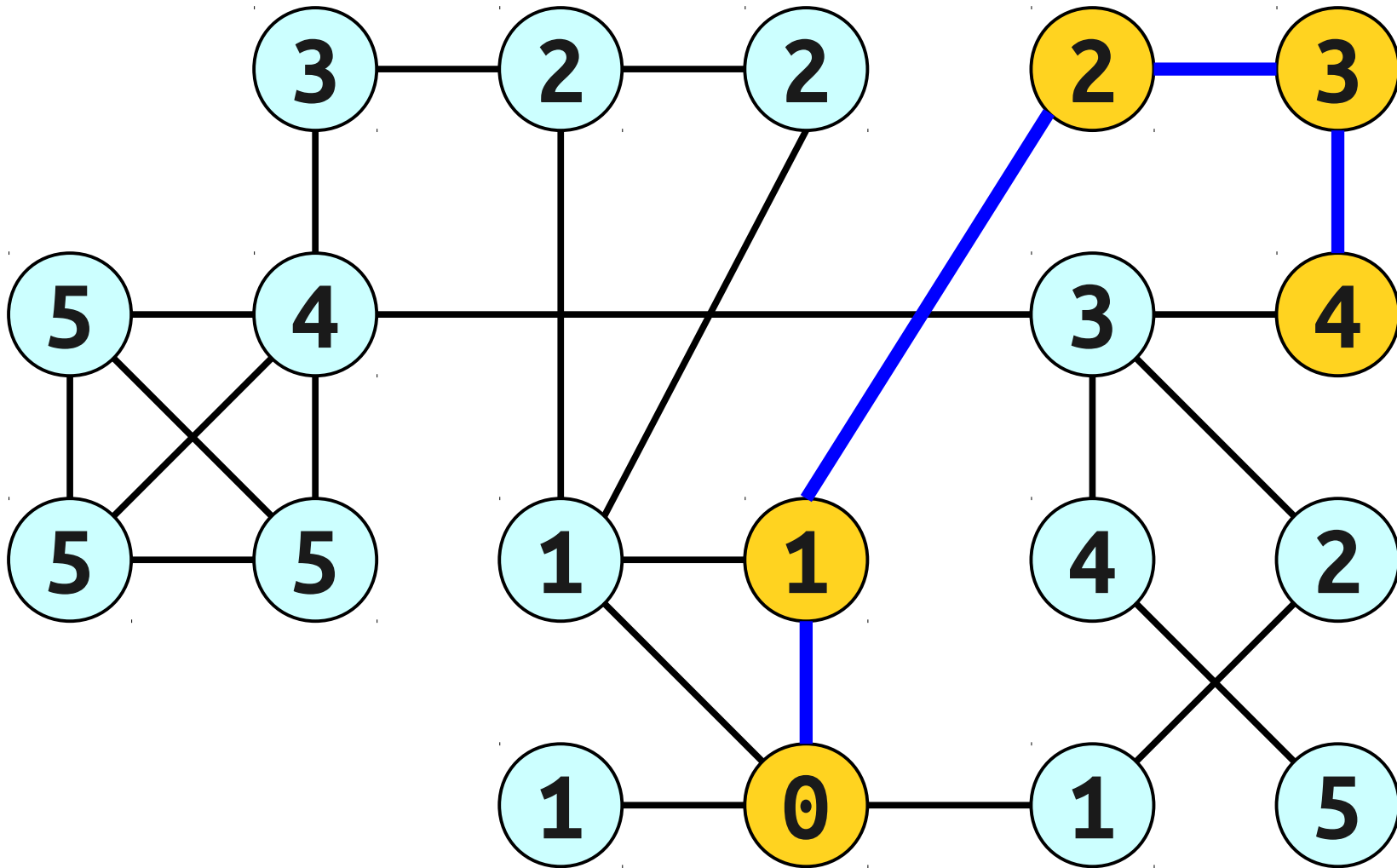




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# Shortest Paths

- The **length** of a path  $P$  (denoted  $|P|$ ) in a graph is the number of edges it contains.
- A **shortest path** between  $u$  and  $v$  is a path  $P$  where  $|P| \leq |P'|$  for any path  $P'$  from  $u$  to  $v$ .
- For any nodes  $u$  and  $v$ , define  **$d(u, v)$**  to be the length of the shortest path from  $u$  to  $v$ , or  $\infty$  if no such path exists.
- What is  $d(v, v)$  for any  $v \in V$ ?

# The Shortest Path Problem

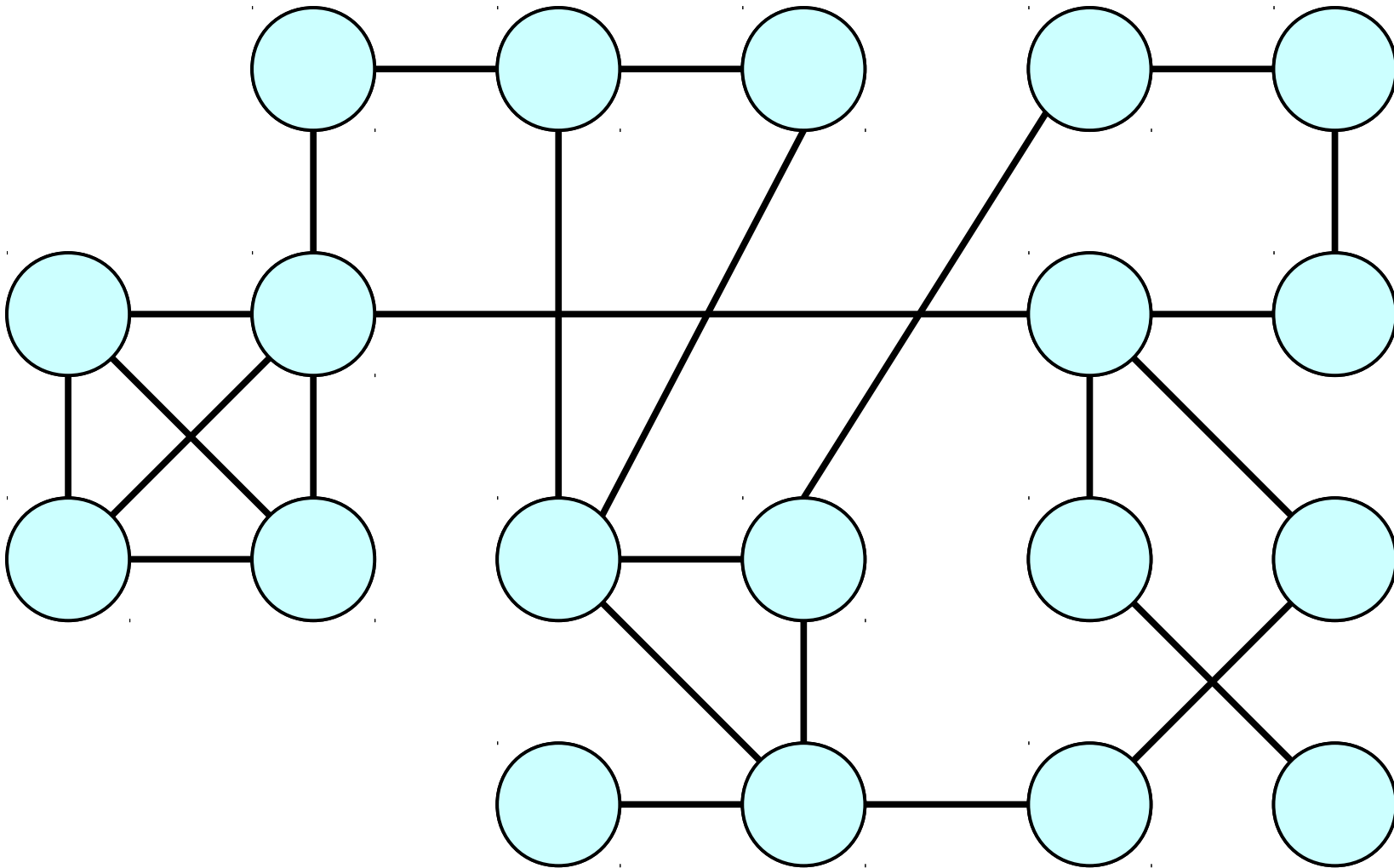
- **Input:**

- A graph  $G = (V, E)$ , which may be directed or undirected.
- A start node  $s \in V$ .

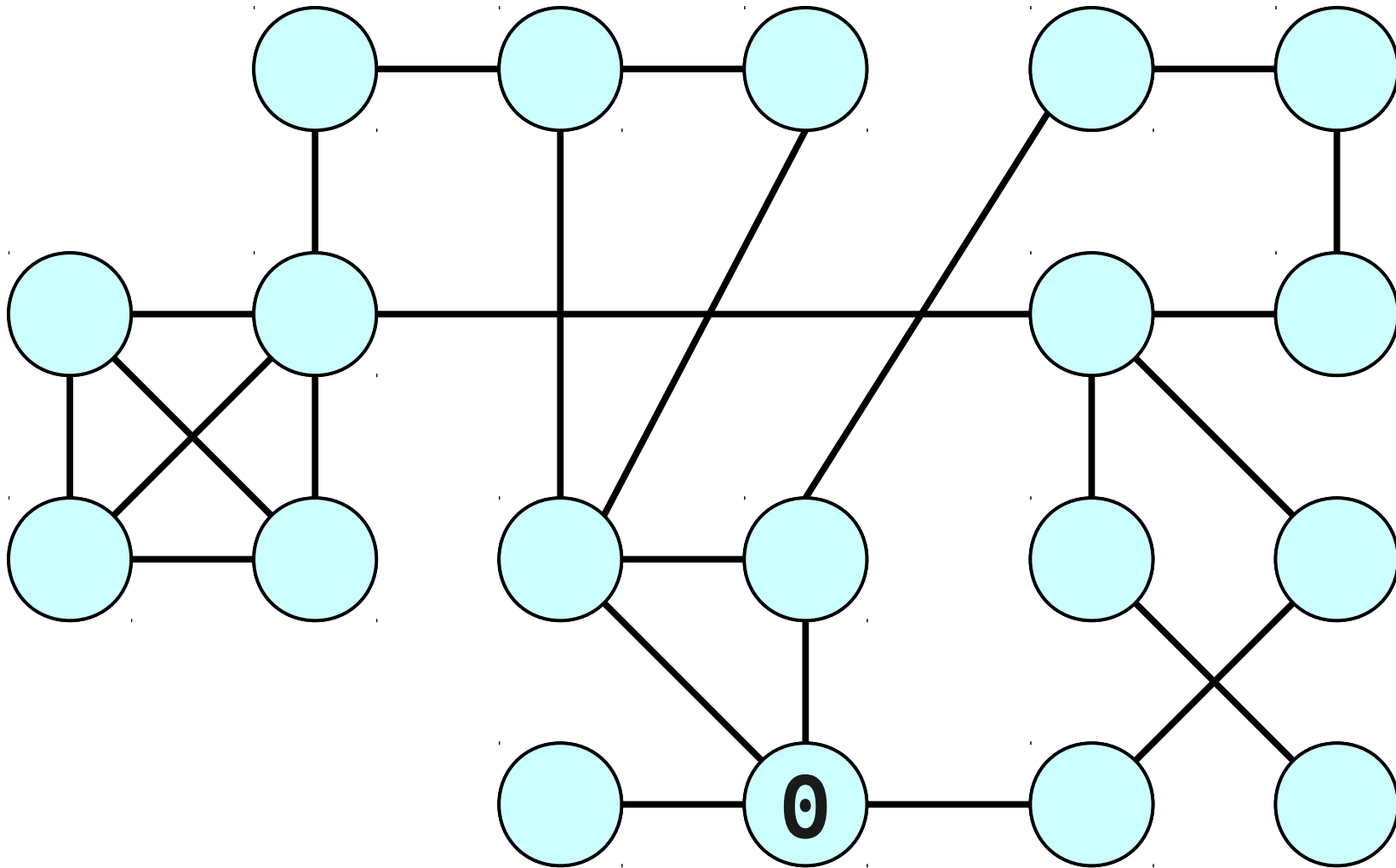
- **Output:**

- A table  $\text{dist}[v]$ , where  $\text{dist}[v] = d(s, v)$  for any  $v \in V$ .

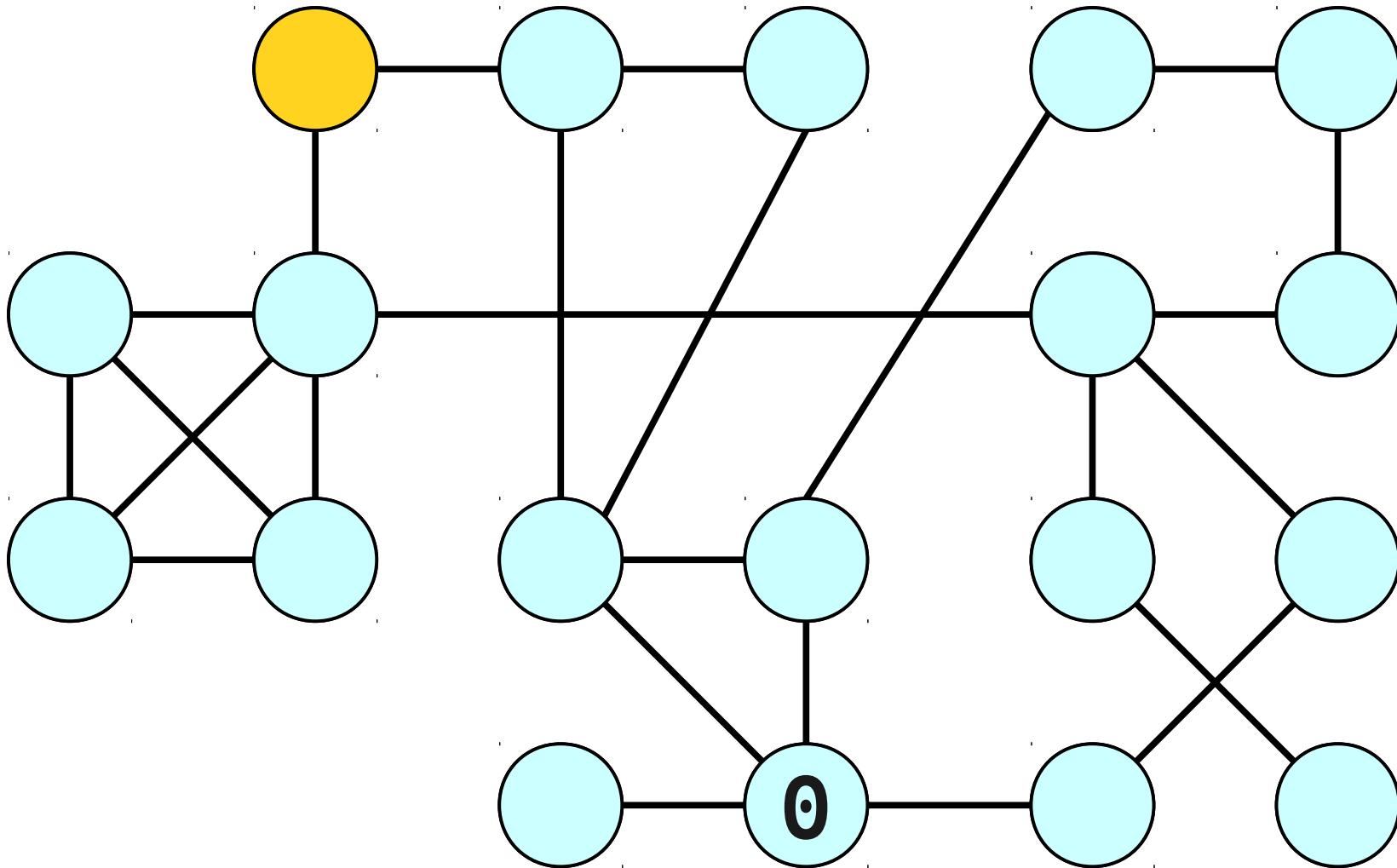
# An Inefficient Algorithm



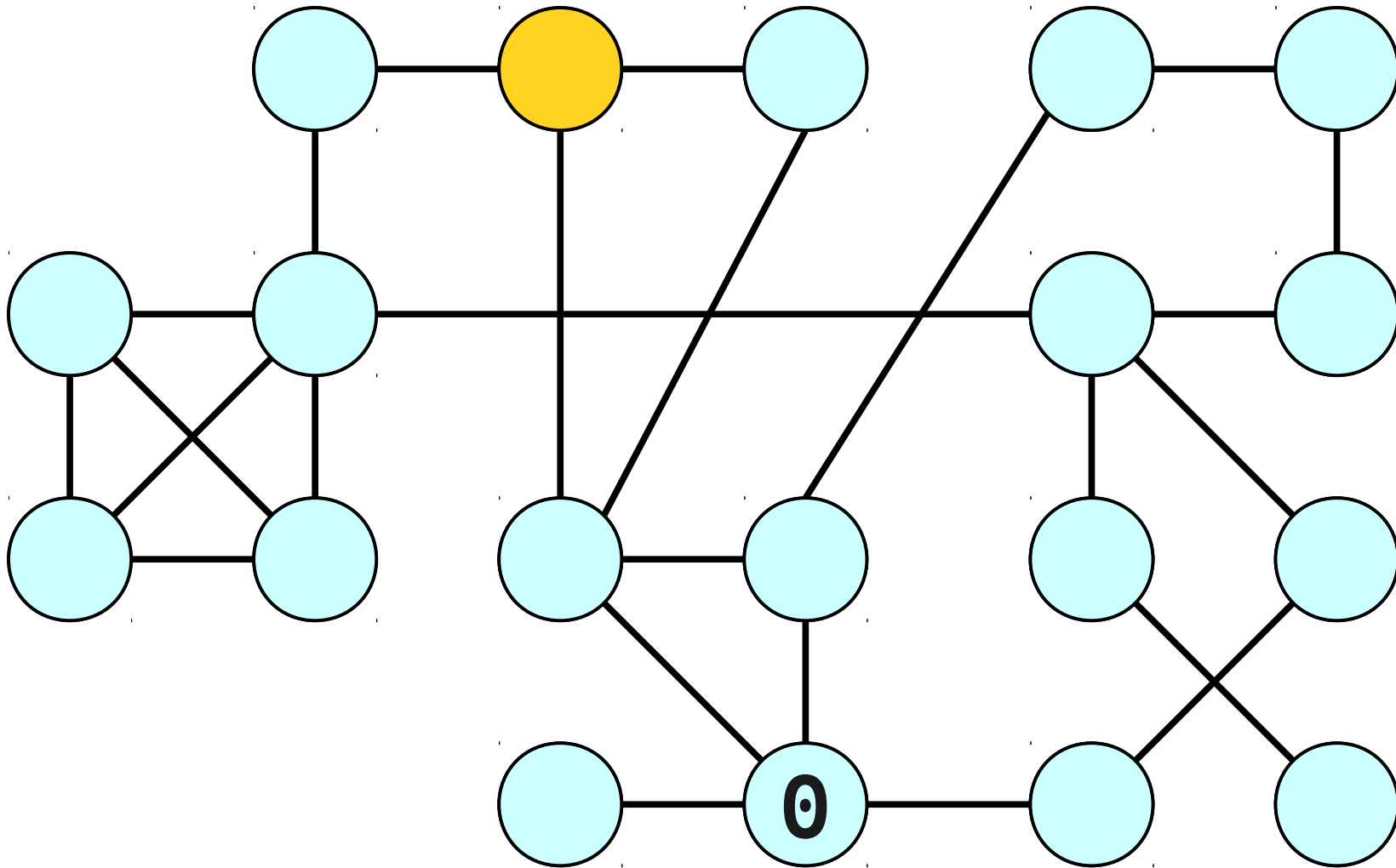
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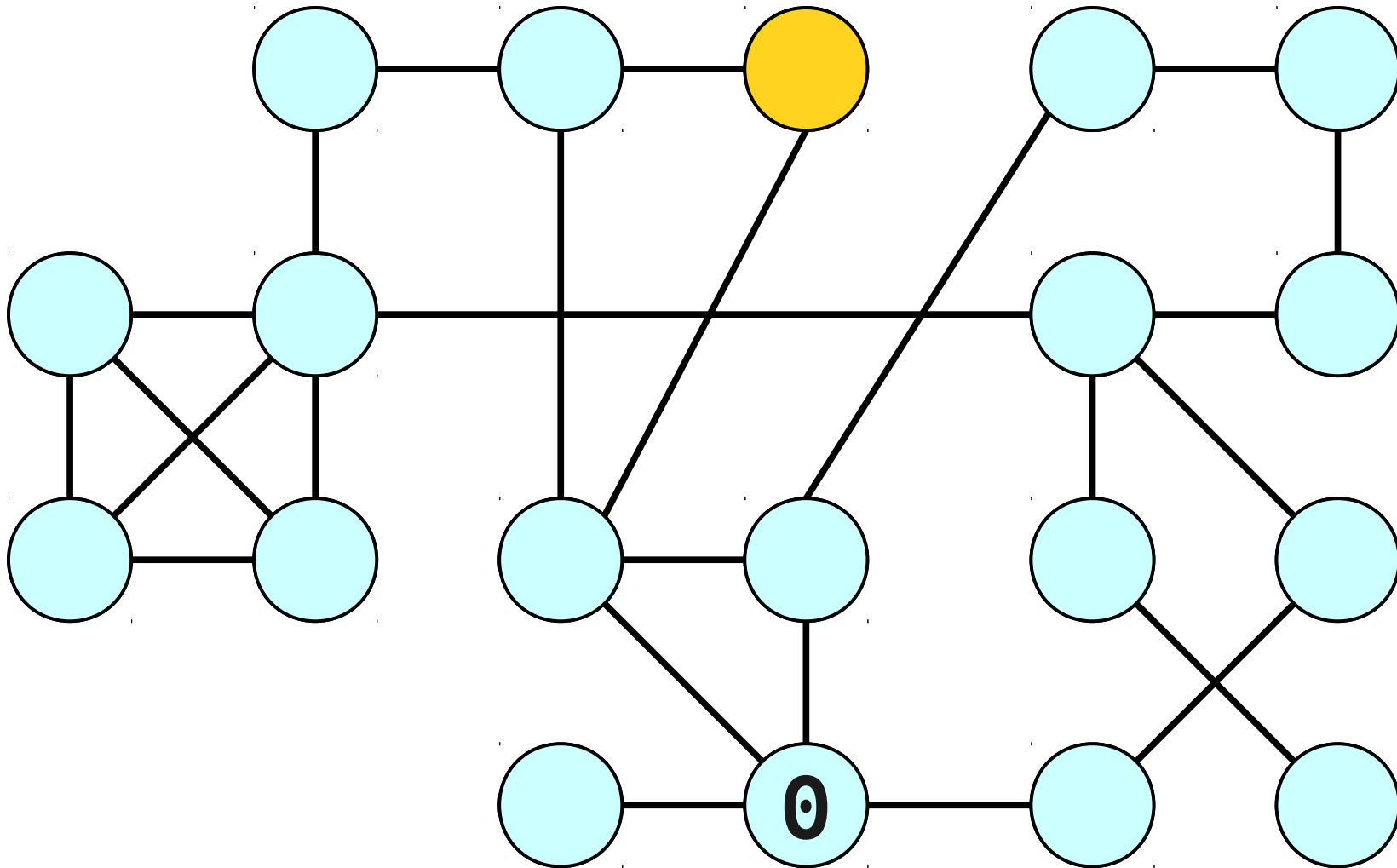


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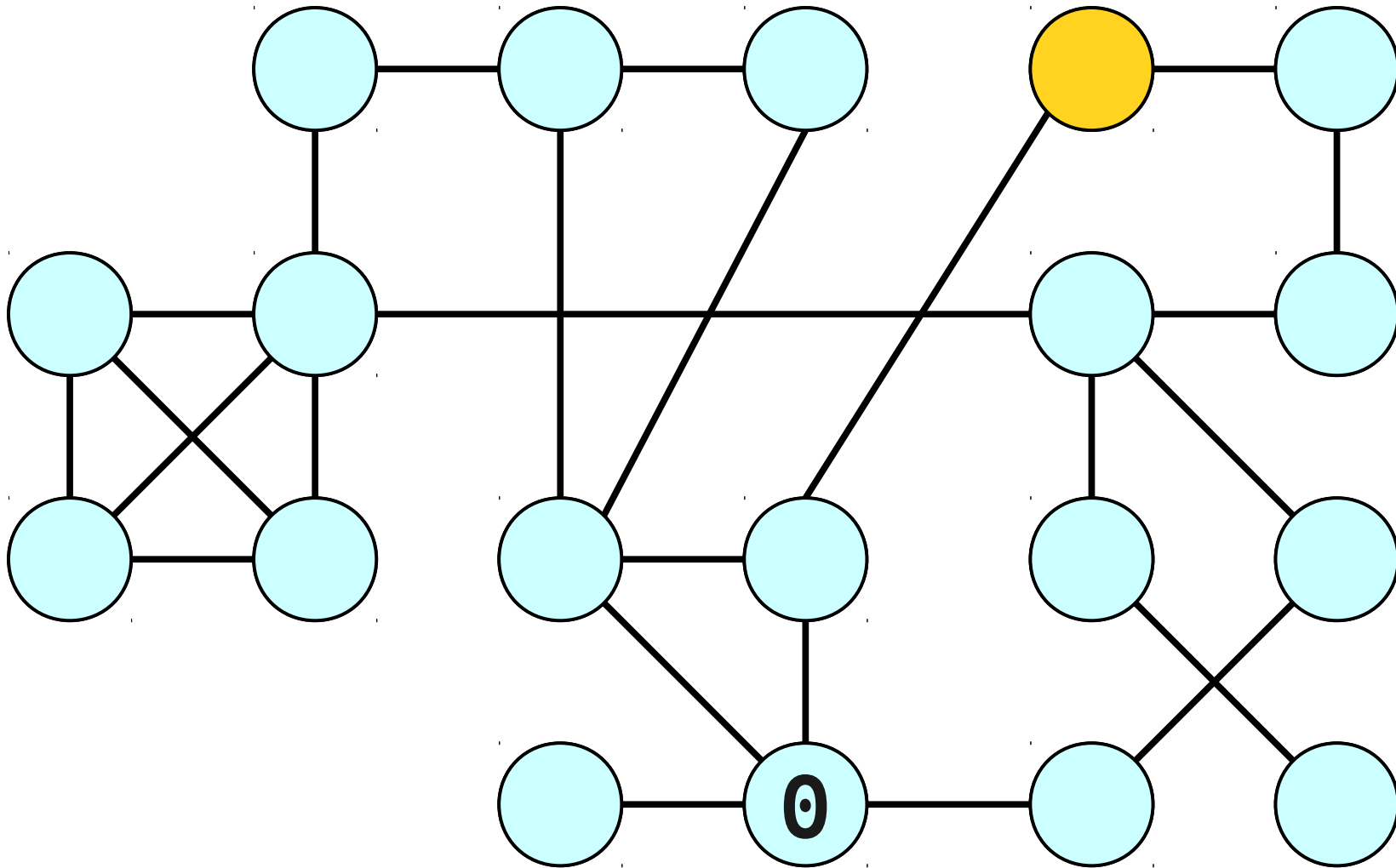




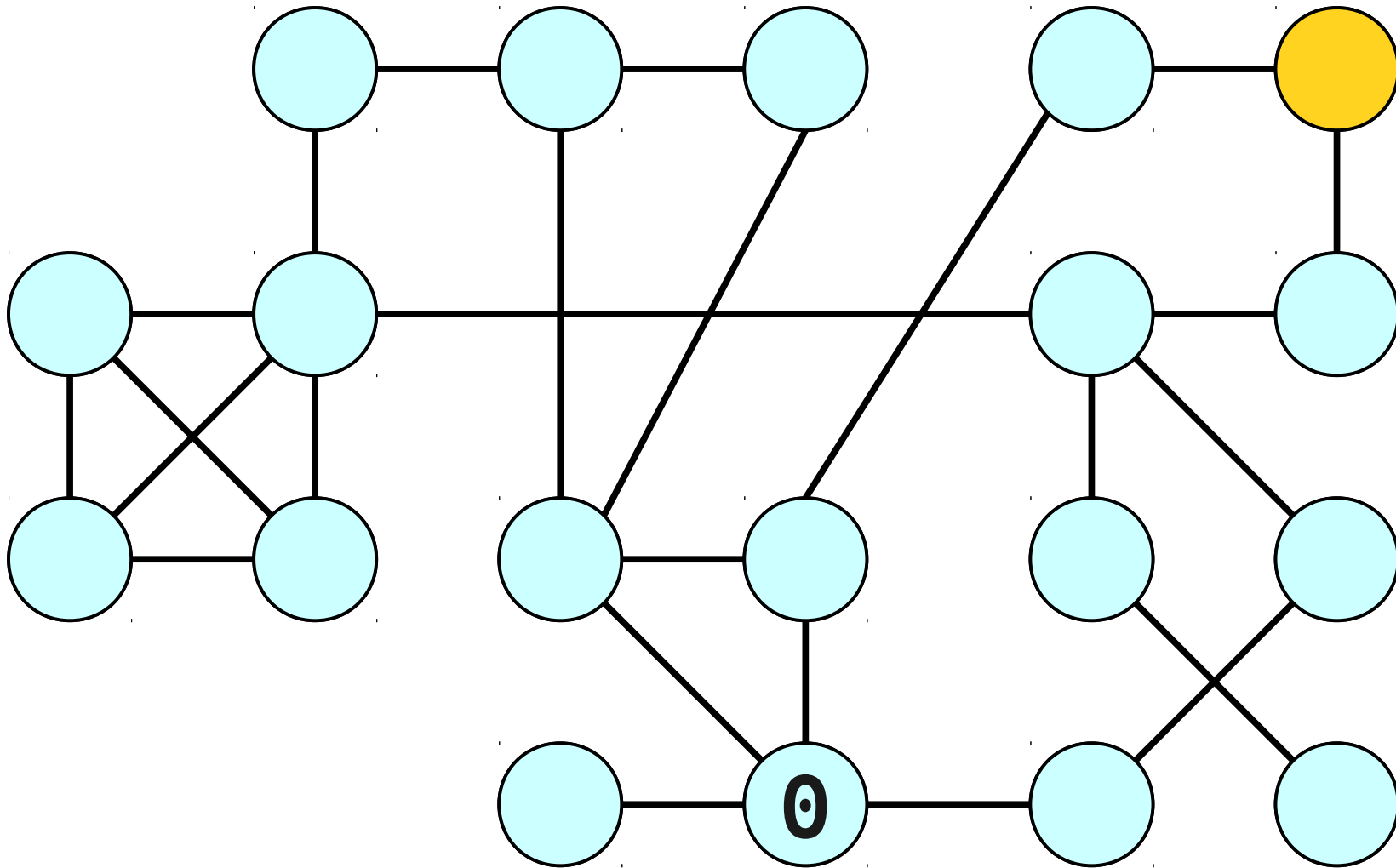
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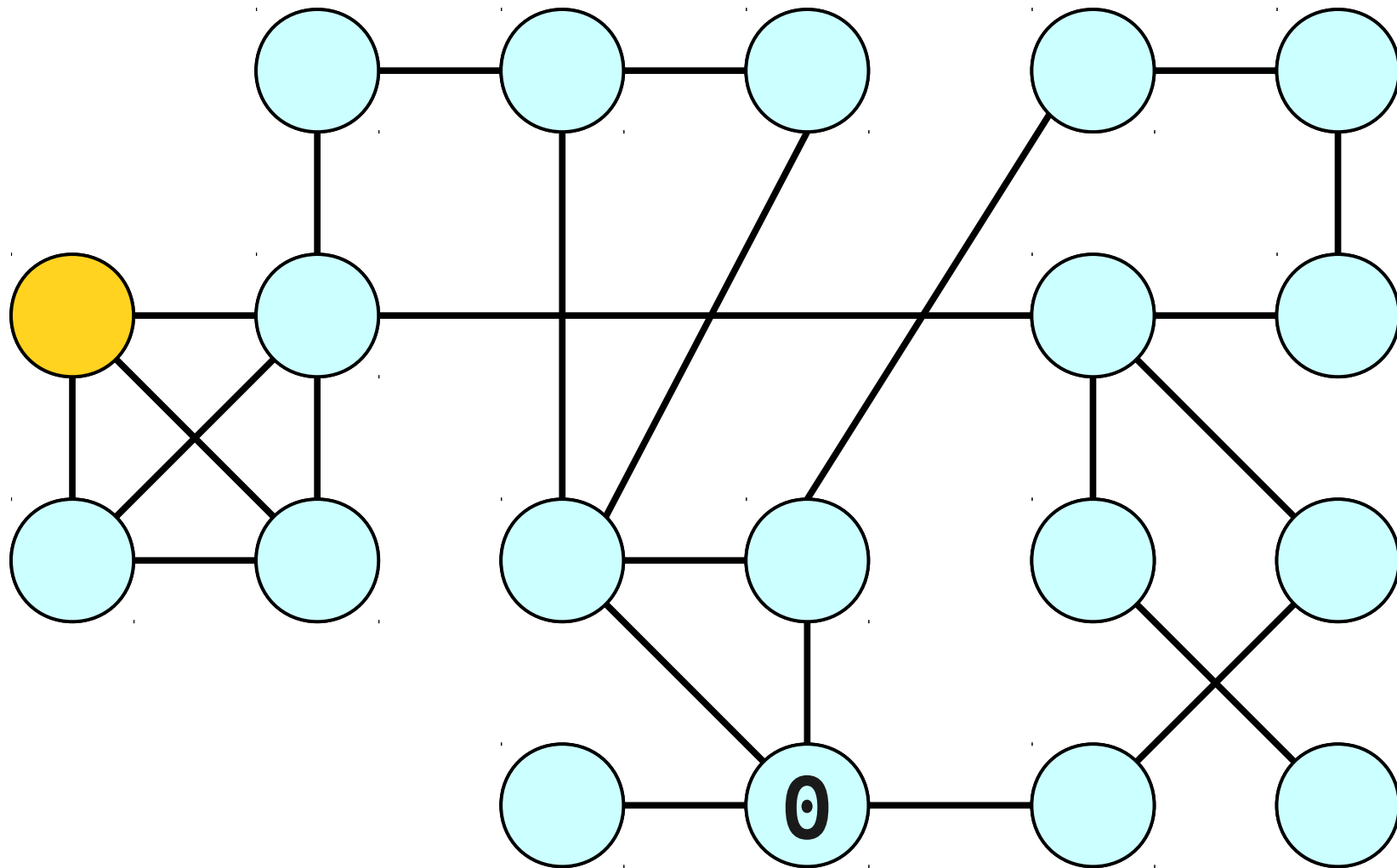
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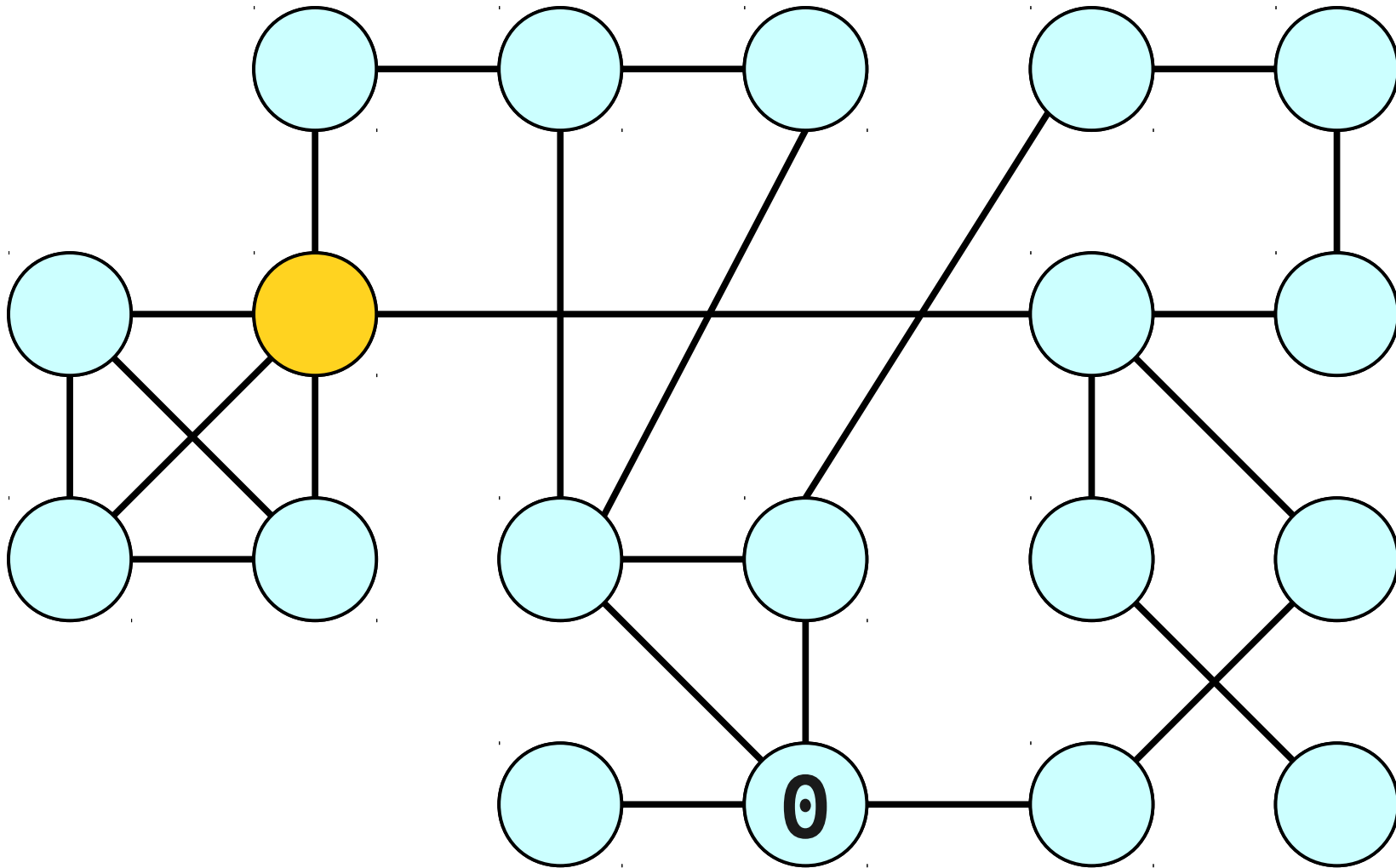
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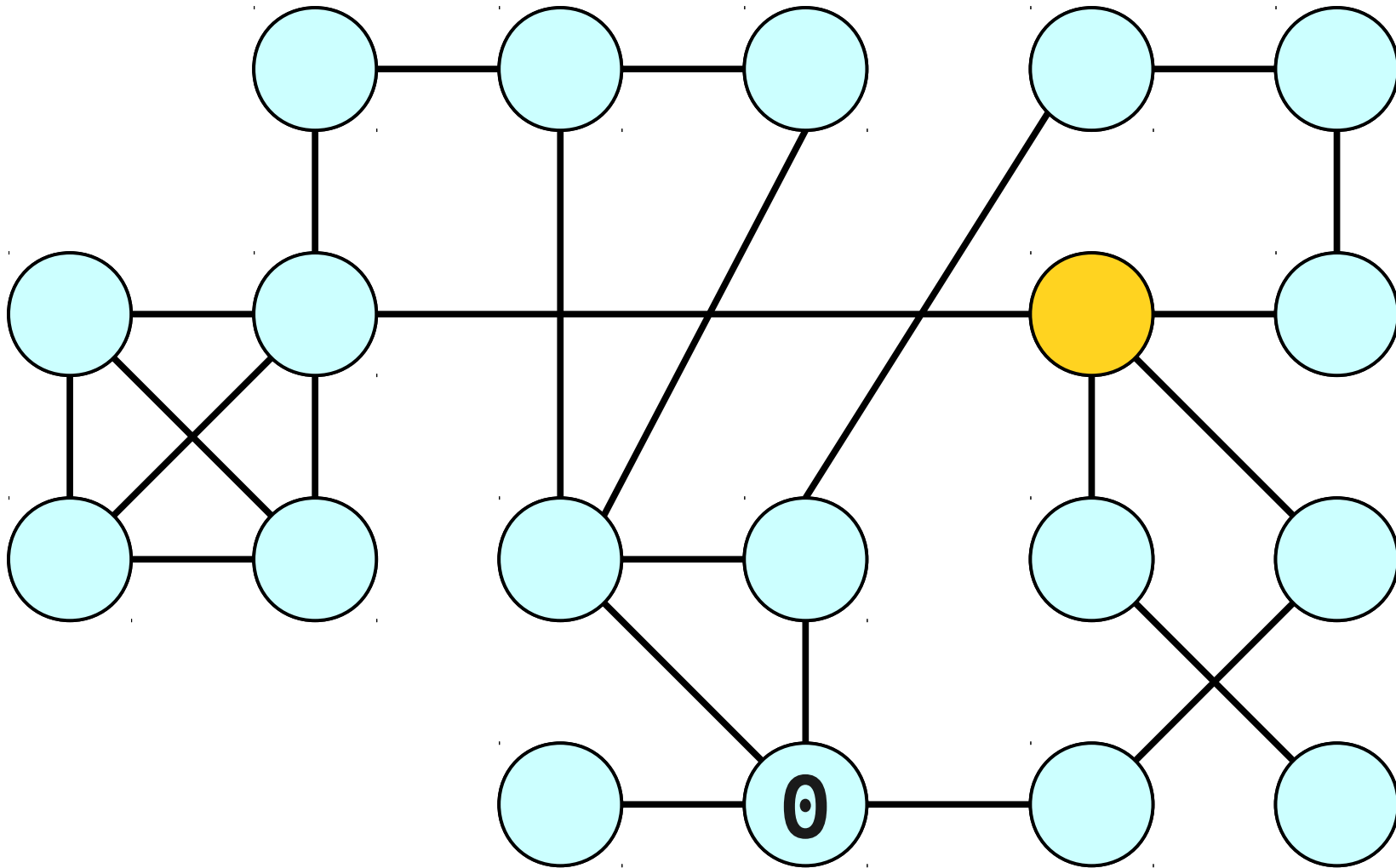
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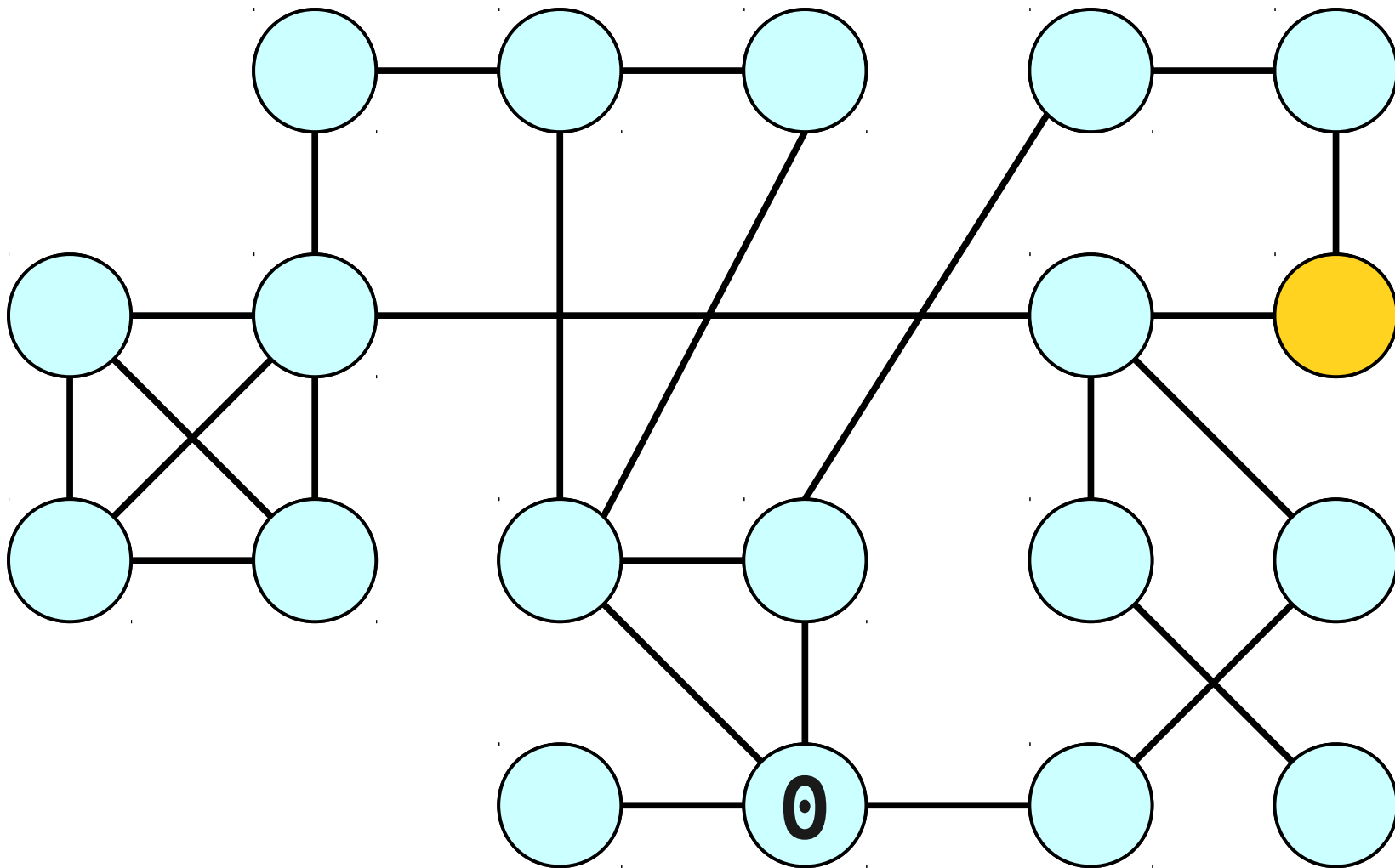
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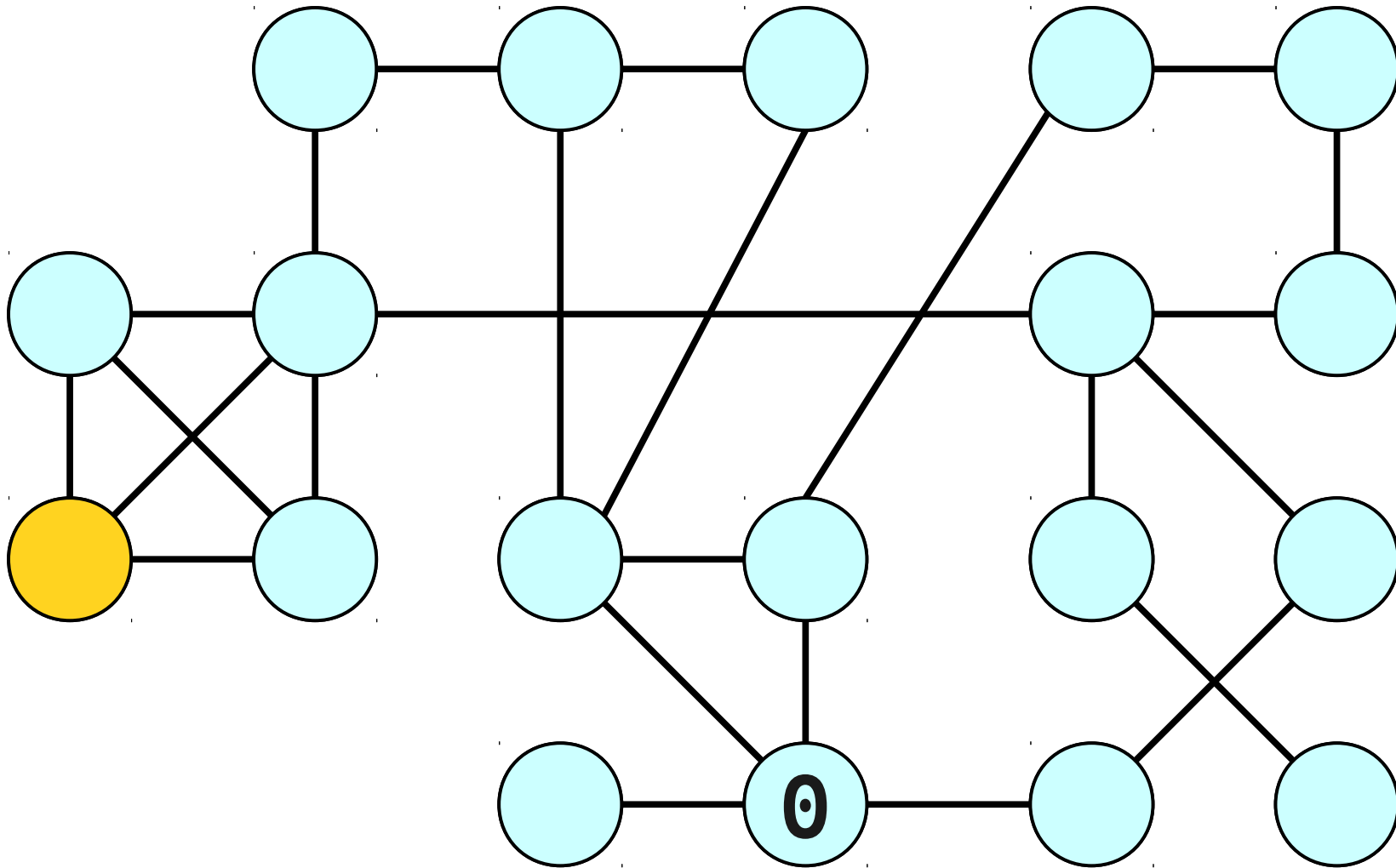
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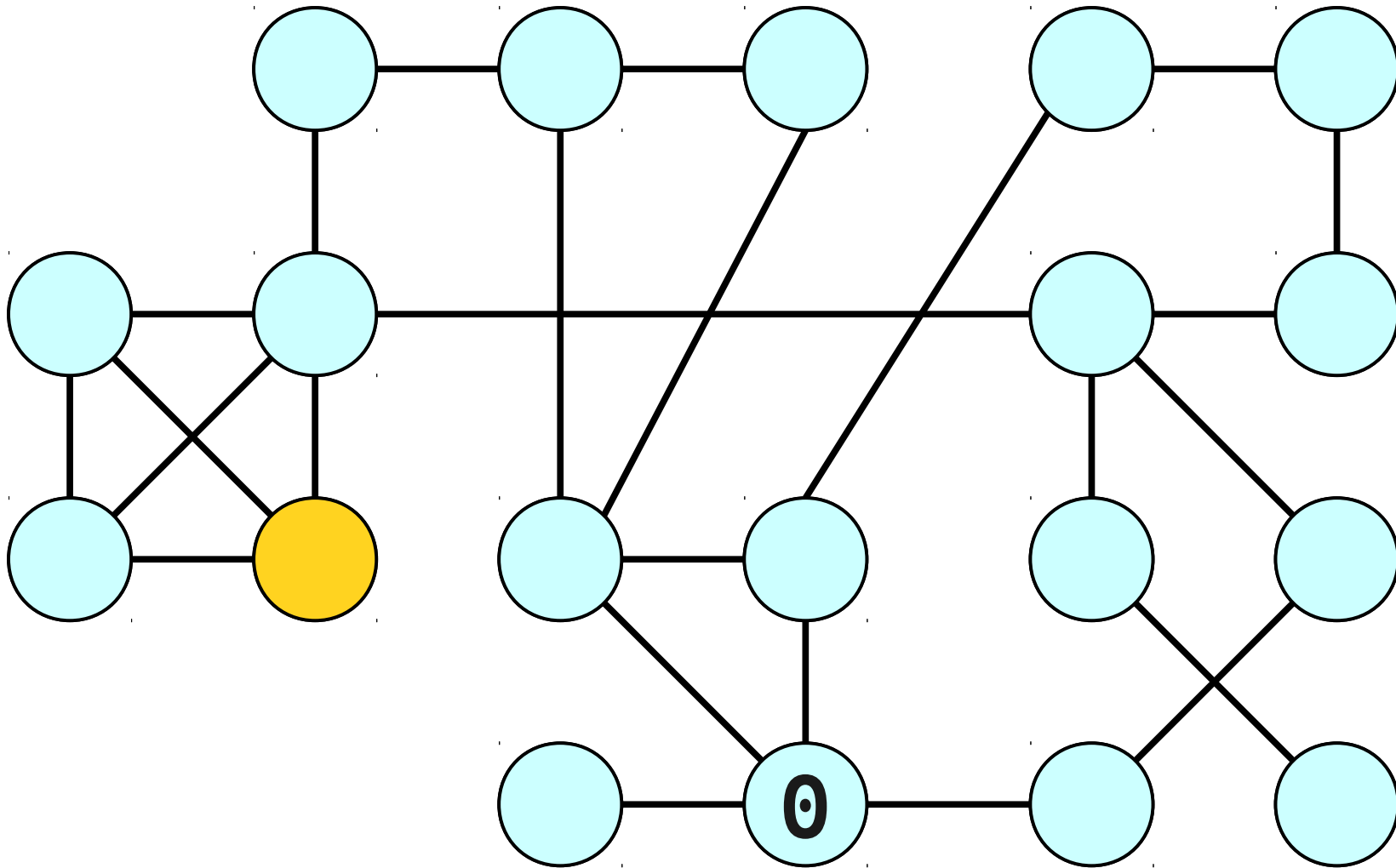


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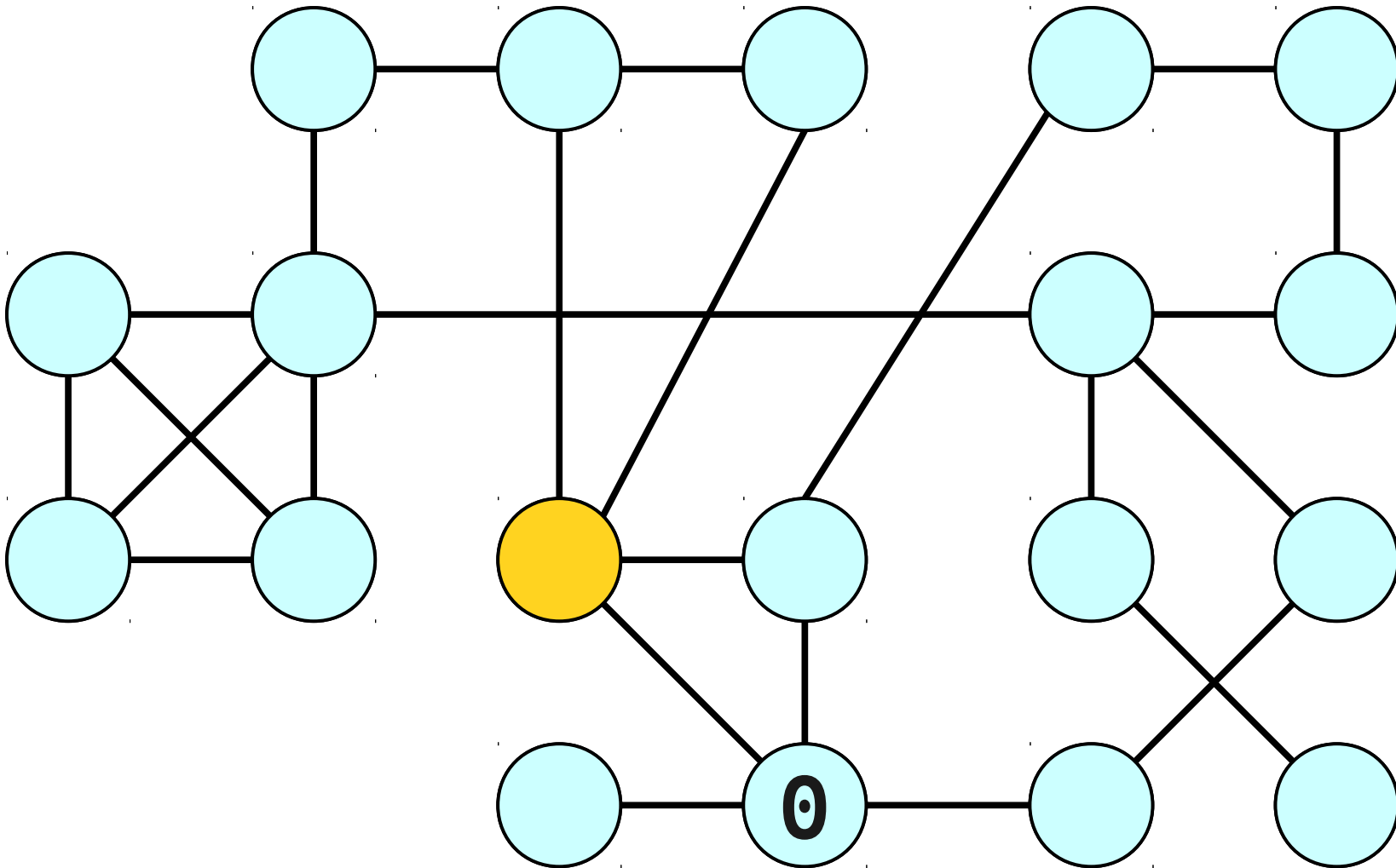




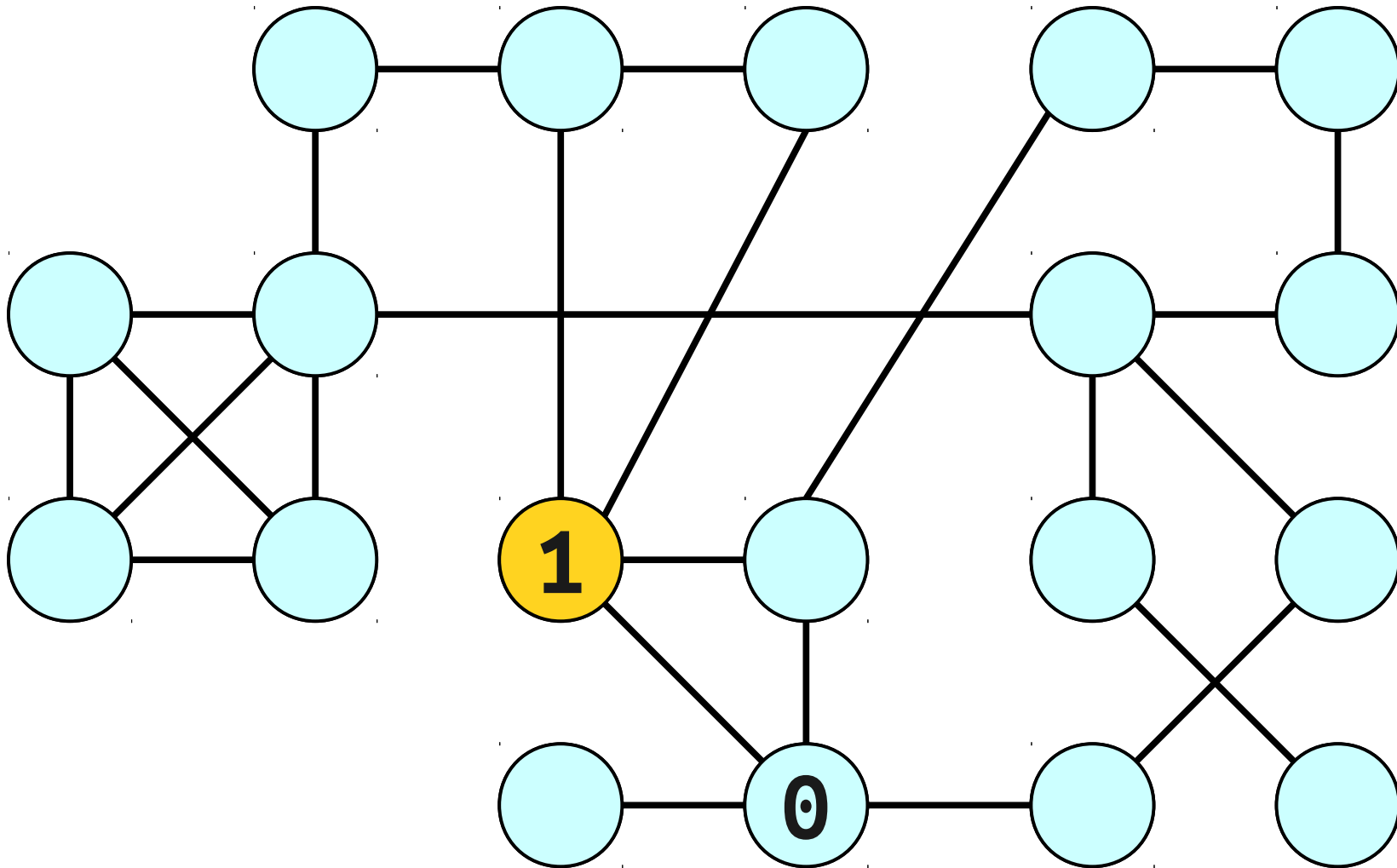
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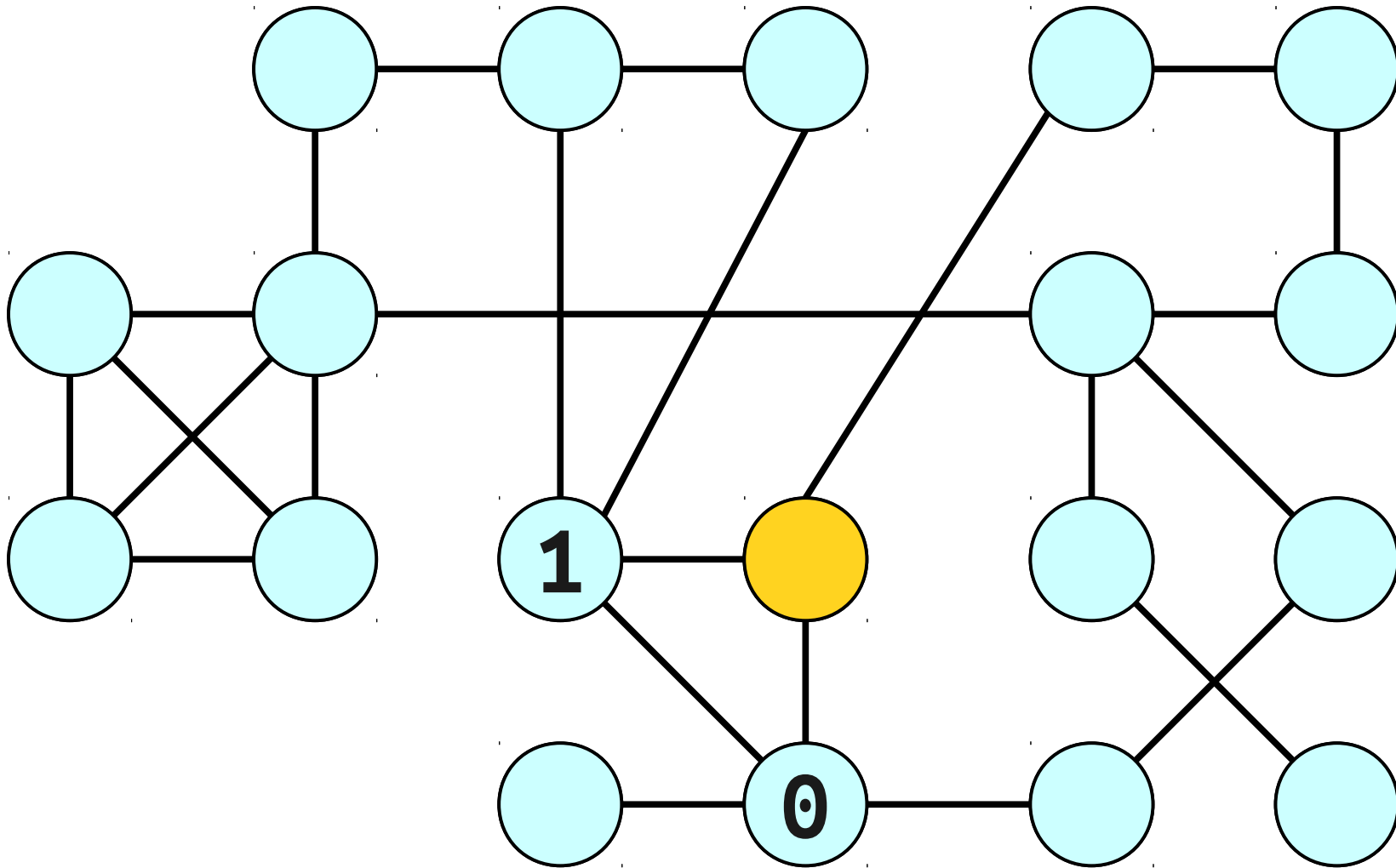
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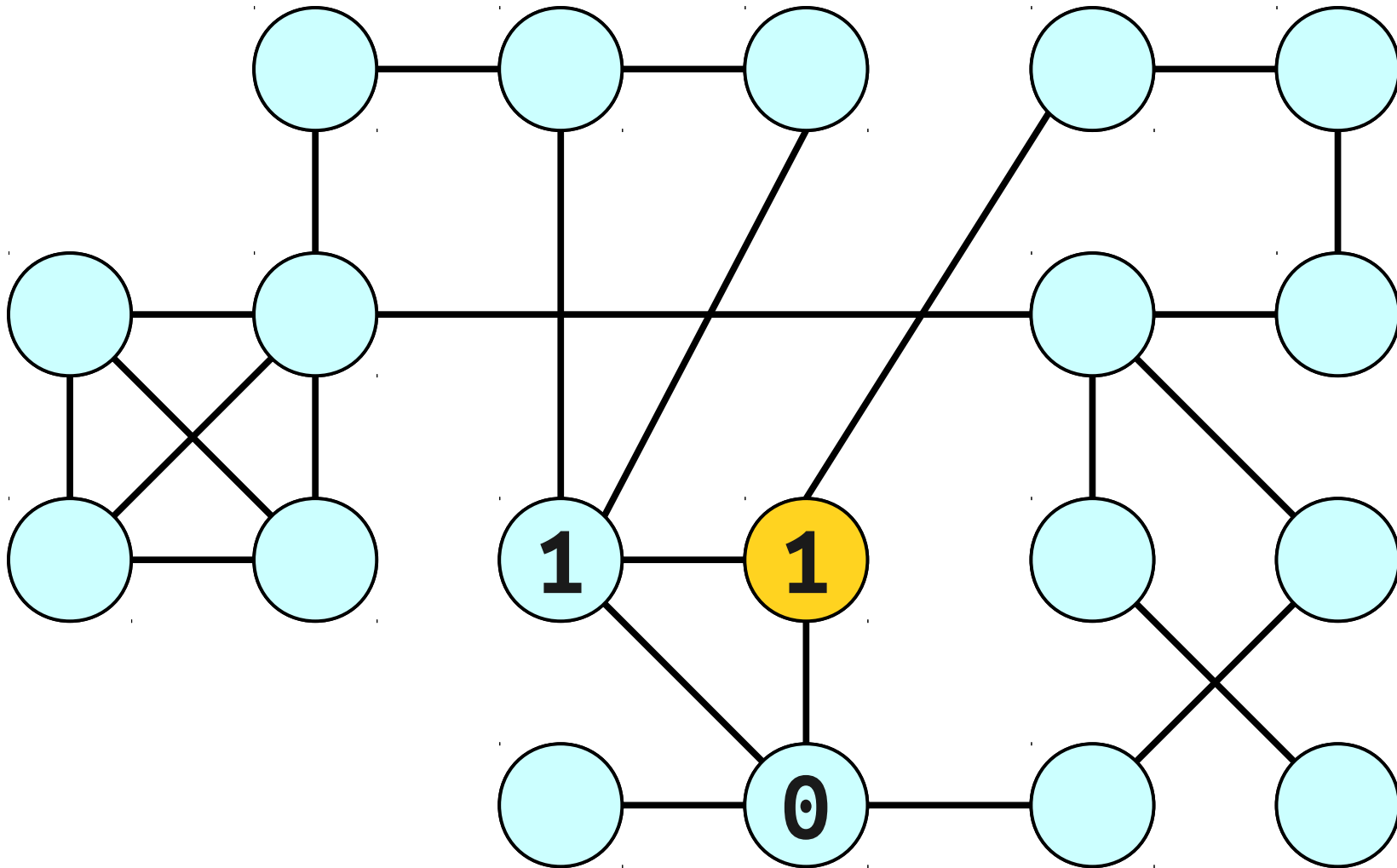
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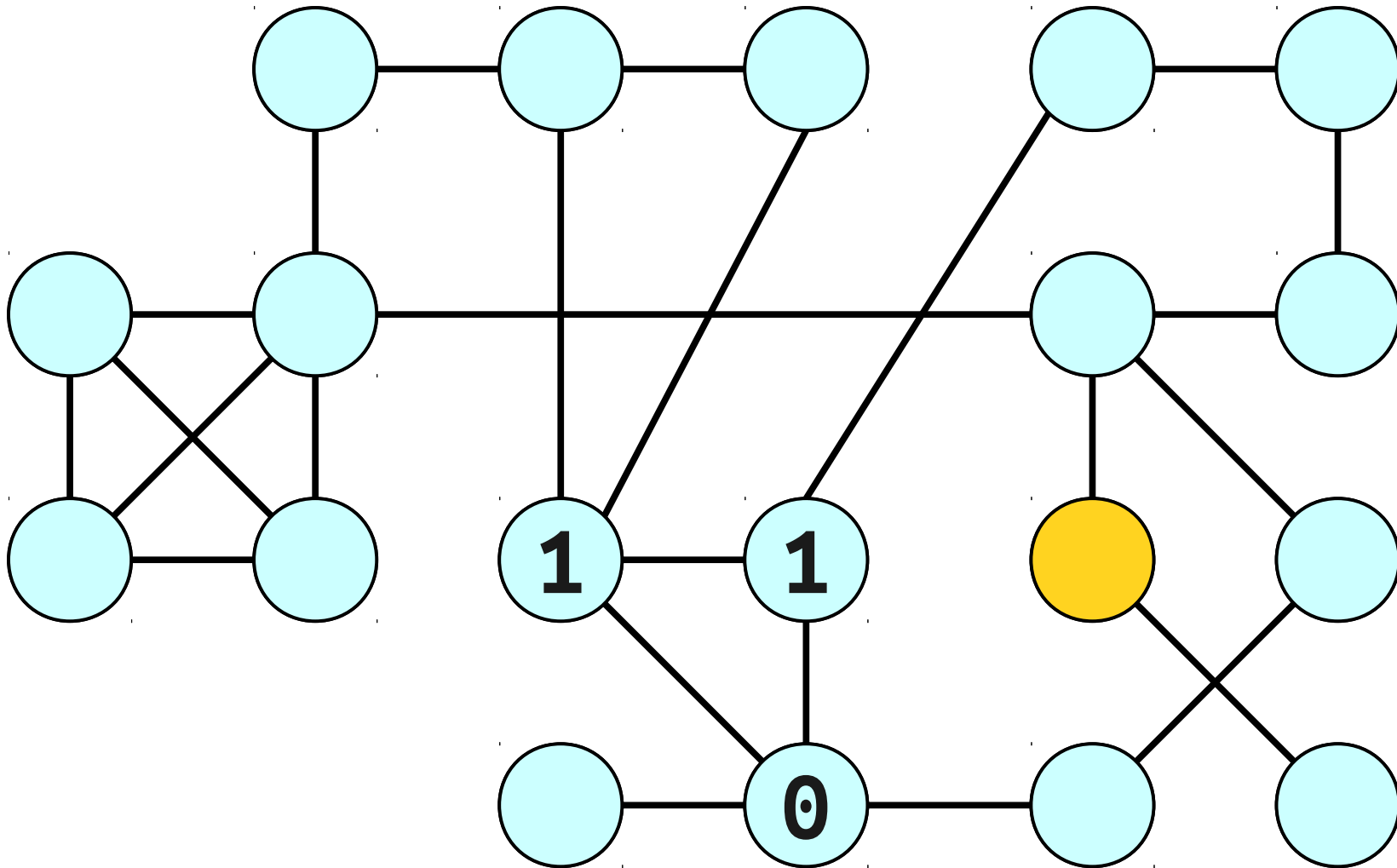
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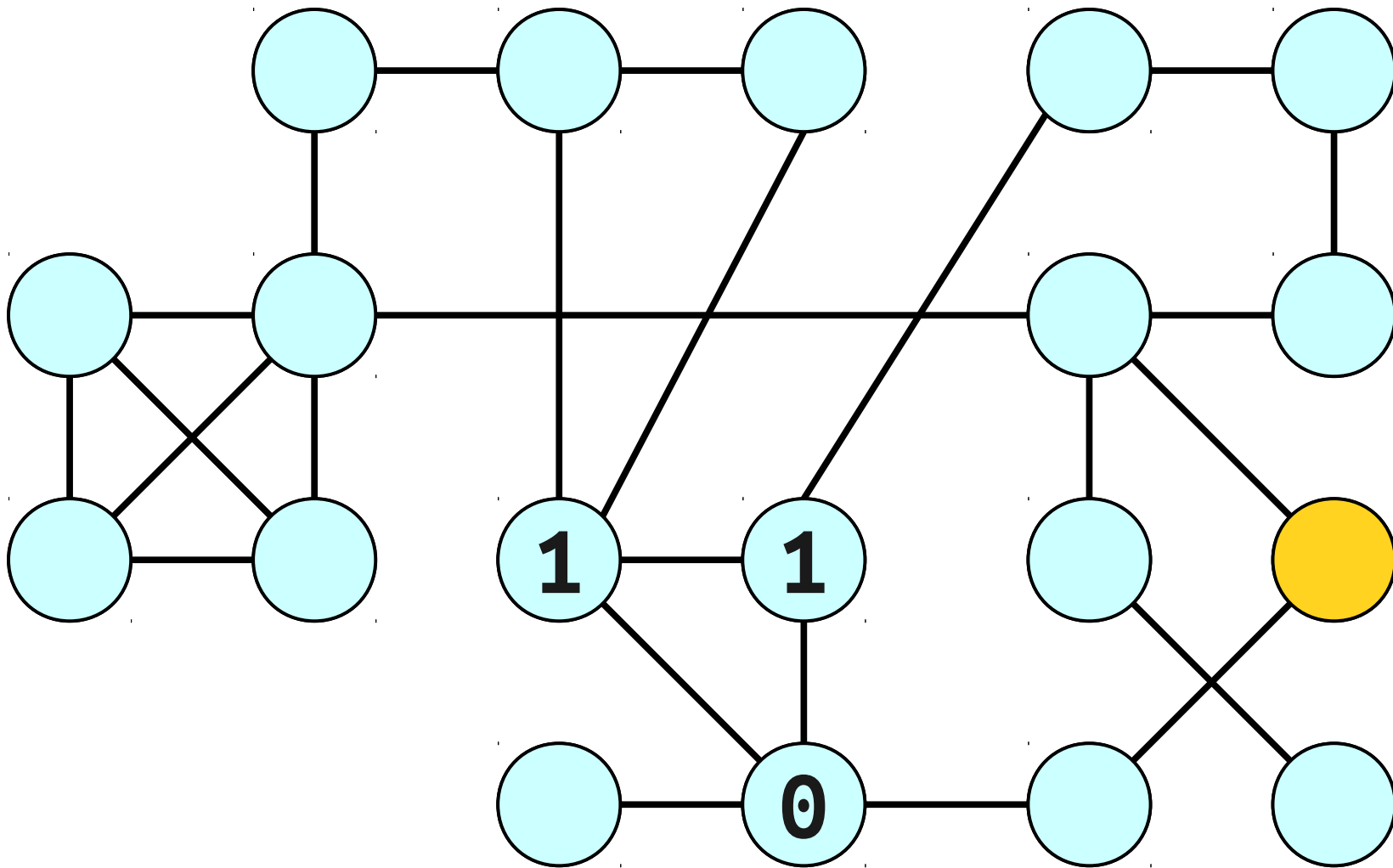
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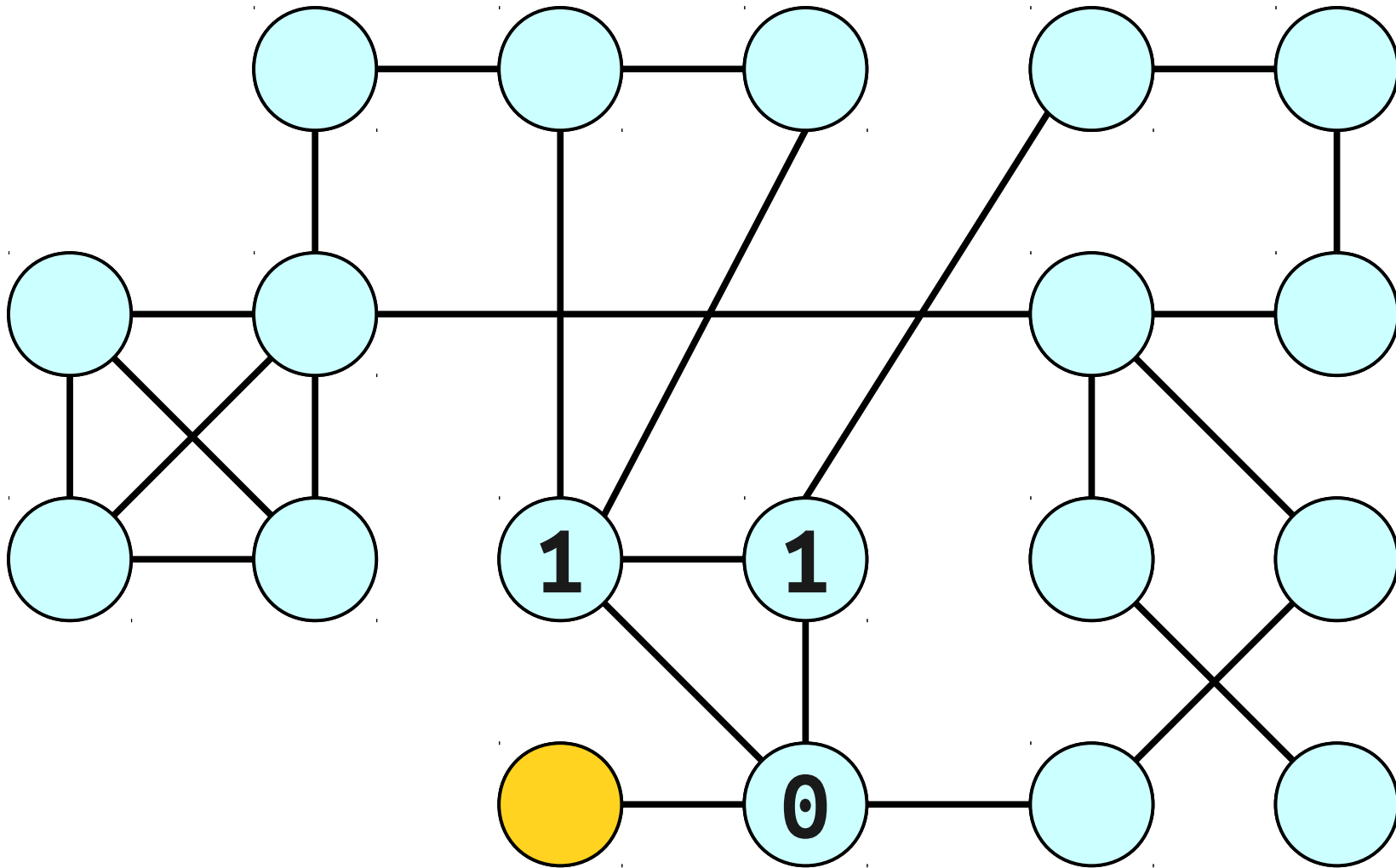
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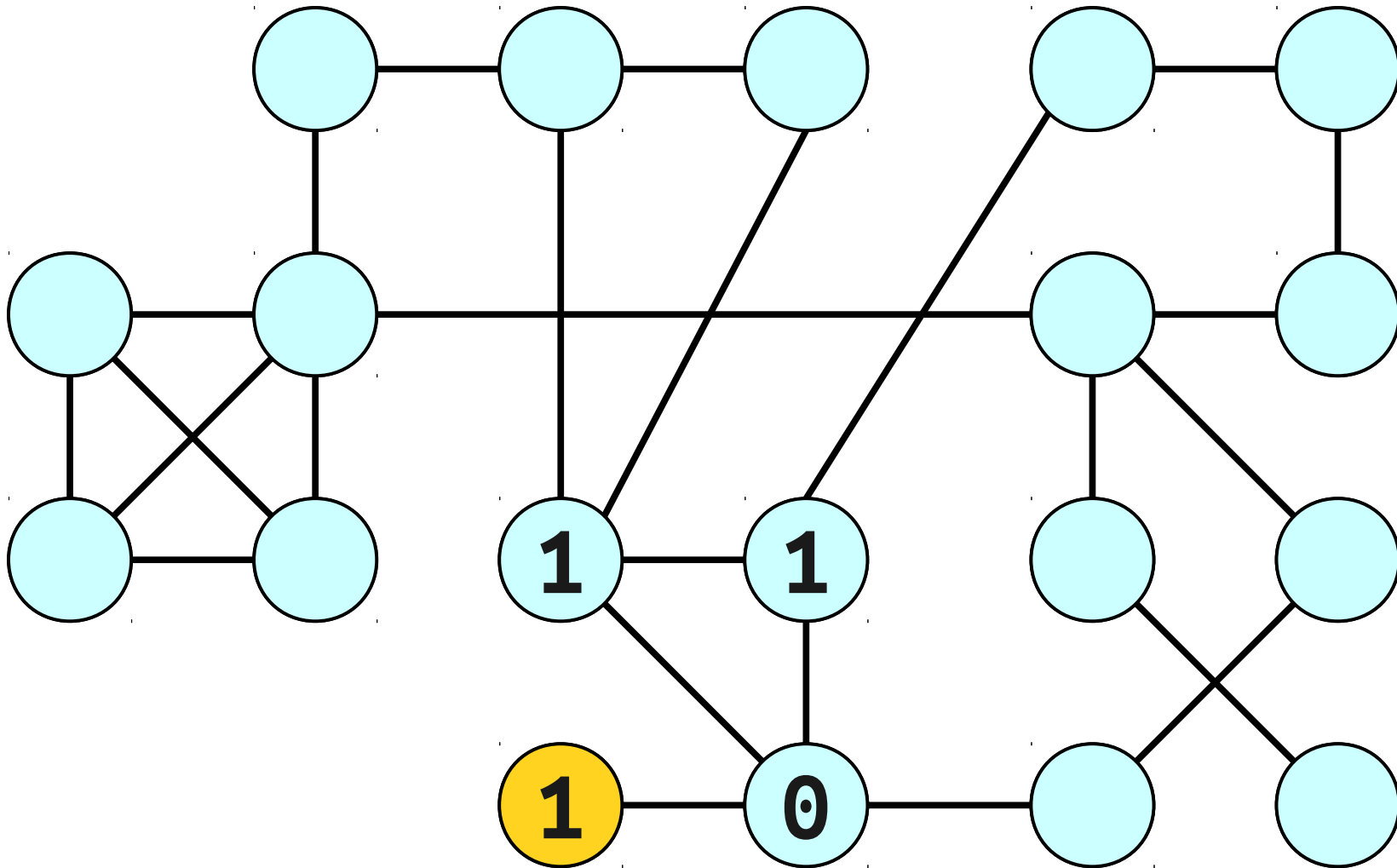


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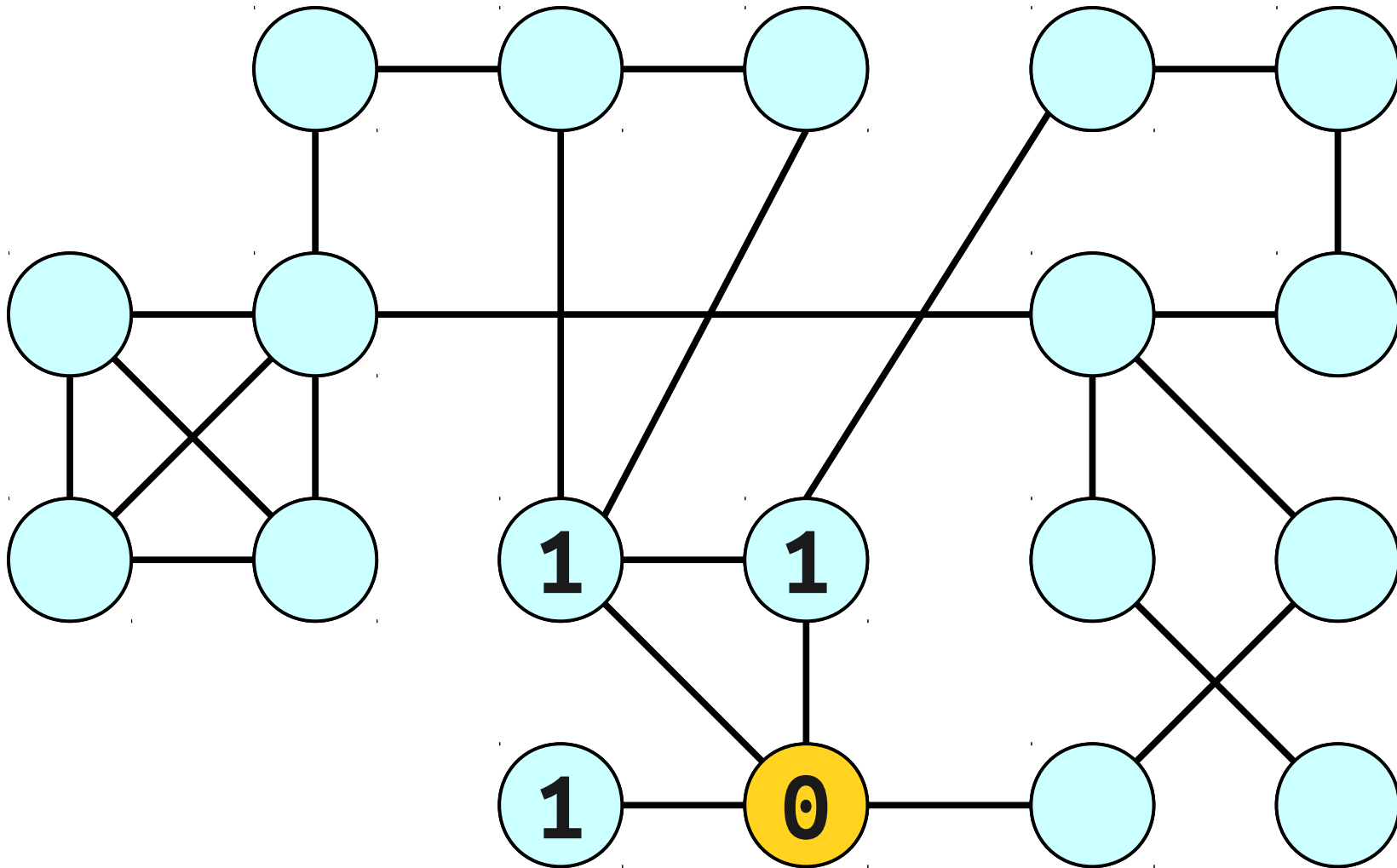




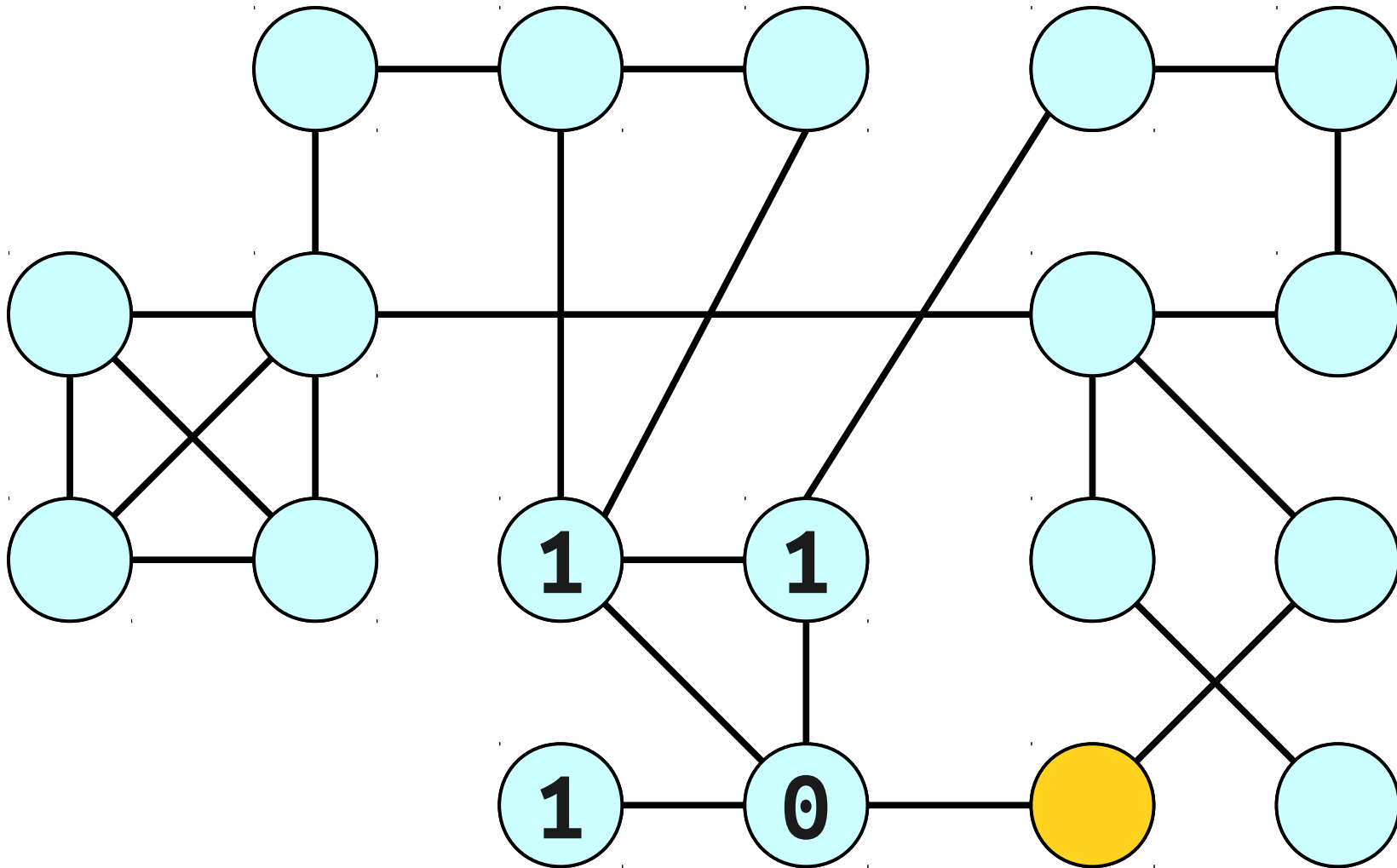
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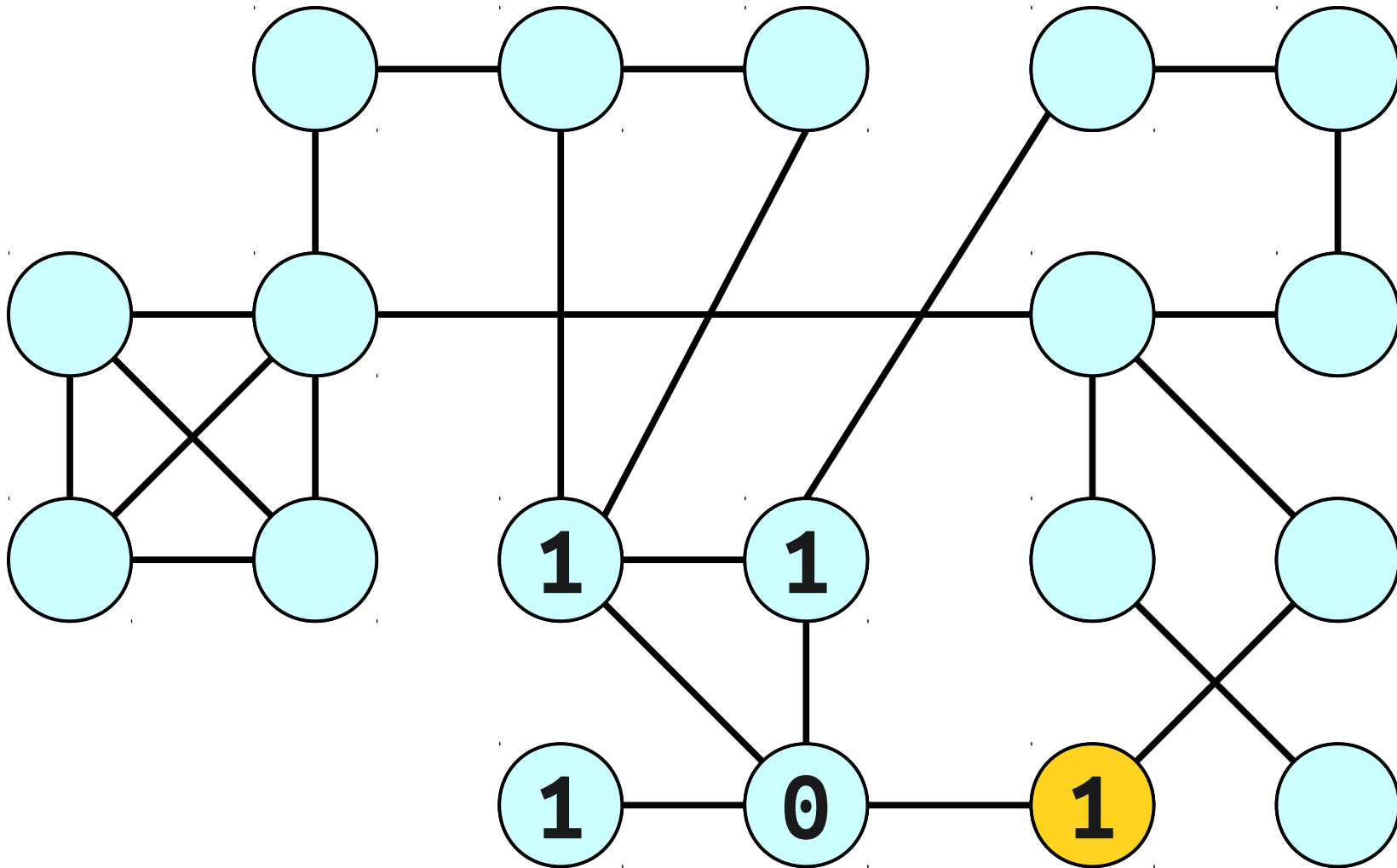
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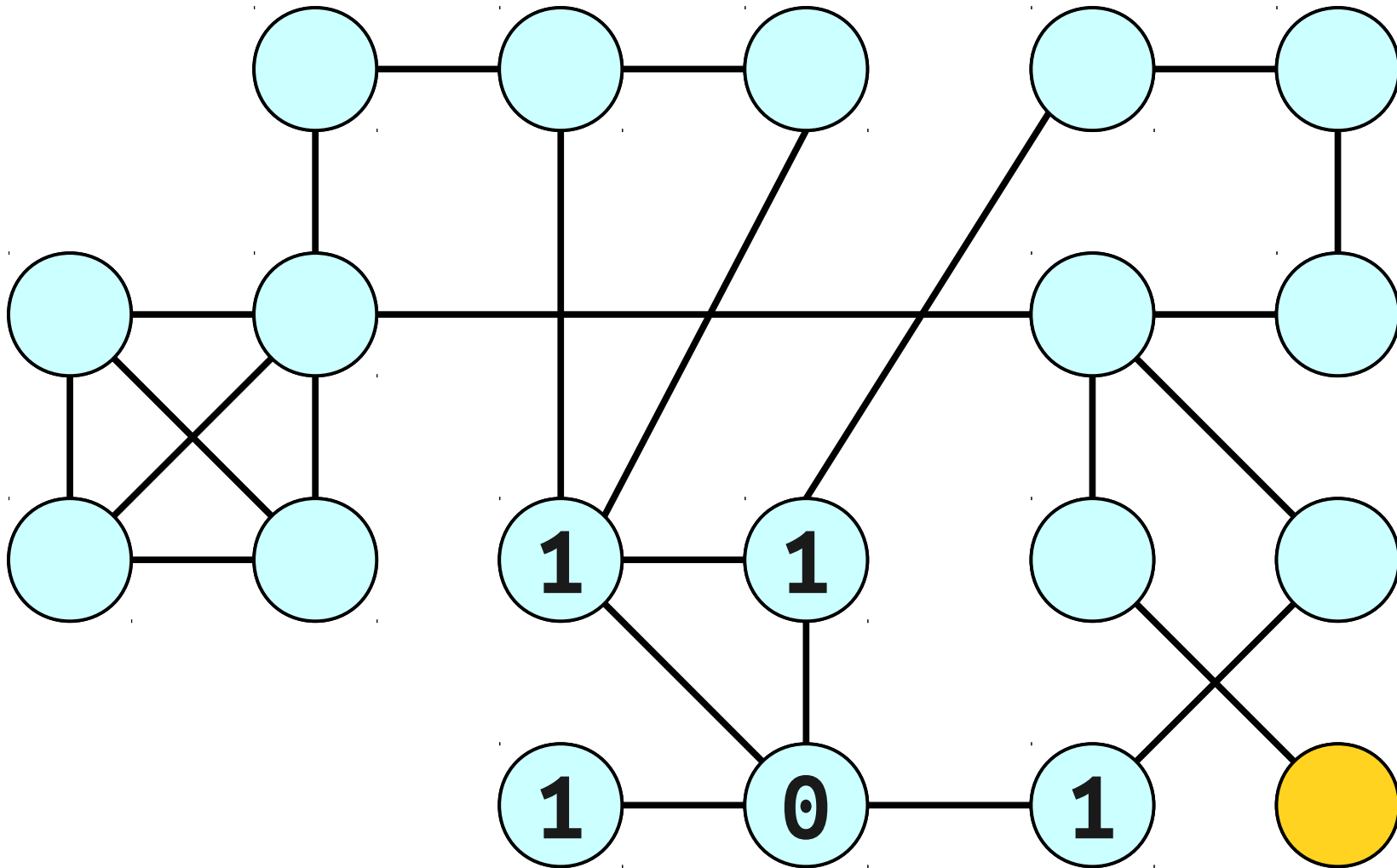
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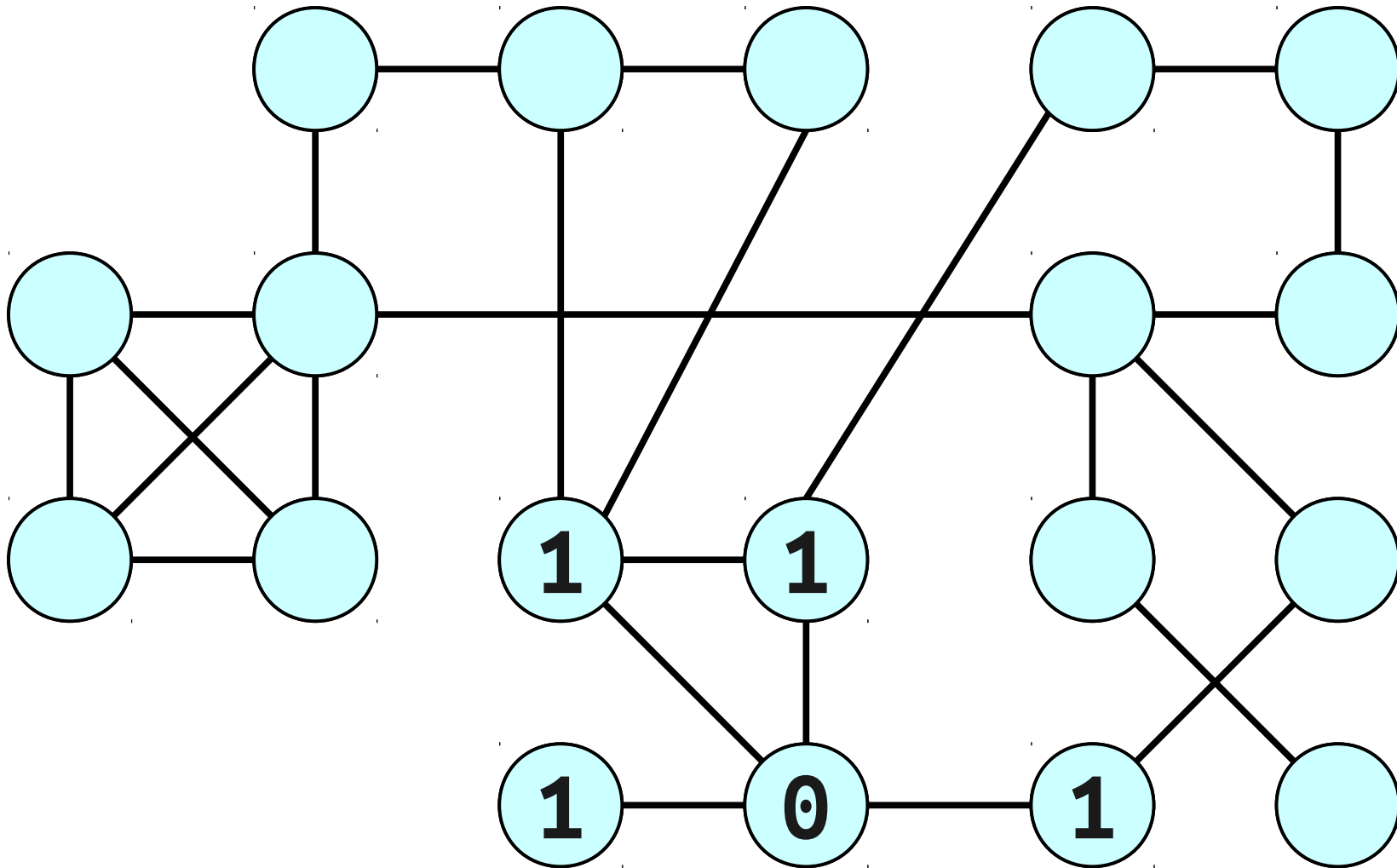
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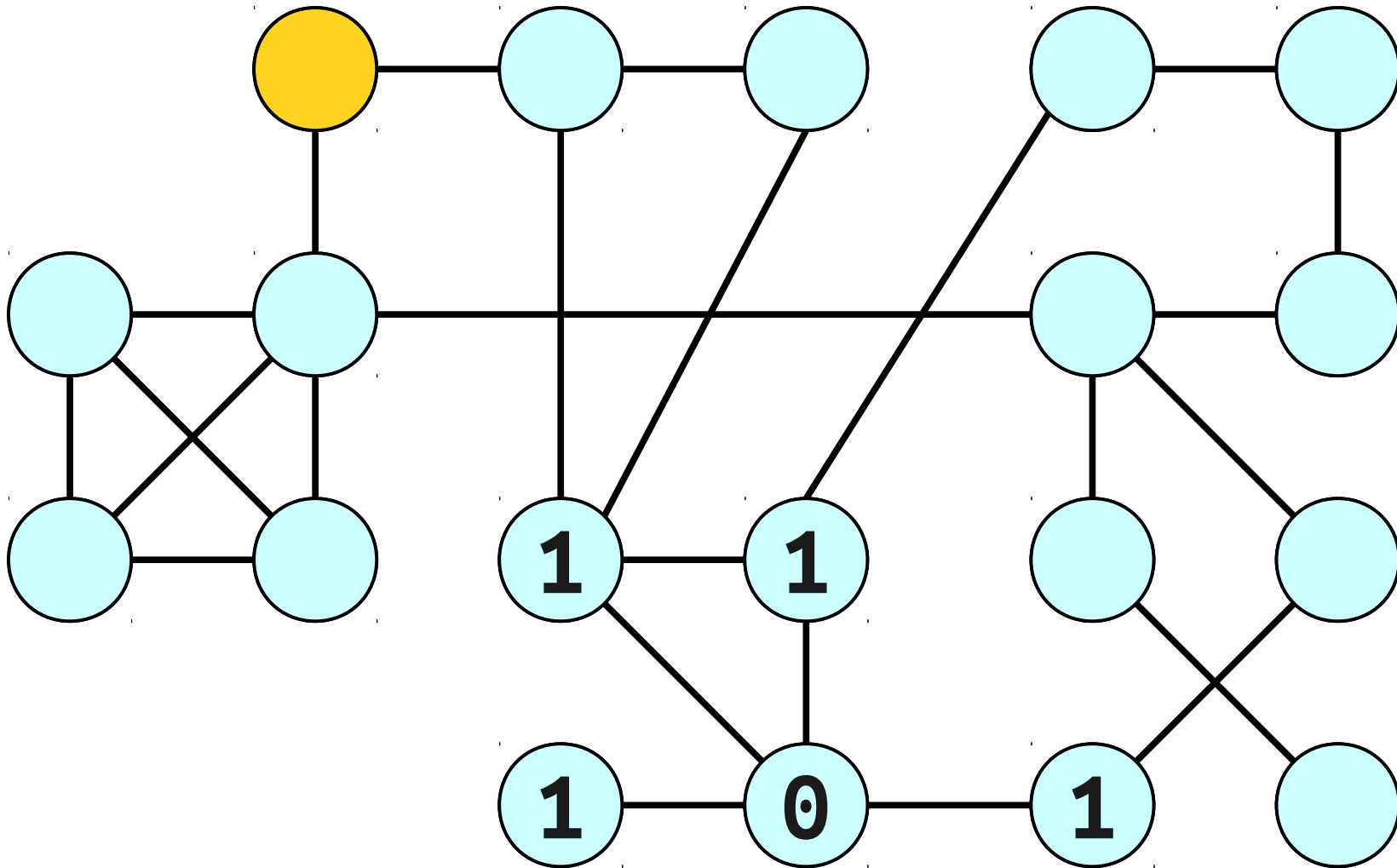
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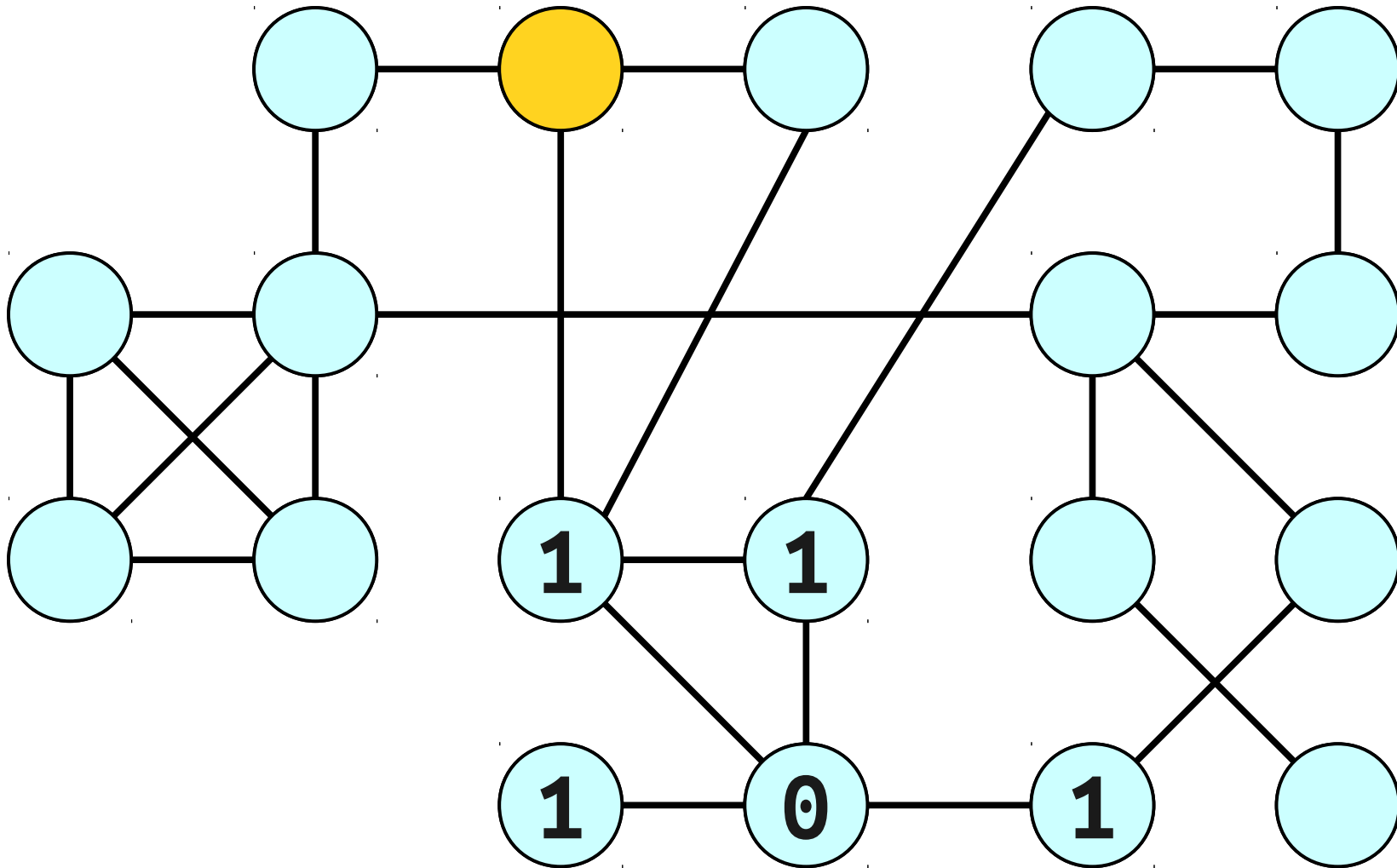
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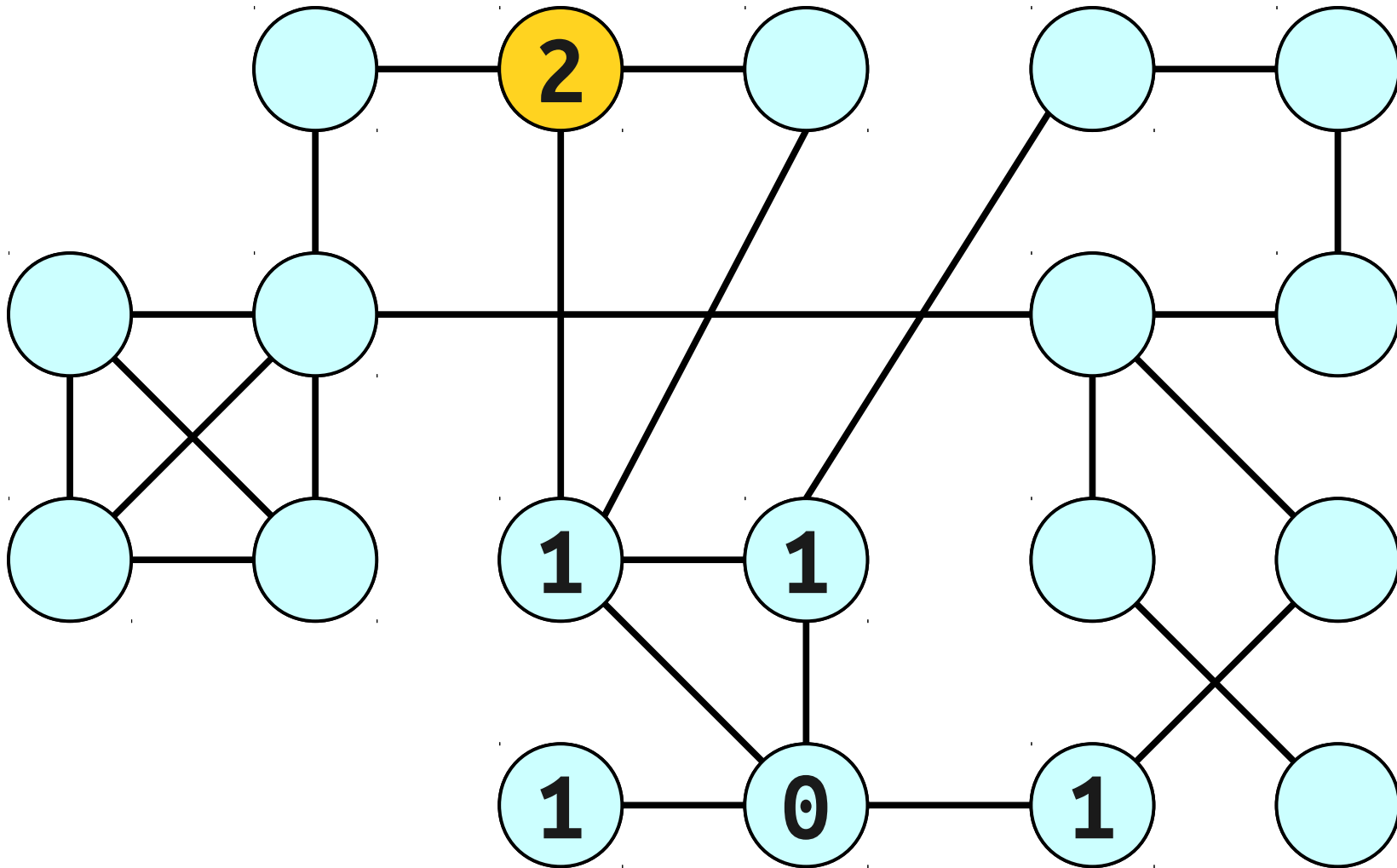


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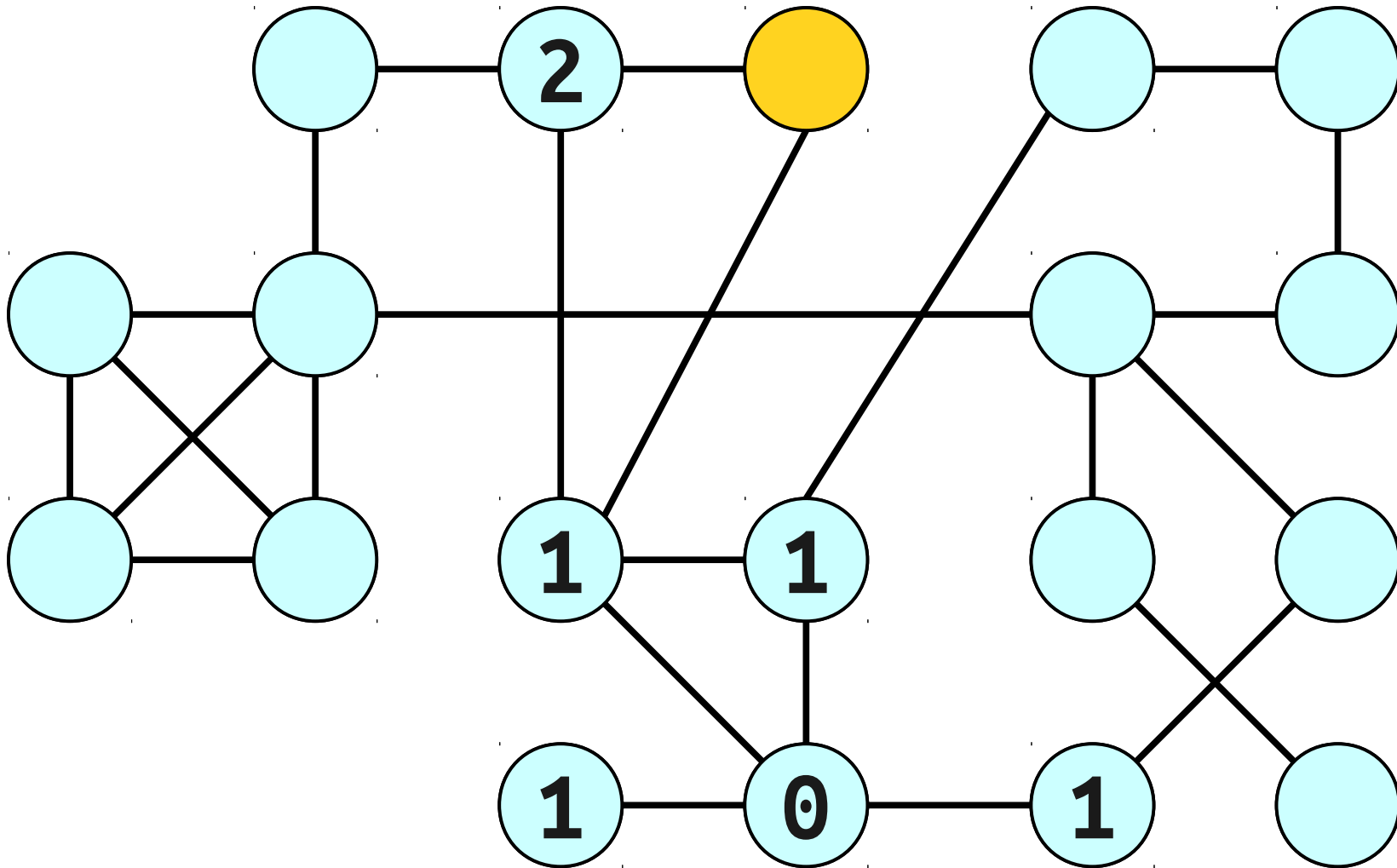




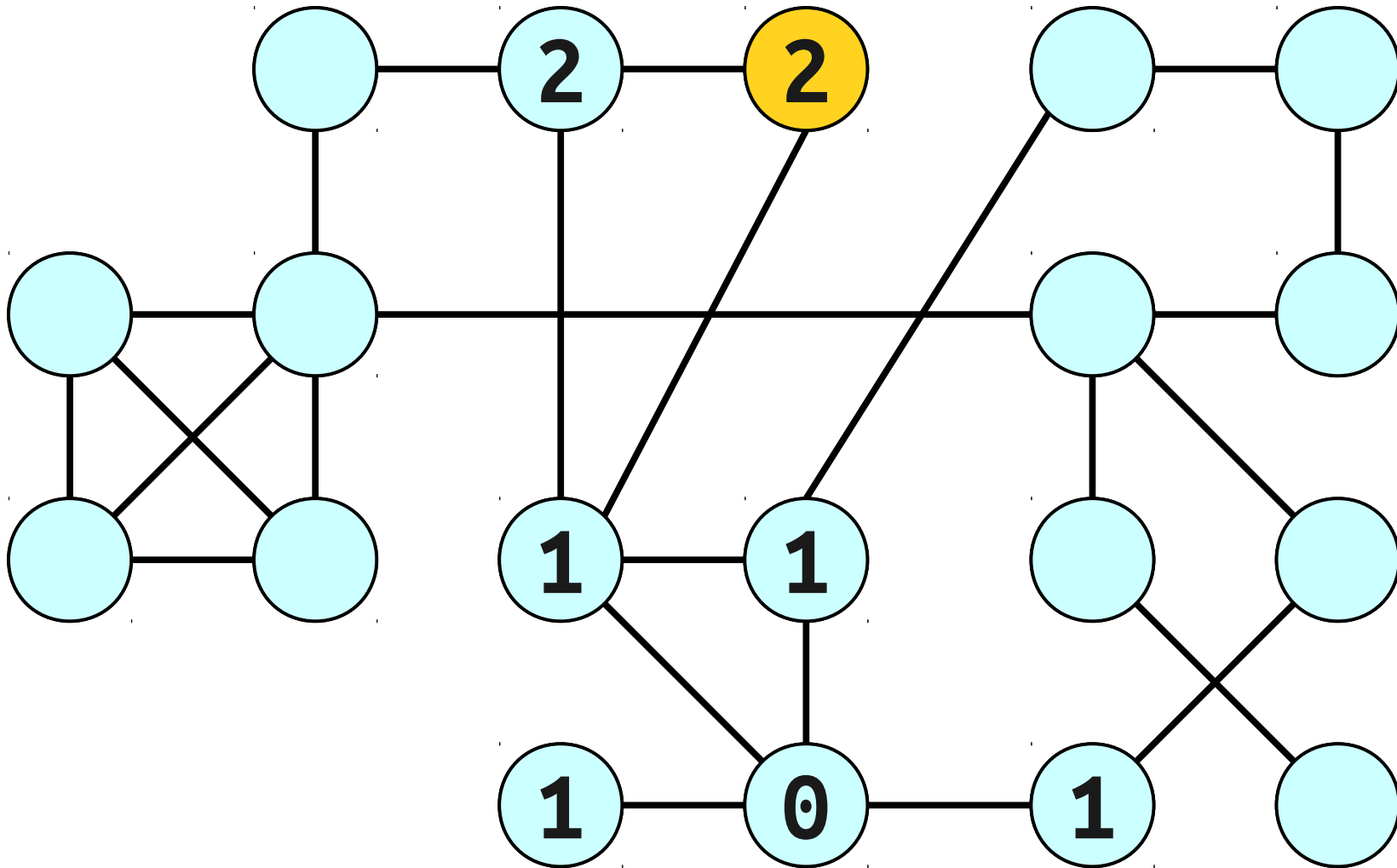
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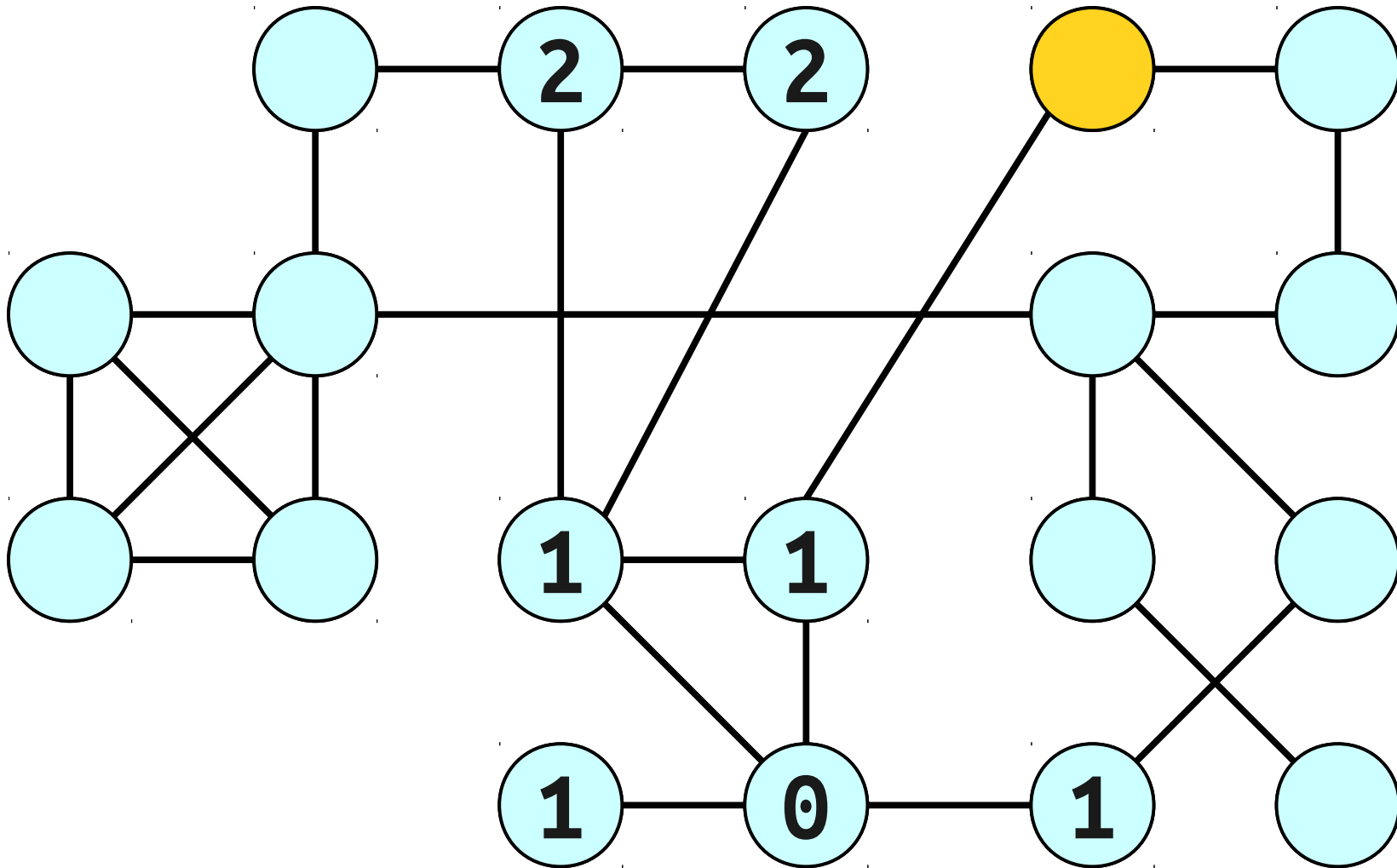
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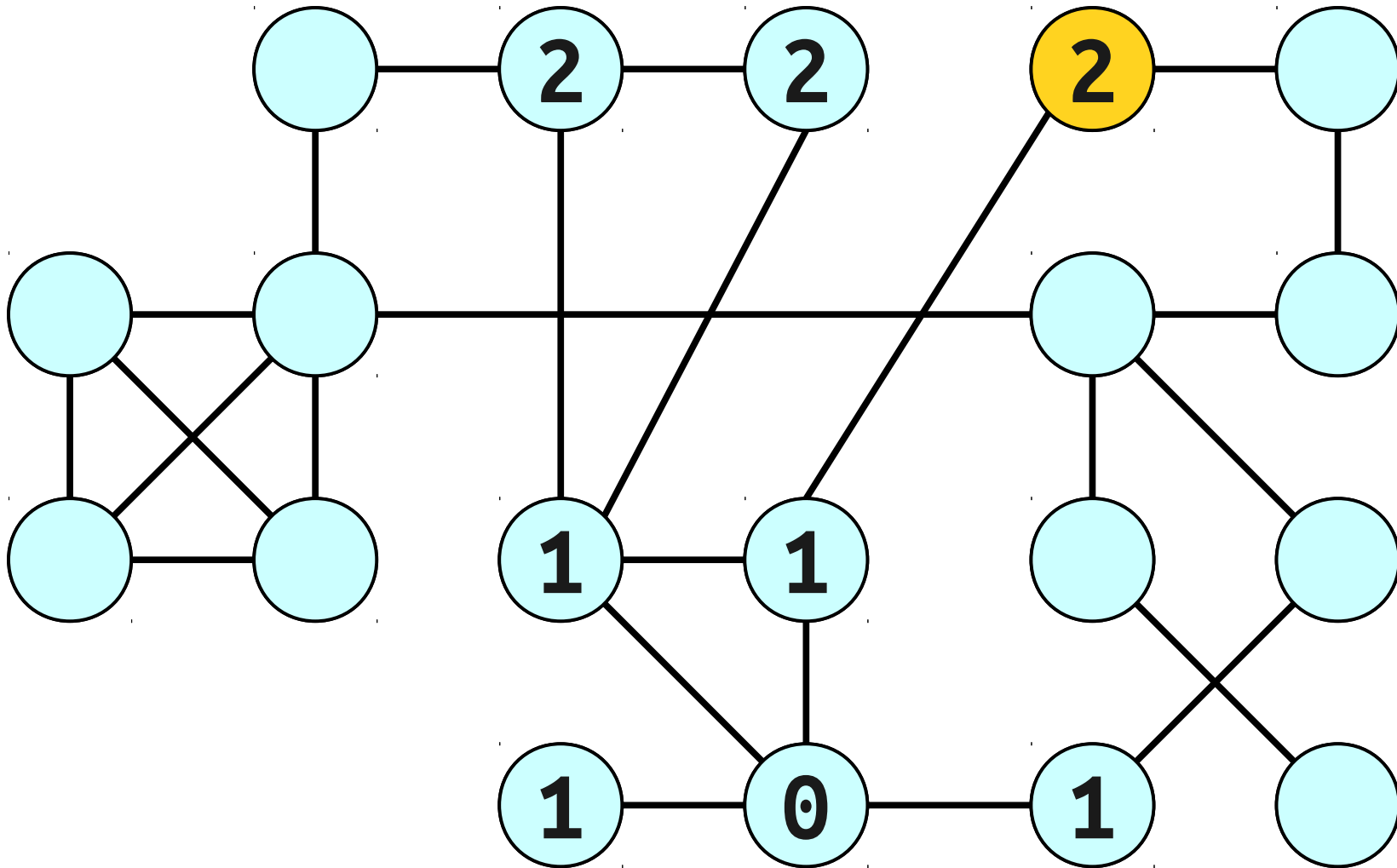
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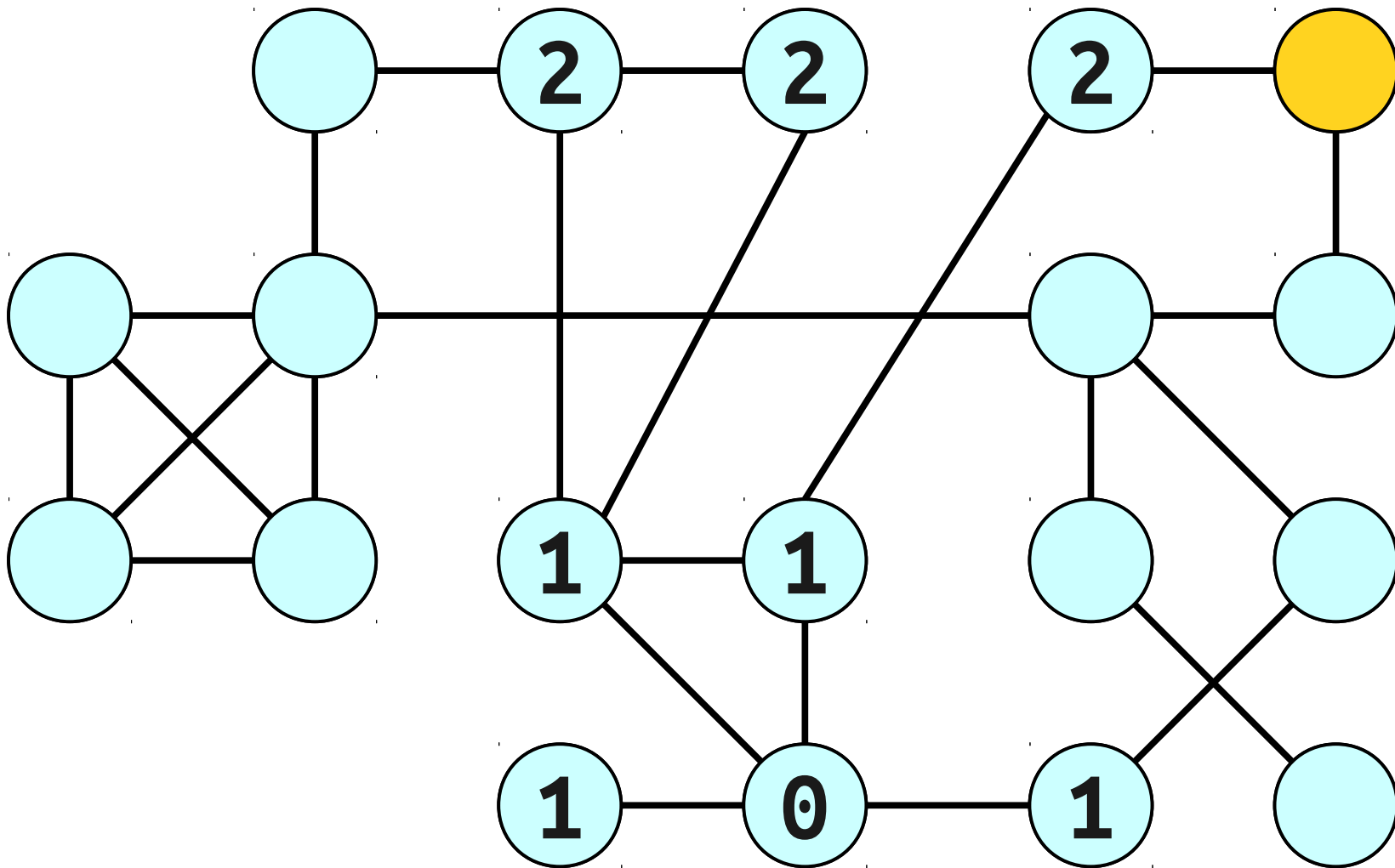
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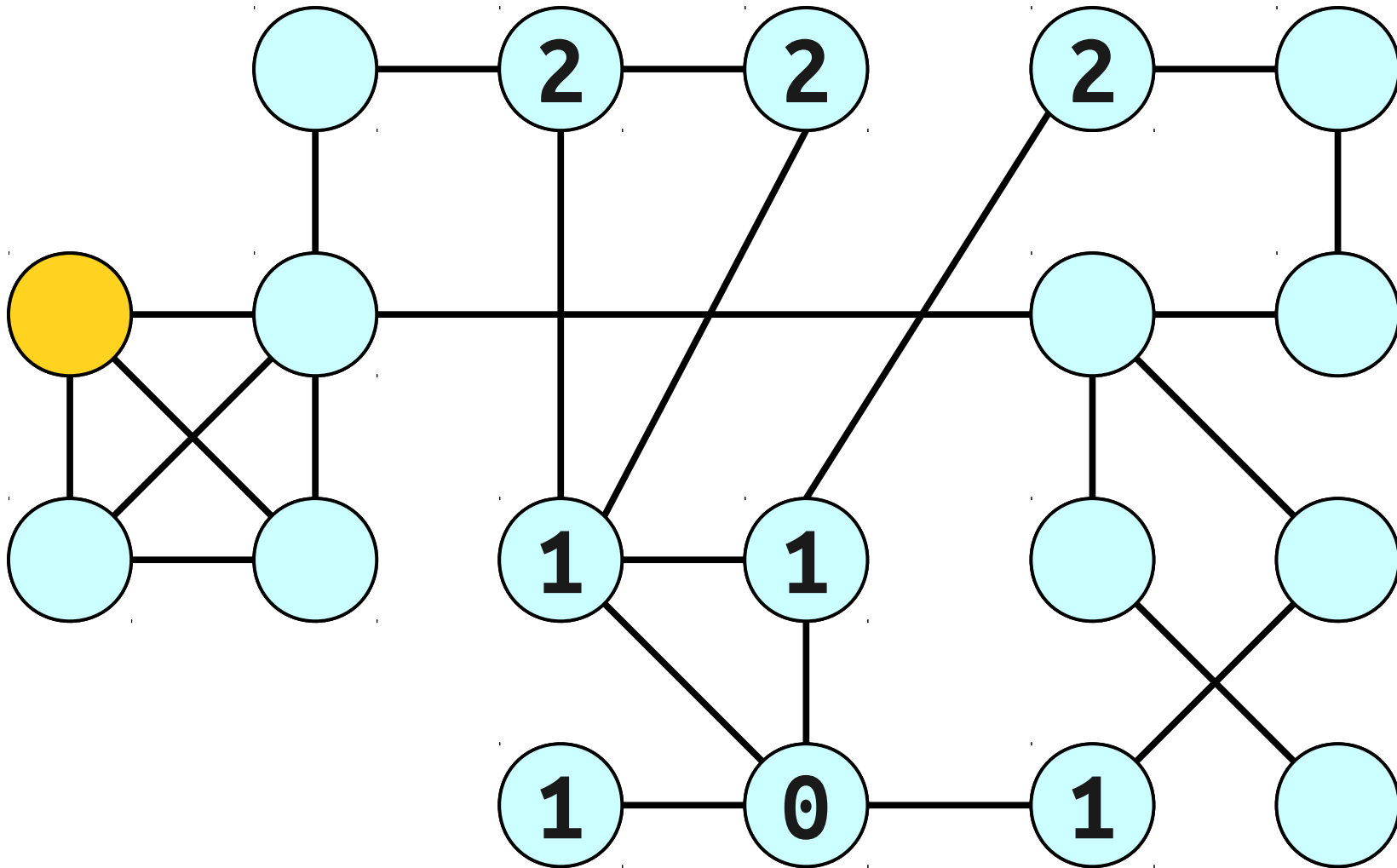
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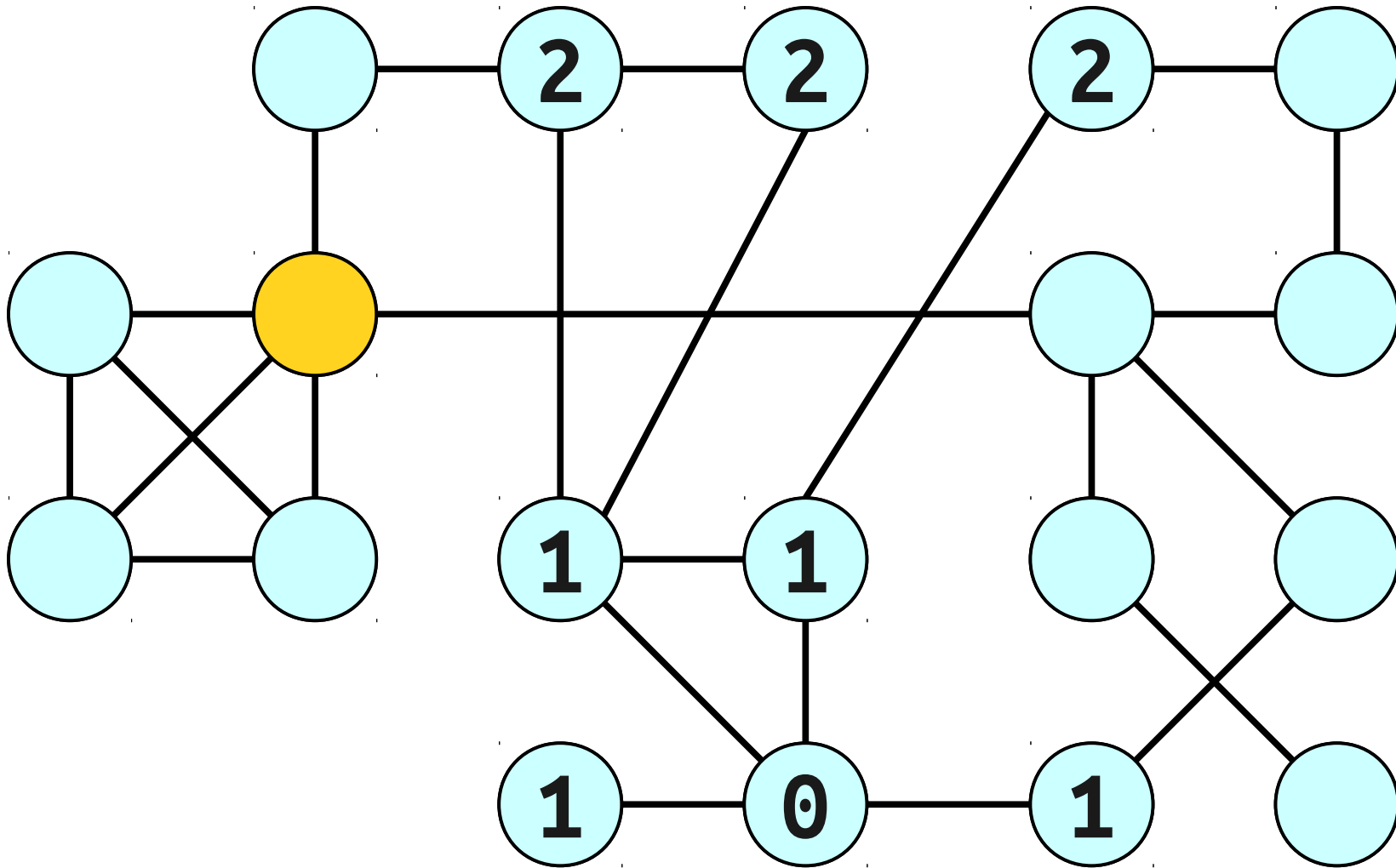
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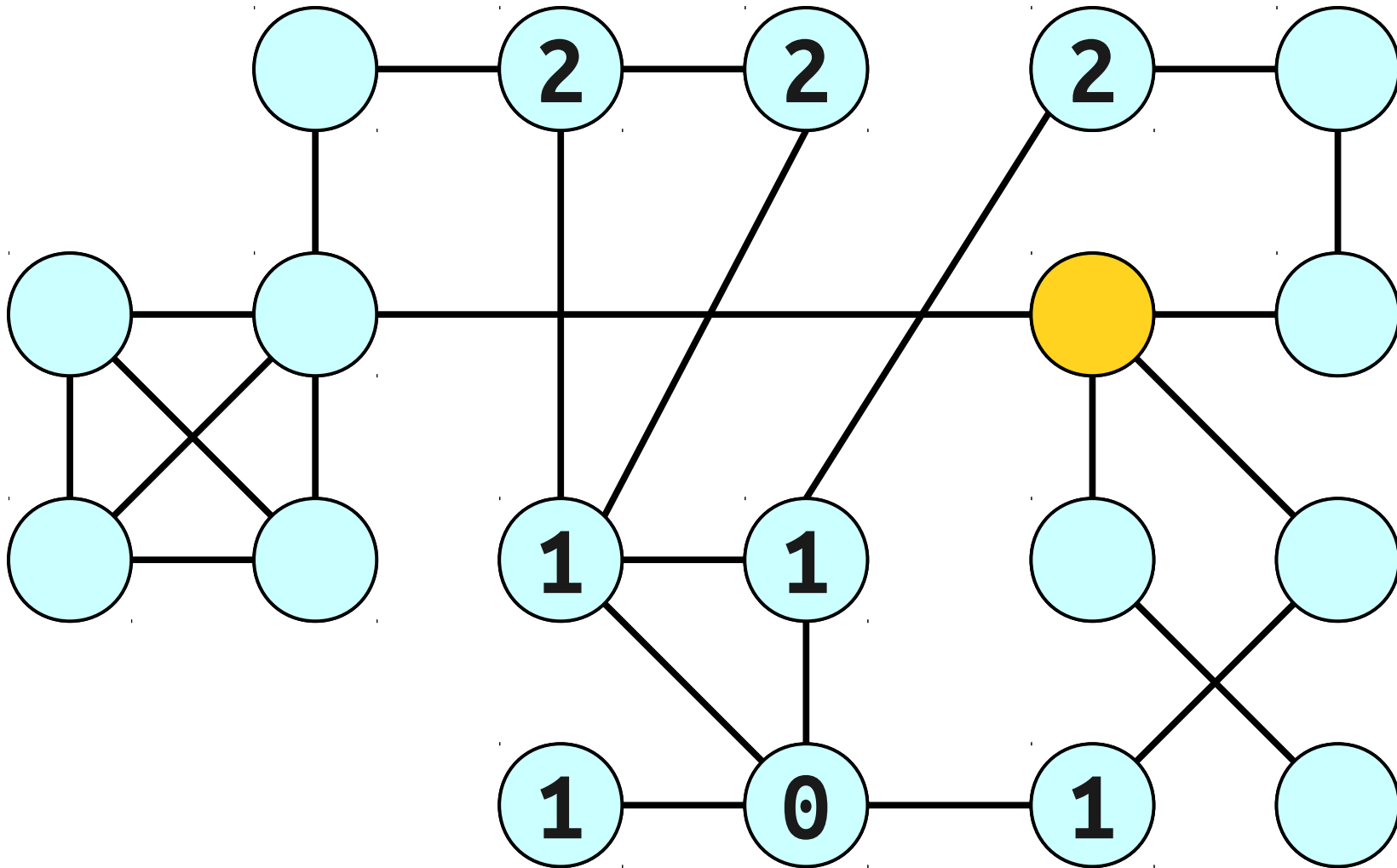


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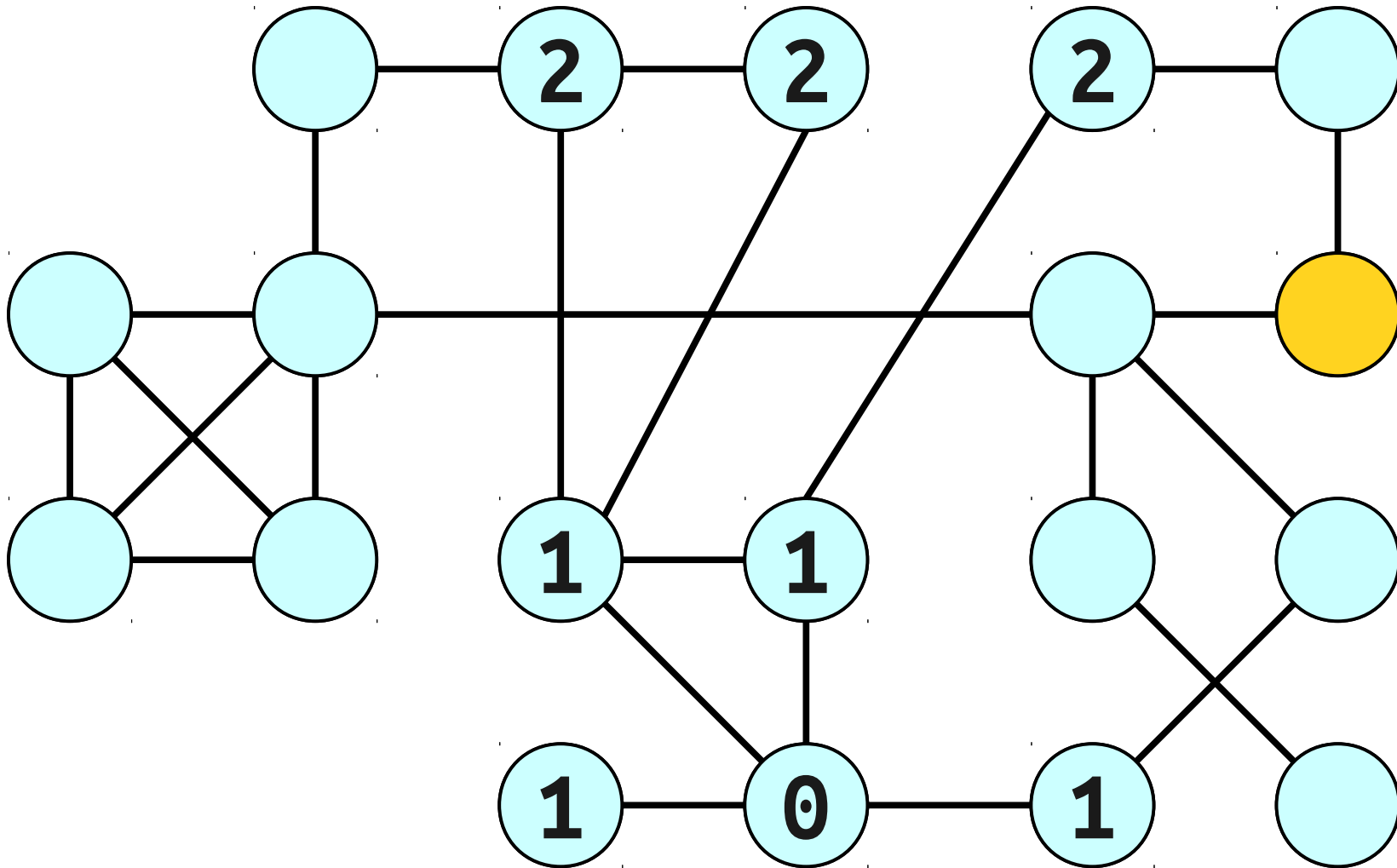




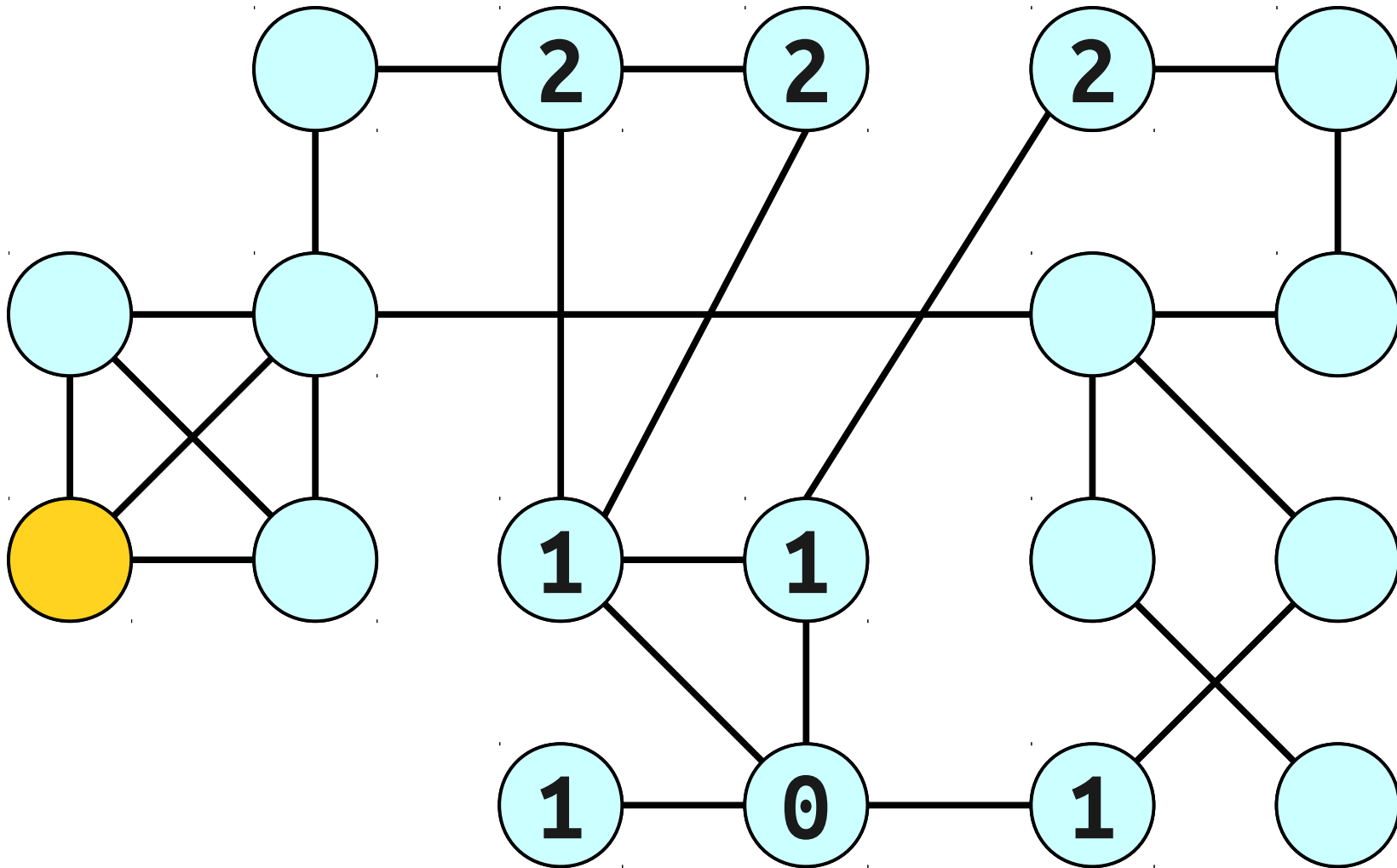
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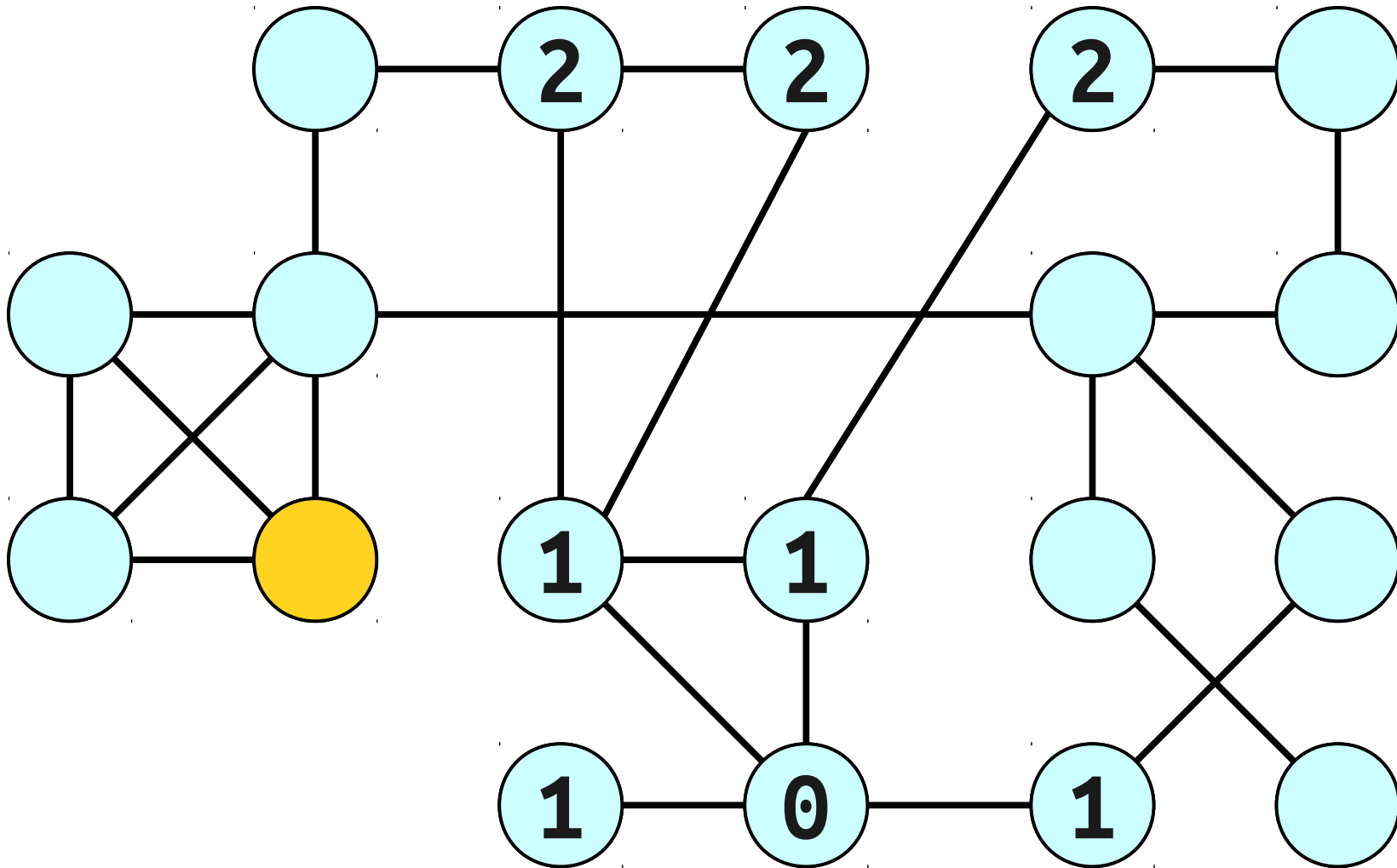
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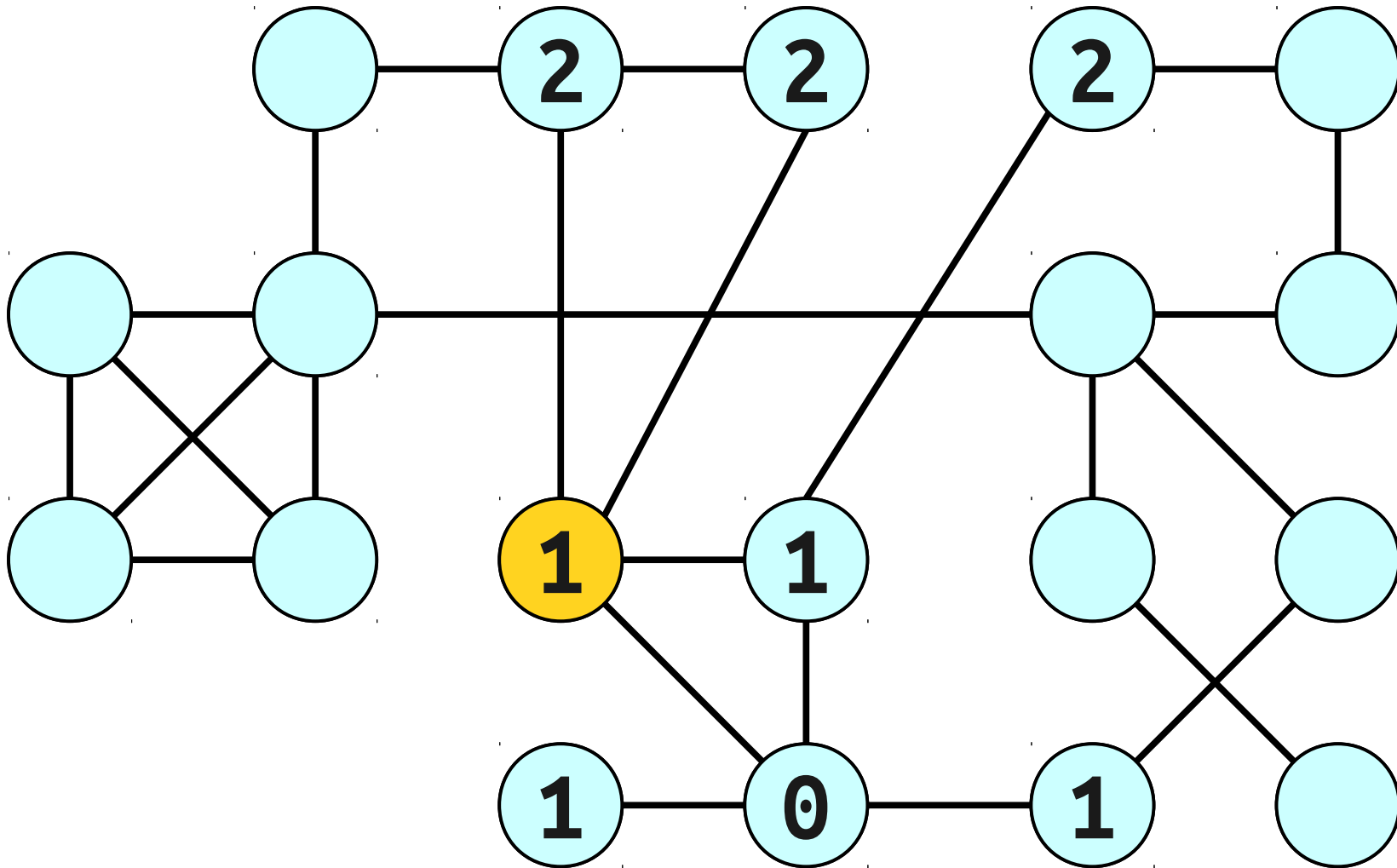
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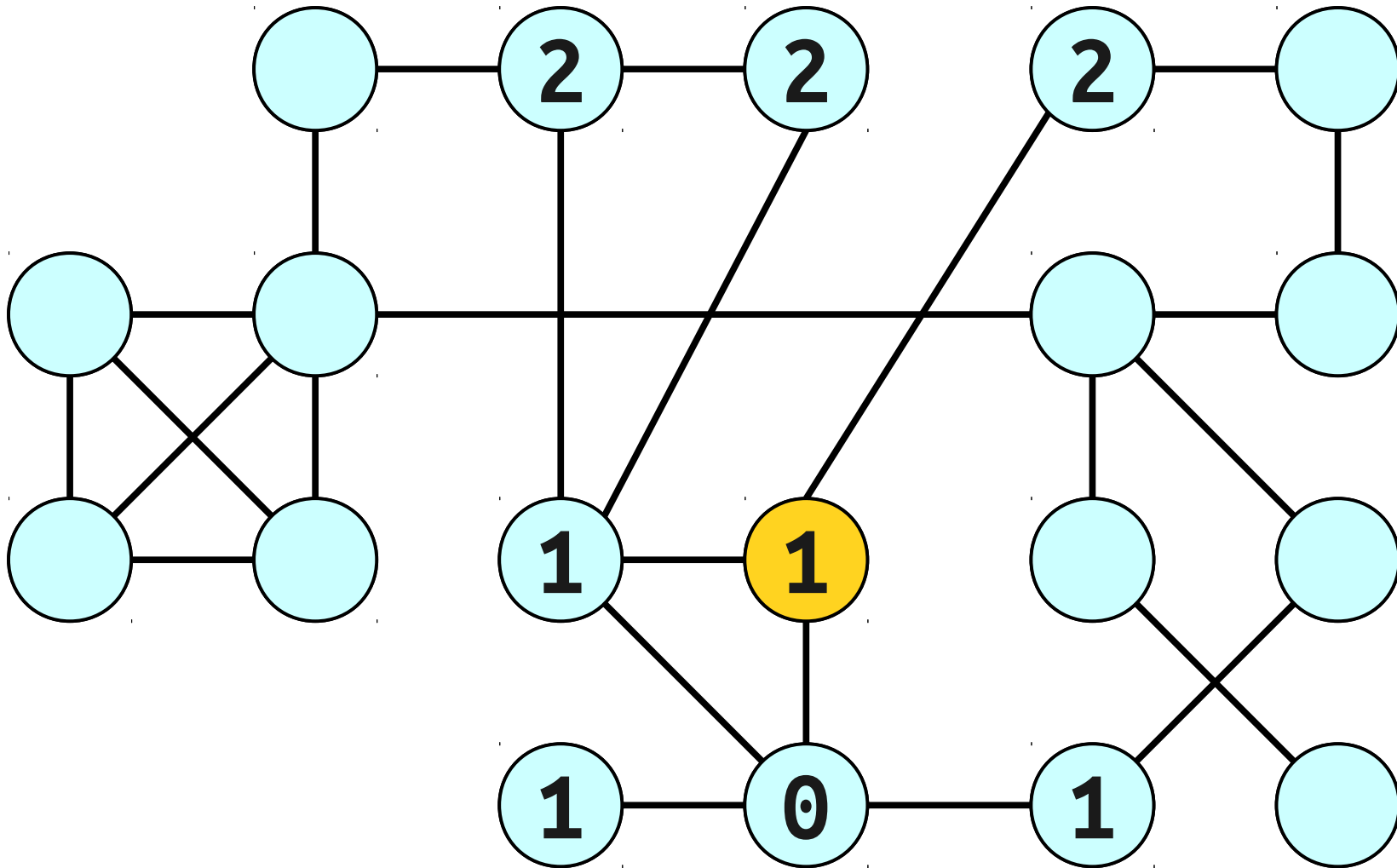
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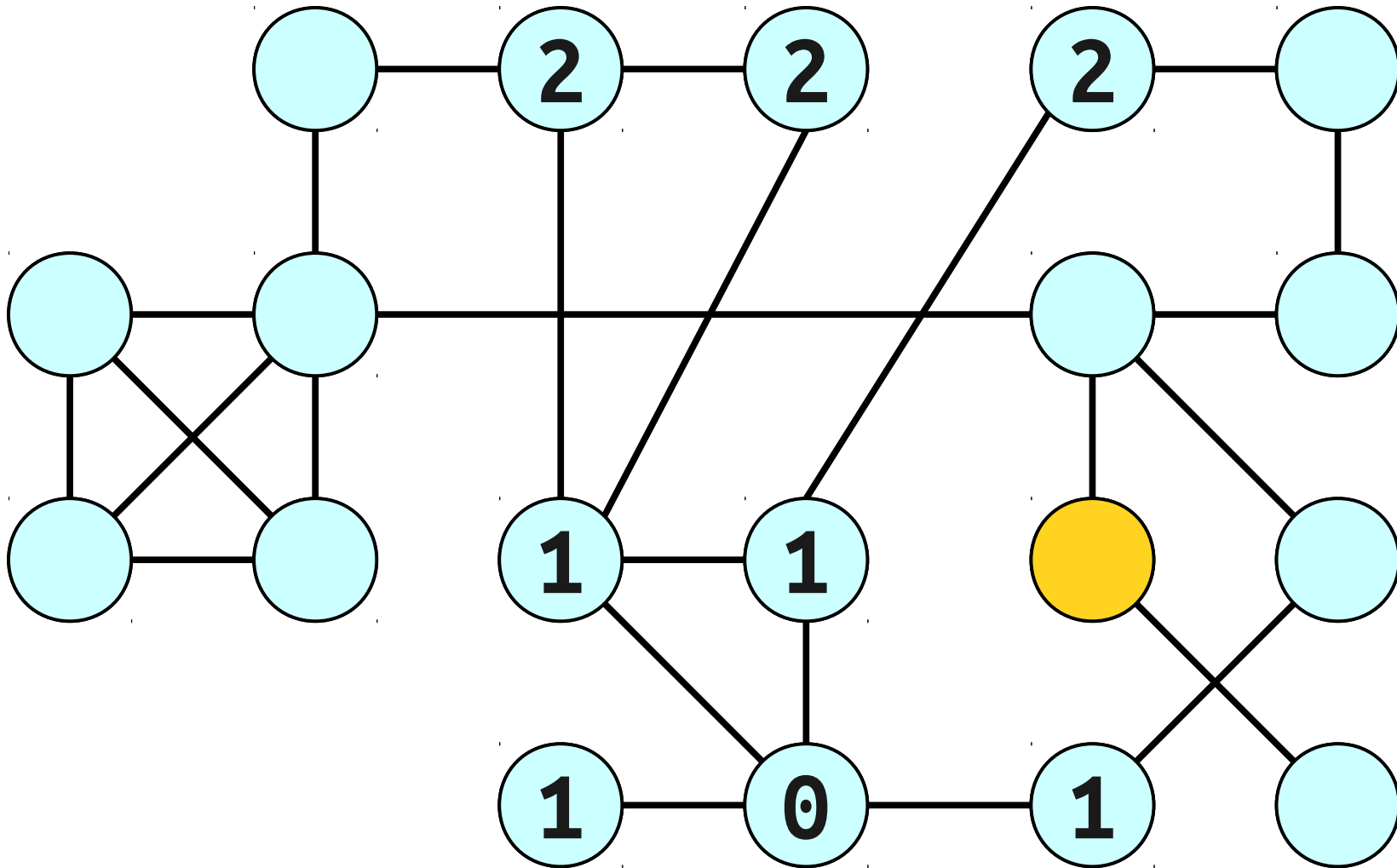
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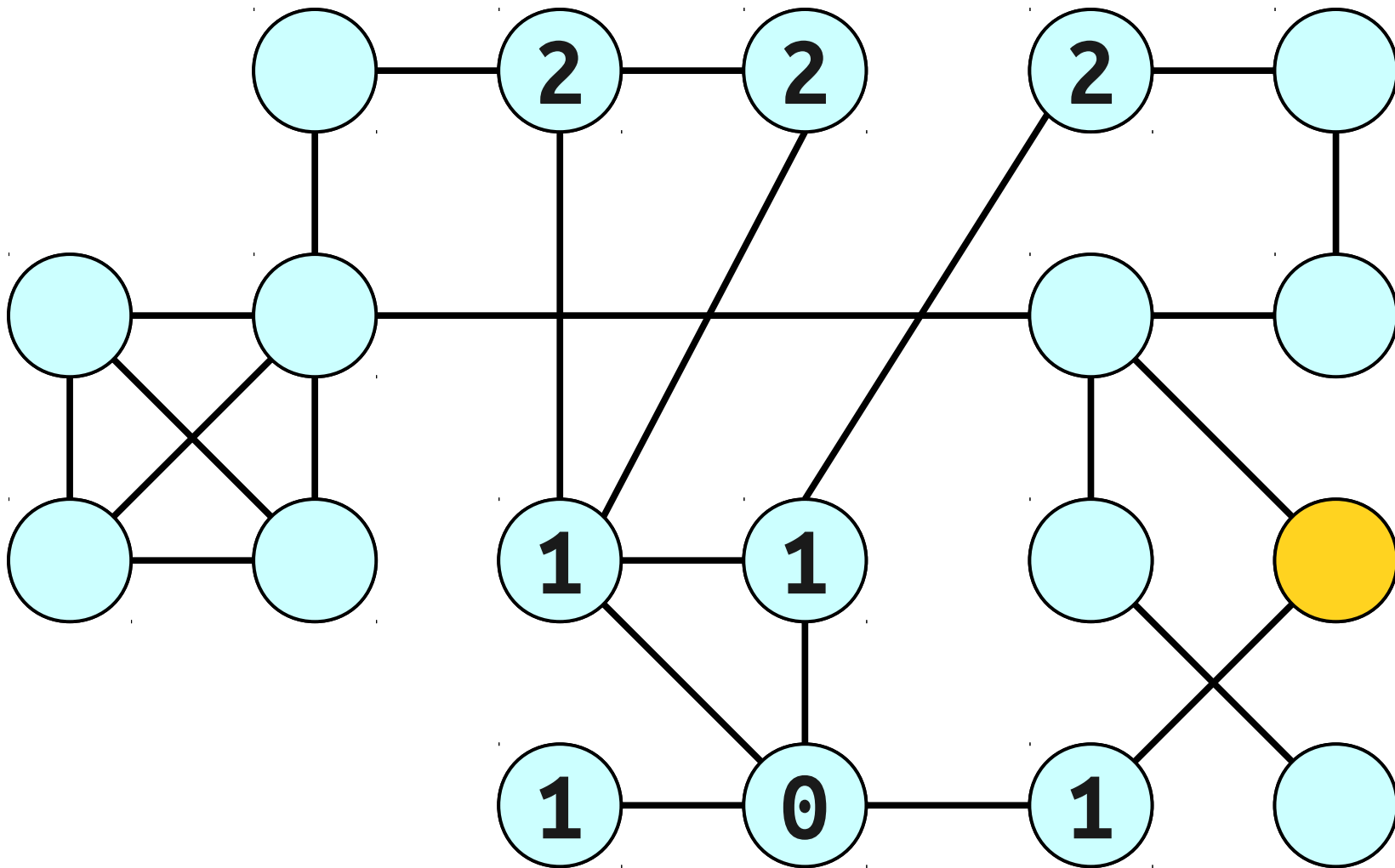
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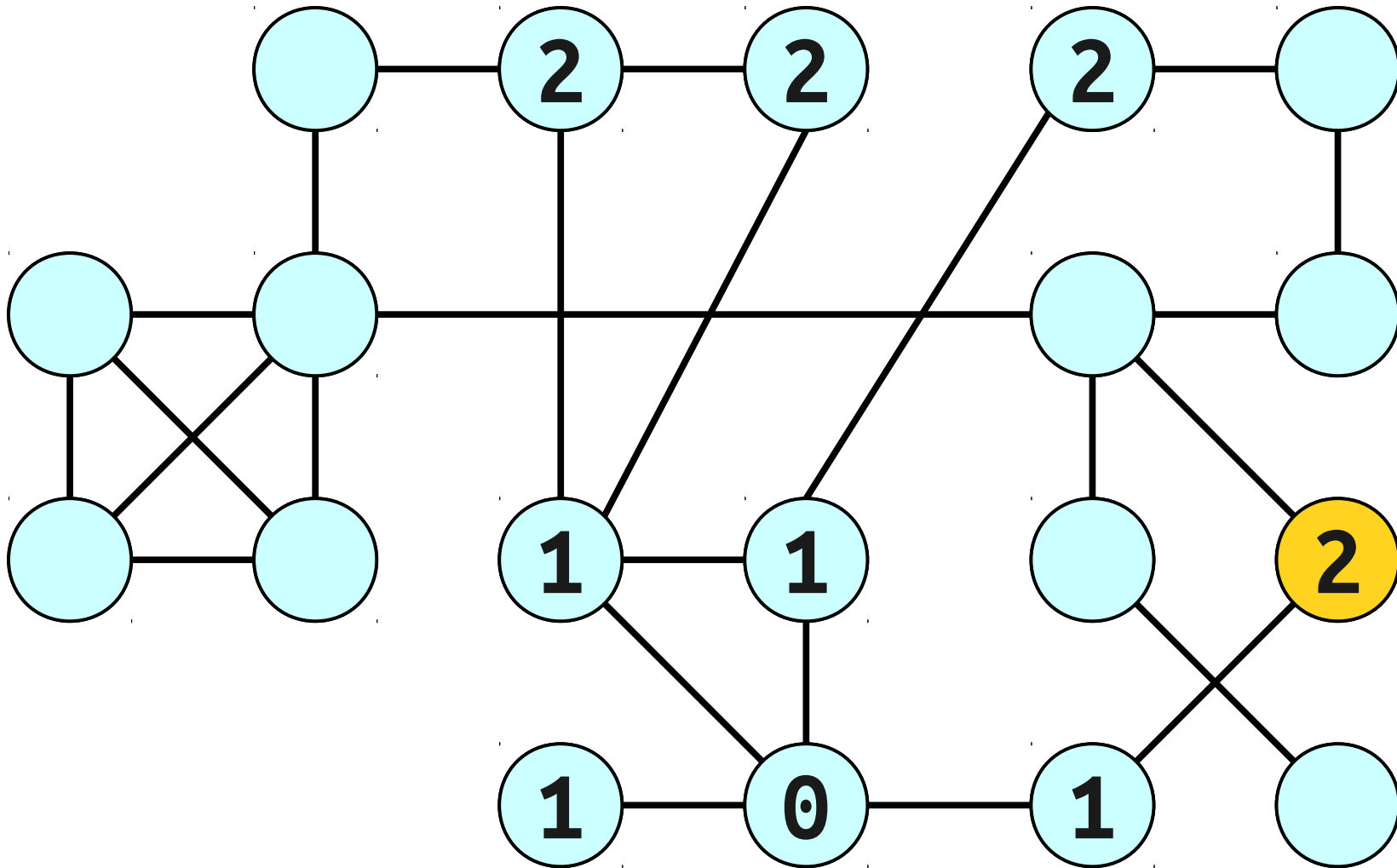


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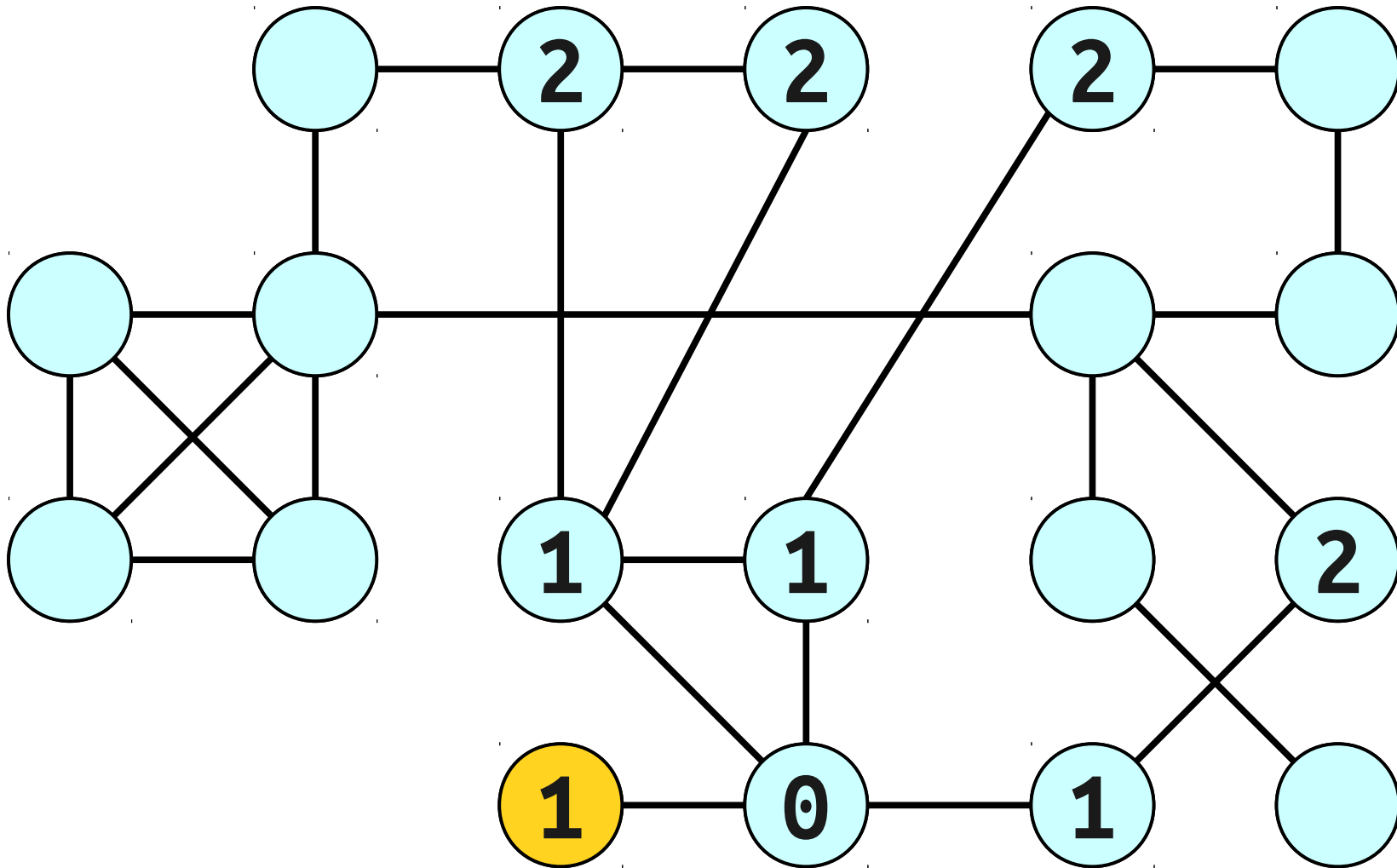




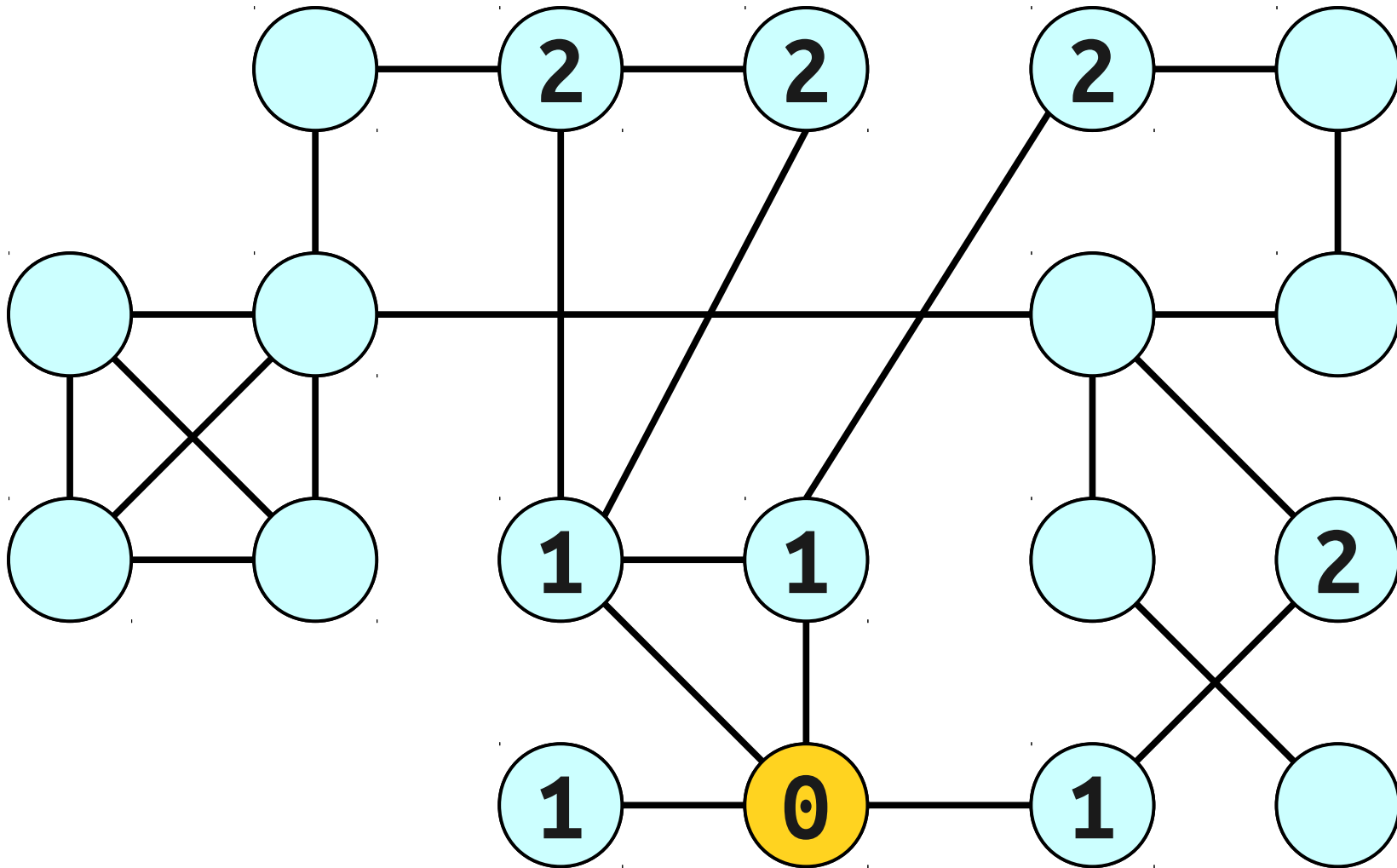
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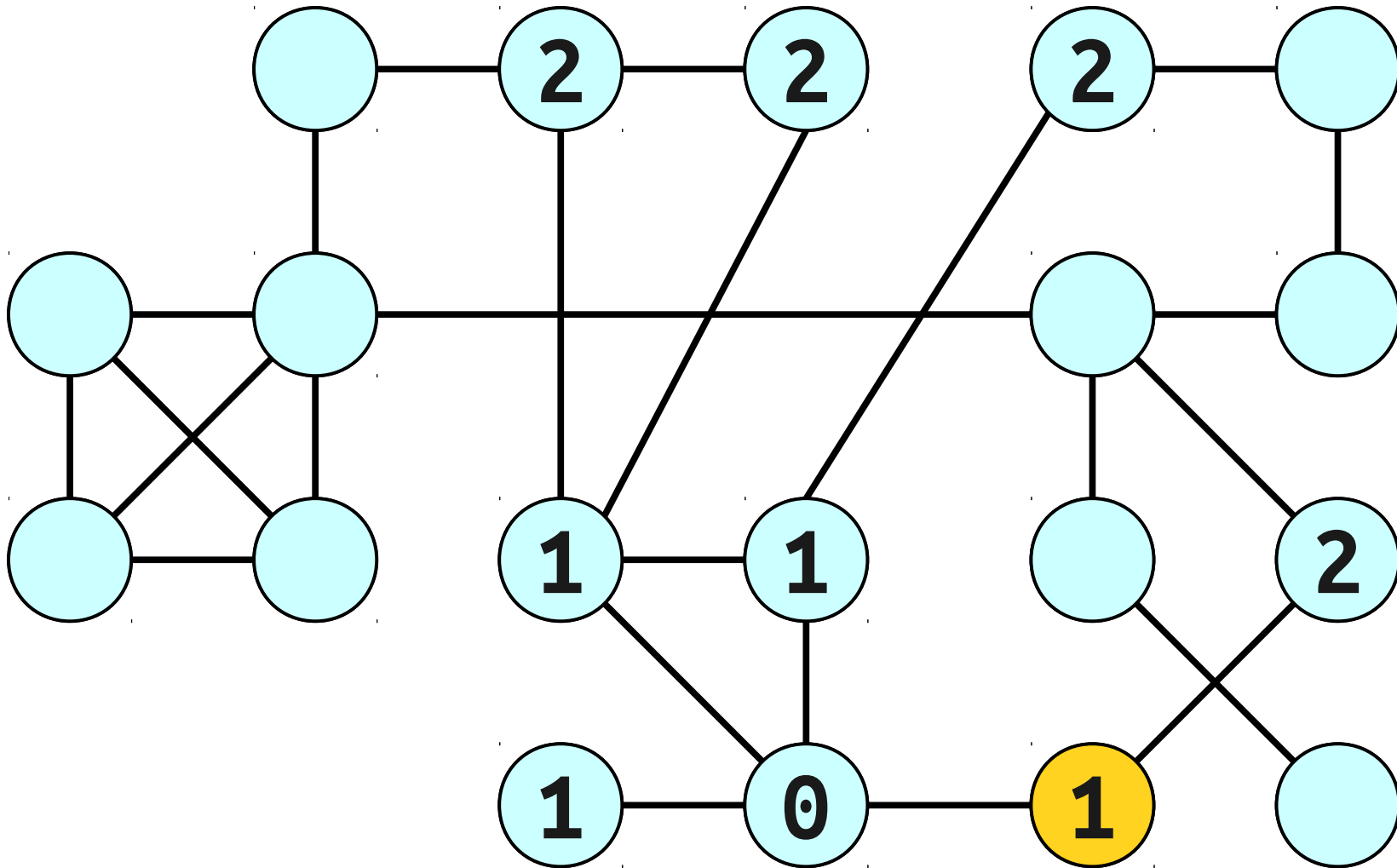
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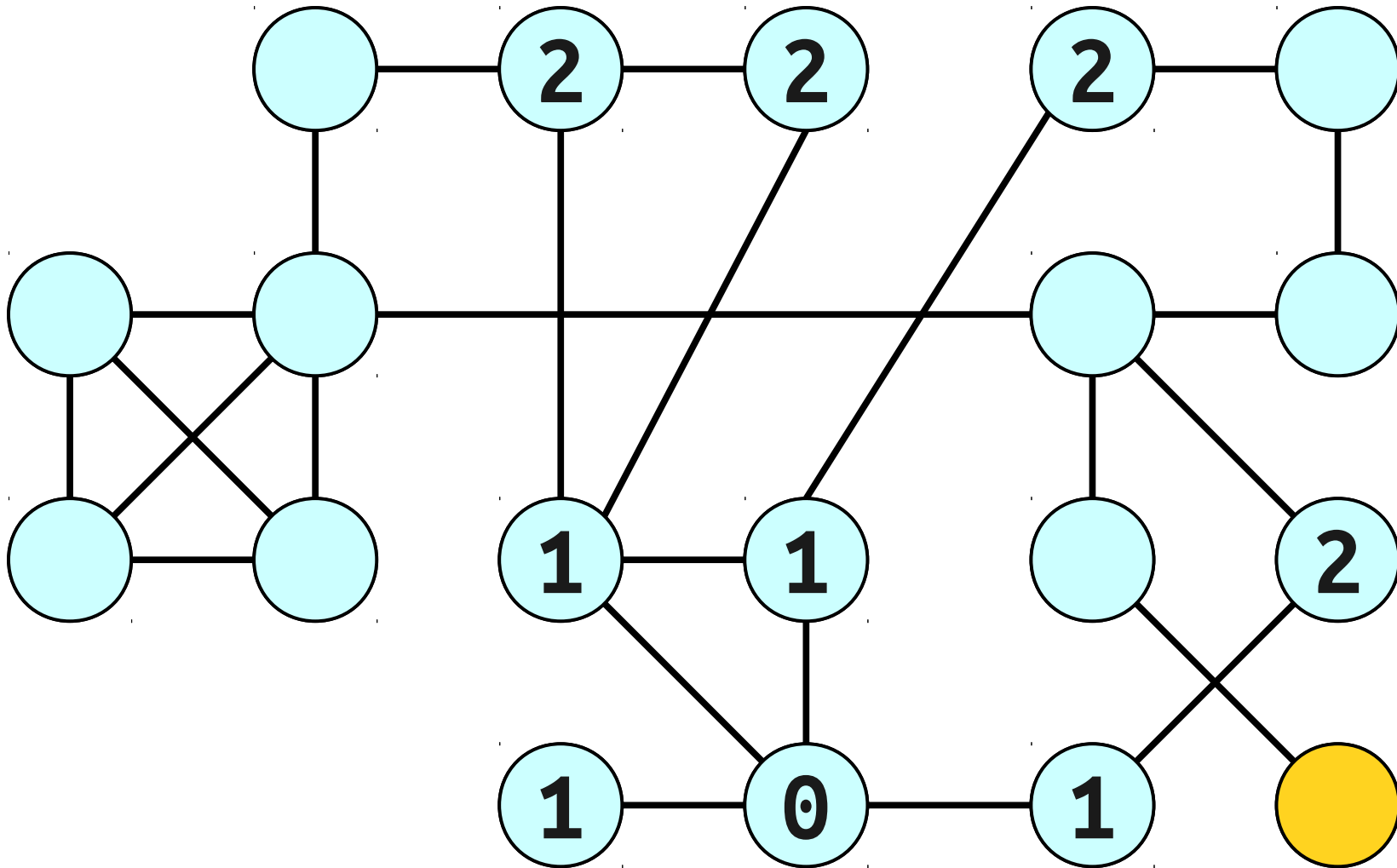
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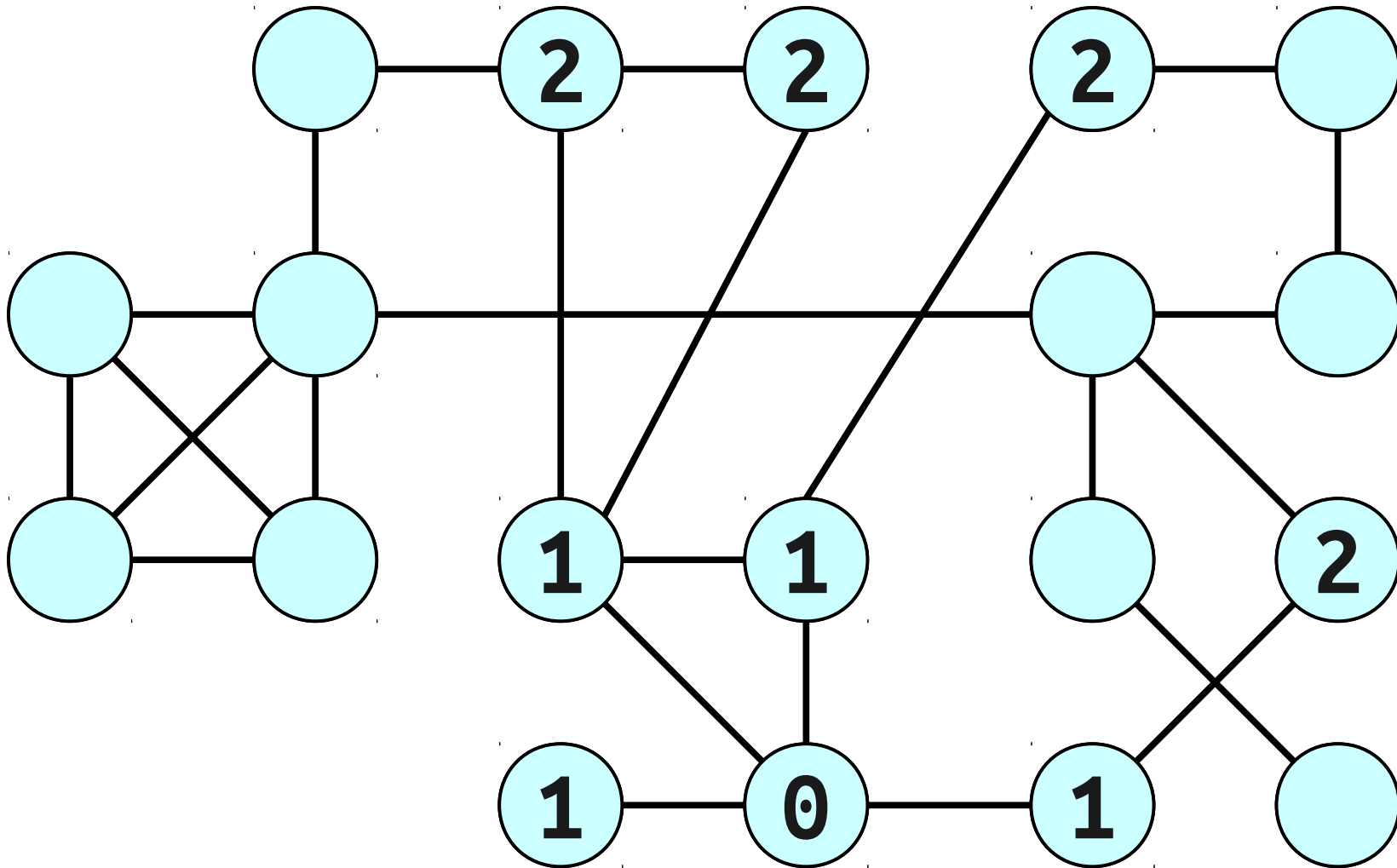
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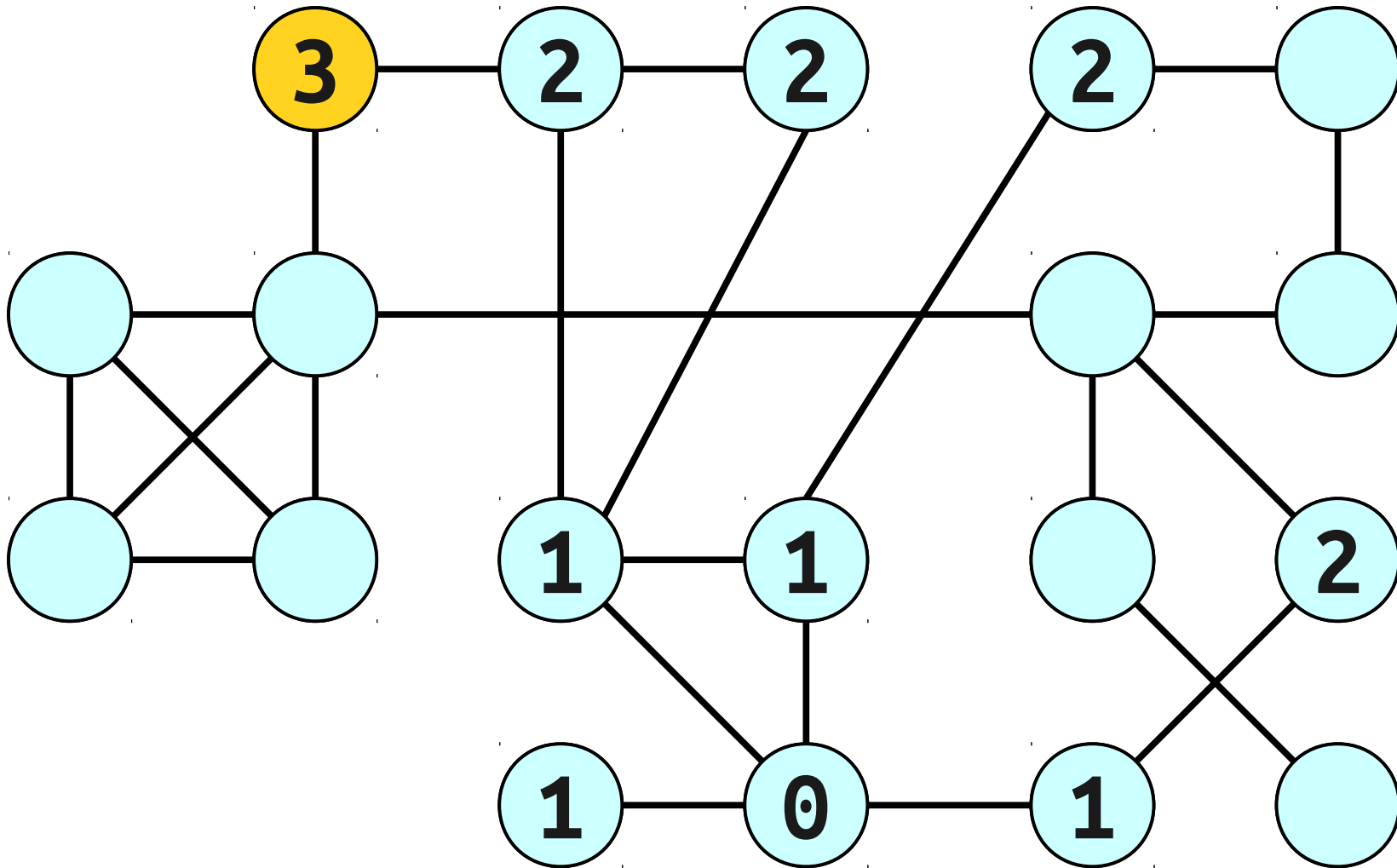


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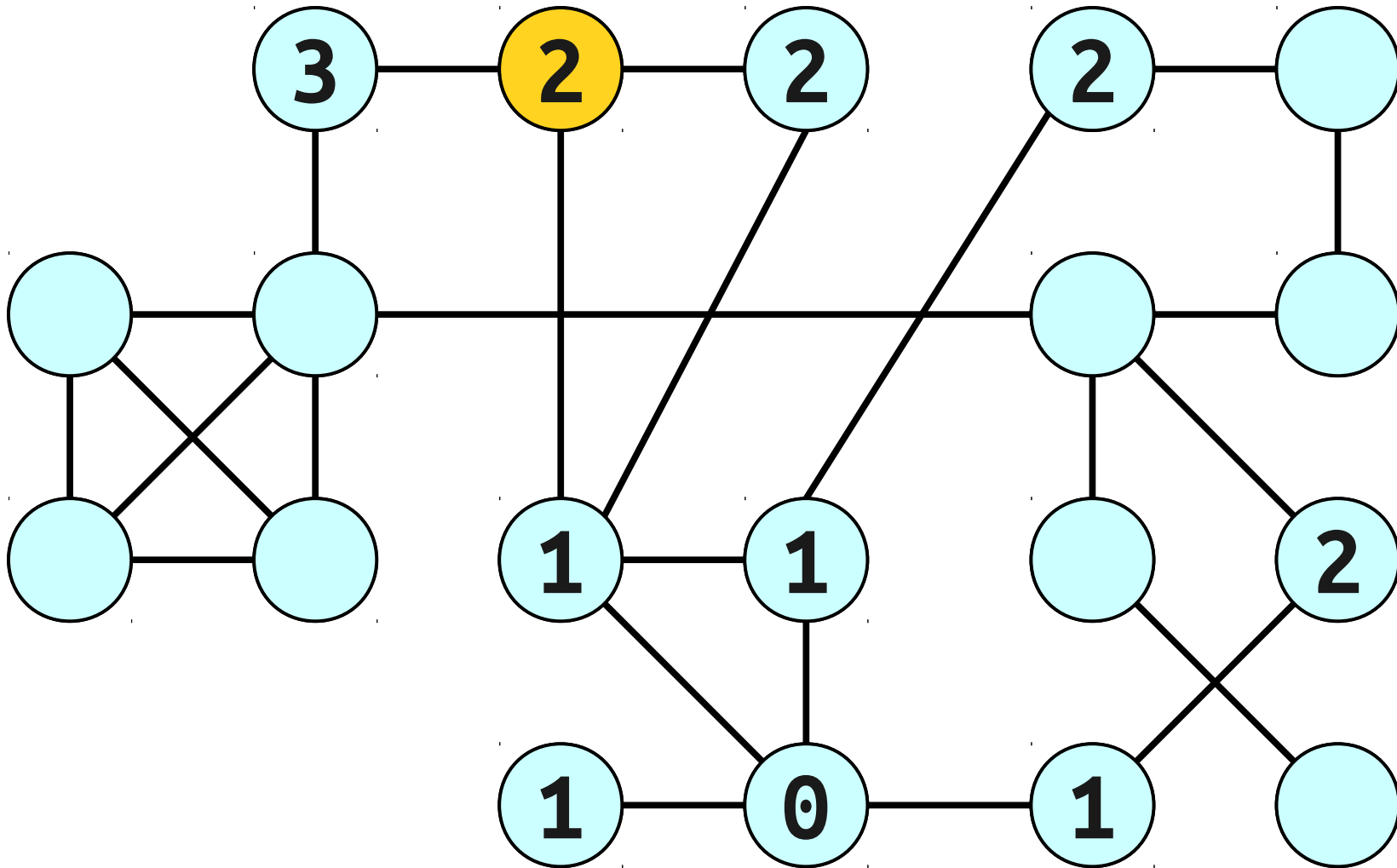


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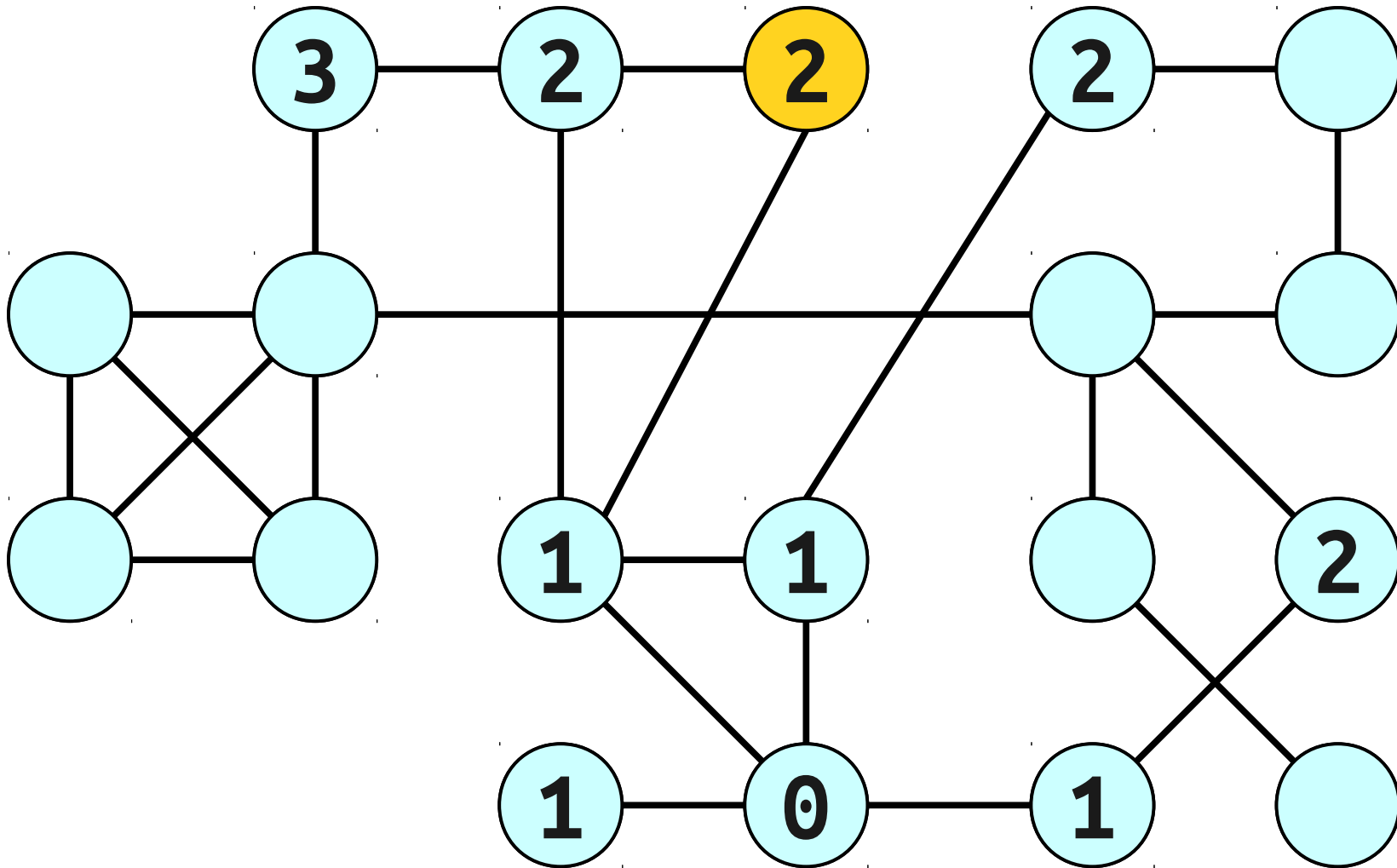




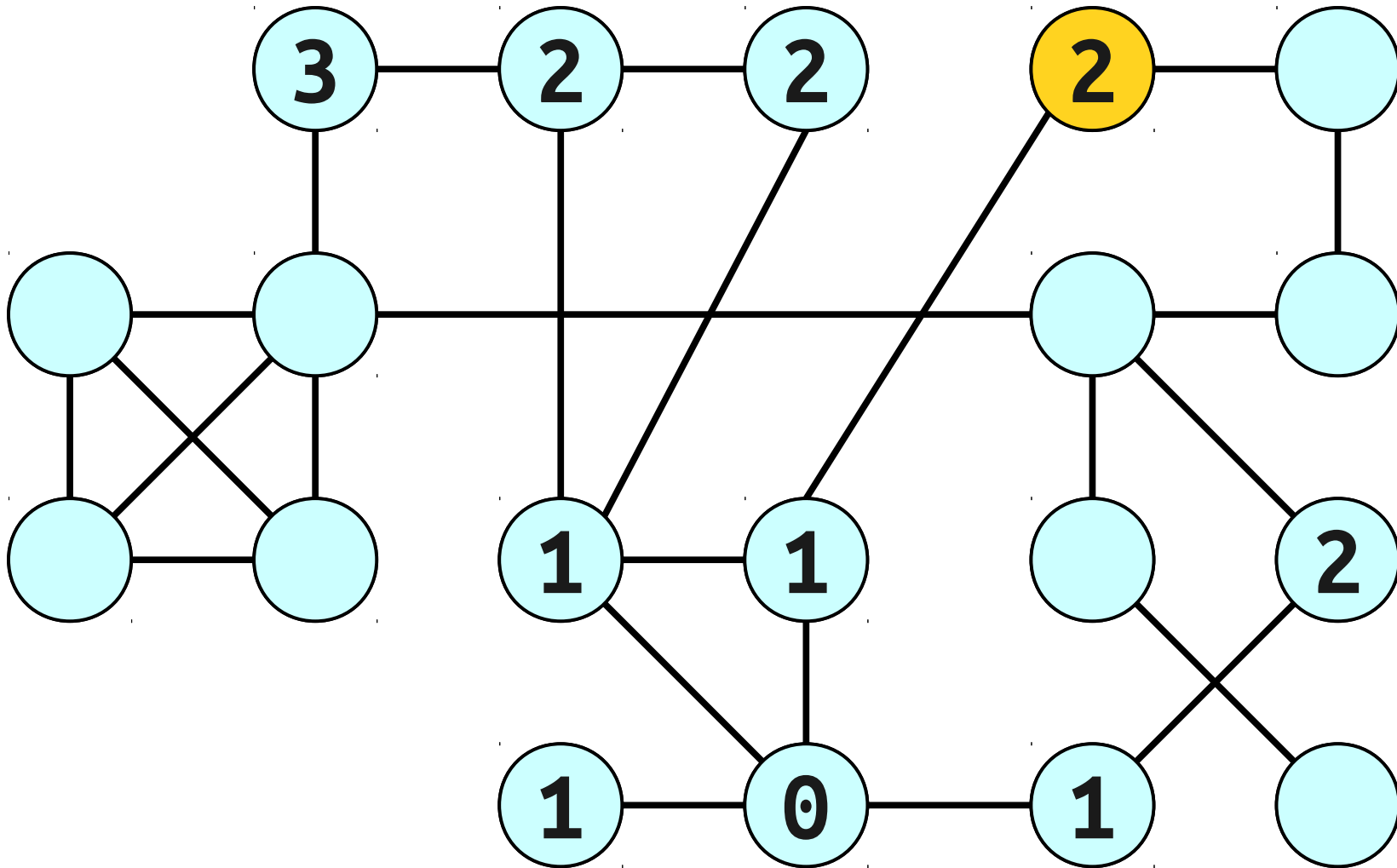
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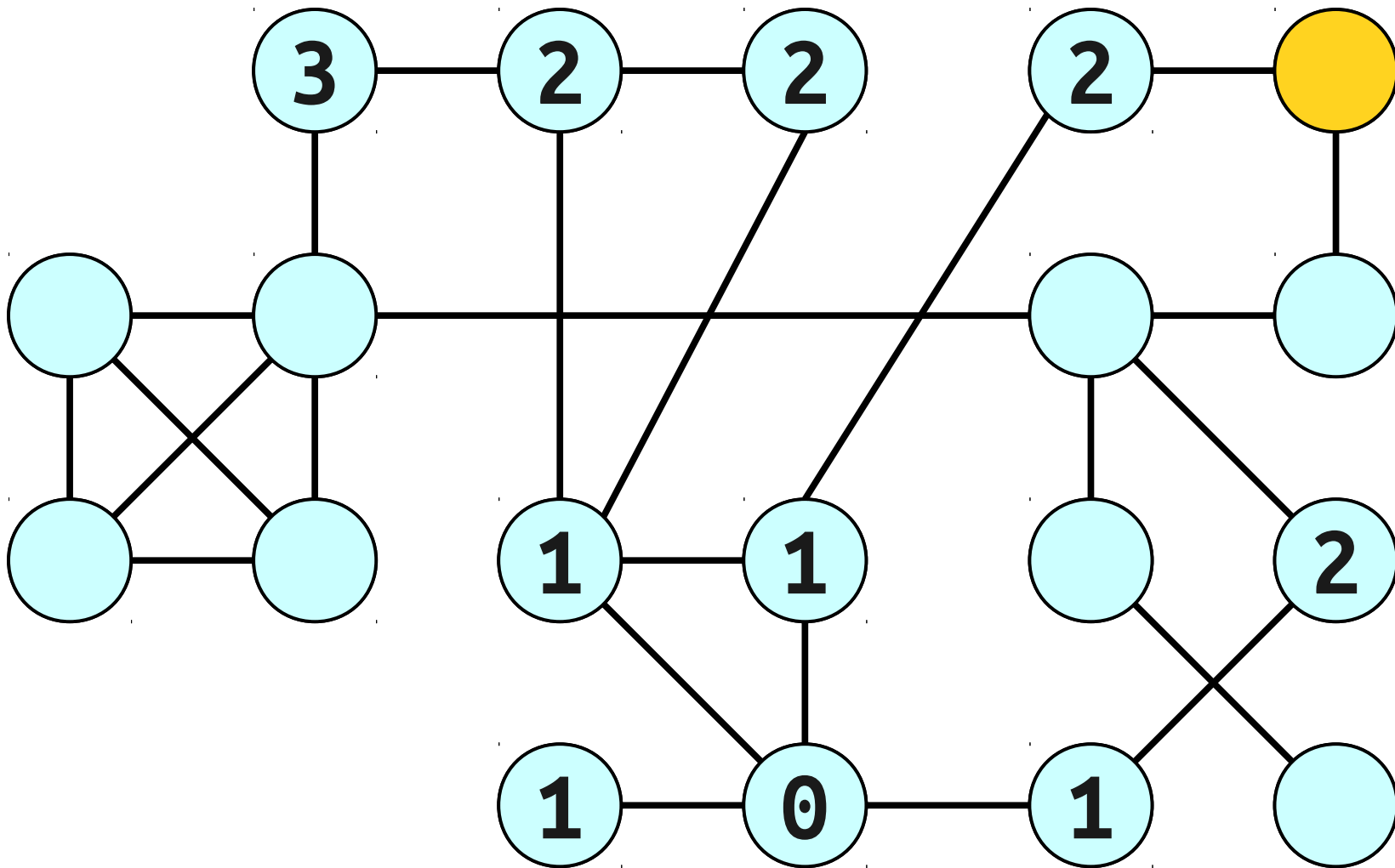
# An Inefficient Algorithm



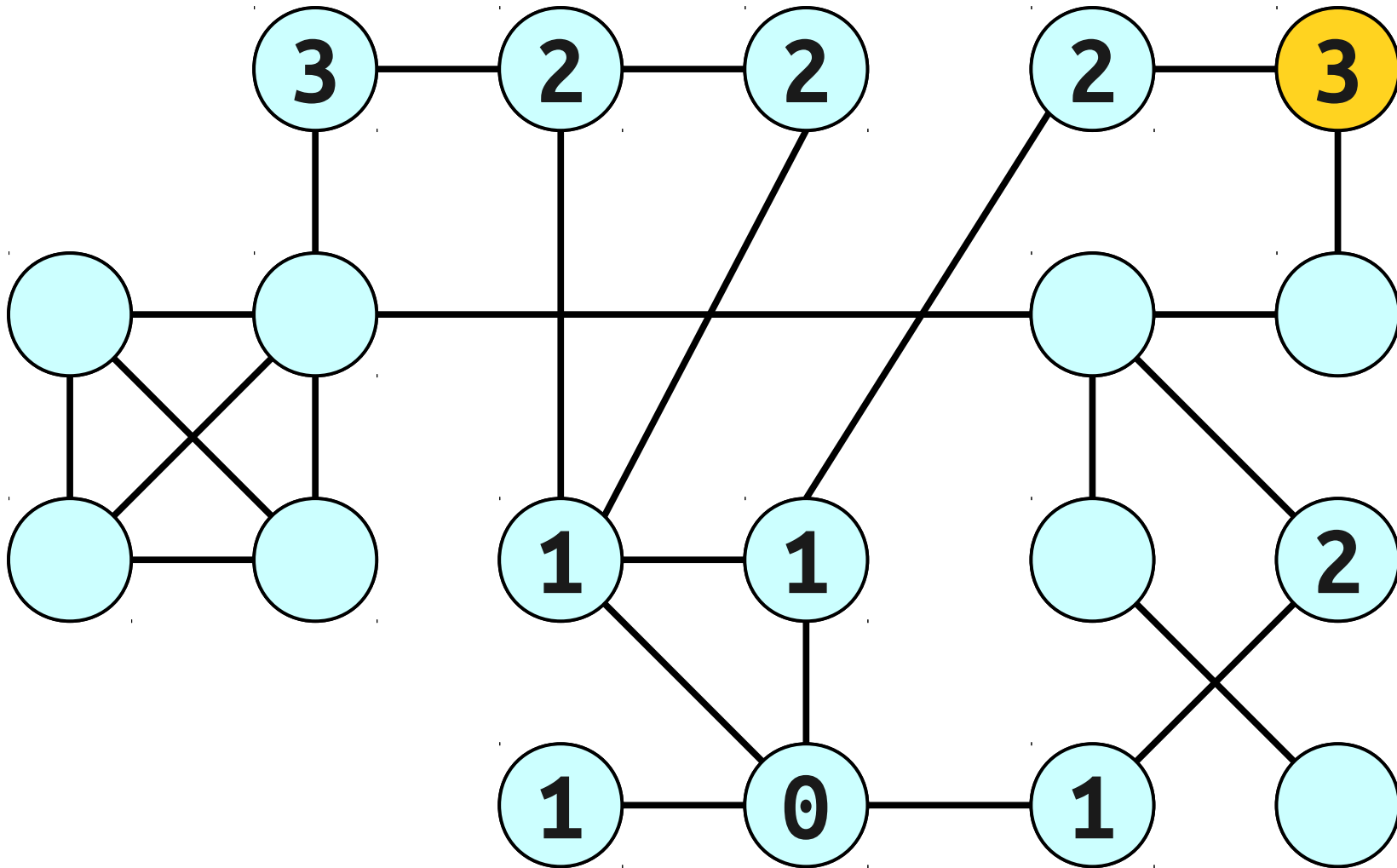
# An Inefficient Algorithm



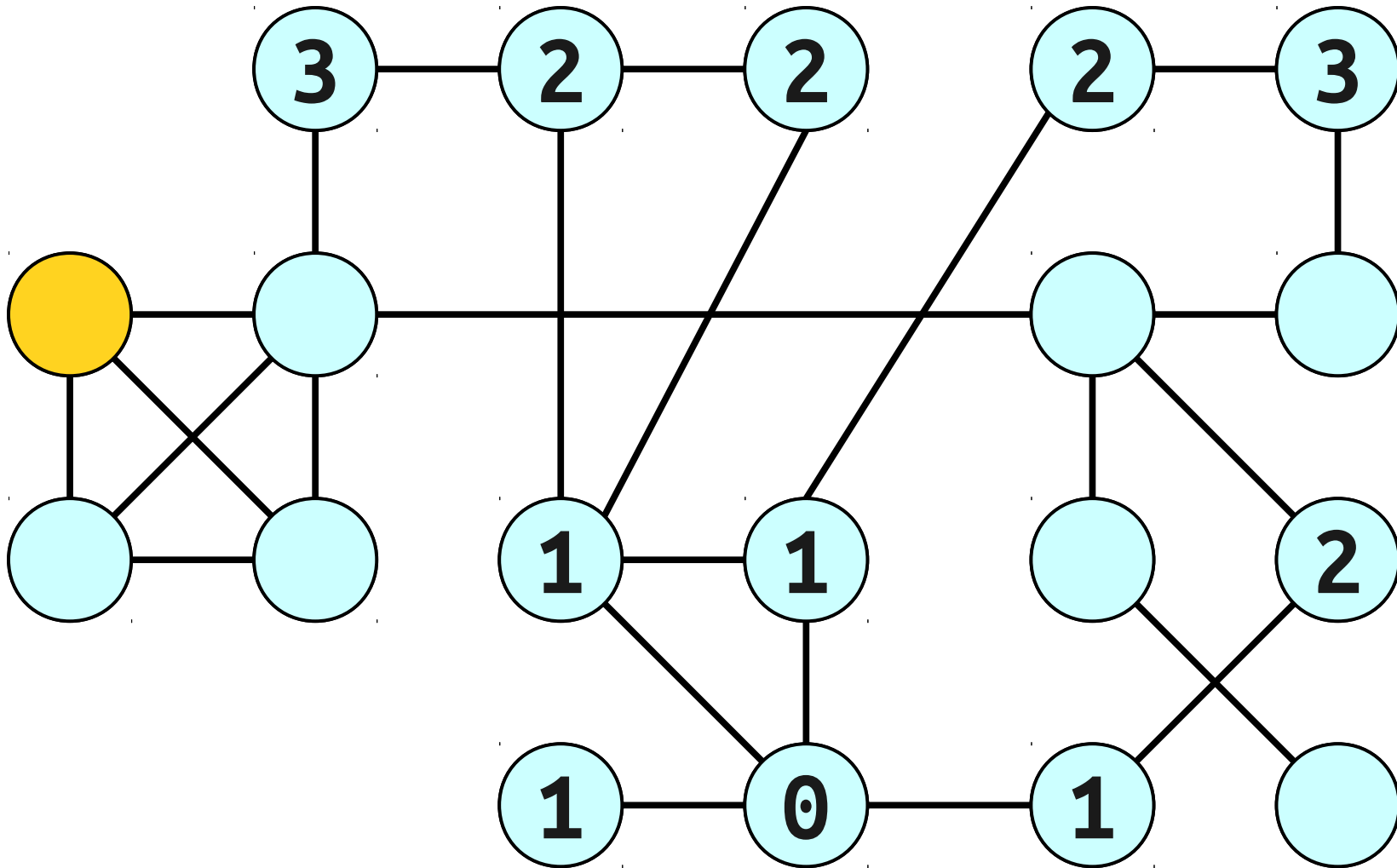
# An Inefficient Algorithm



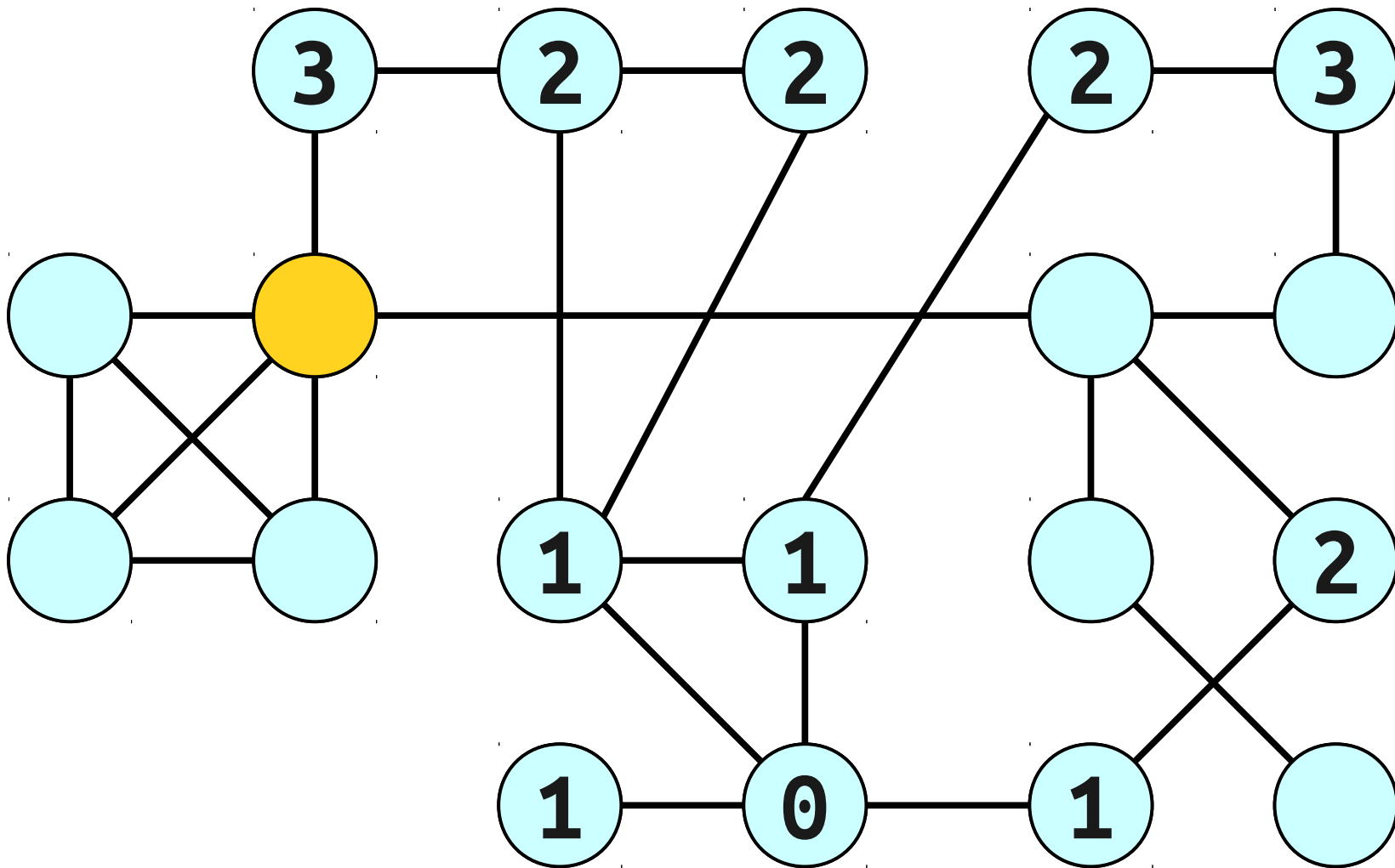
# An Inefficient Algorithm



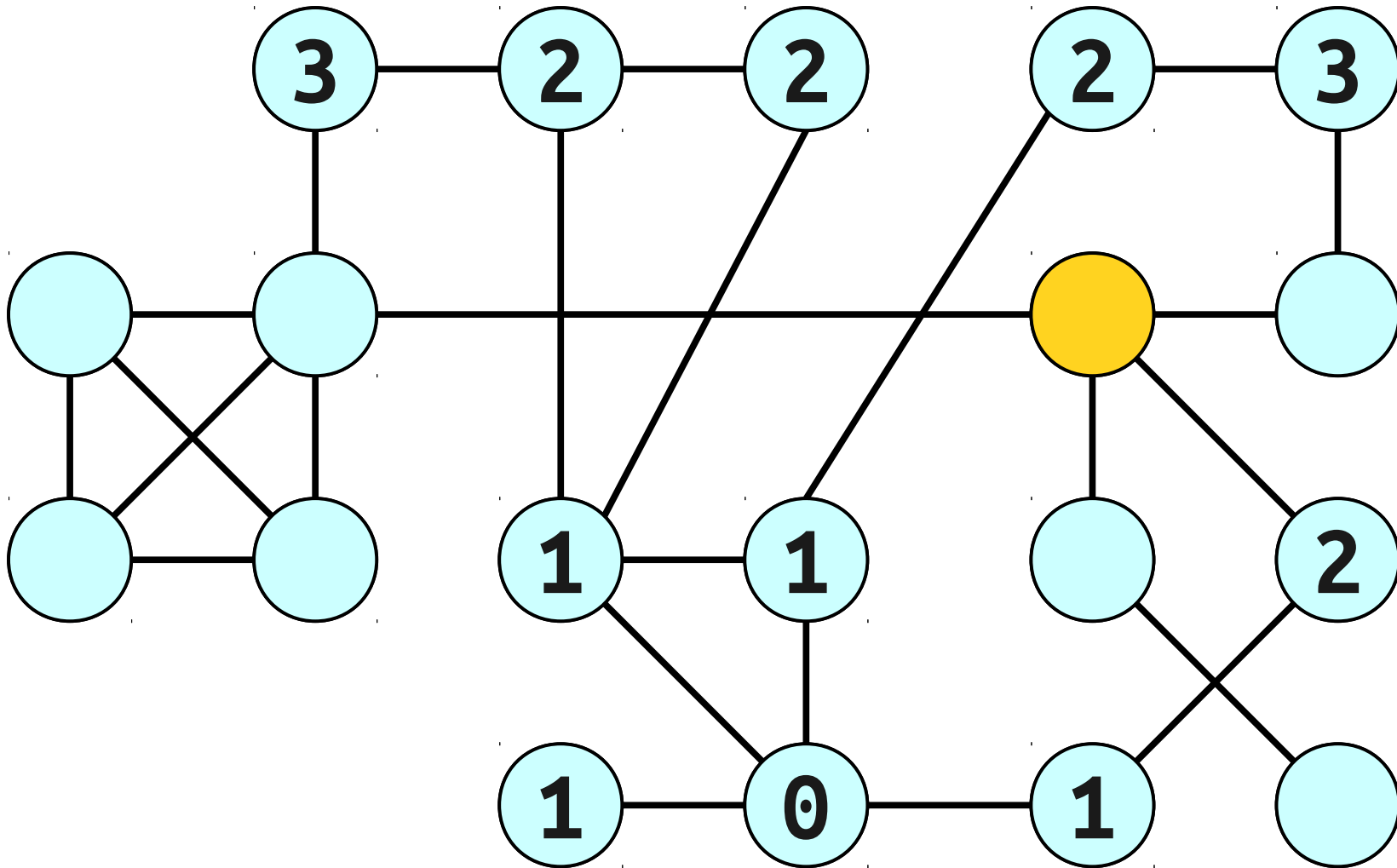
# An Inefficient Algorithm



# An Inefficient Algorithm

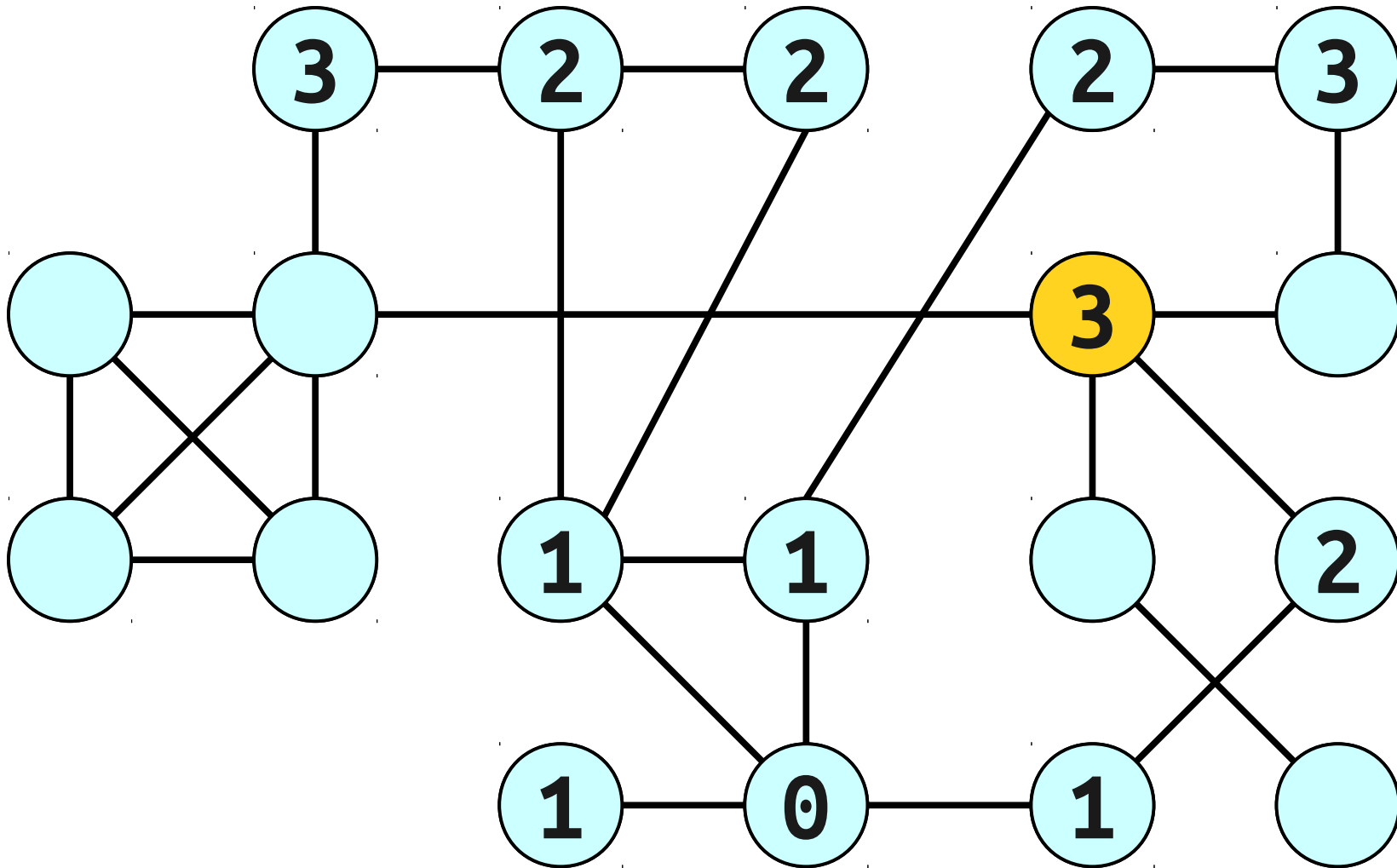


# An Inefficient Algorithm

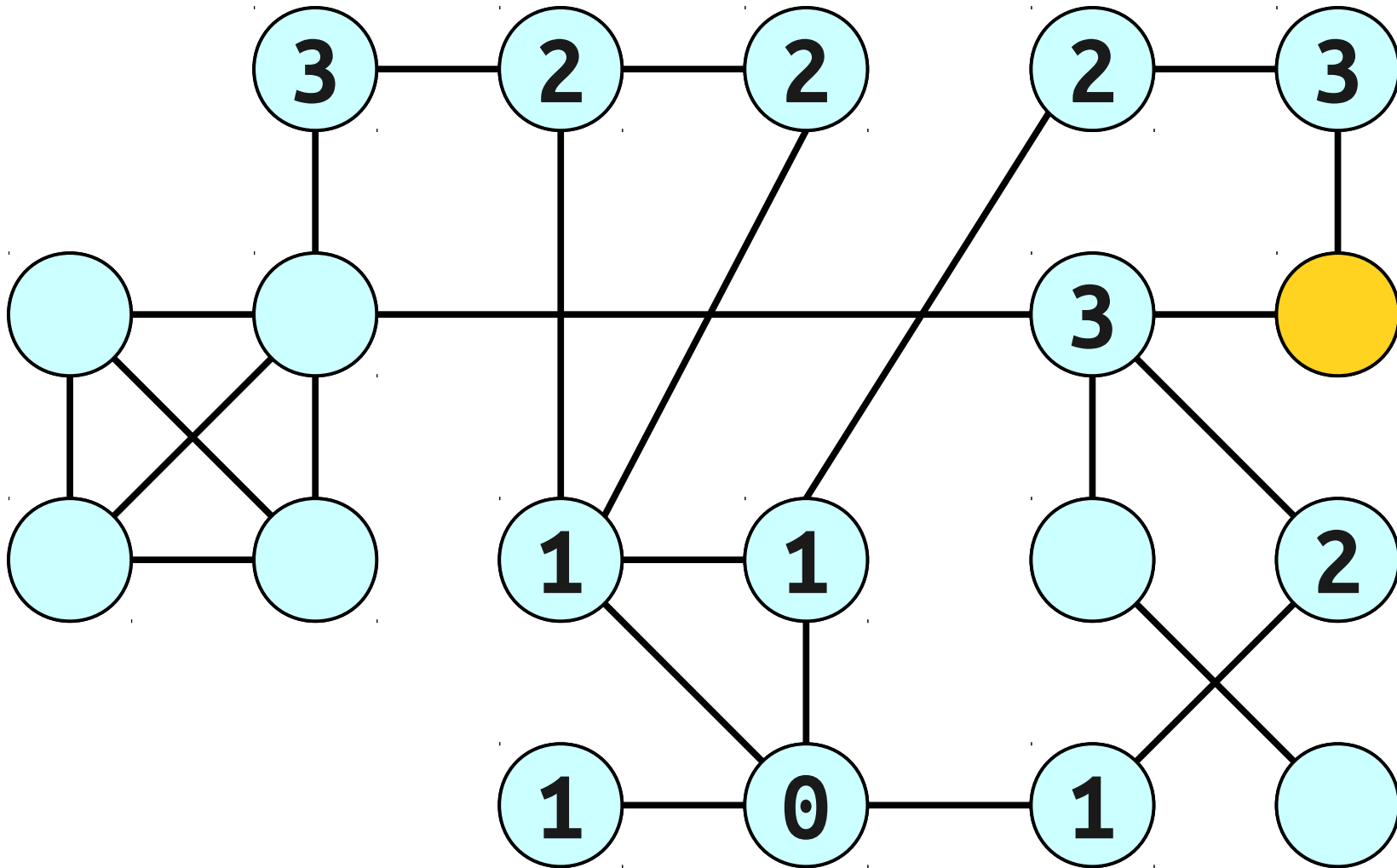




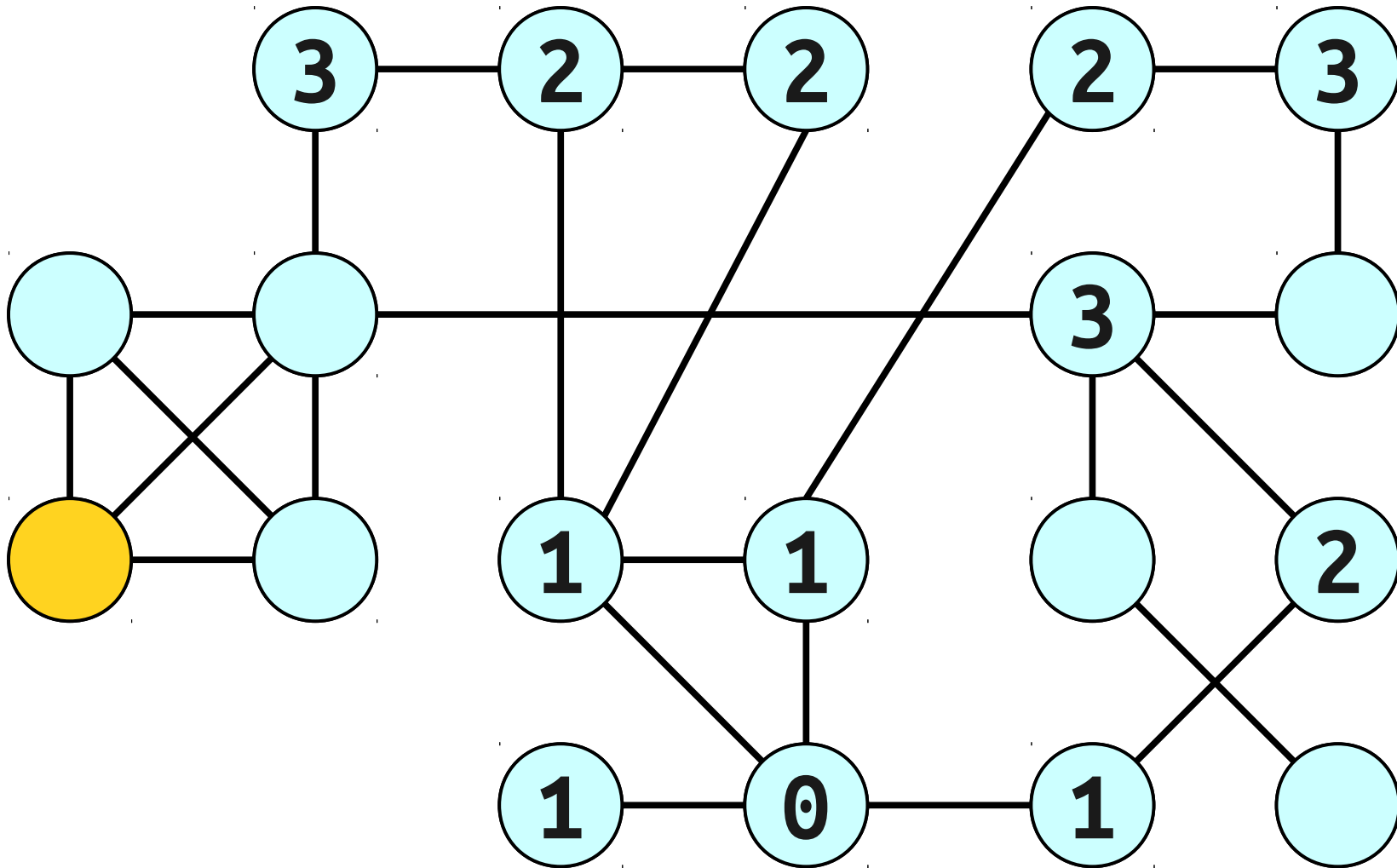
# An Inefficient Algorithm



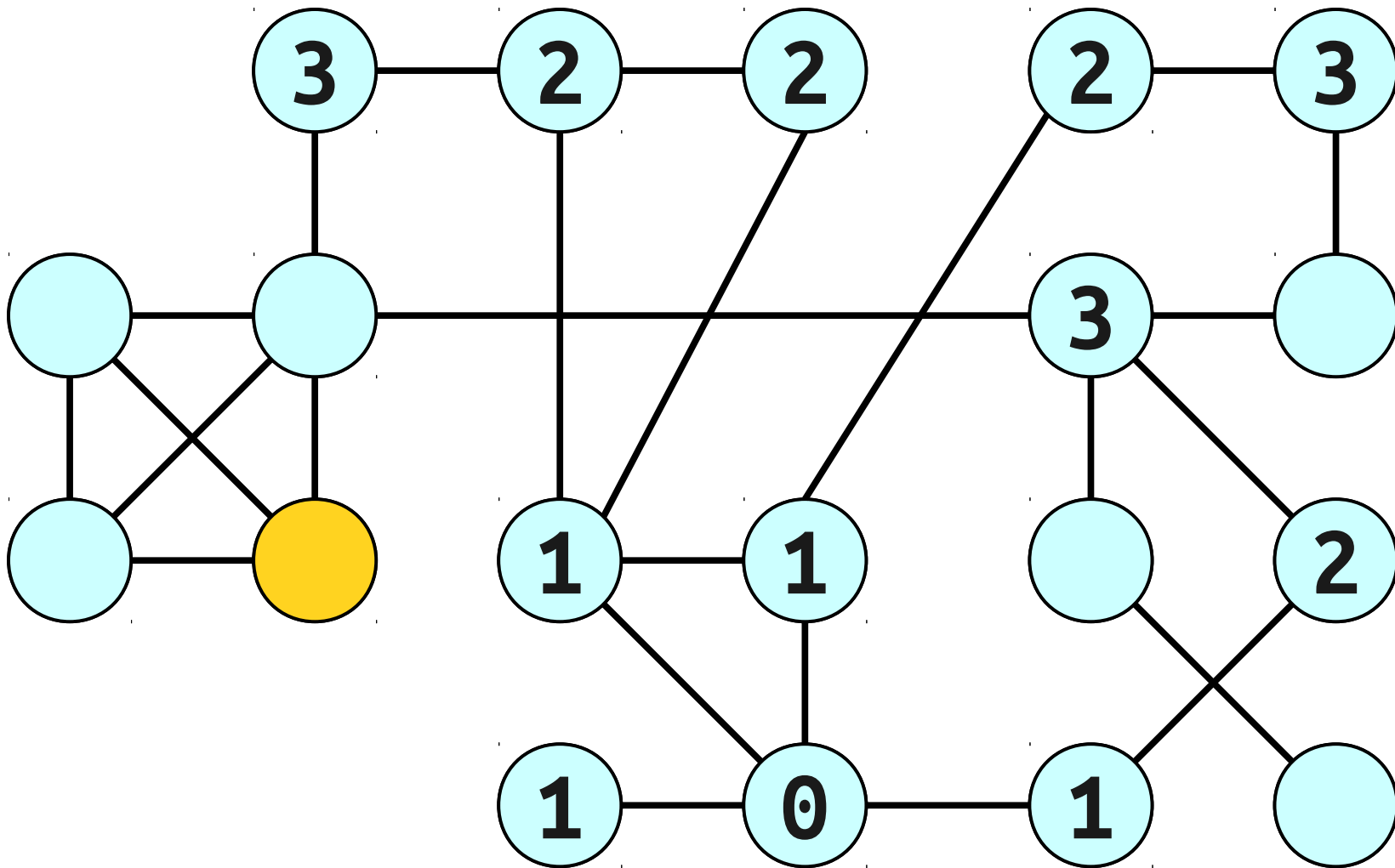
# An Inefficient Algorithm



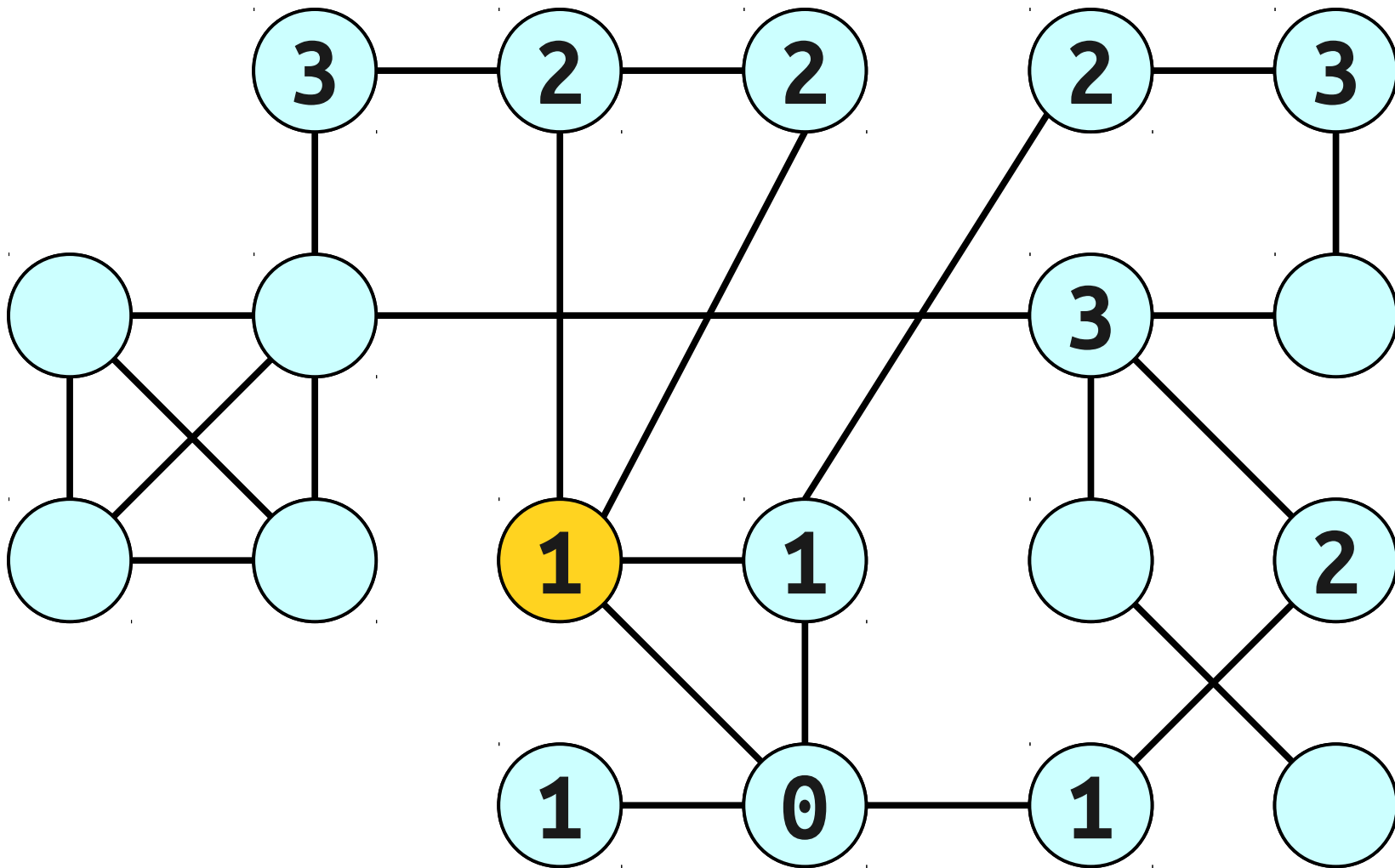
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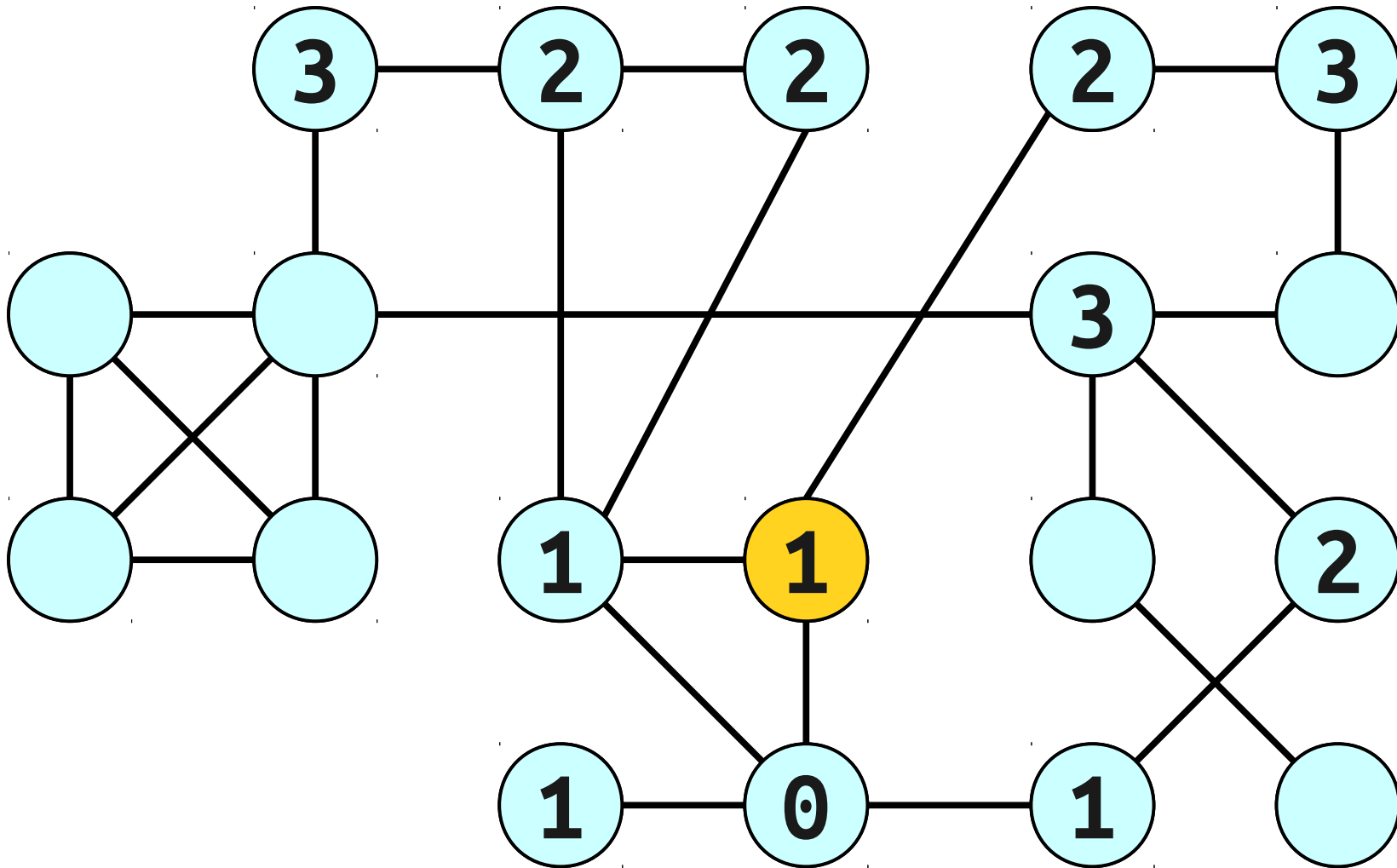
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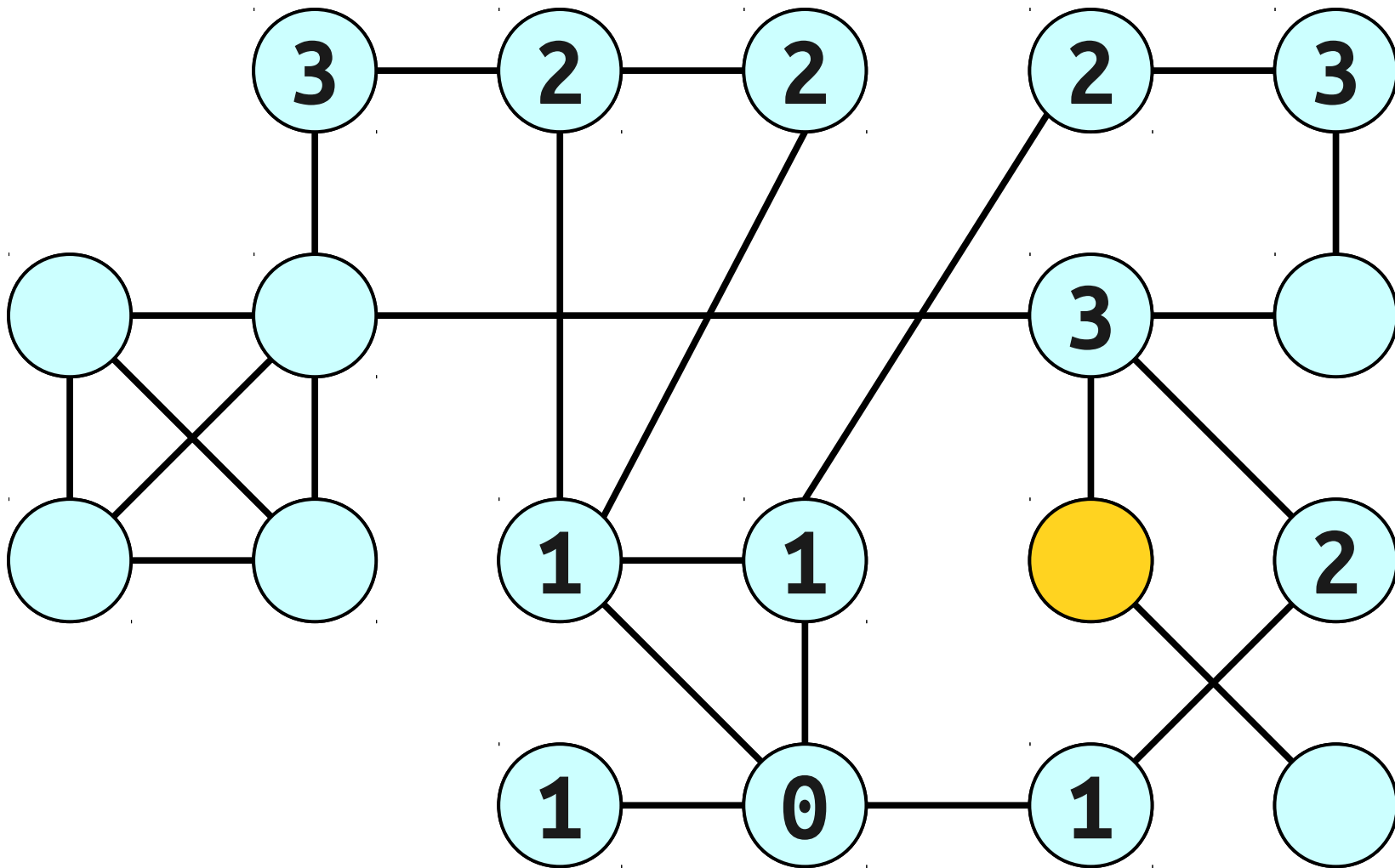
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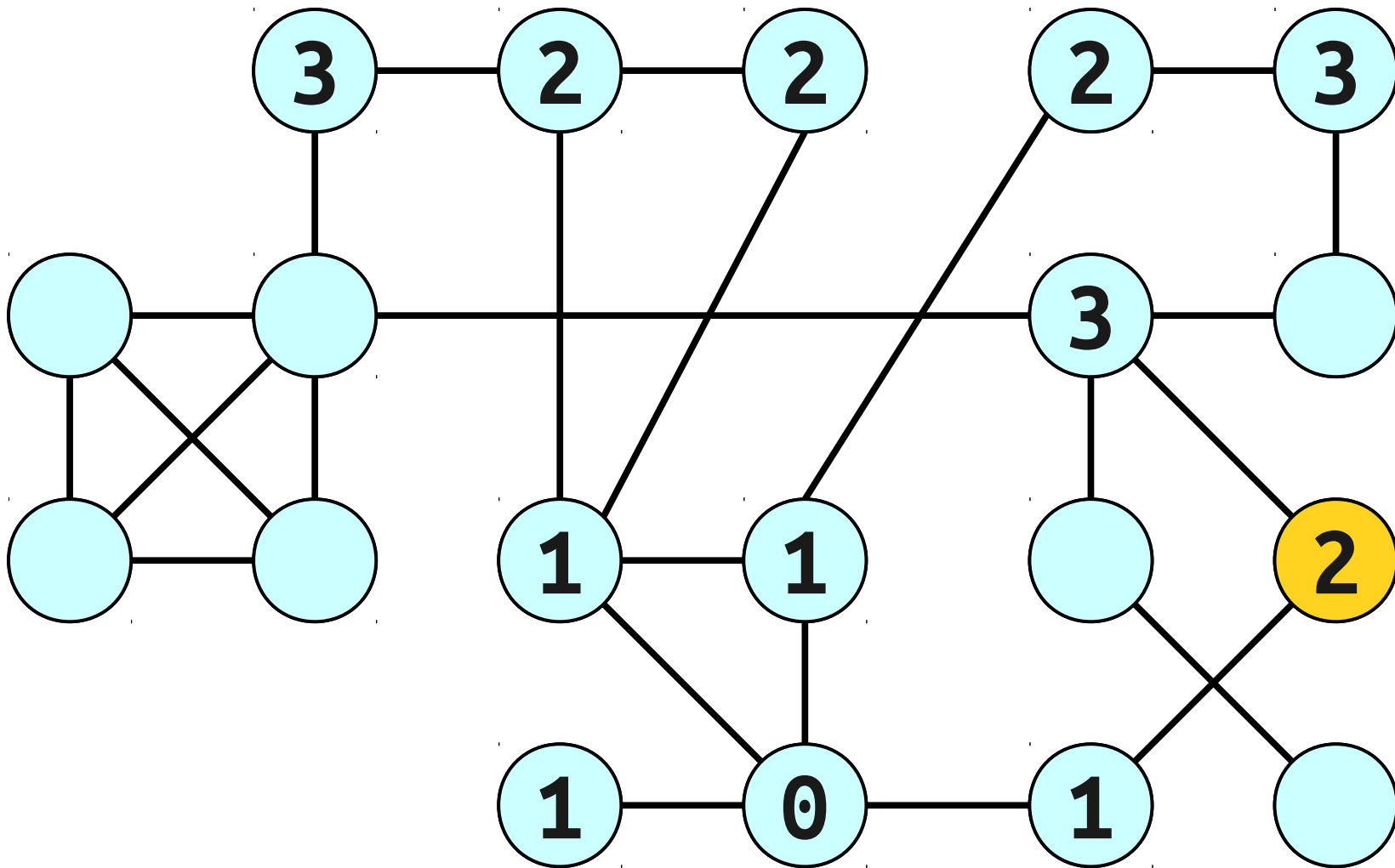
# An Inefficient Algorithm



# An Inefficient Algorithm

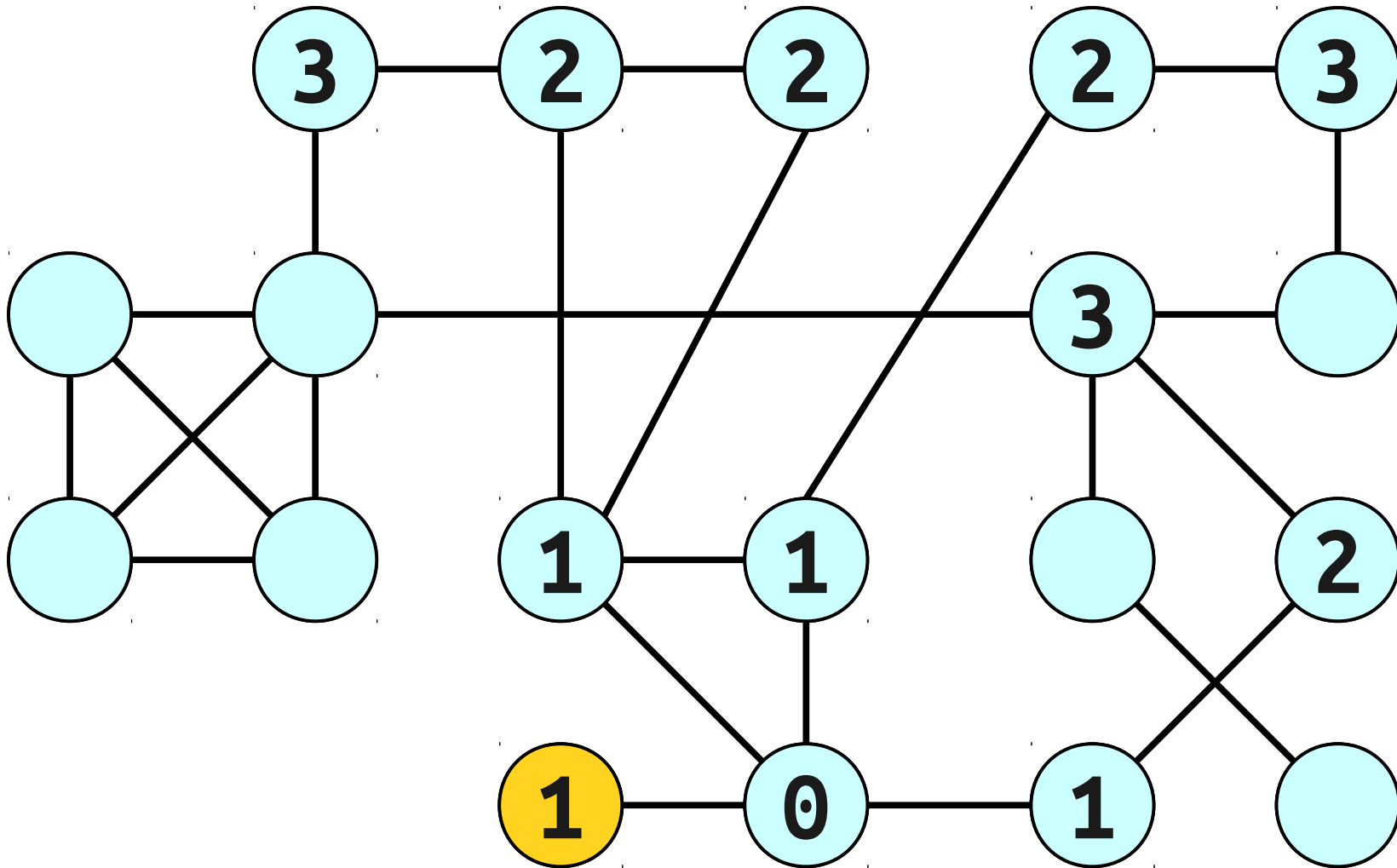


# An Inefficient Algorithm

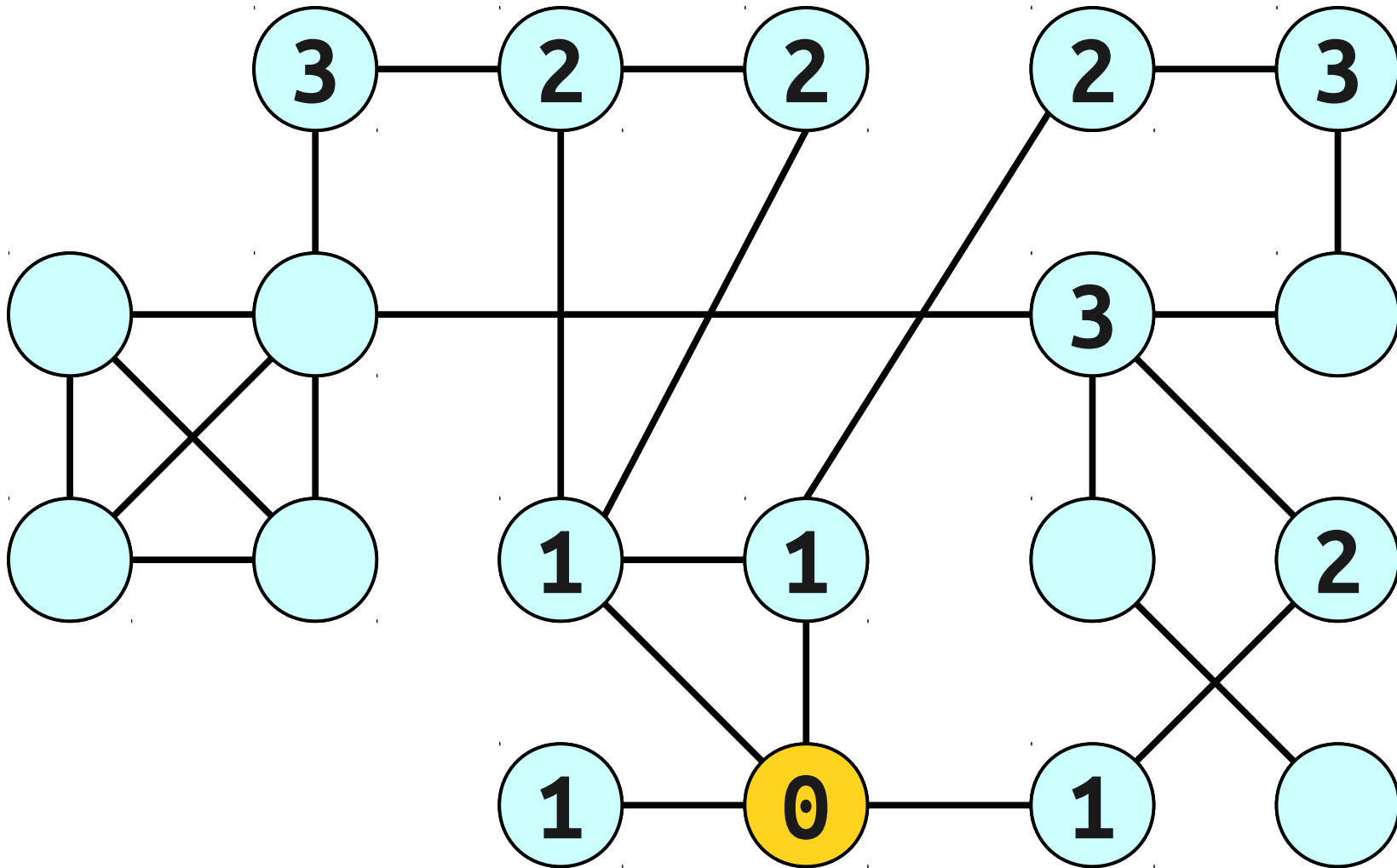




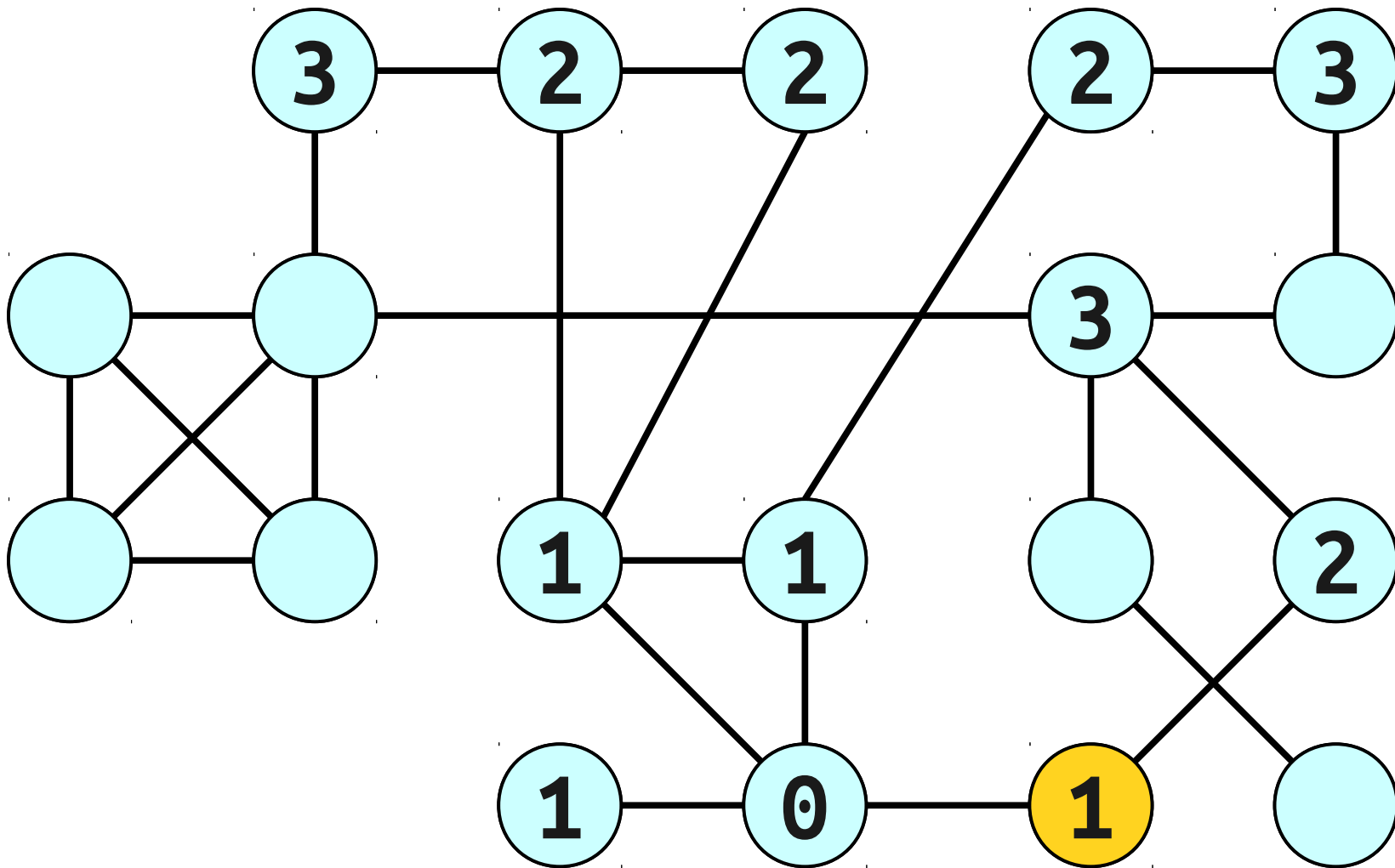
# An Inefficient Algorithm



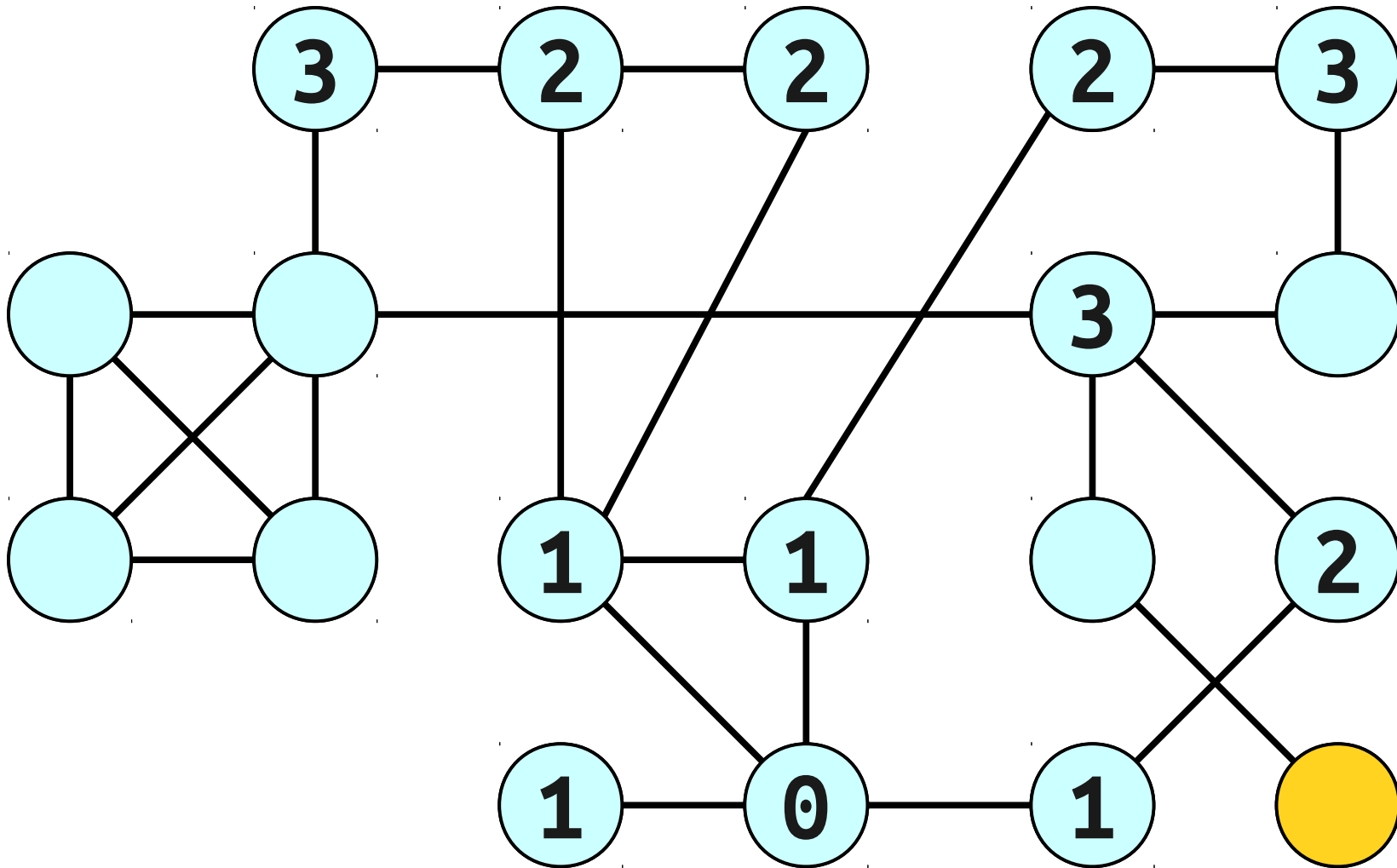
# An Inefficient Algorithm



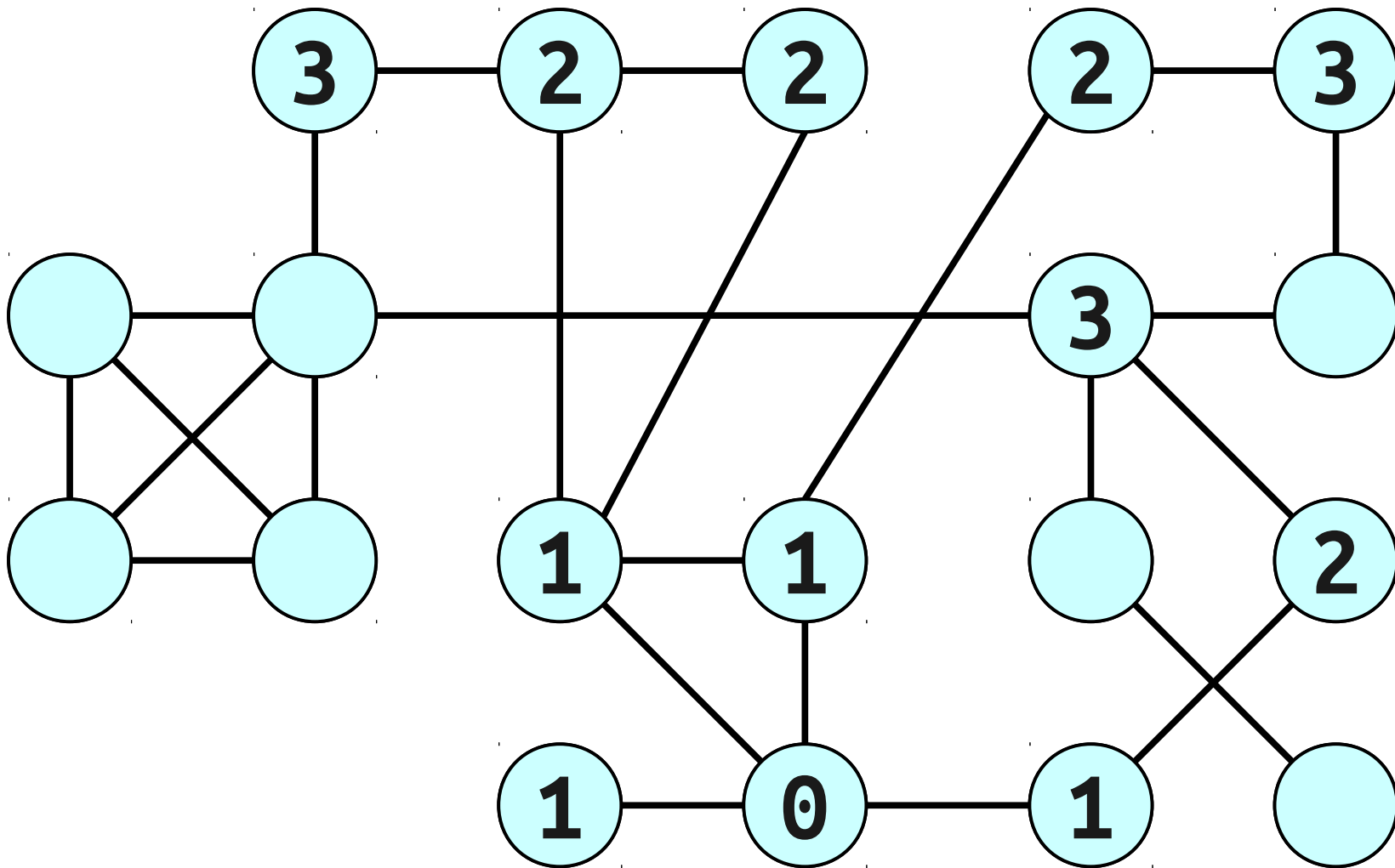
# An Inefficient Algorithm



# An Inefficient Algorithm

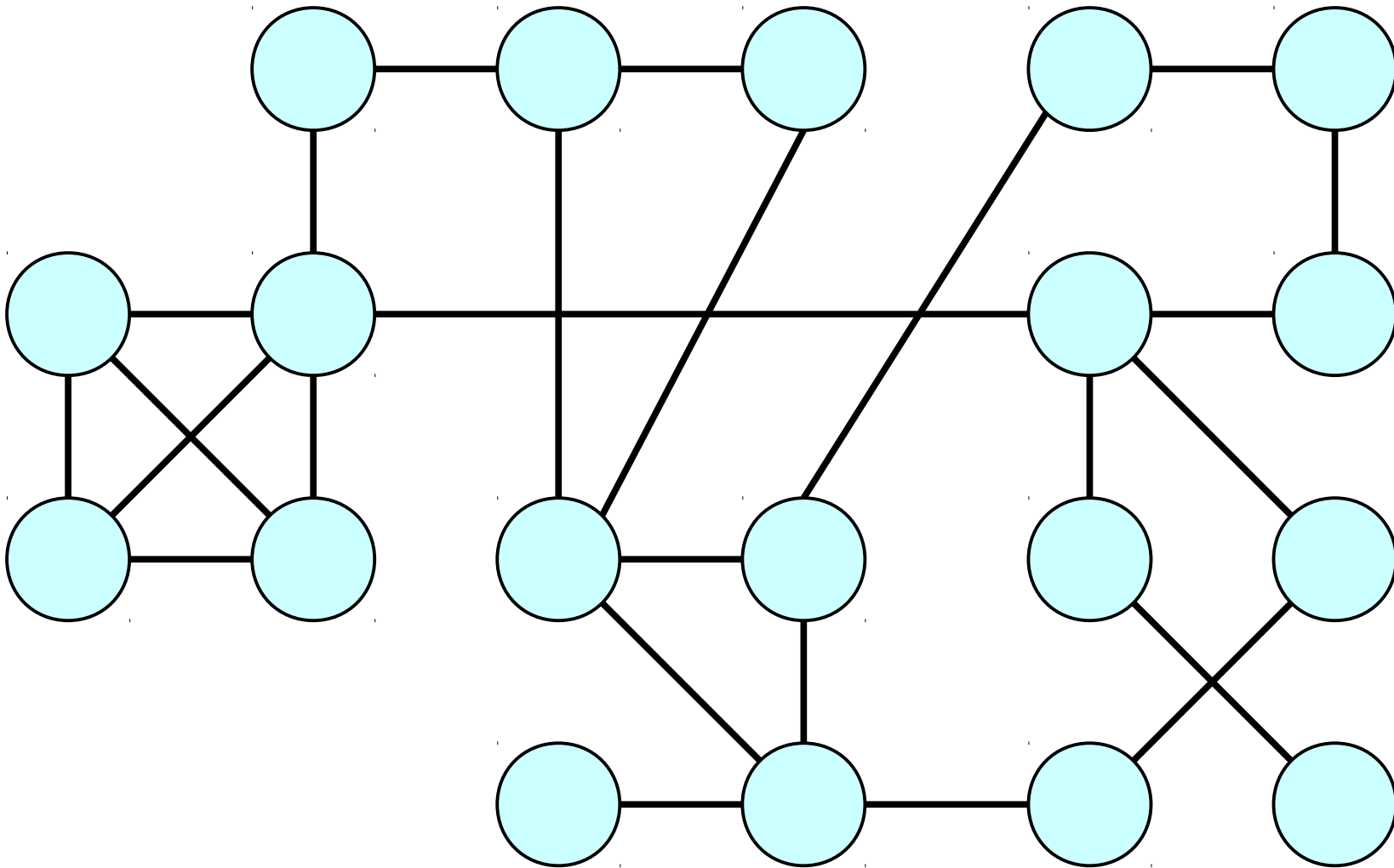


# An Inefficient Algorithm

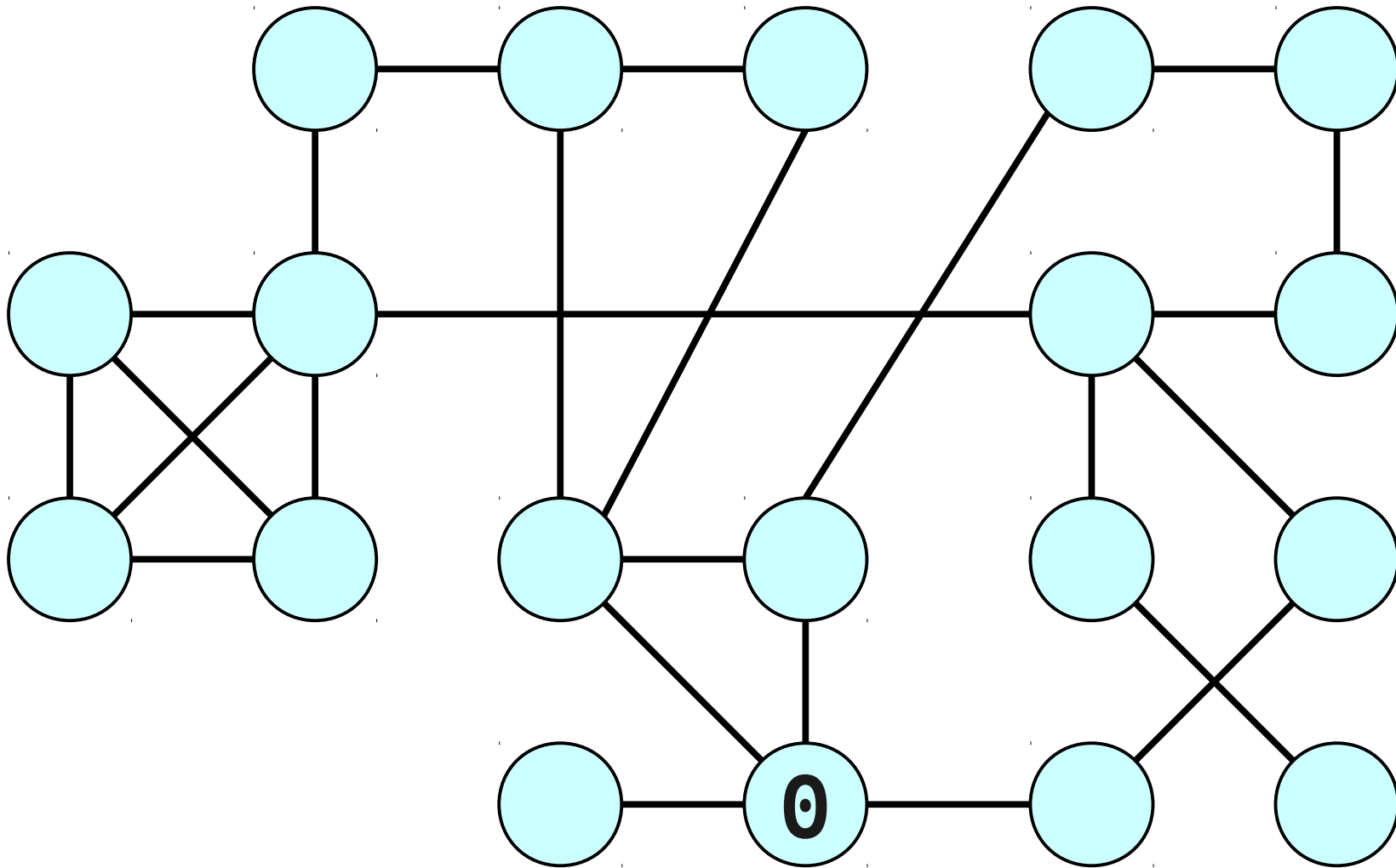


# A Better Approach

# Radiating Outward



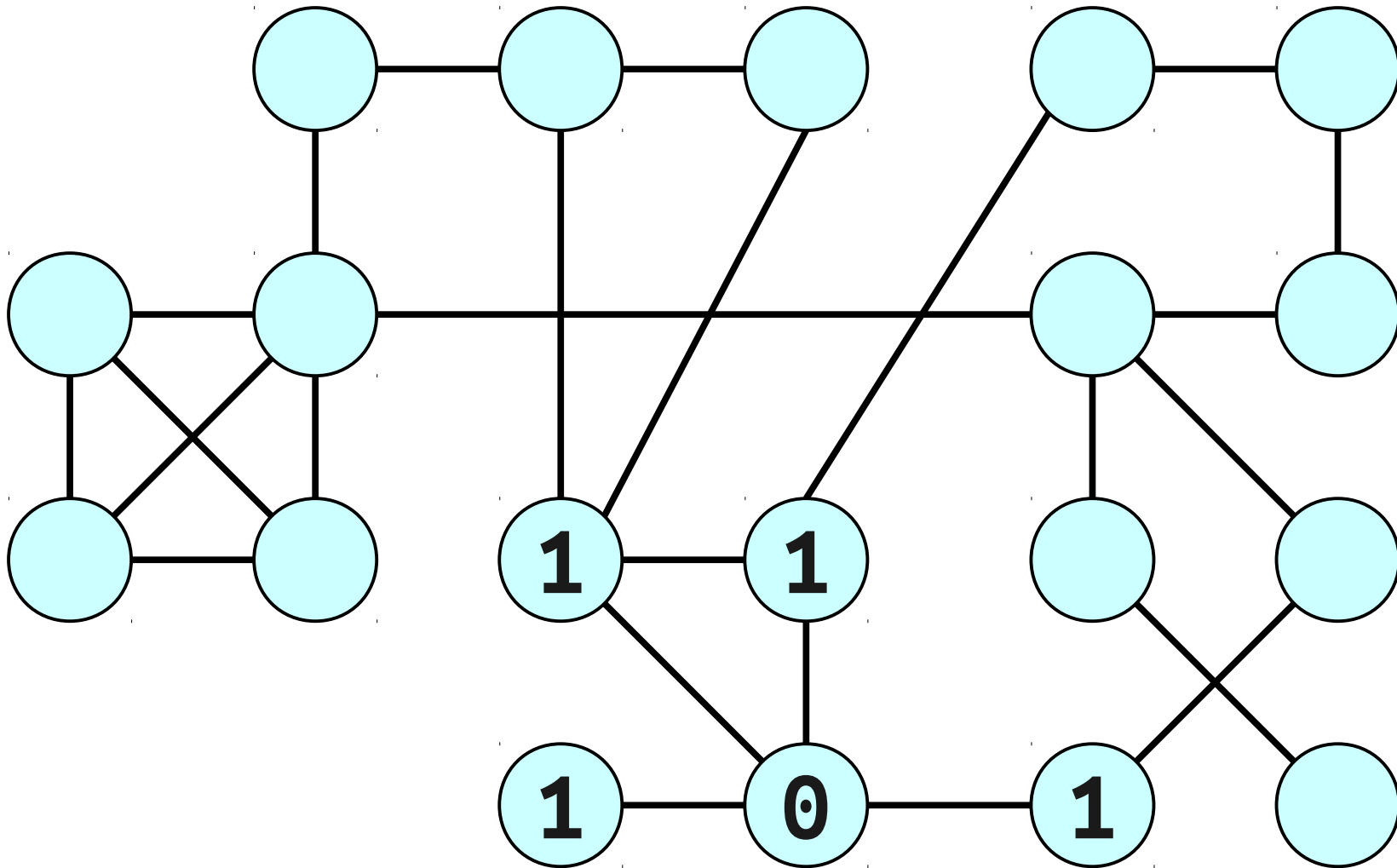
# Radiating Outward



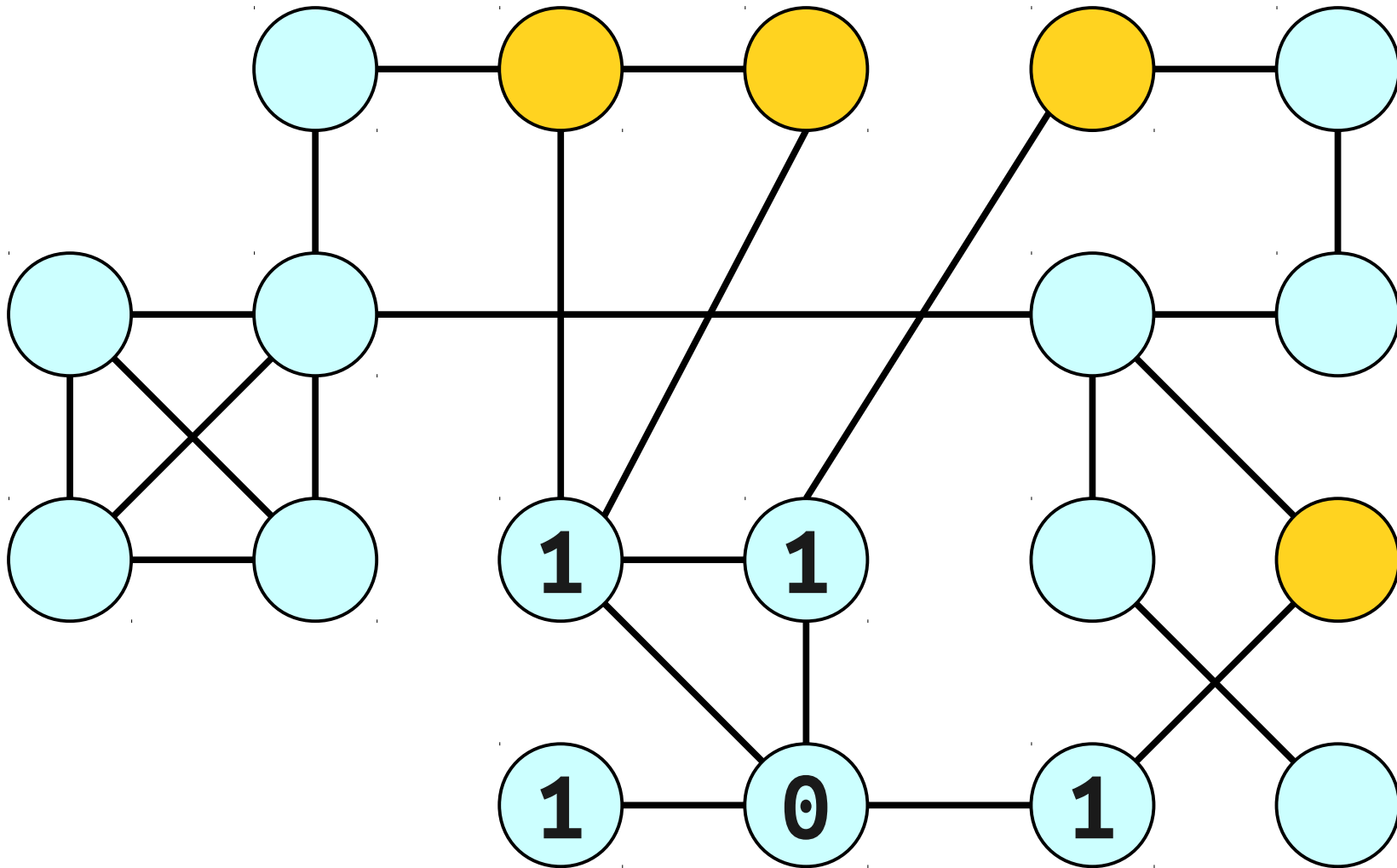




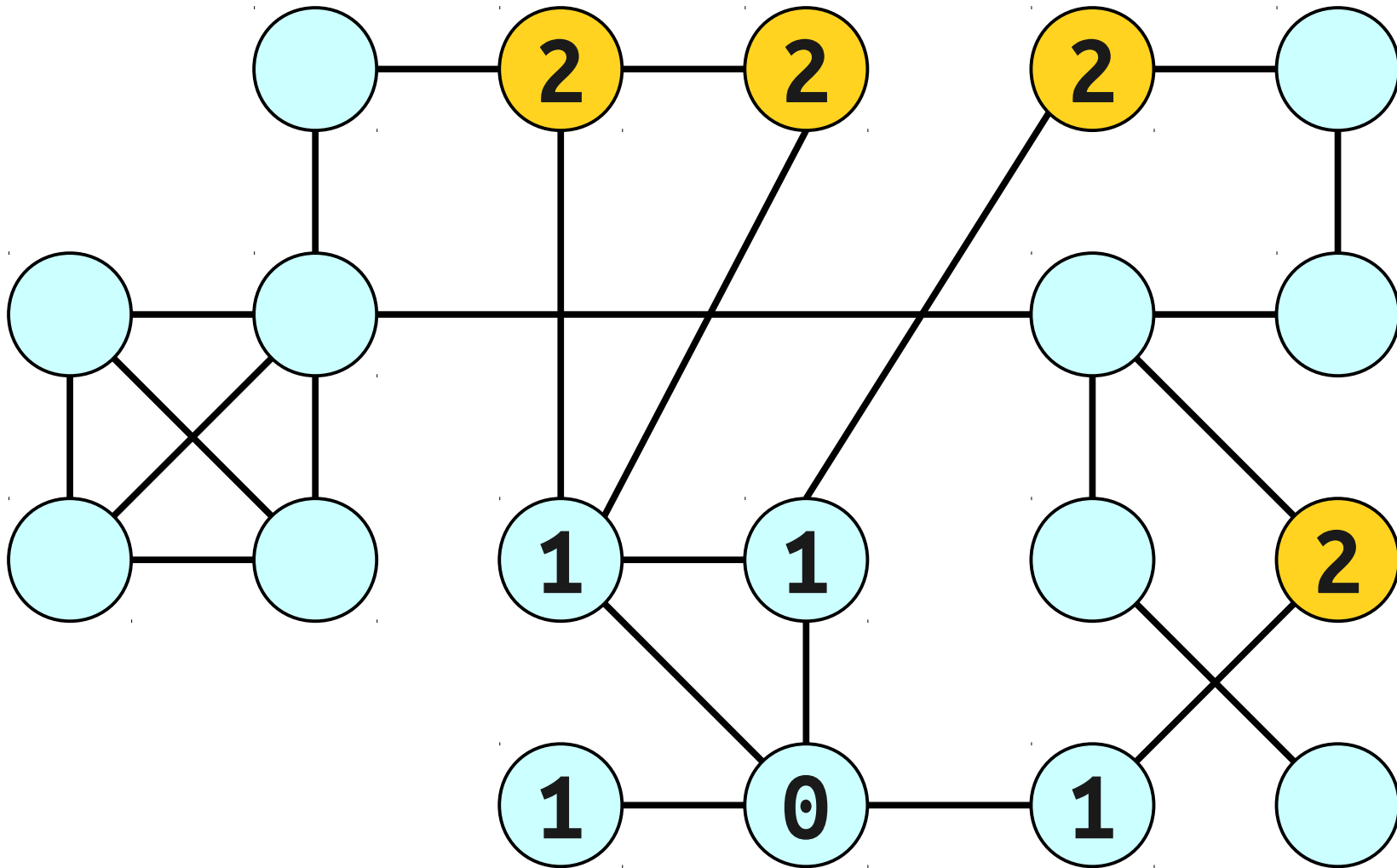
# Radiating Outward



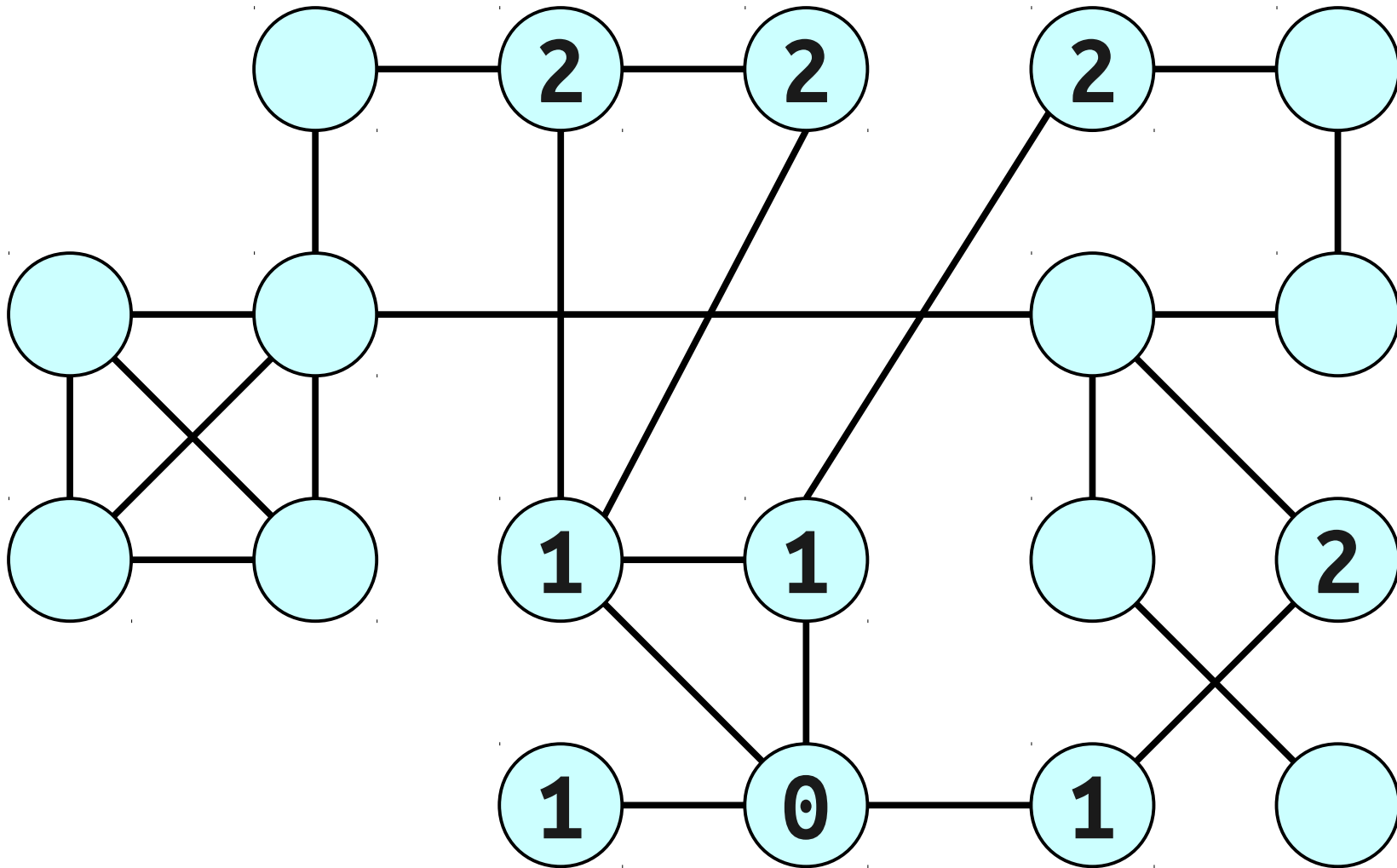
# Radiating Outward



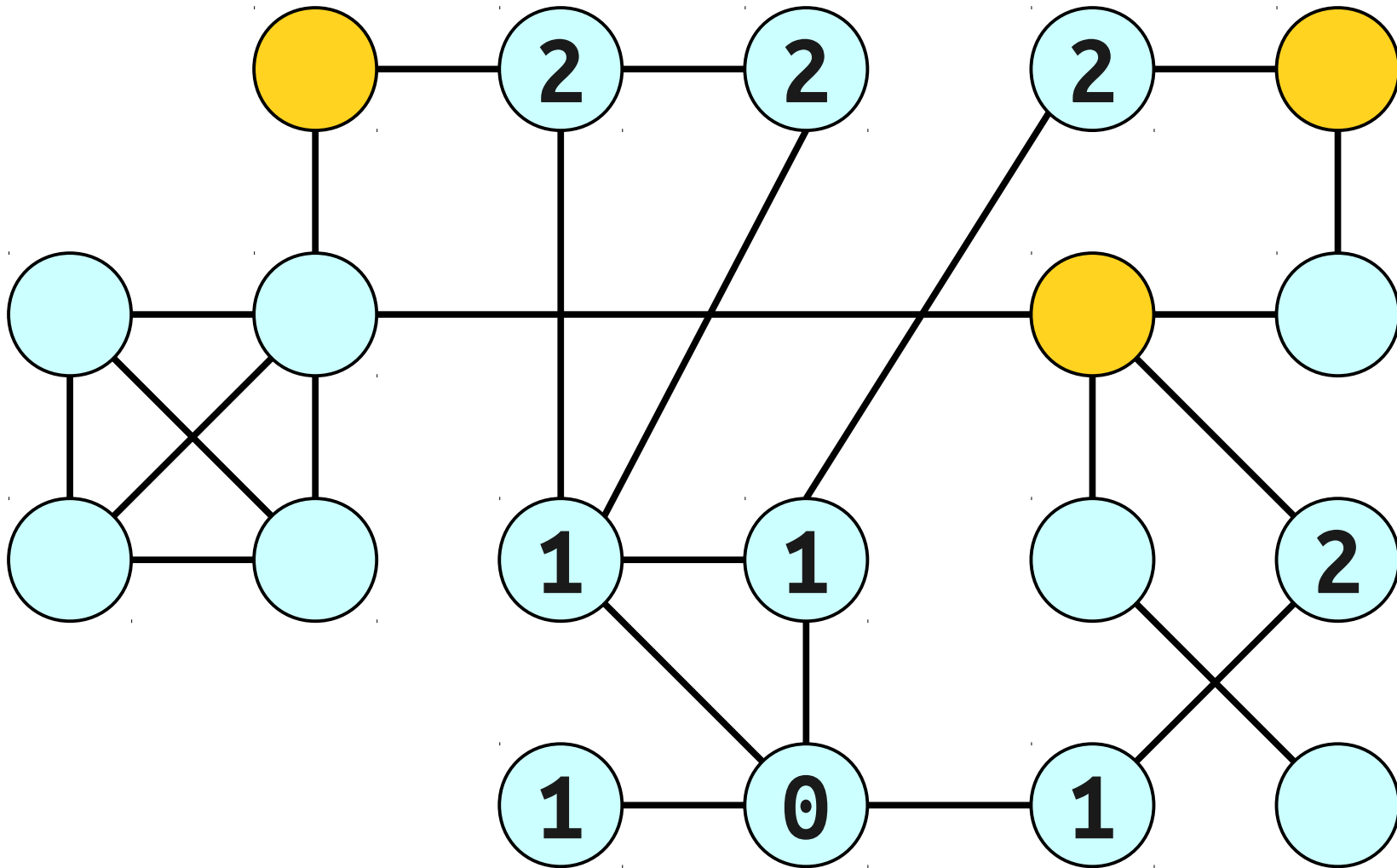
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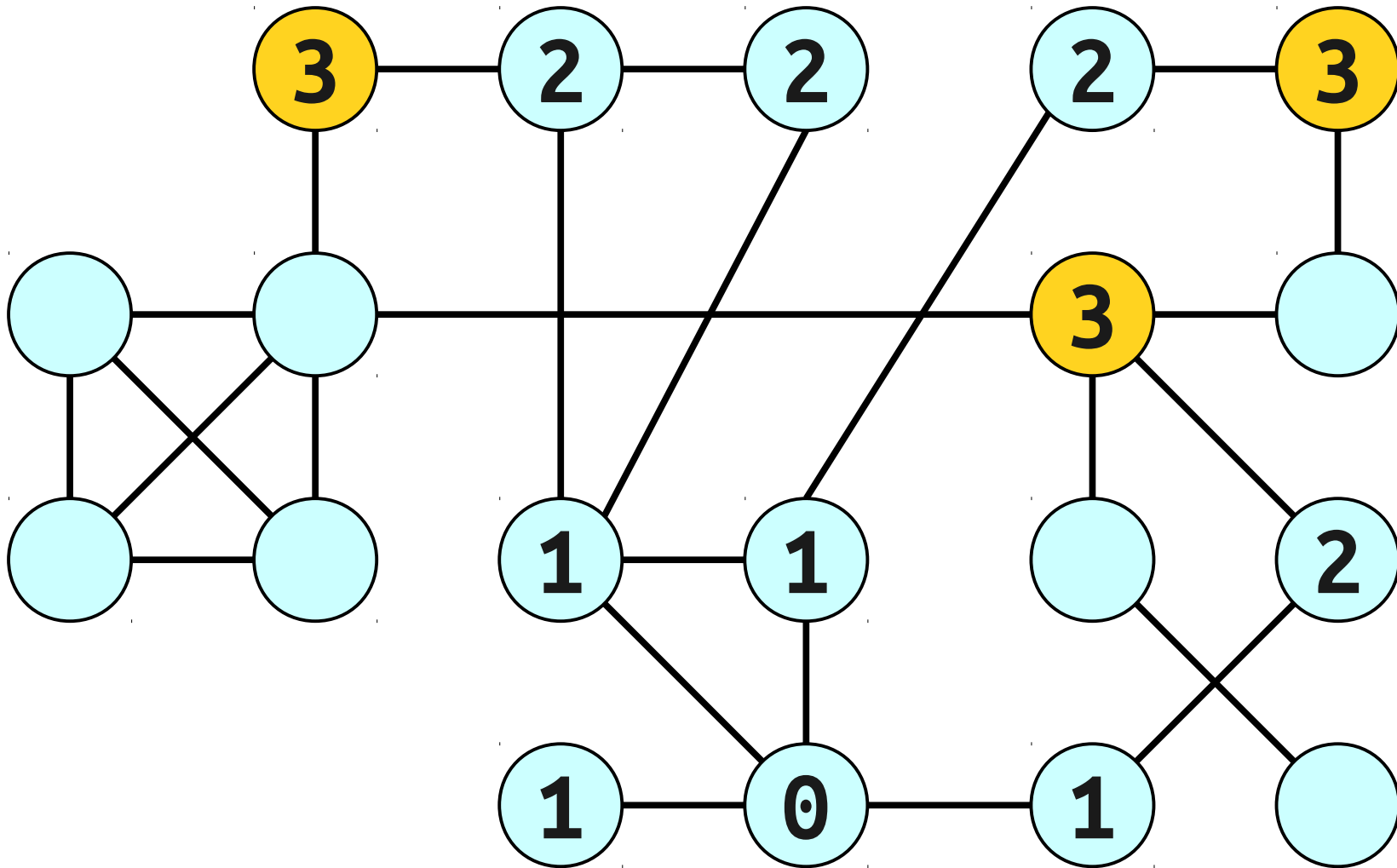
# Radiating Outward



# Radiating Outward



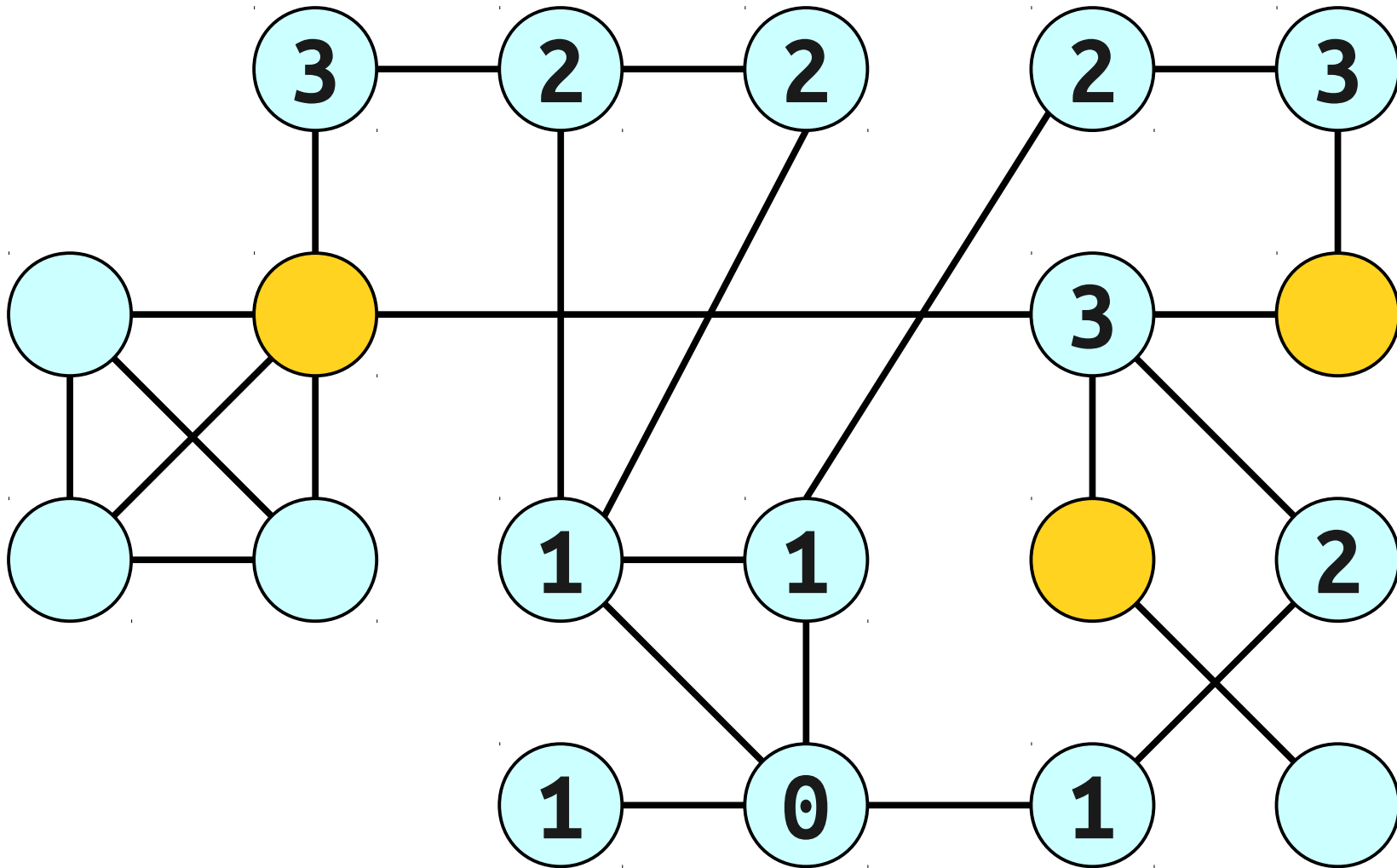
# Radiating Outward



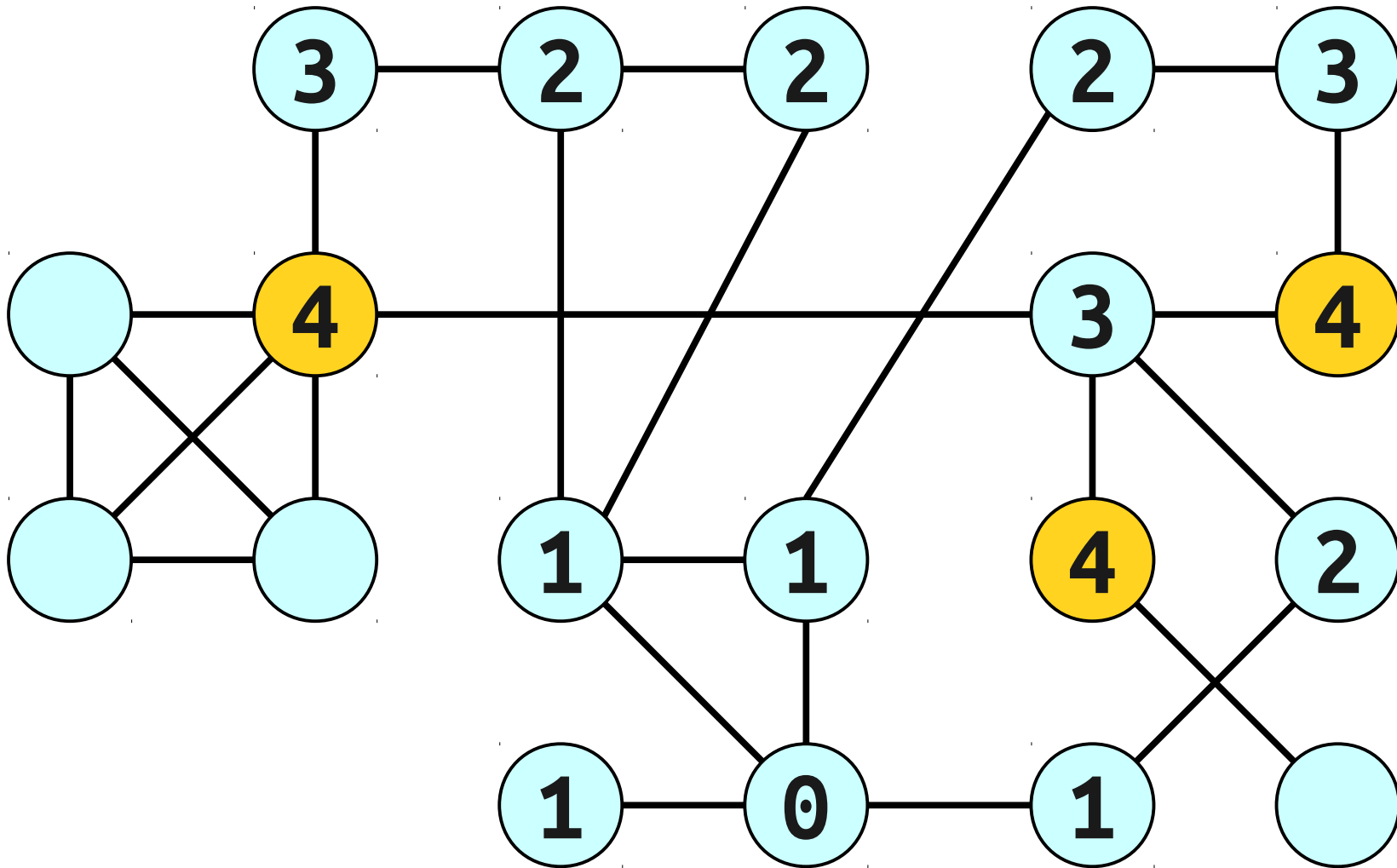




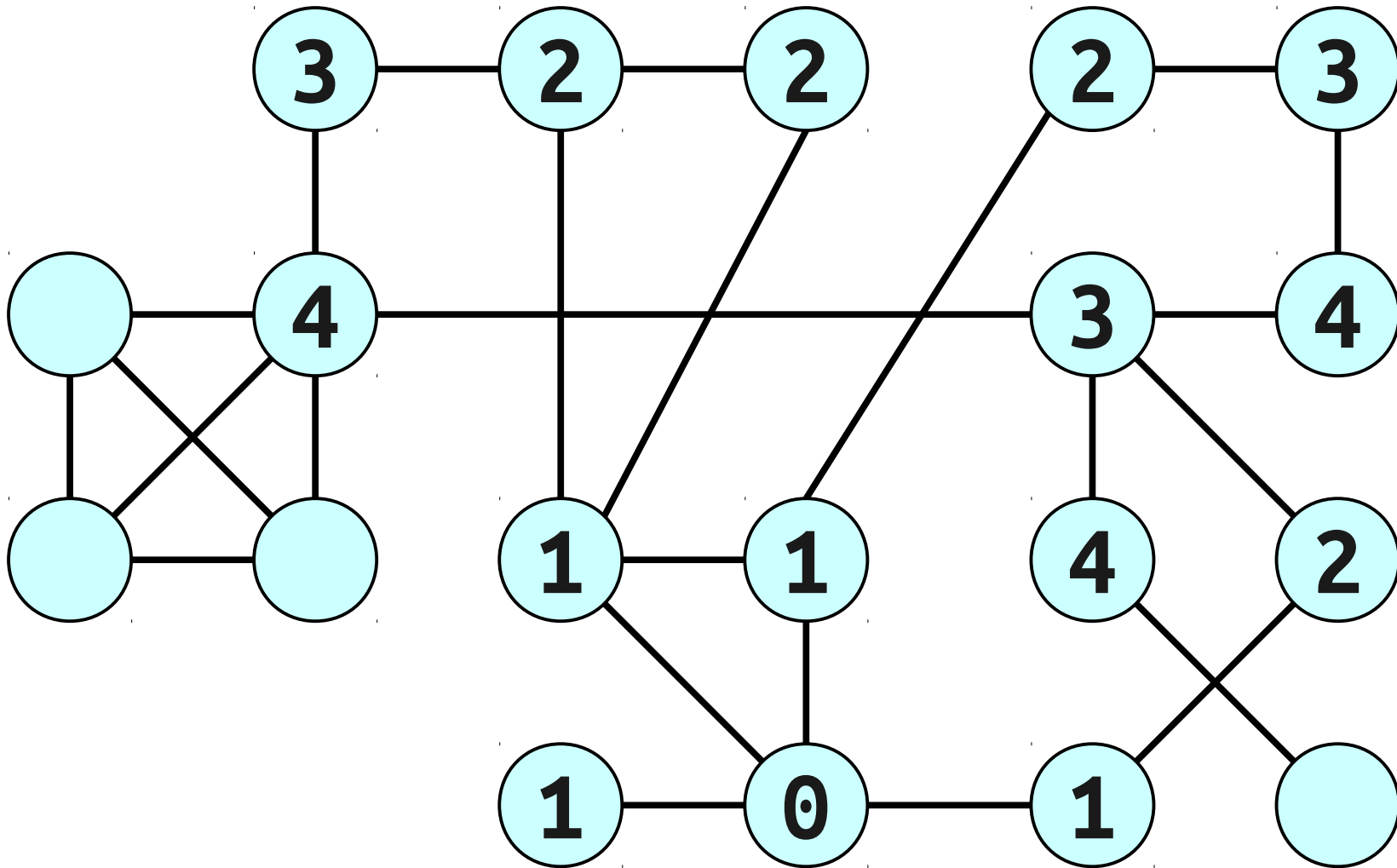
# Radiating Outward



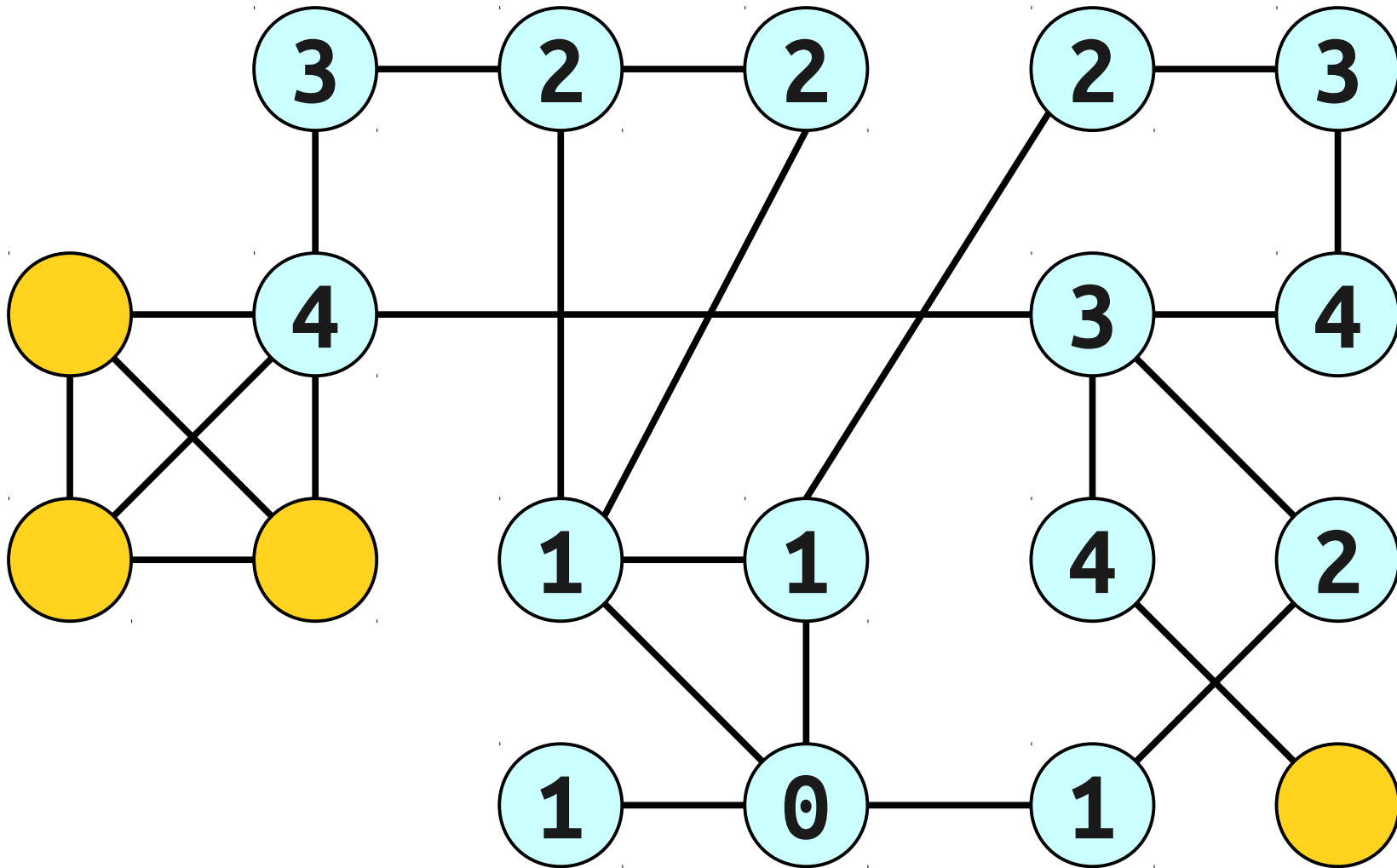
# Radiating Outward



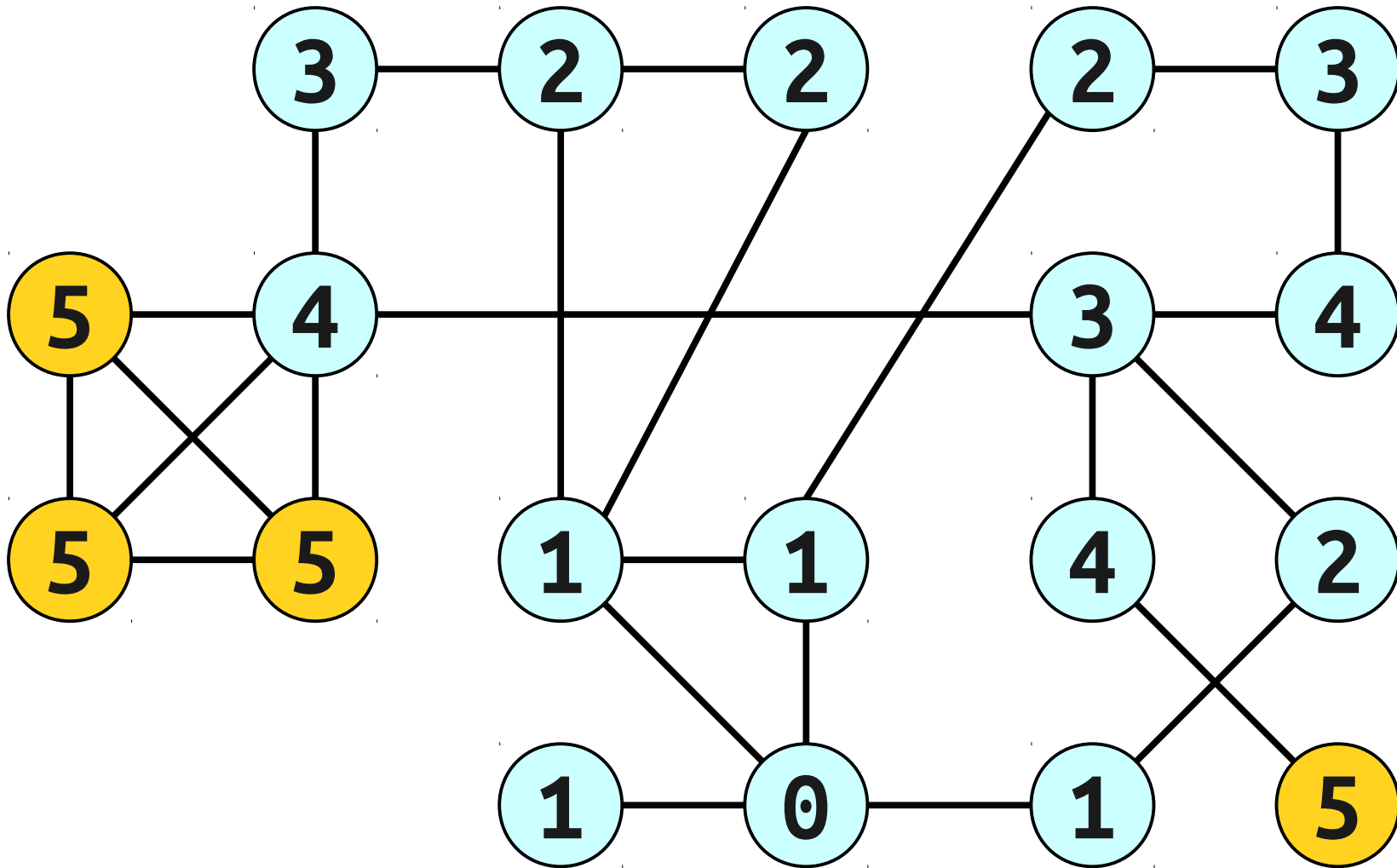
# Radiating Outward



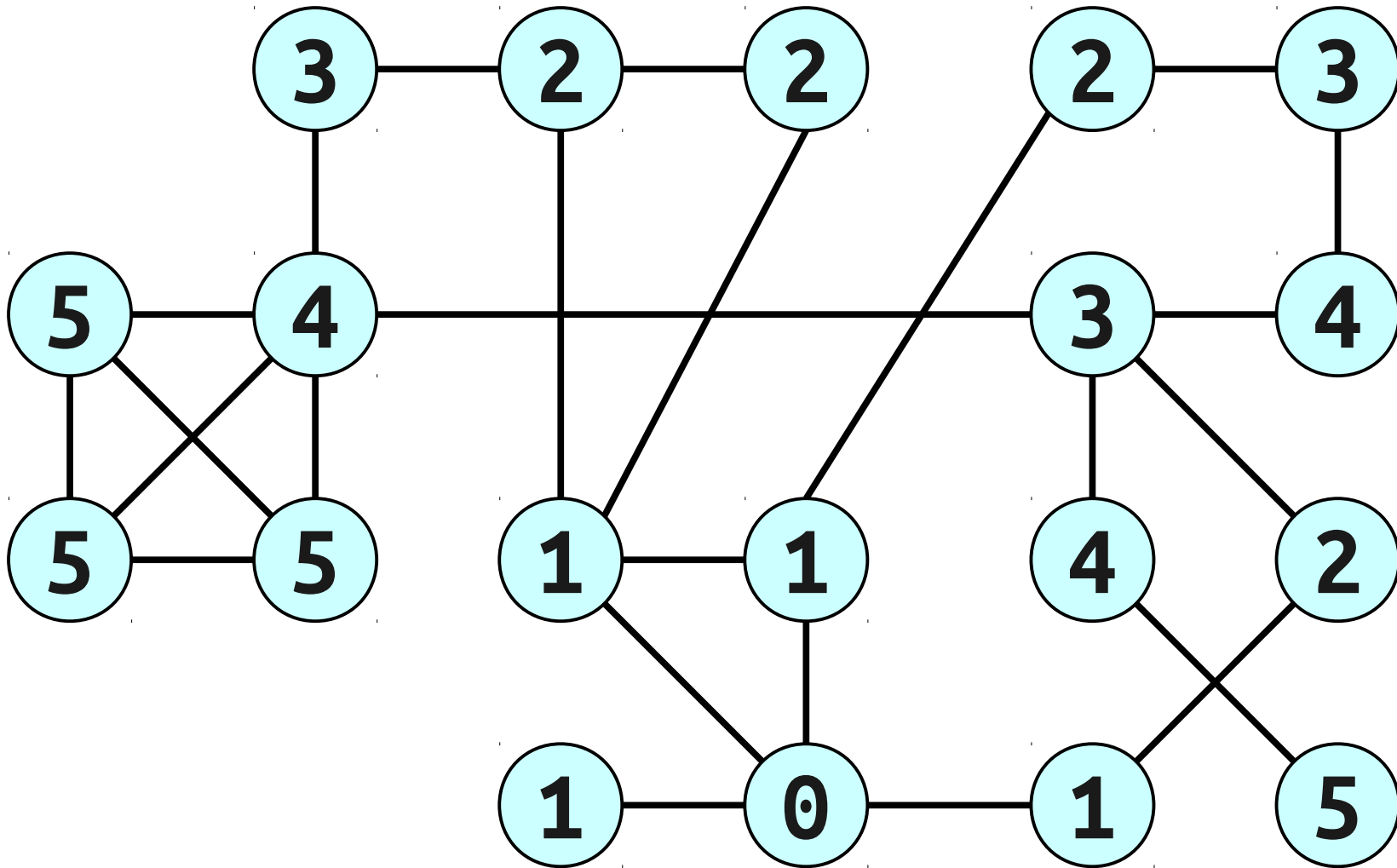
# Radiating Outward



# Radiating Outward



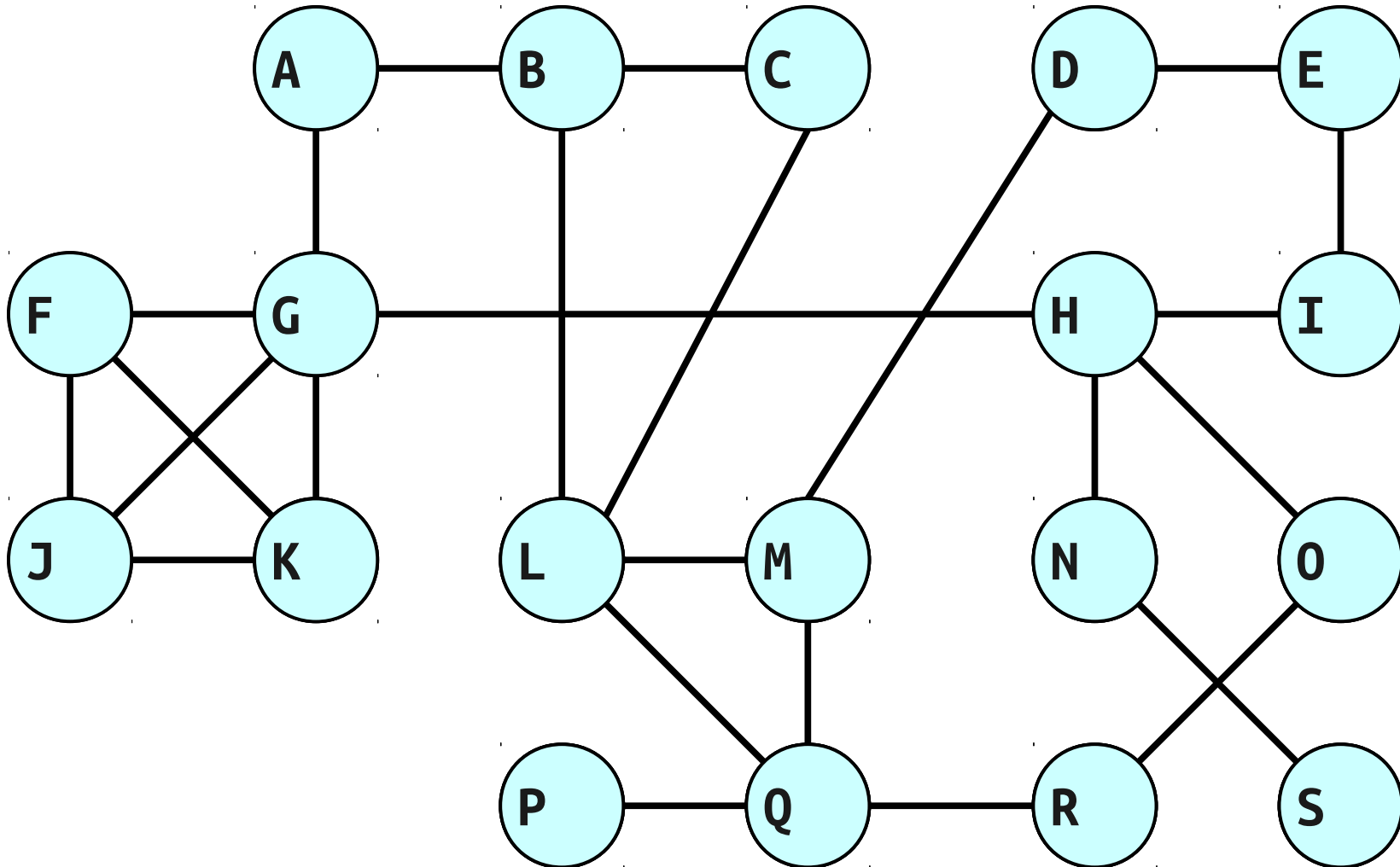
# Radiating Outward



# A Secondary Idea

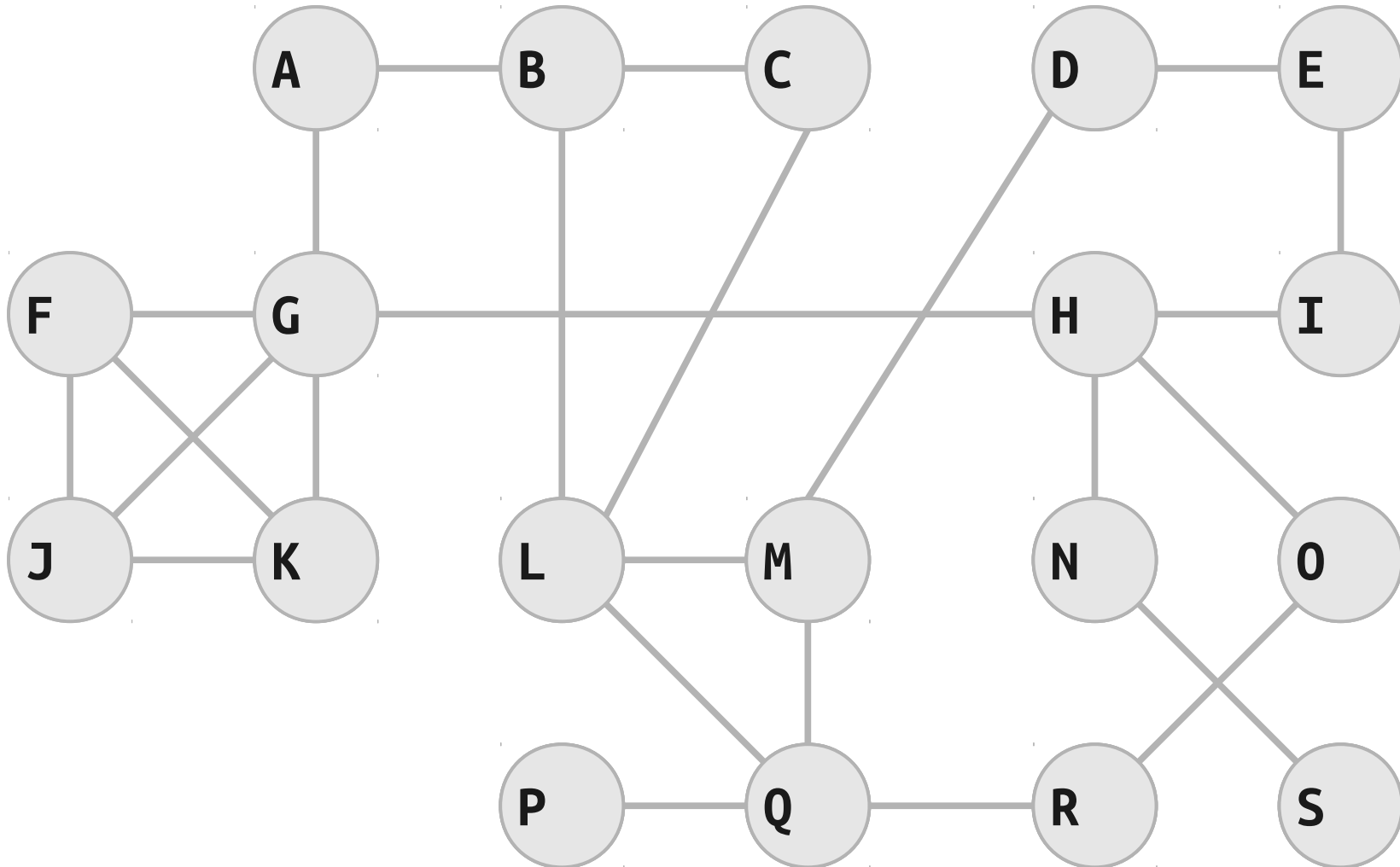
- Proceed outward from the source node  $s$  in “layers.”
  - The first layer is all nodes of distance 0.
  - The second layer is all nodes of distance 1.
  - The third layer is all nodes of distance 2.
  - etc.
- This gives rise to **breadth-first search**.

# Breadth-First Search

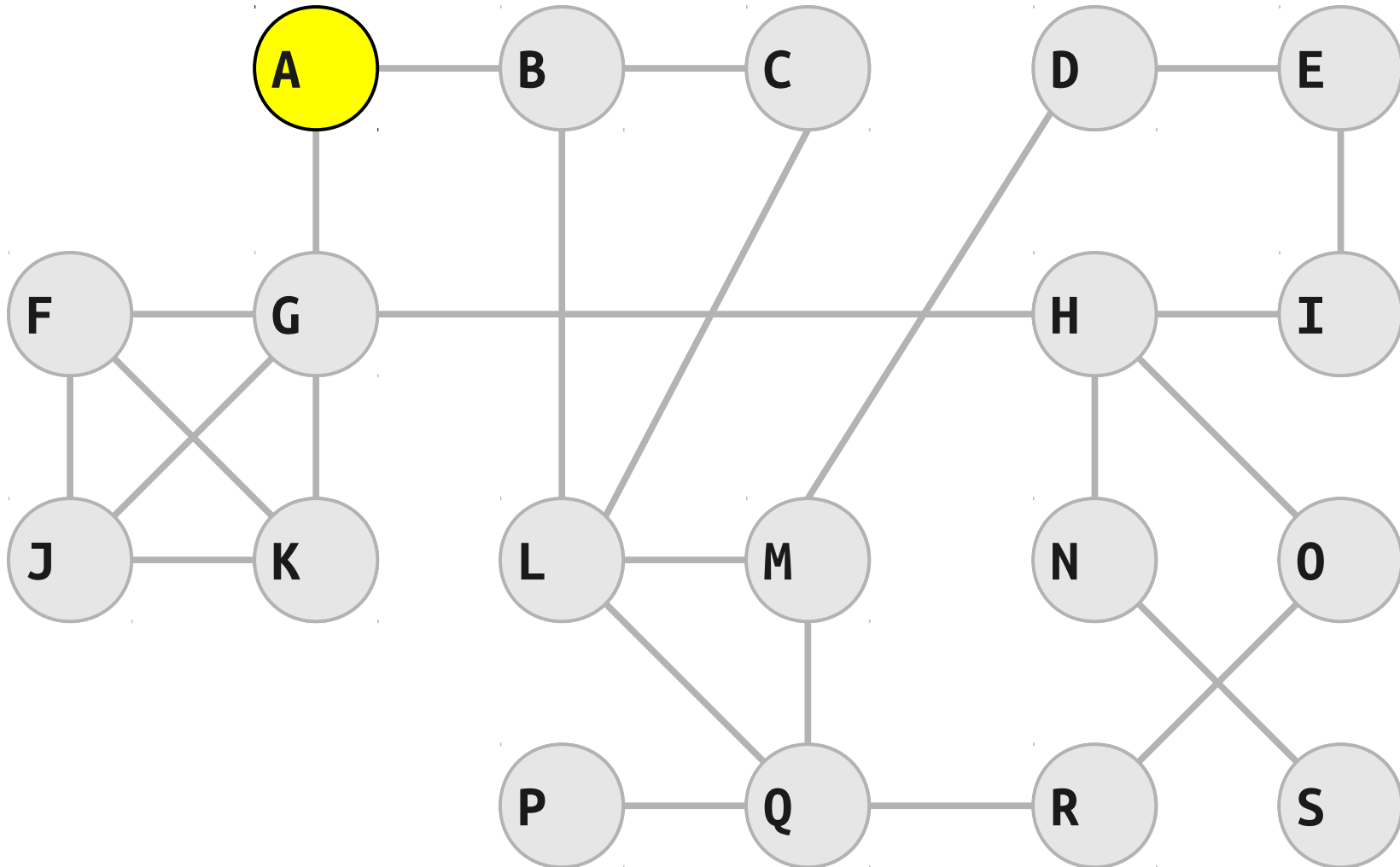




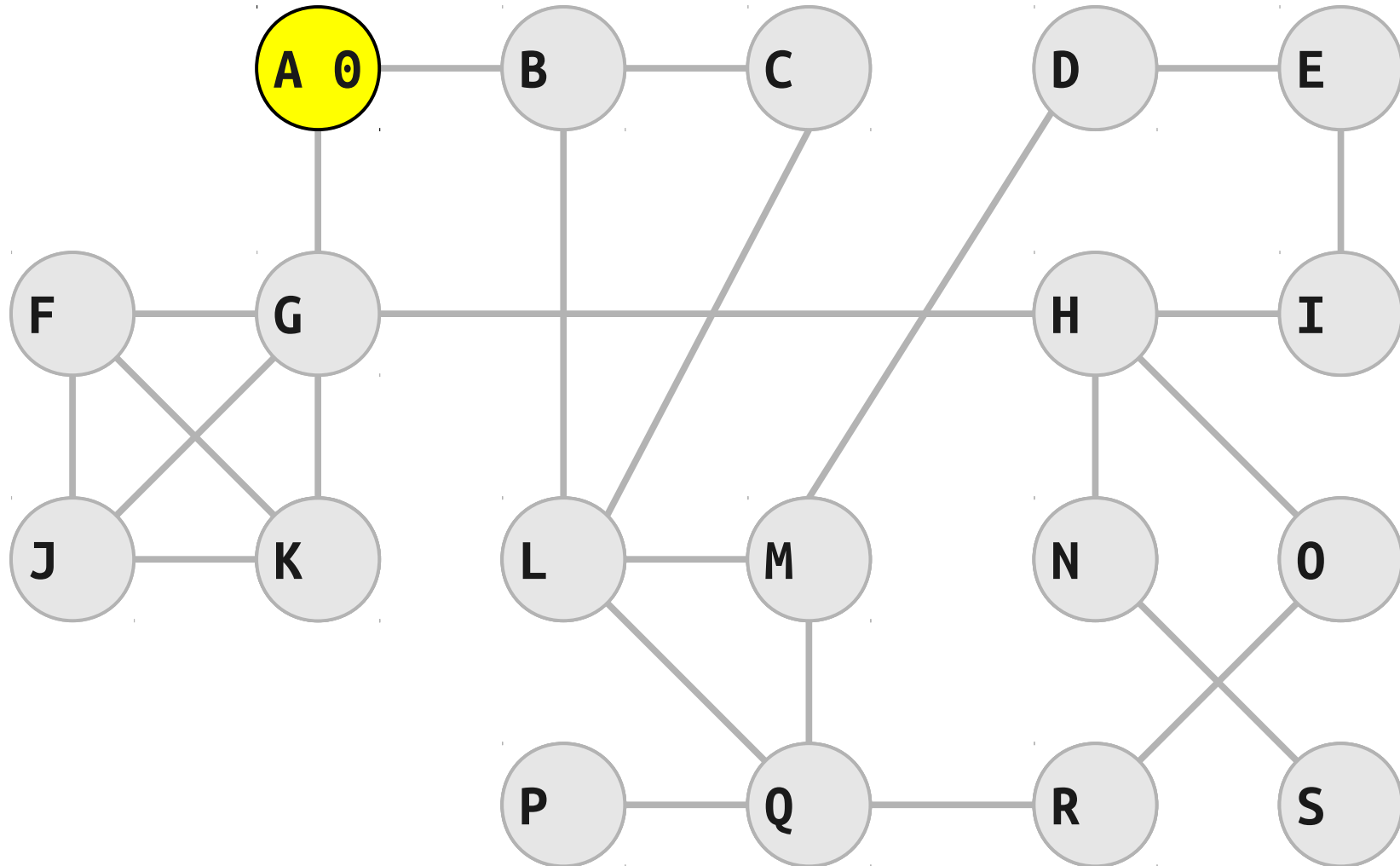
# Breadth-First Search



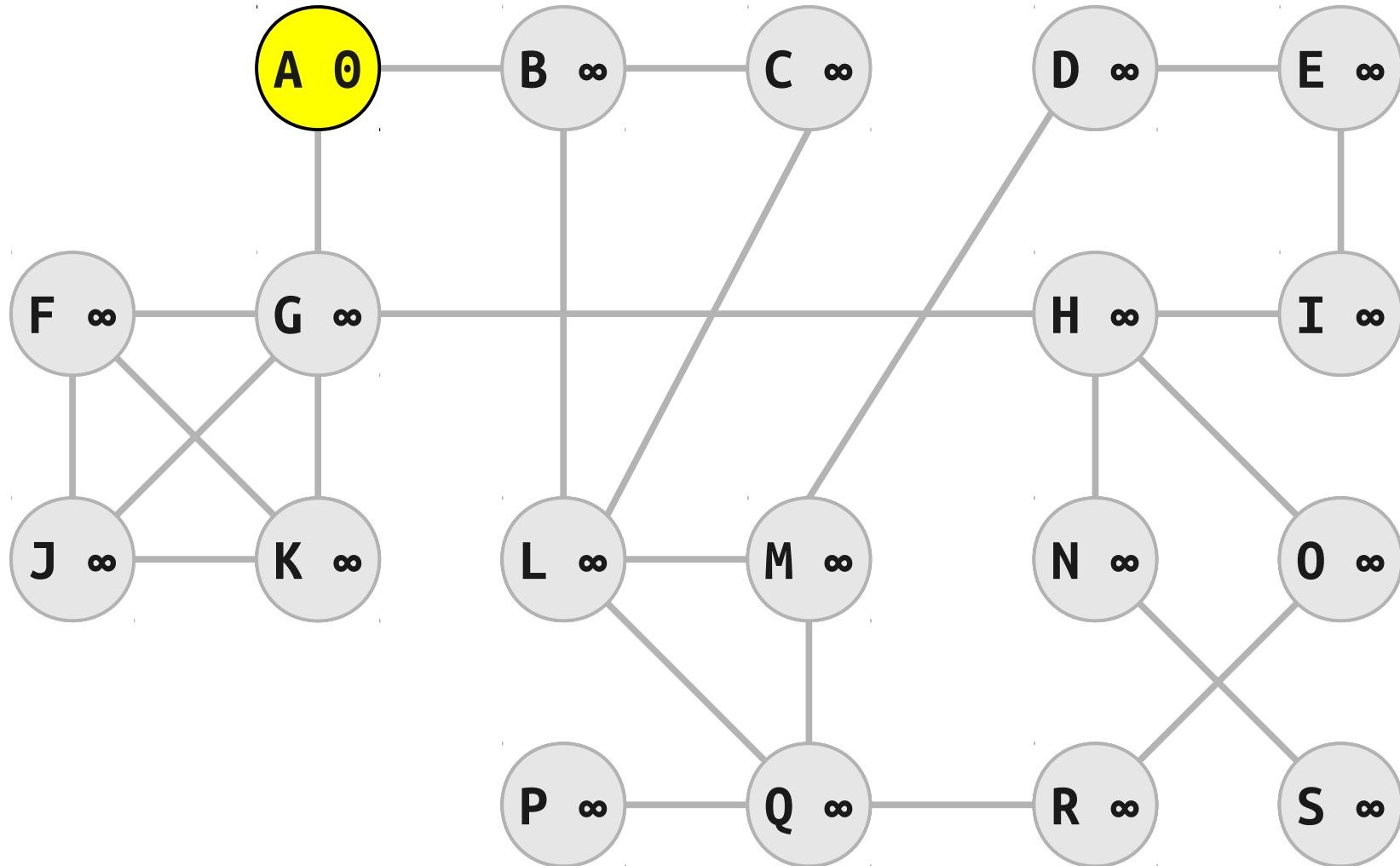
# Breadth-First Search



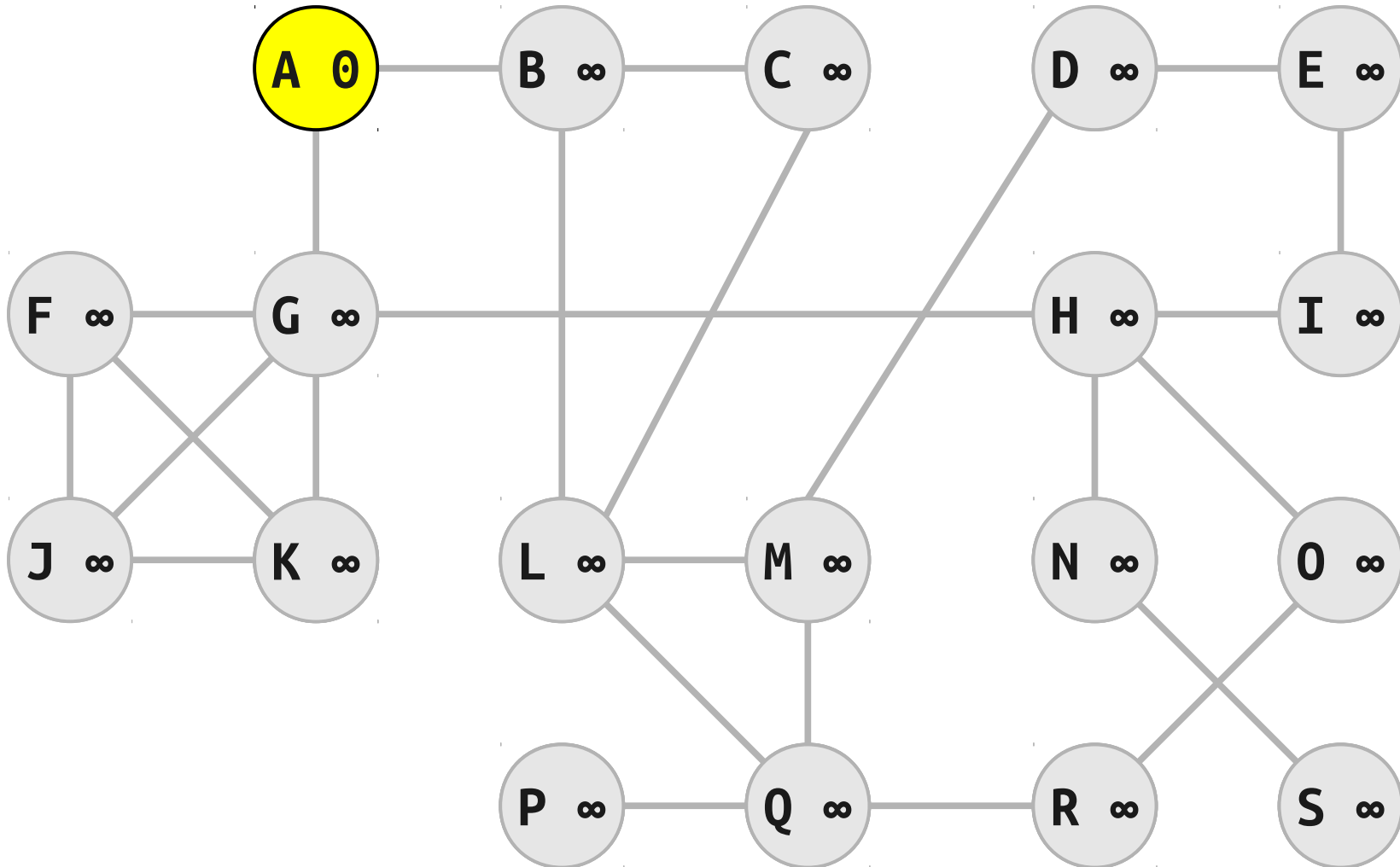
# Breadth-First Search



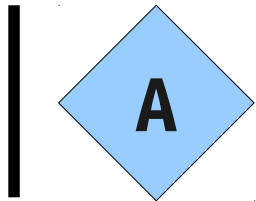
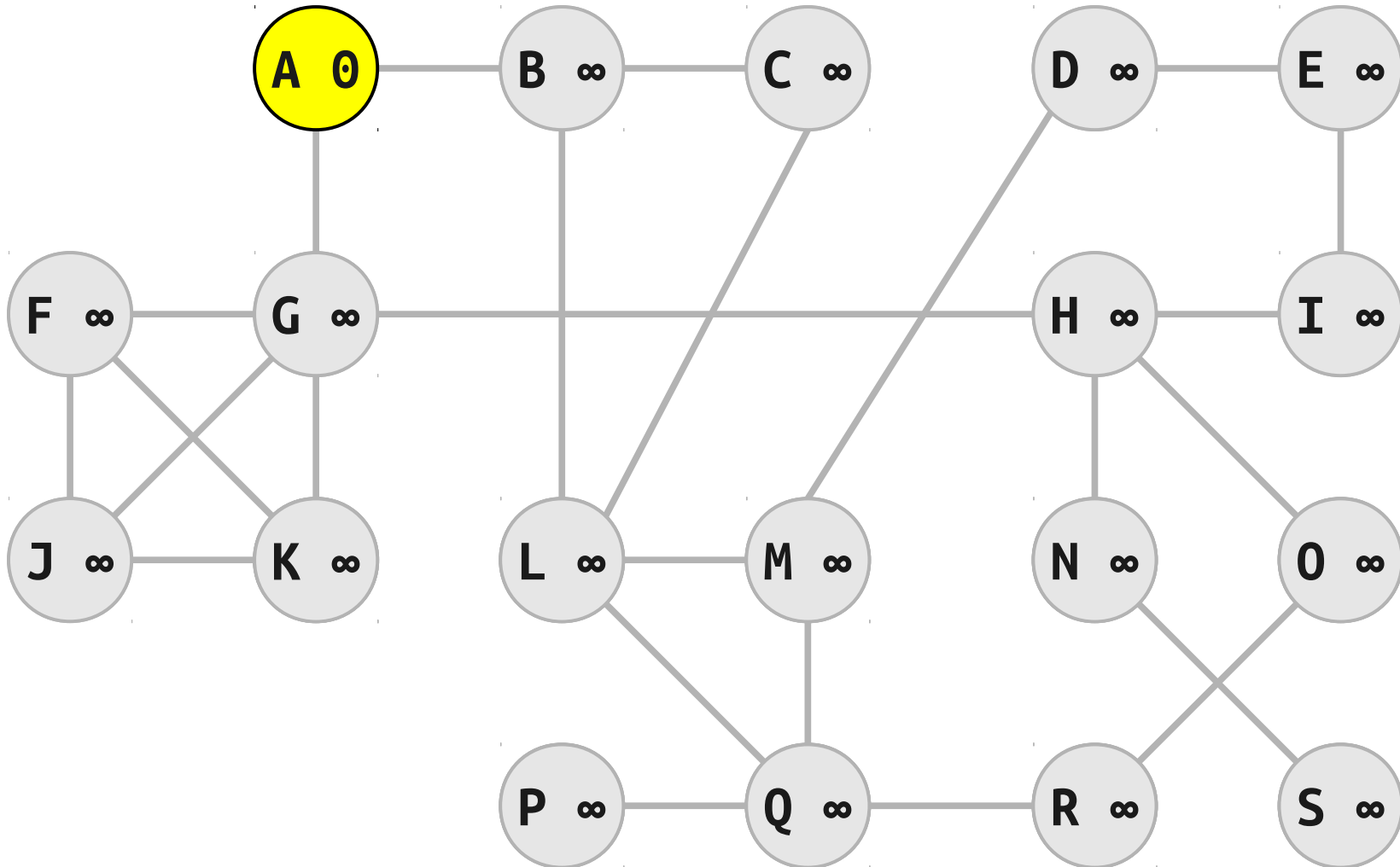
# Breadth-First Search



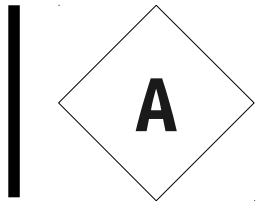
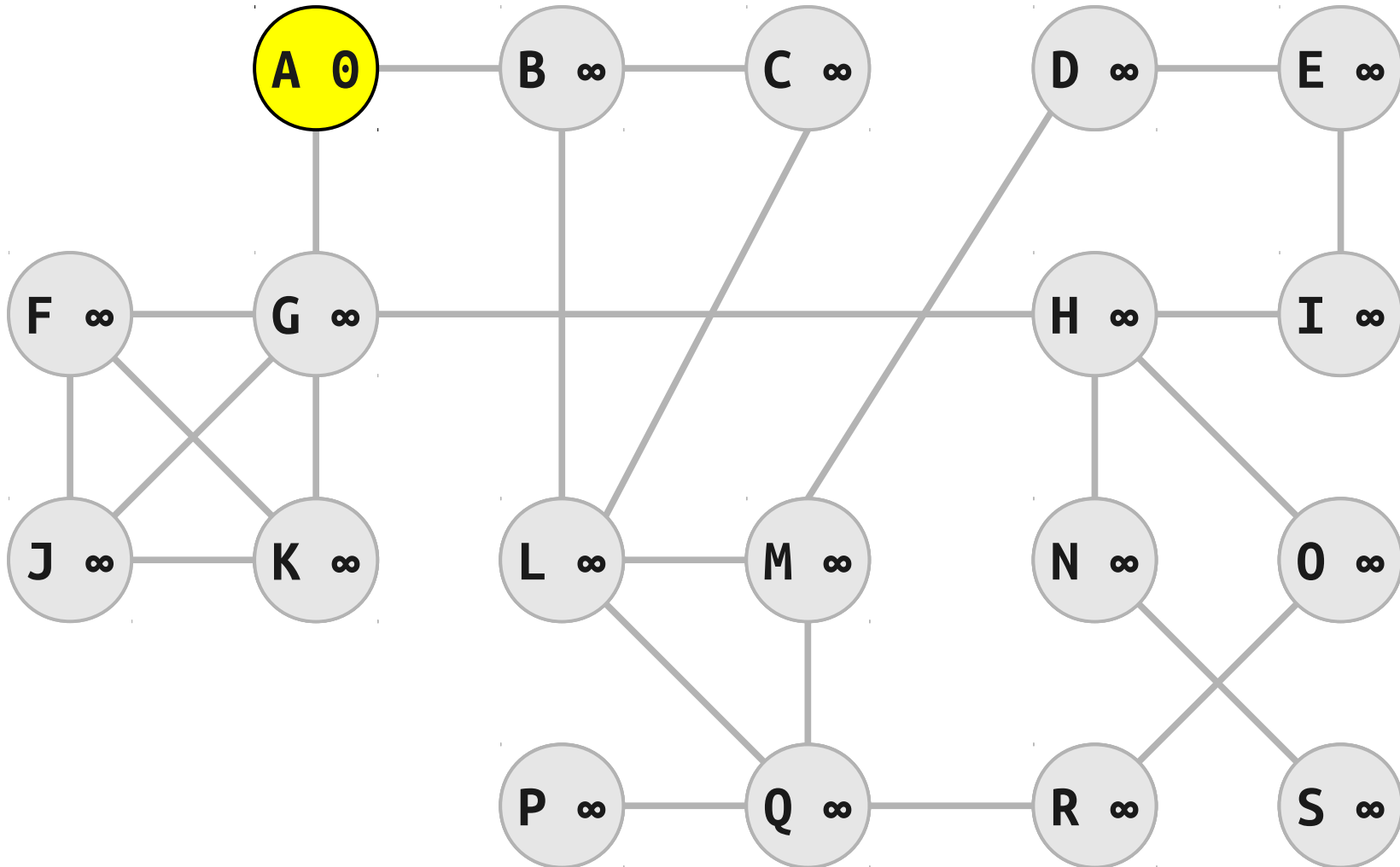
# Breadth-First Search



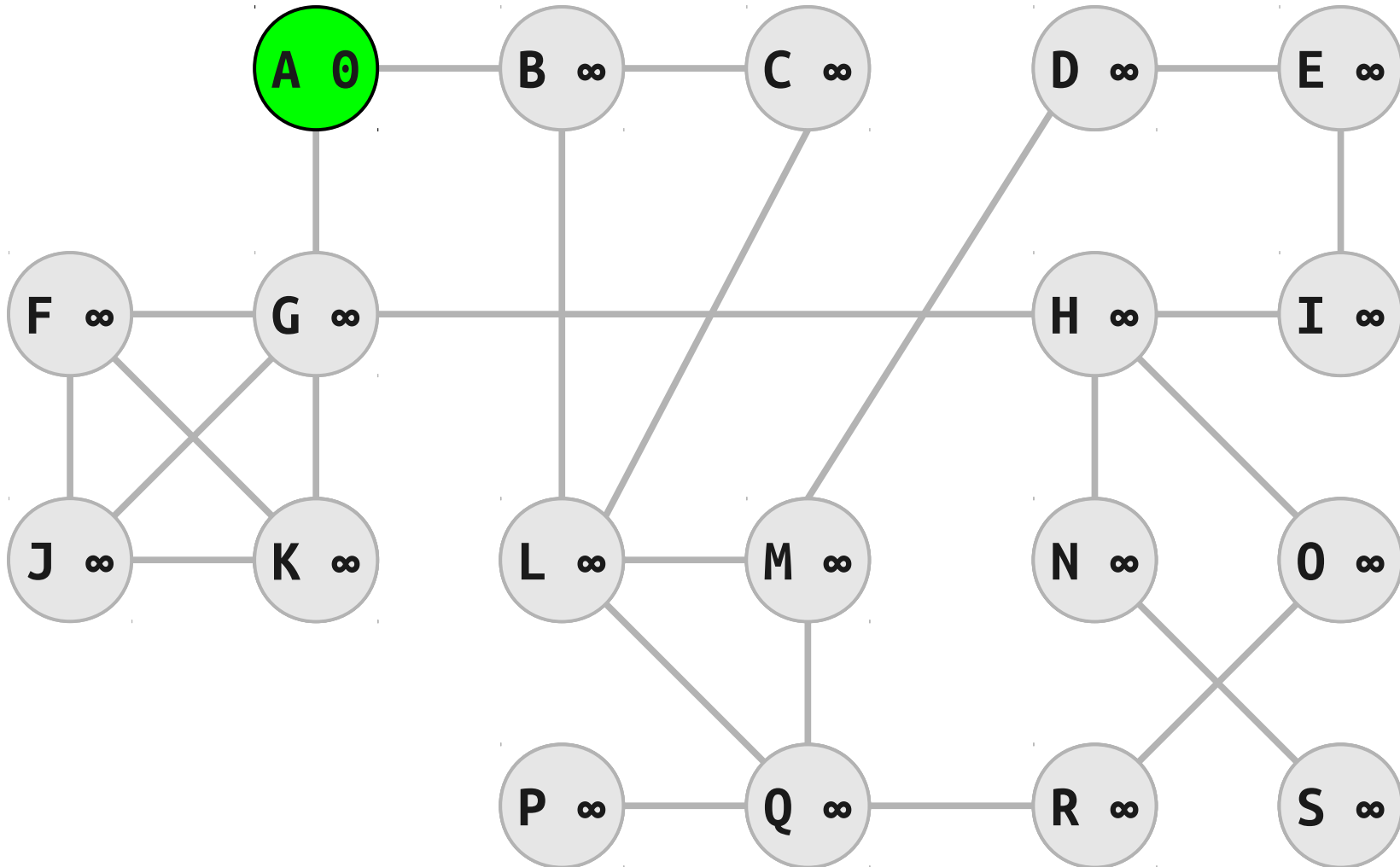
# Breadth-First Search



# Breadth-First Search

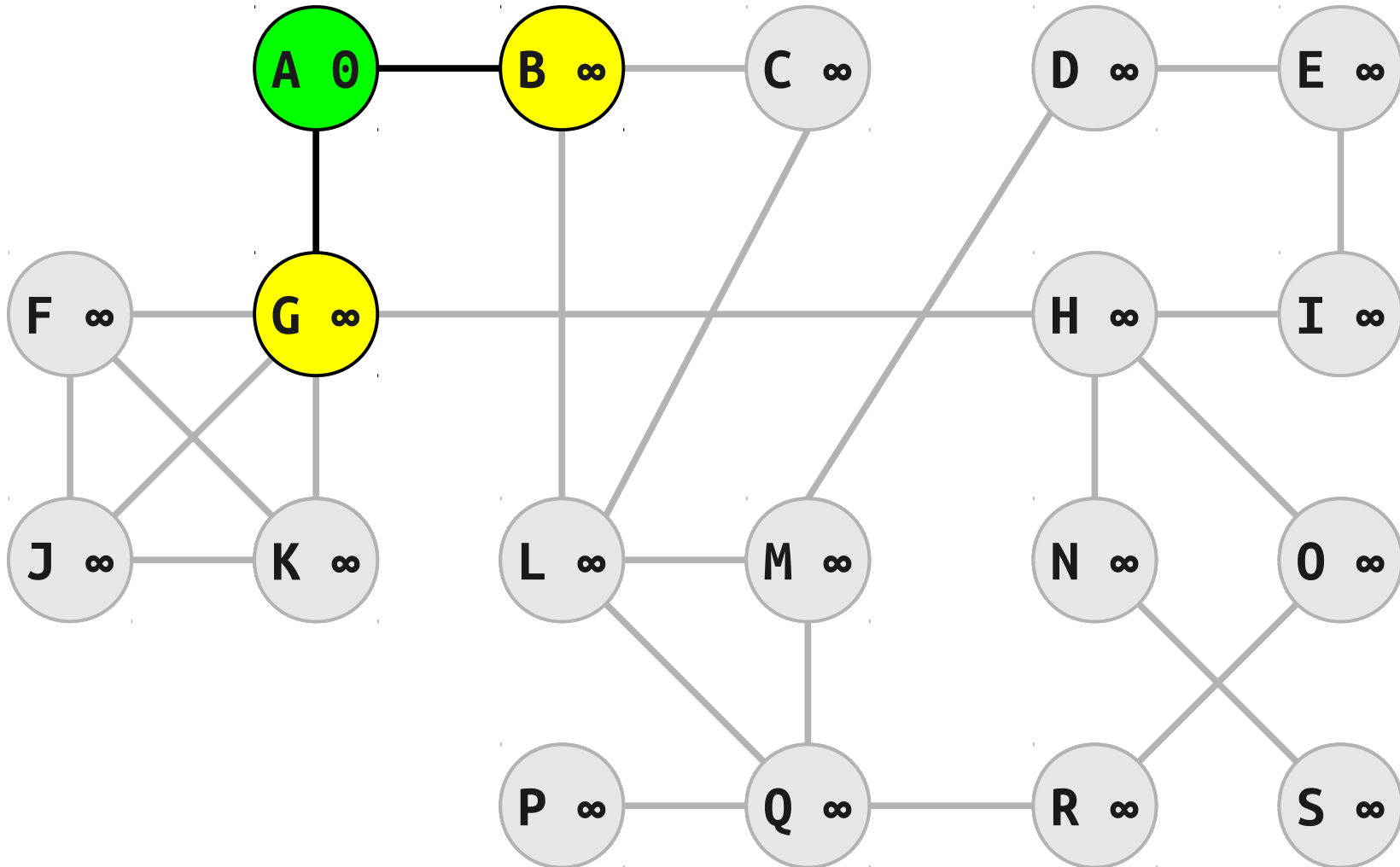


# Breadth-First Search

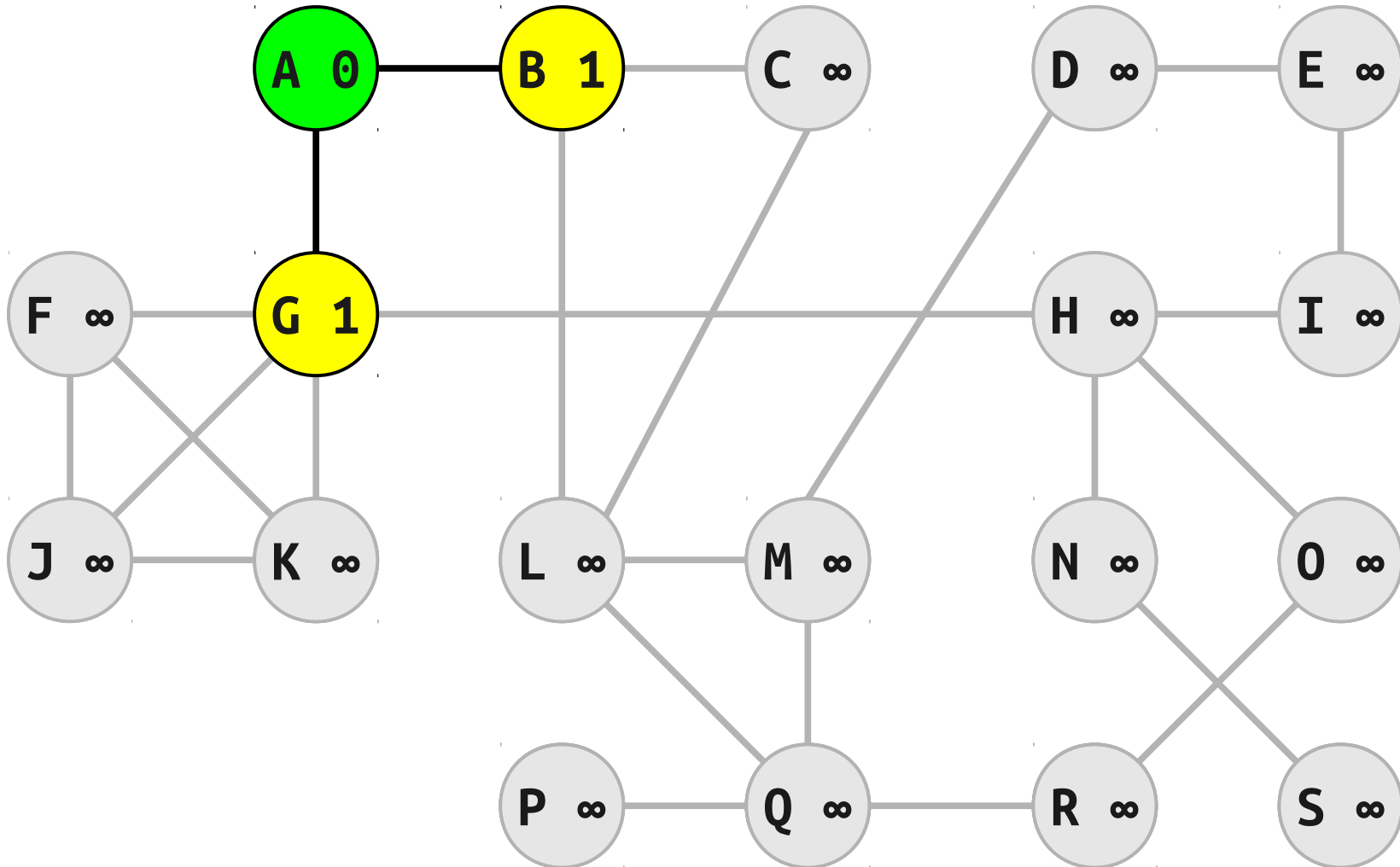




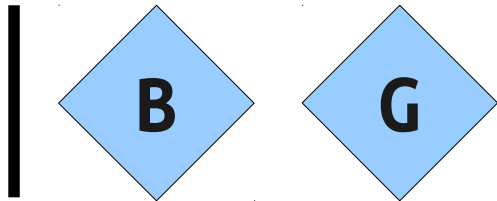
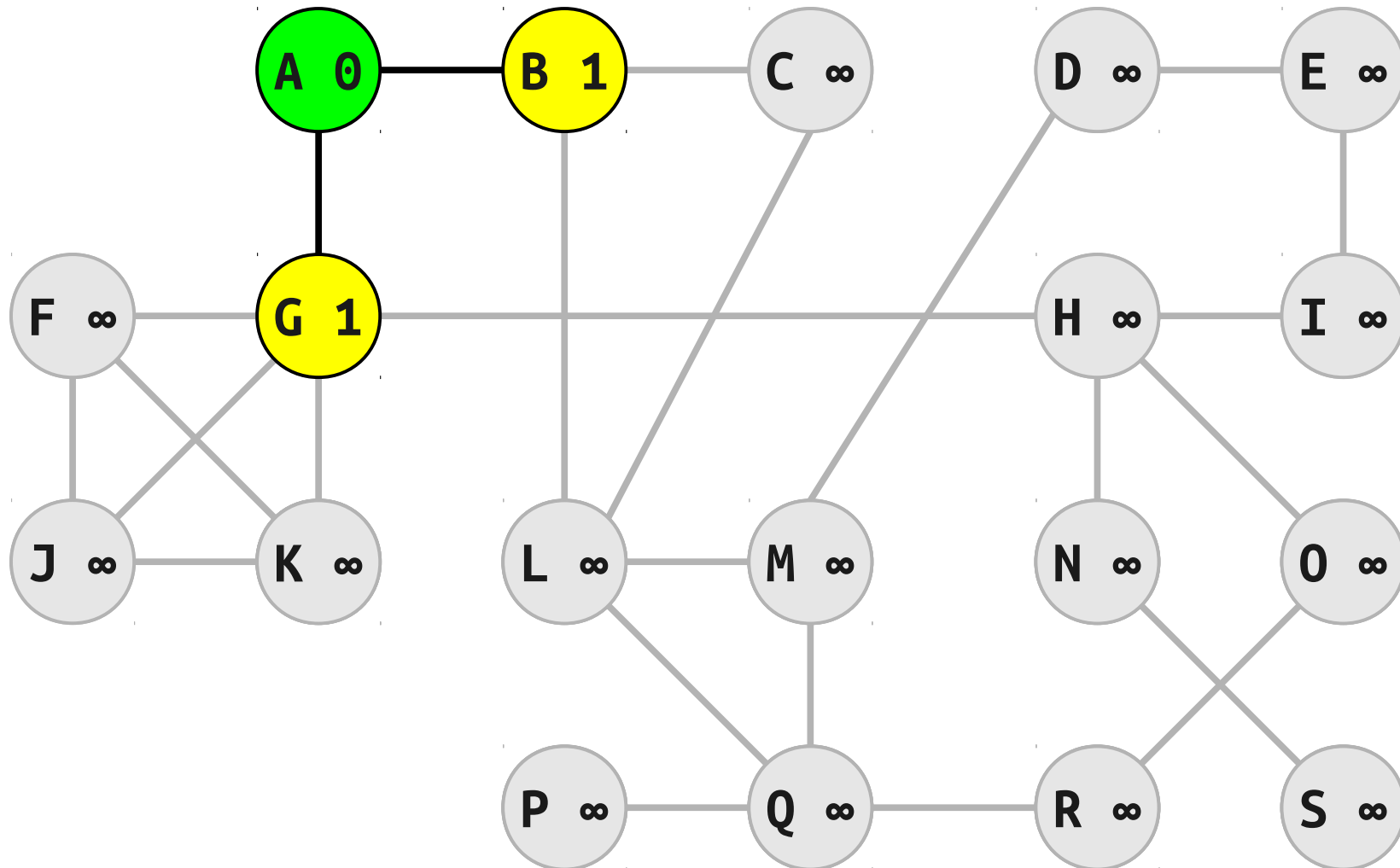
# Breadth-First Search



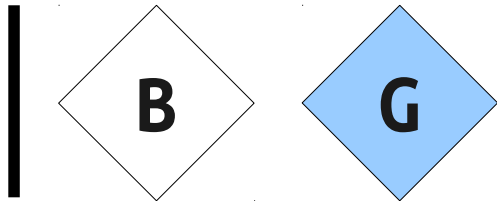
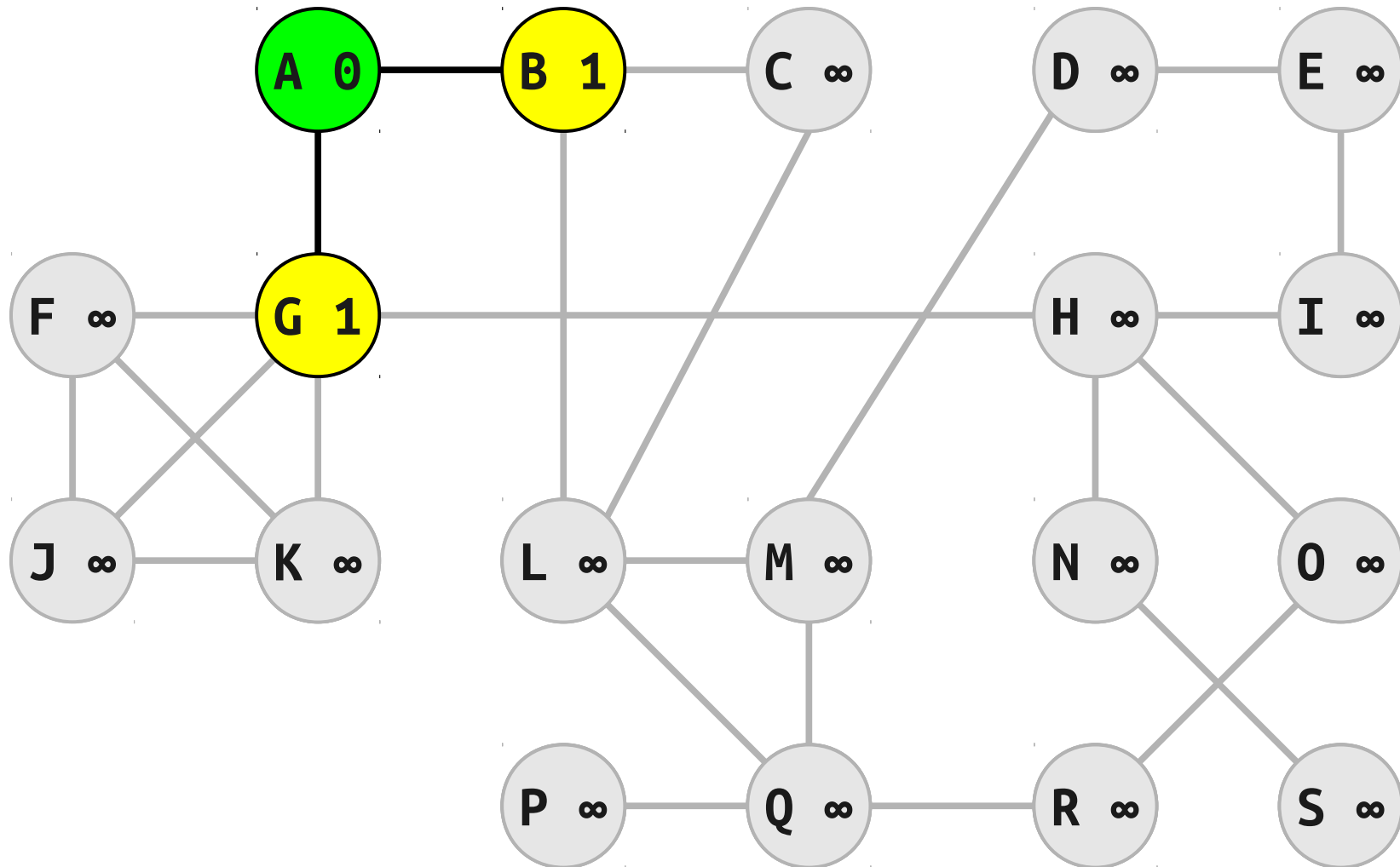
# Breadth-First Search



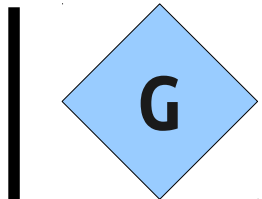
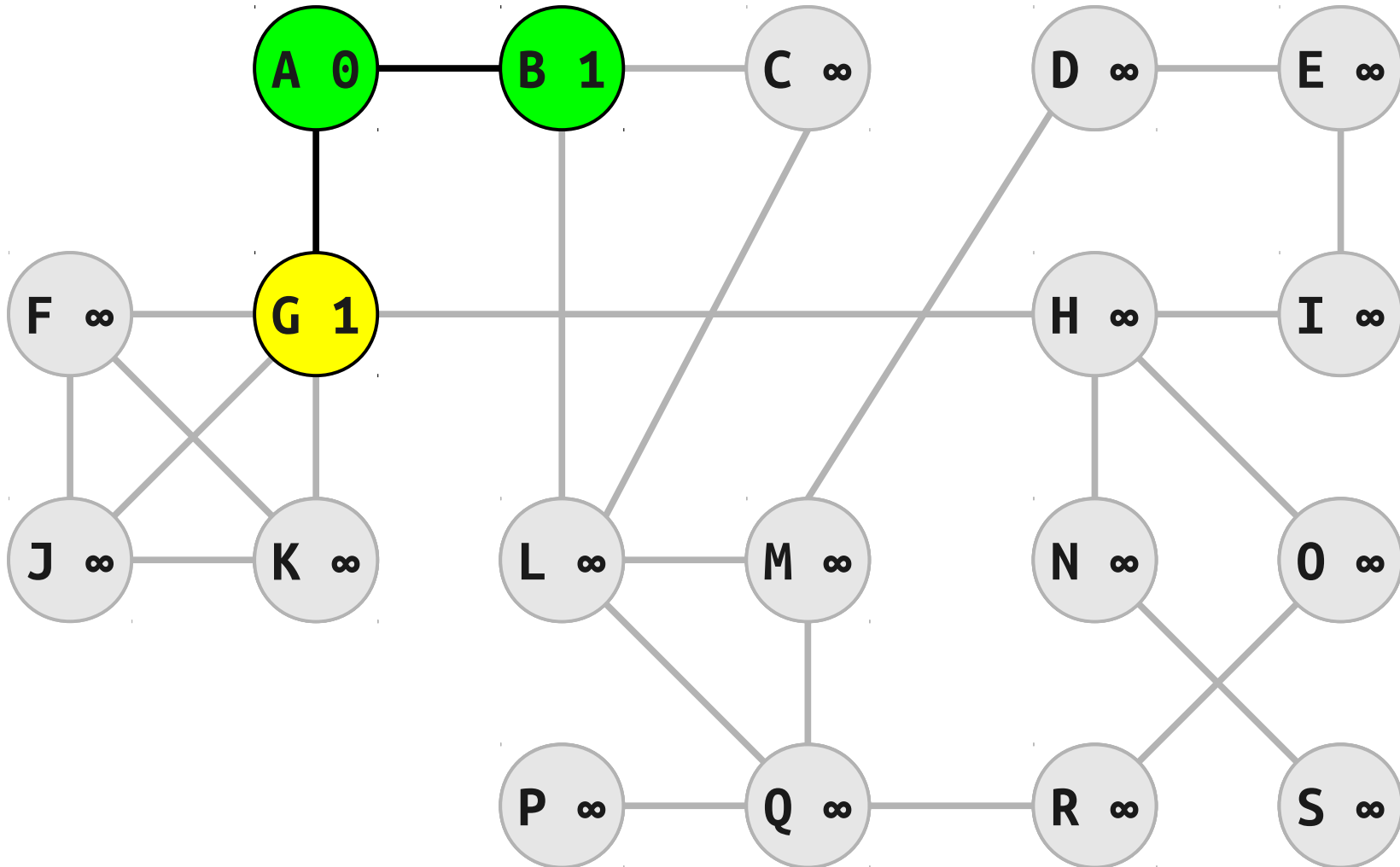
# Breadth-First Search



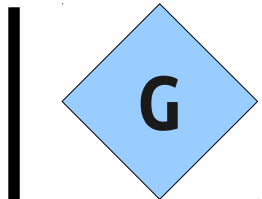
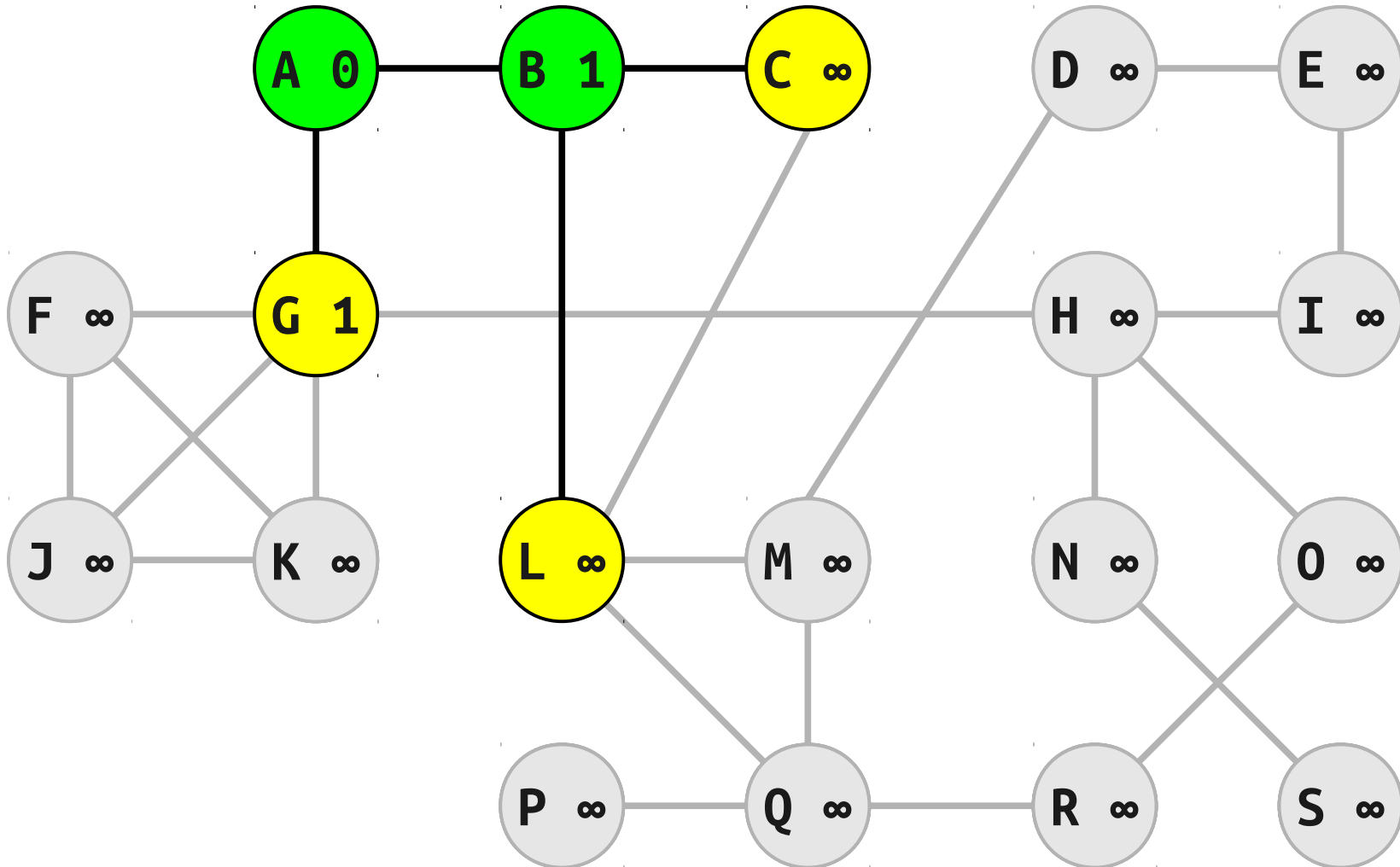
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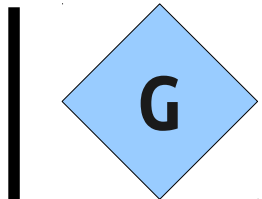
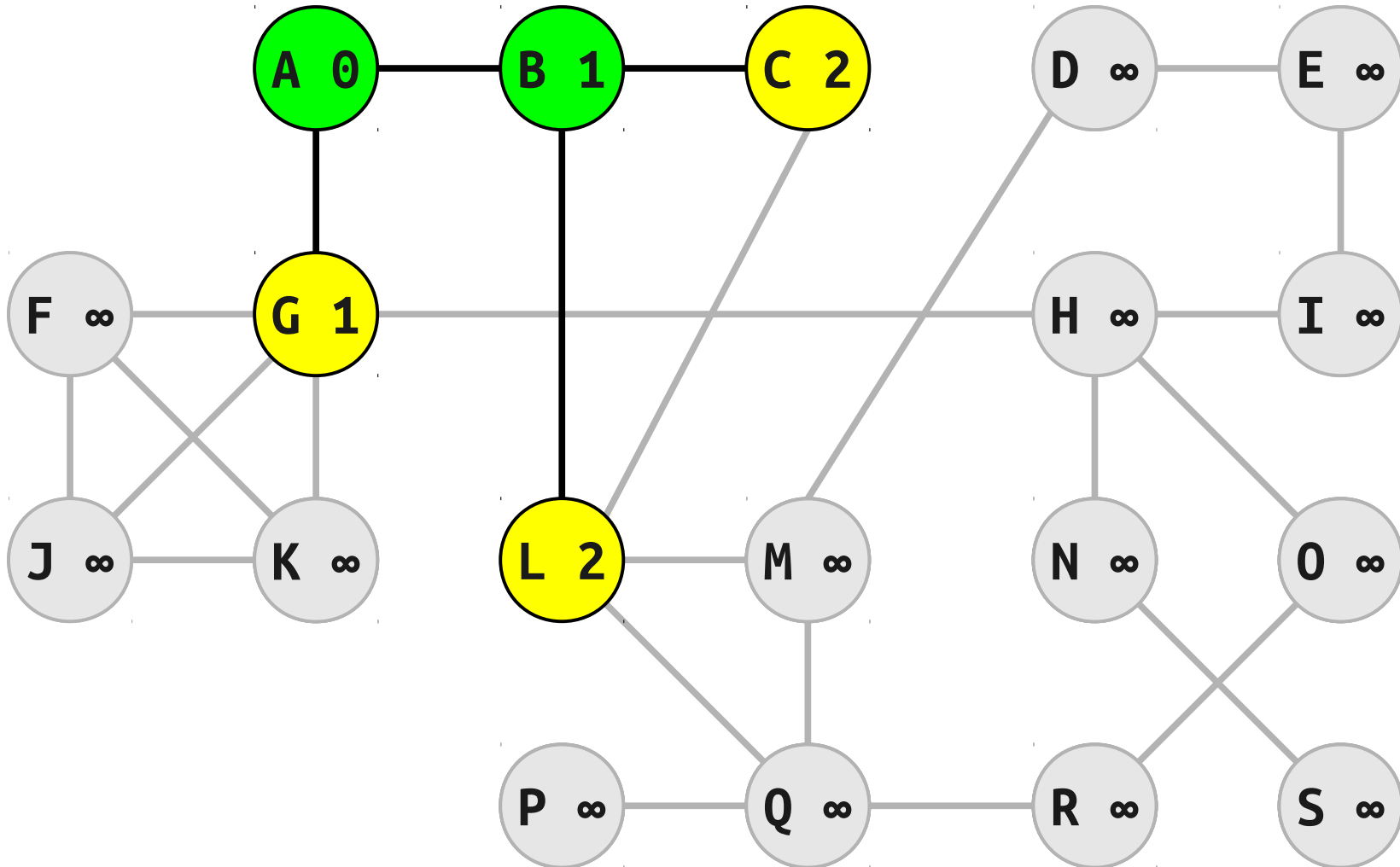
# Breadth-First Search



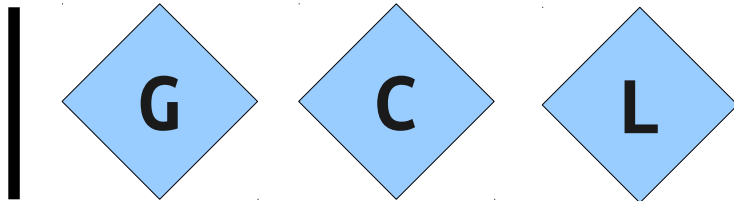
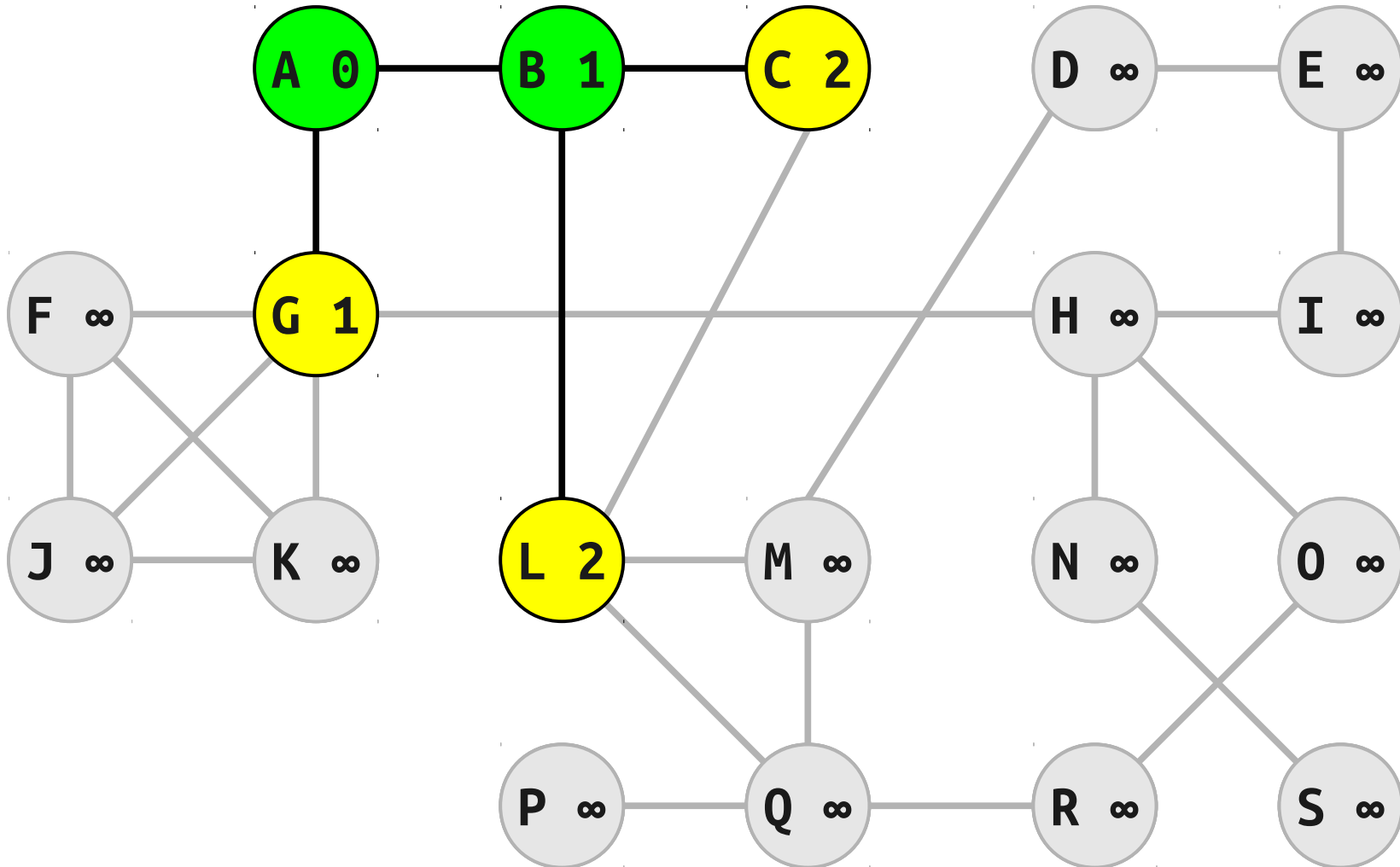
# Breadth-First Search



# Breadth-First Search

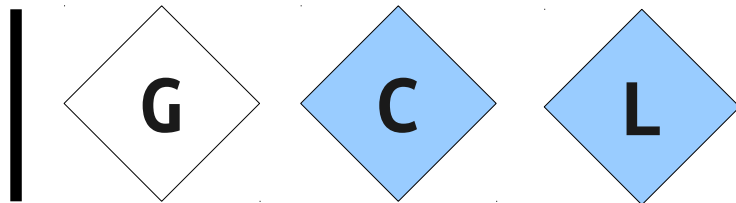
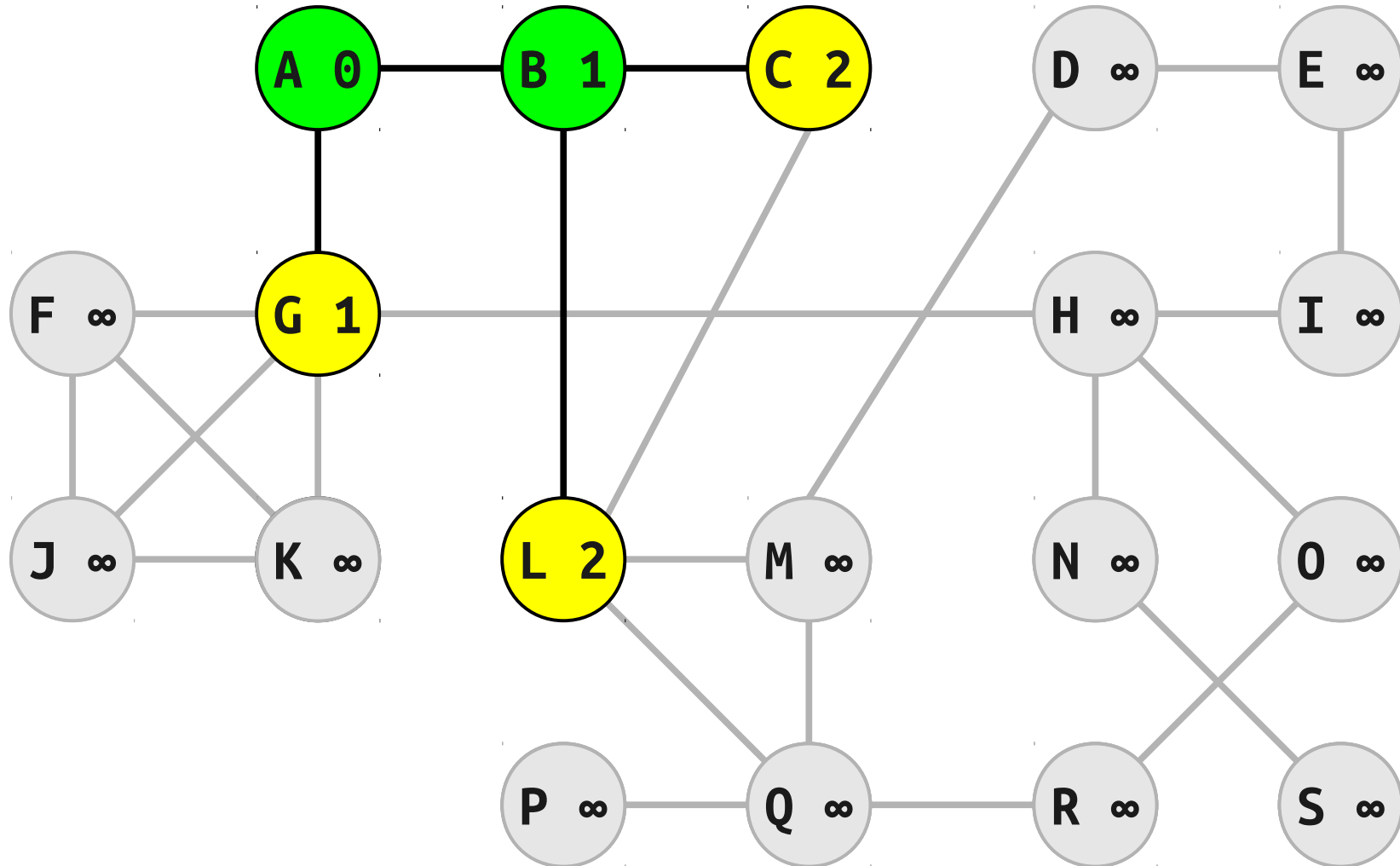


# Breadth-First Search

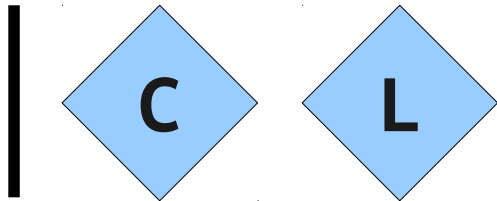
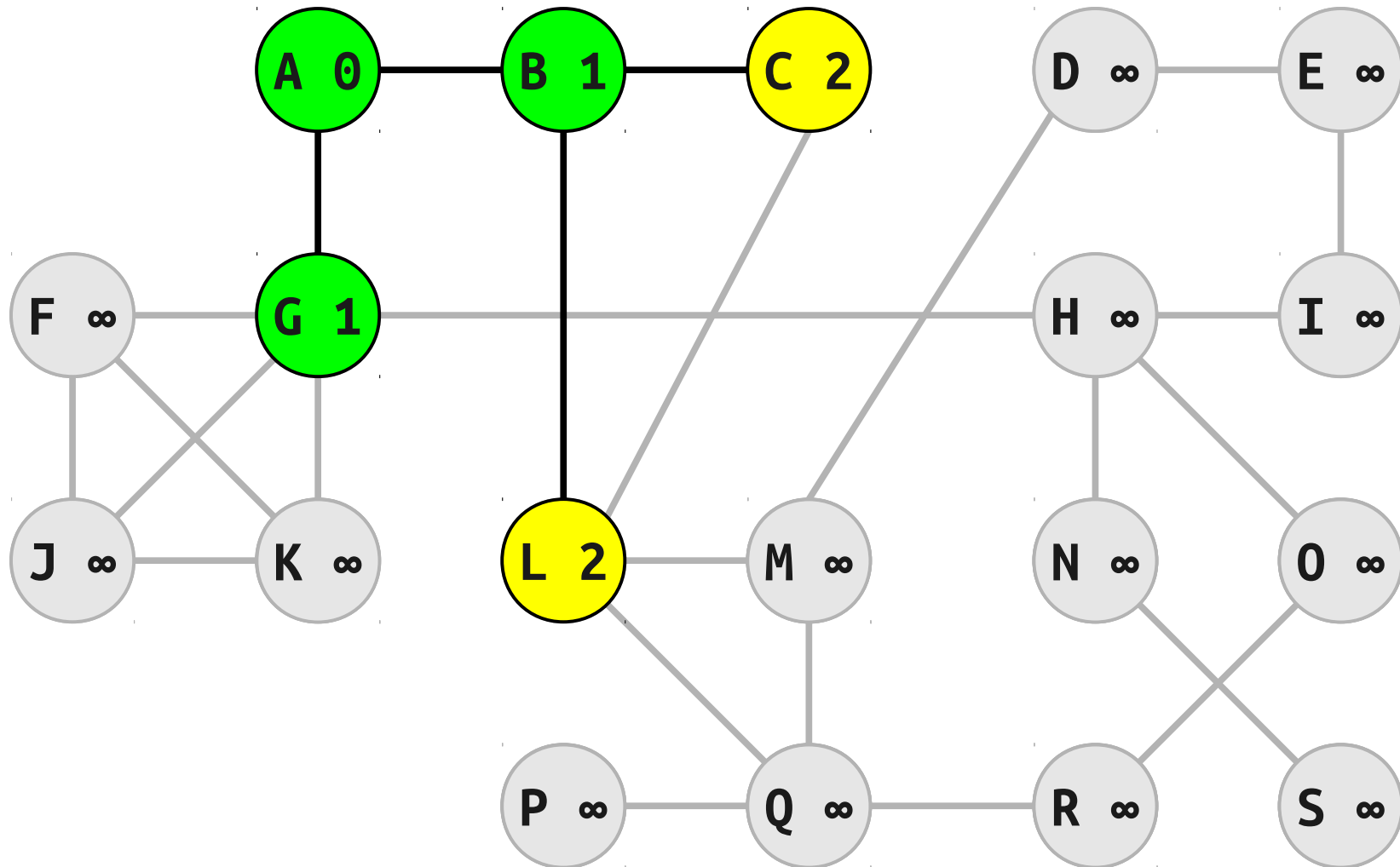




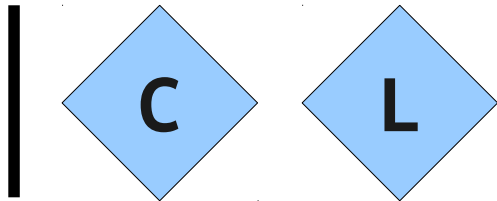
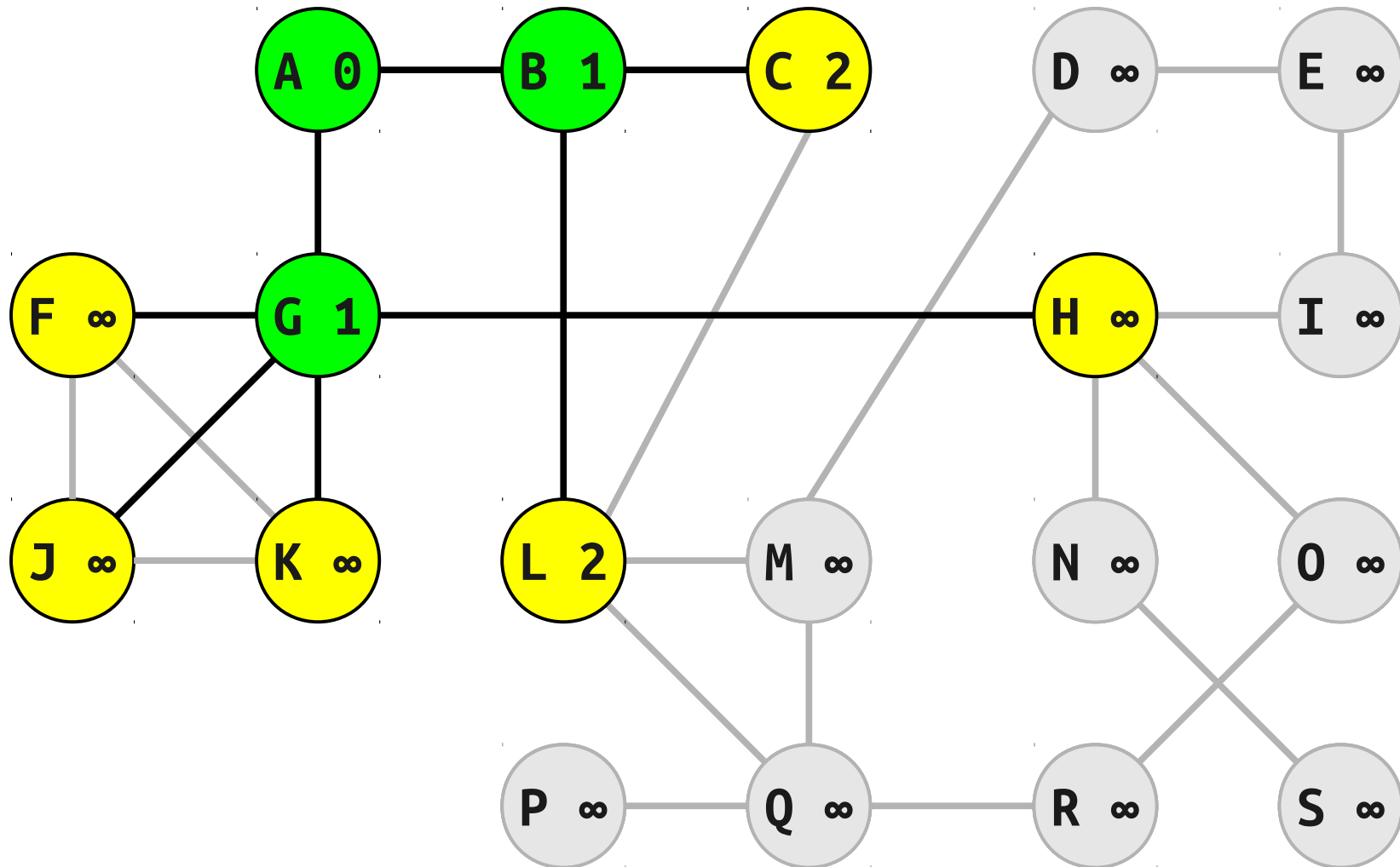
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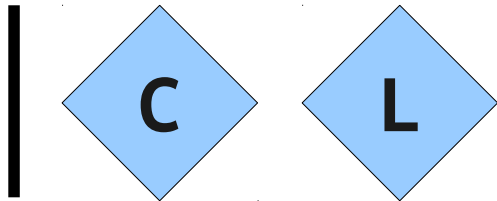
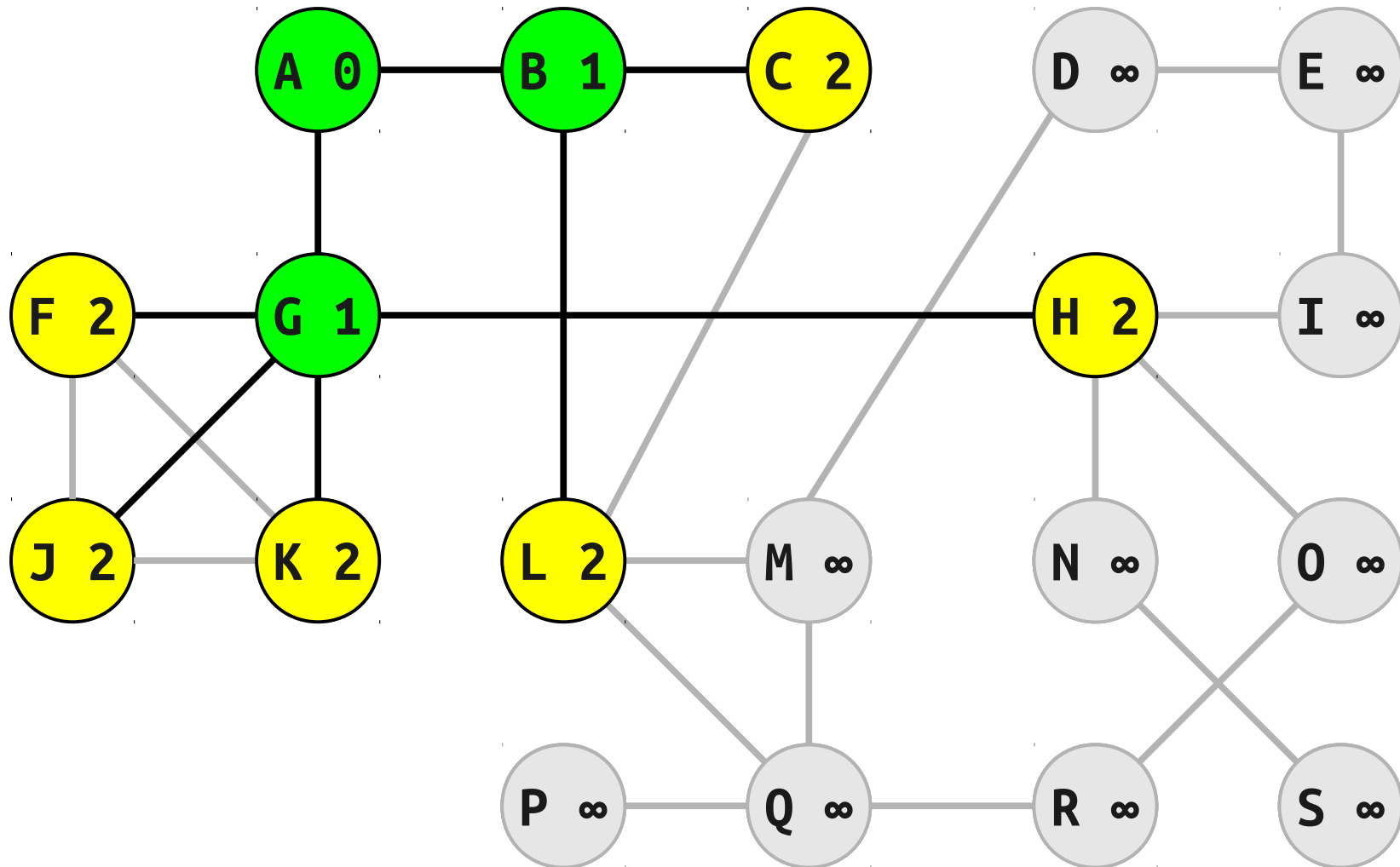
# Breadth-First Search



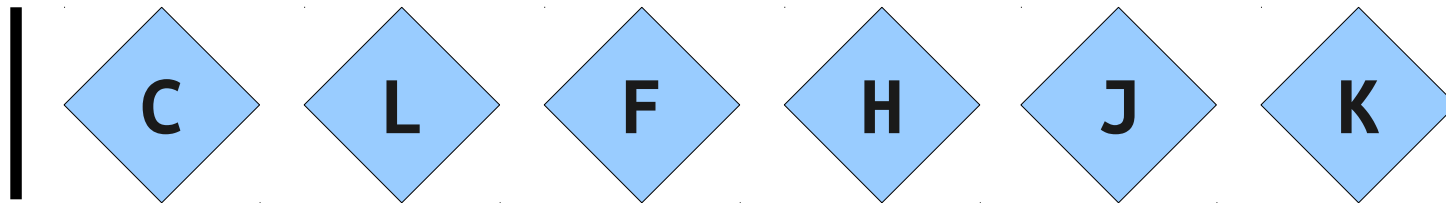
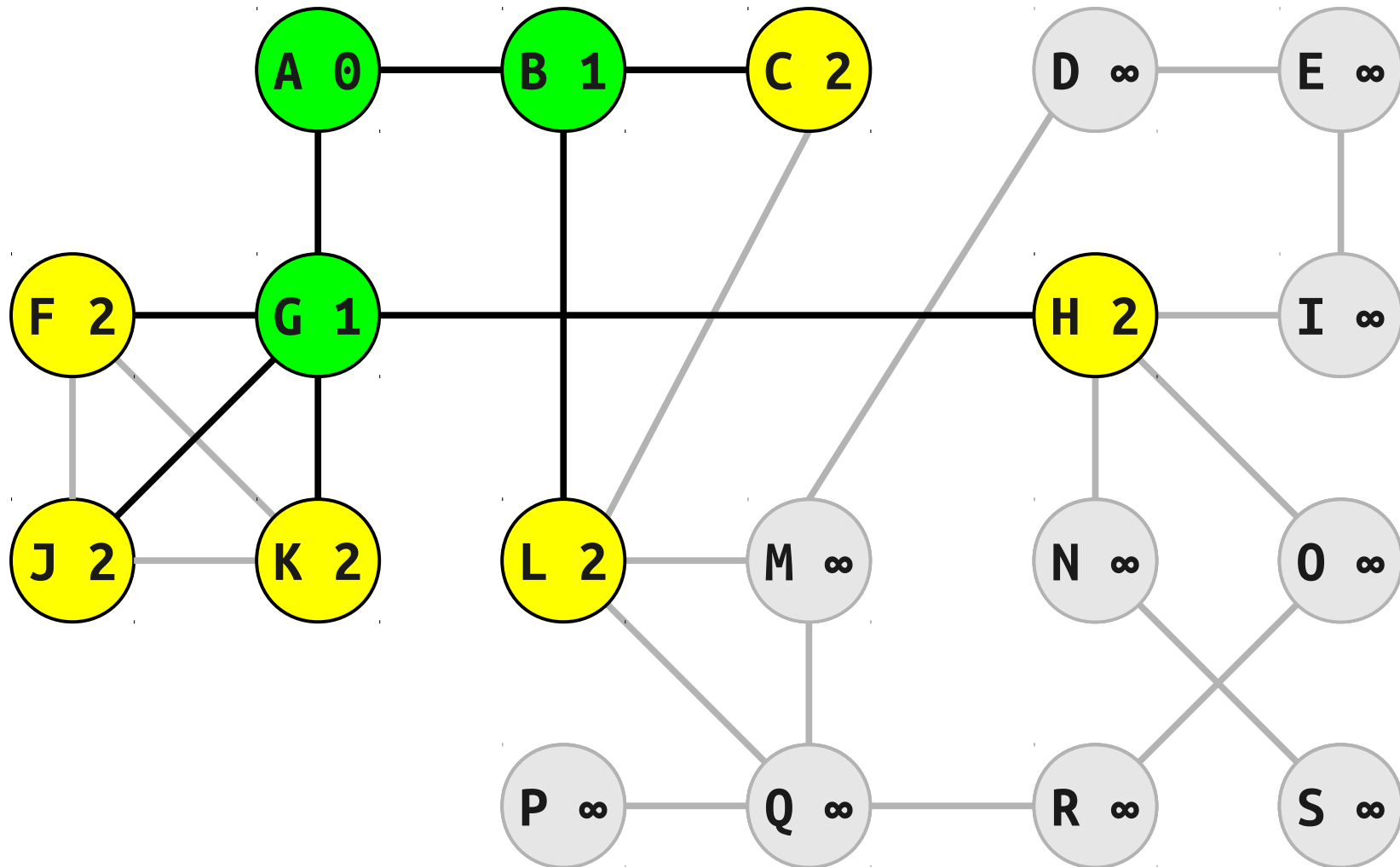
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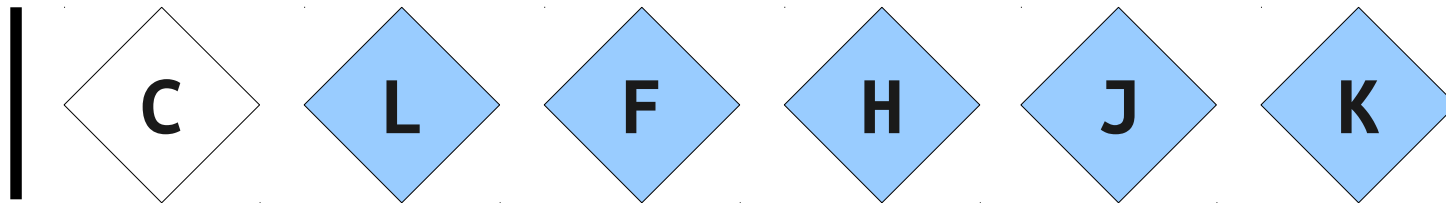
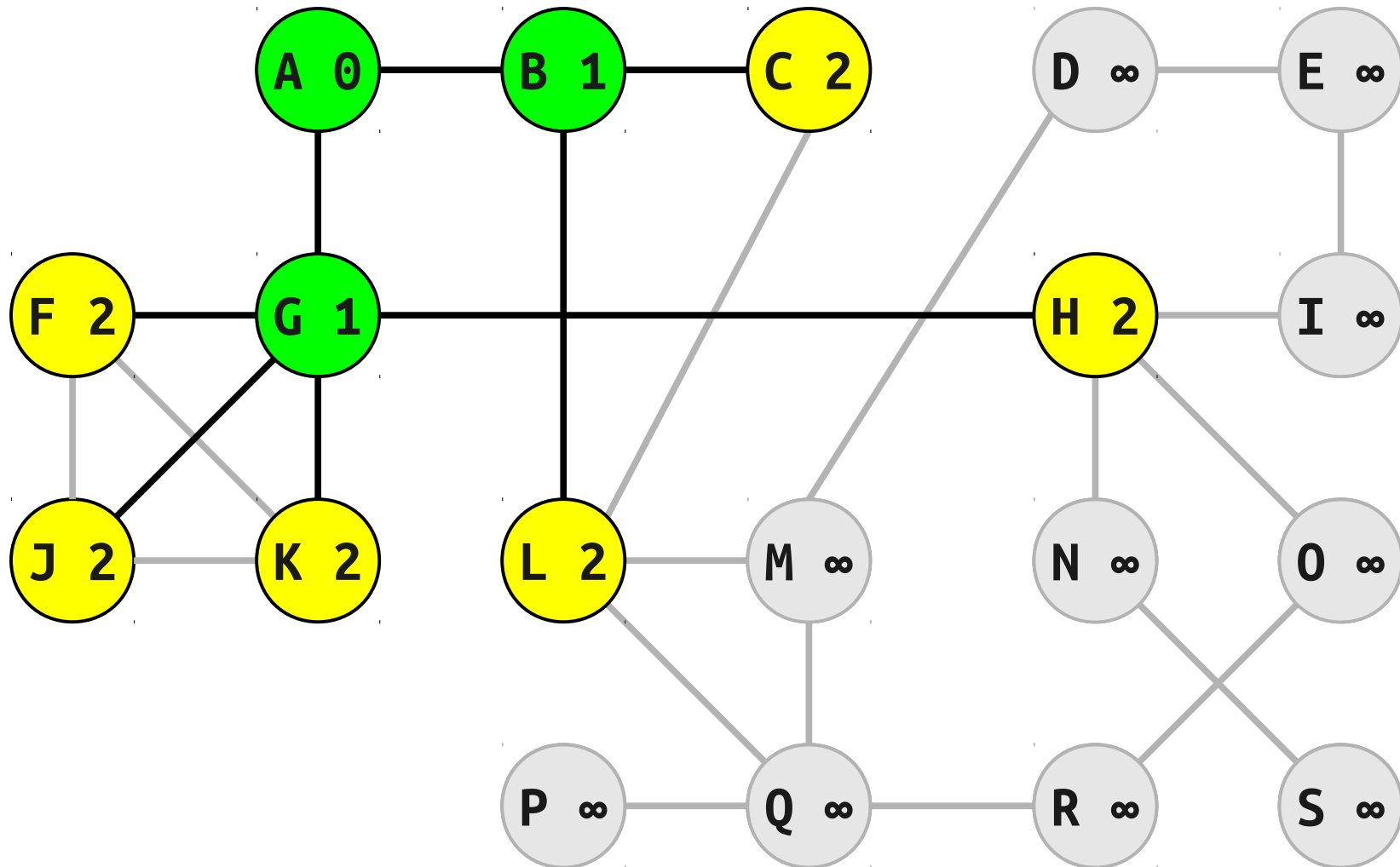
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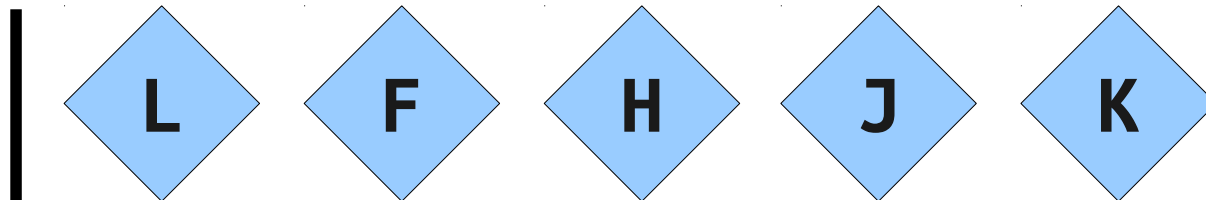
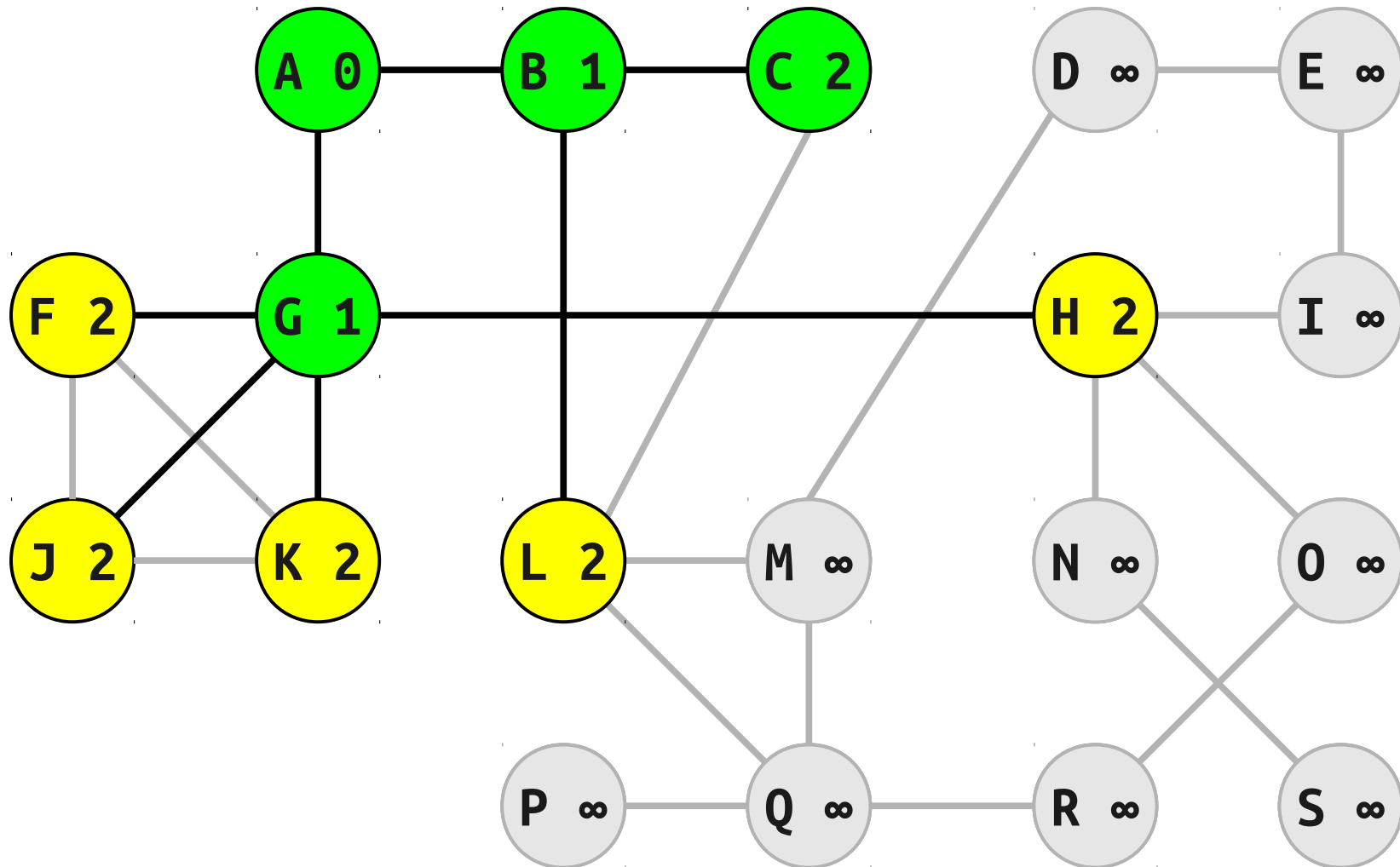
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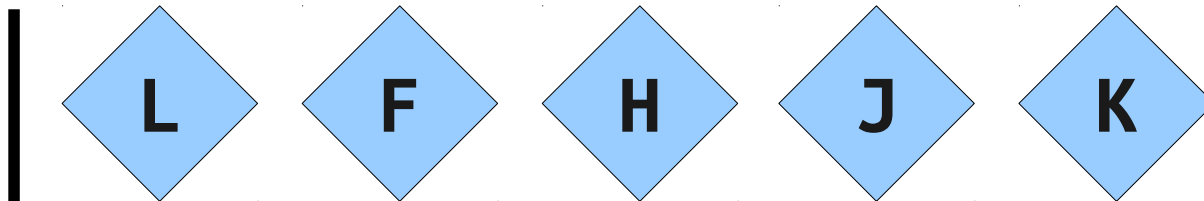
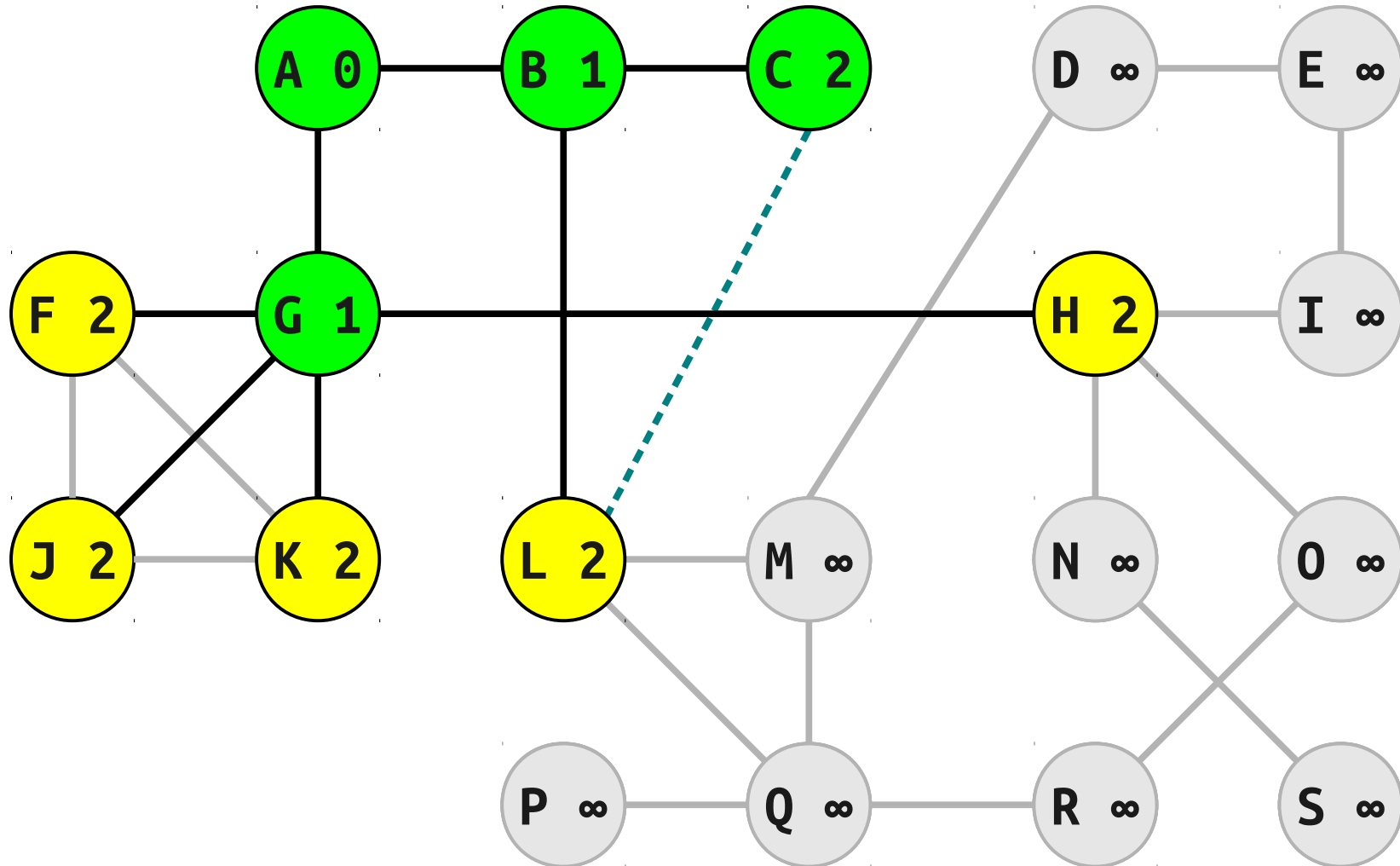
# Breadth-First Search



# Breadth-First Search

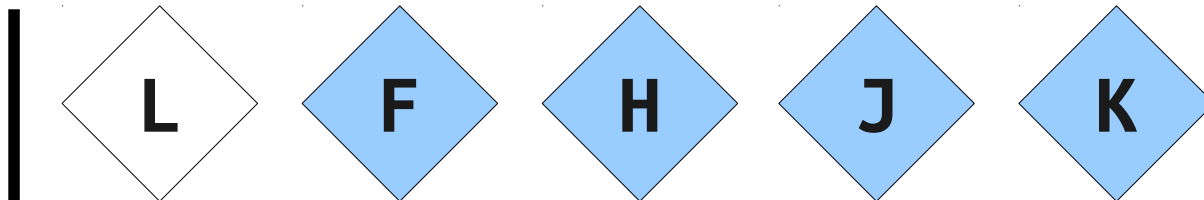
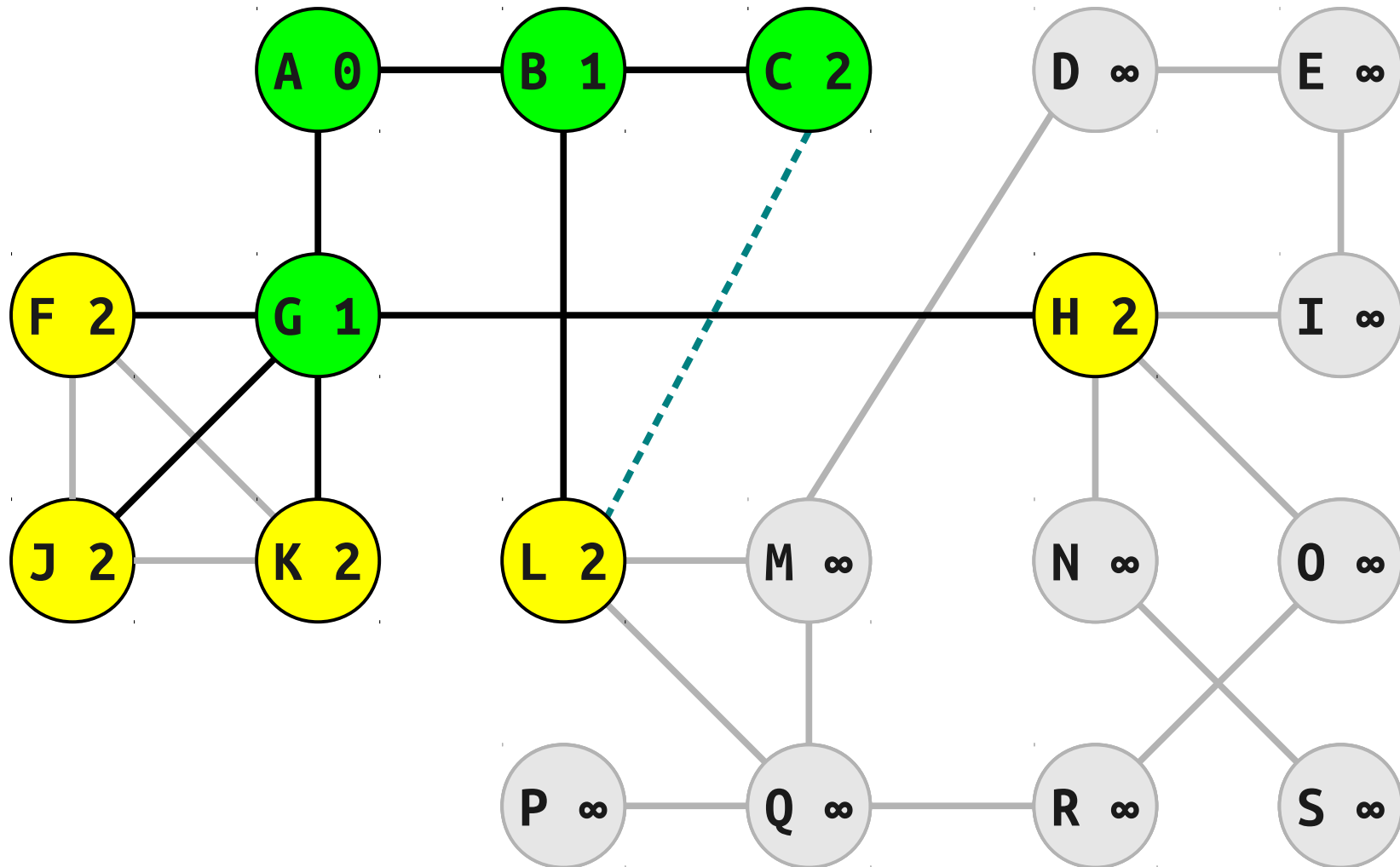


# Breadth-First Search

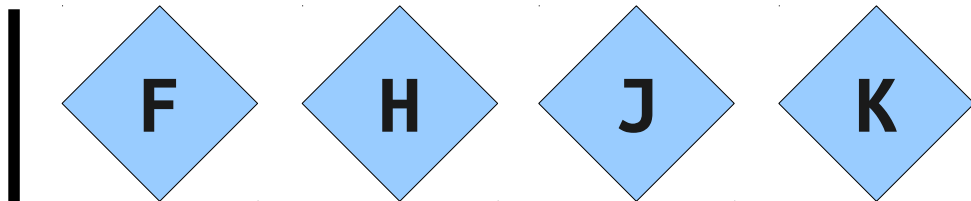
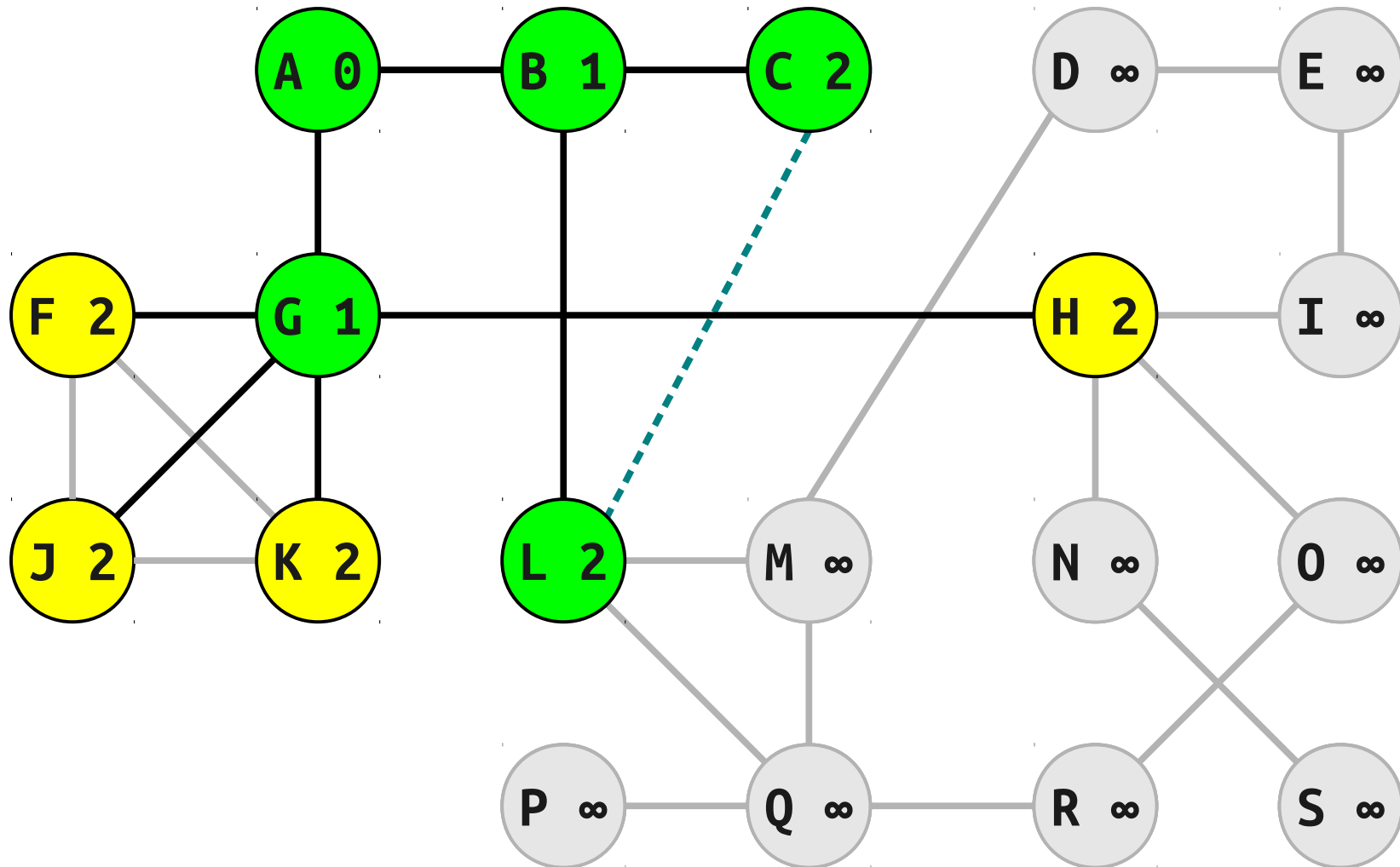




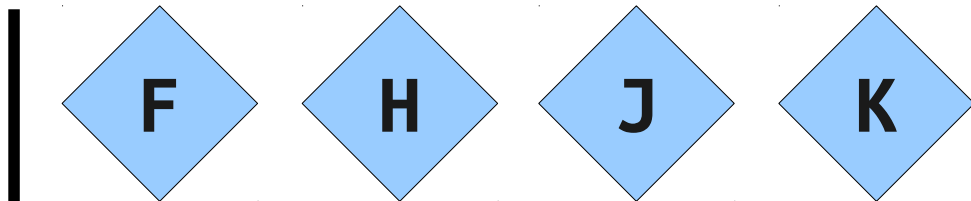
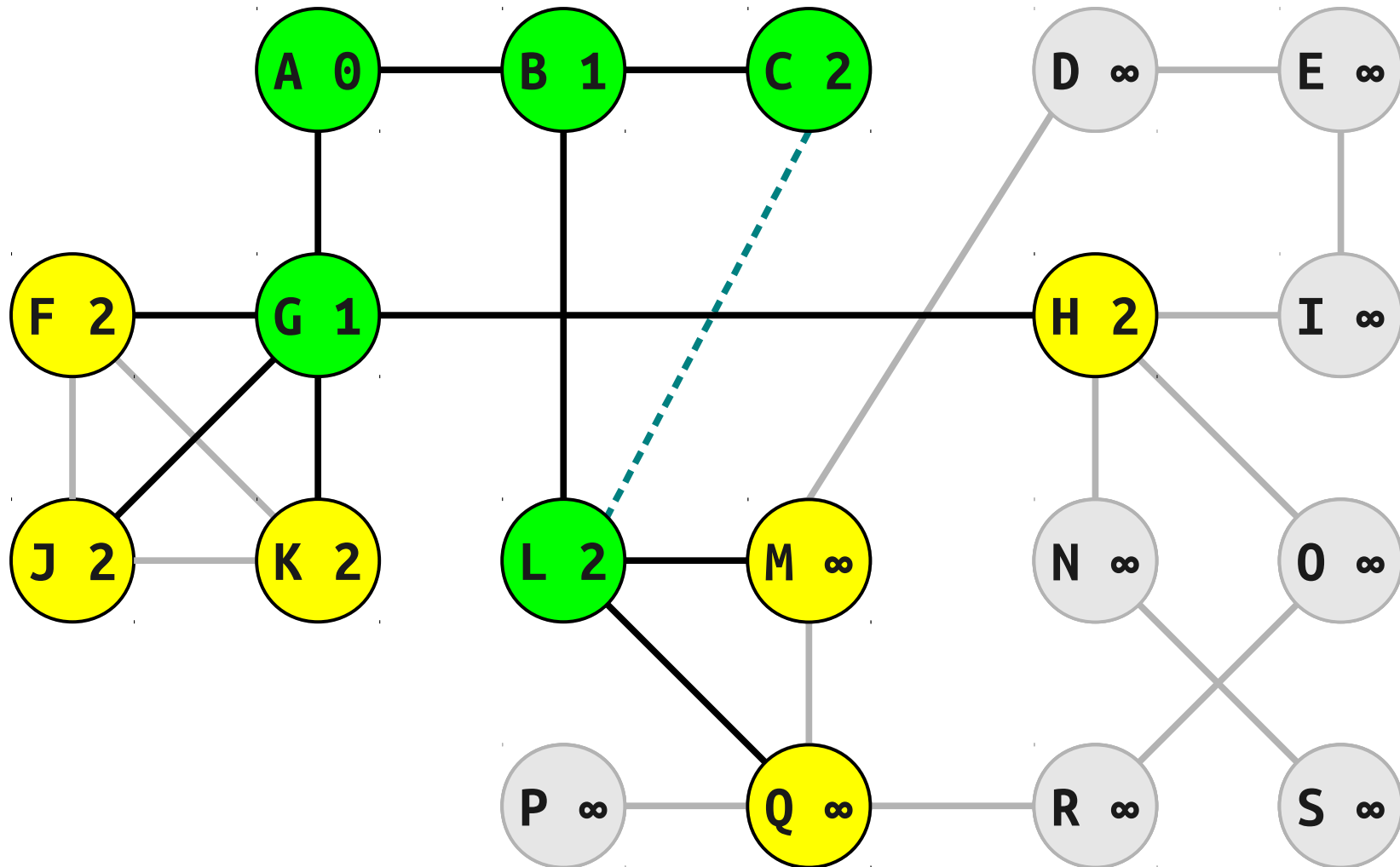
# Breadth-First Search



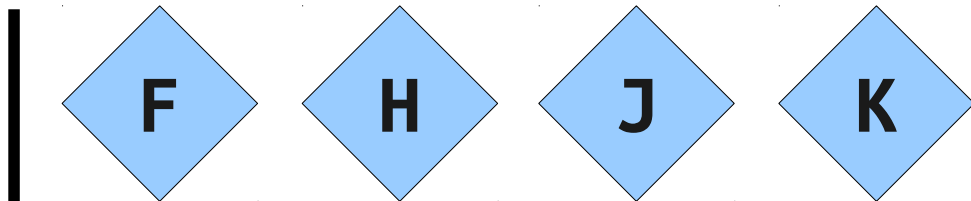
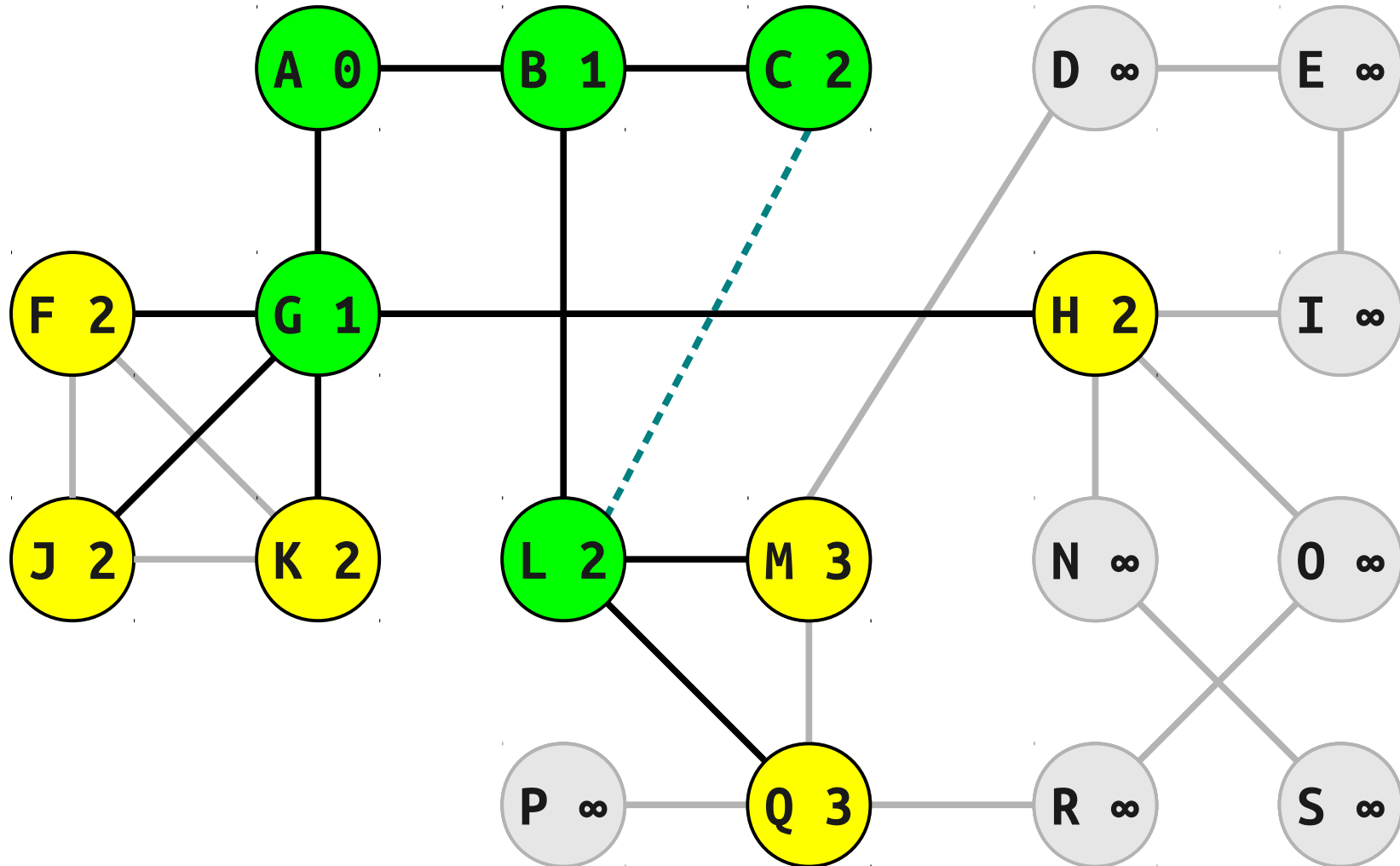
# Breadth-First Search



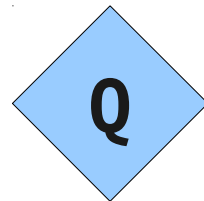
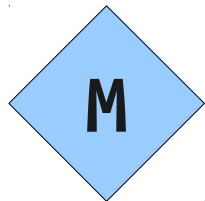
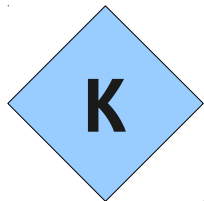
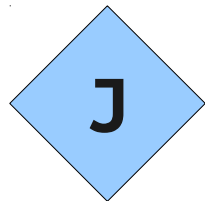
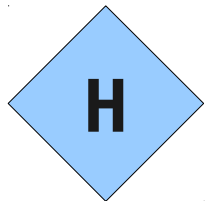
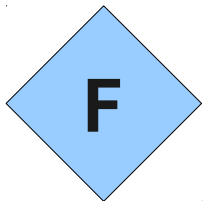
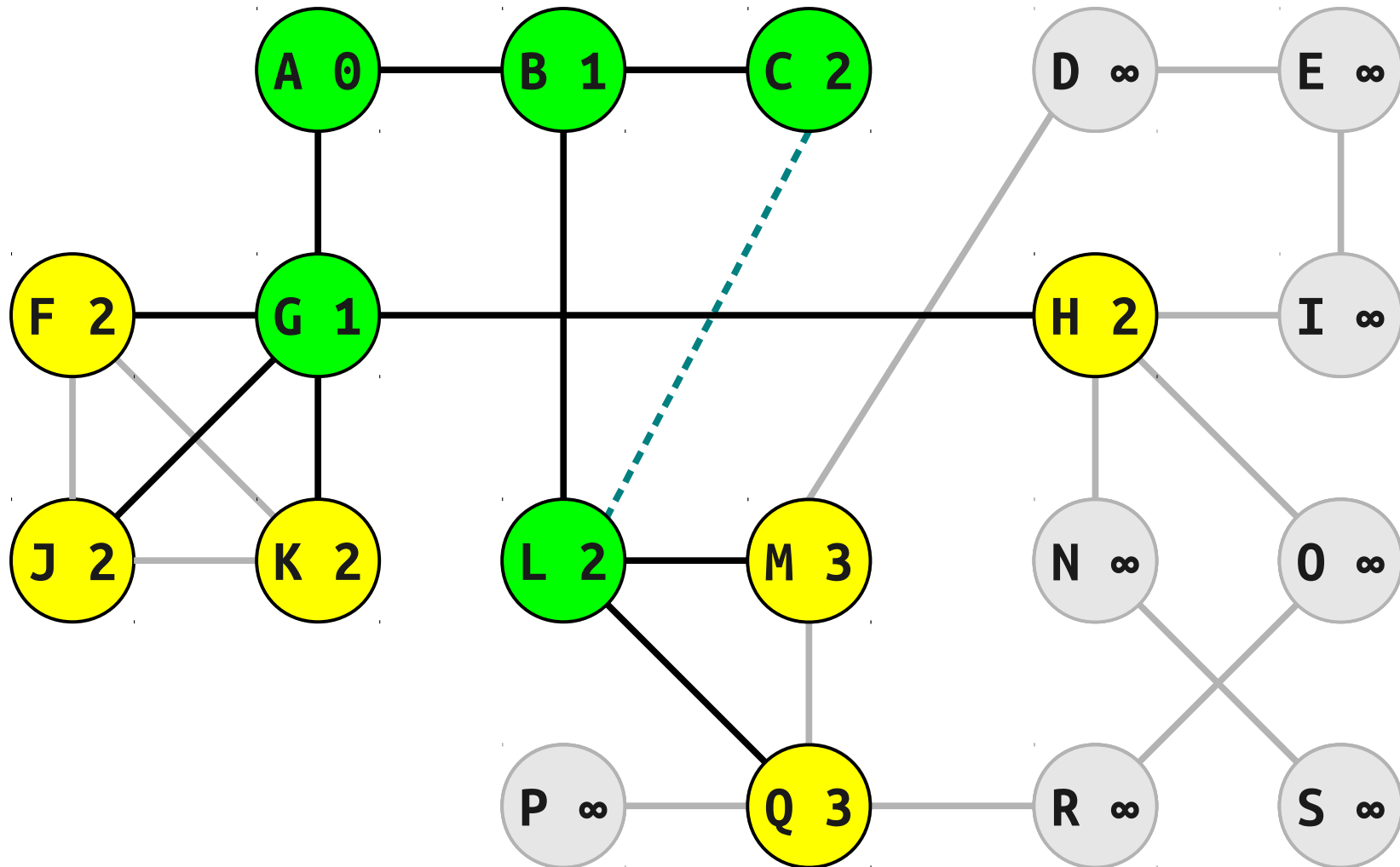
# Breadth-First Search



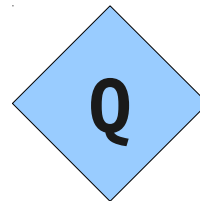
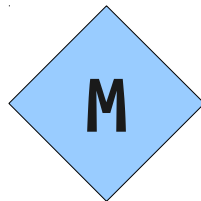
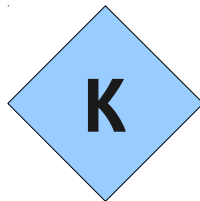
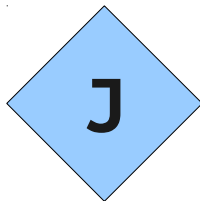
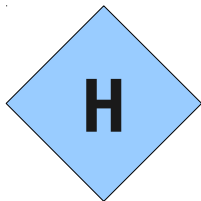
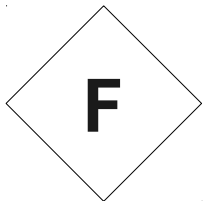
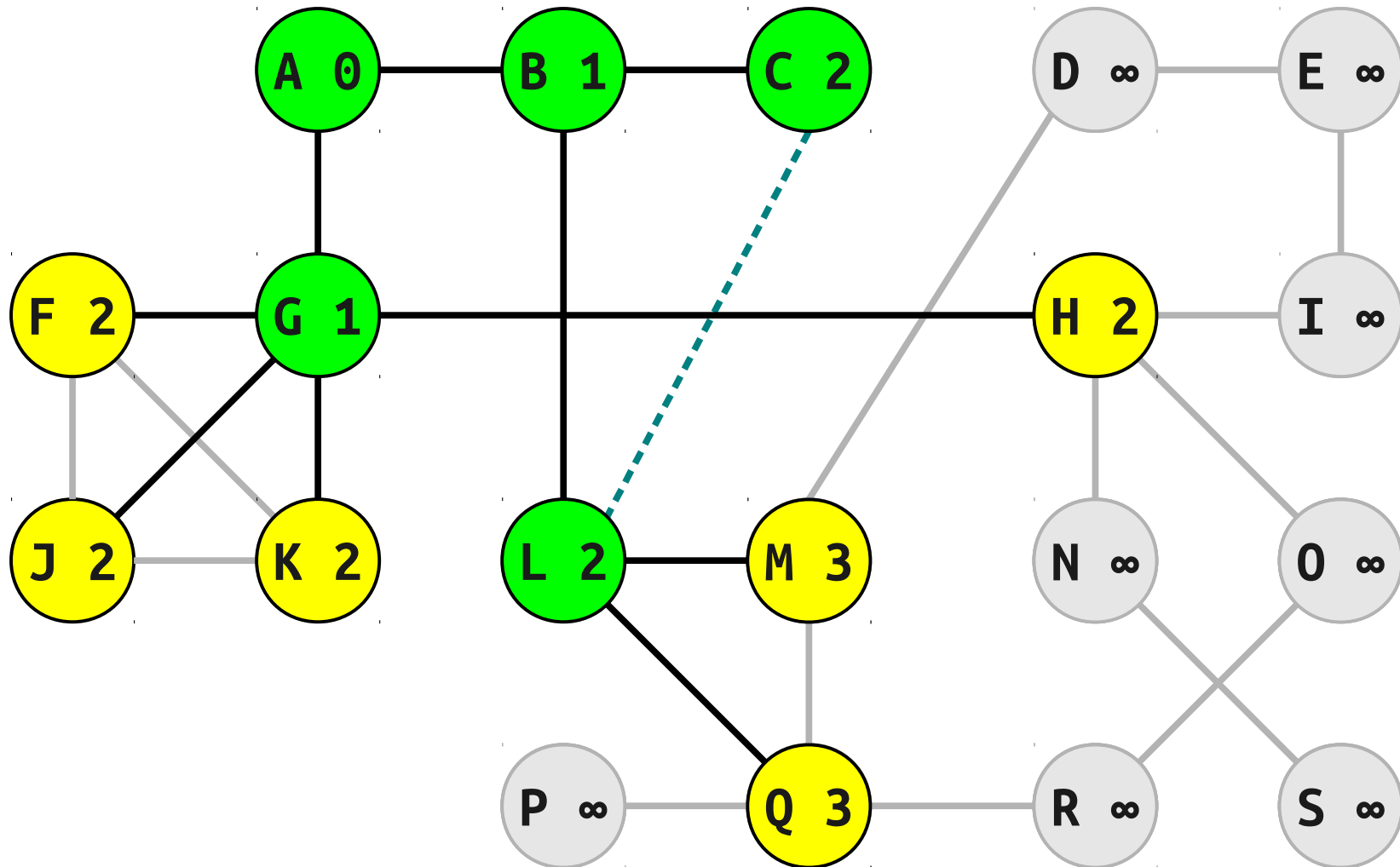
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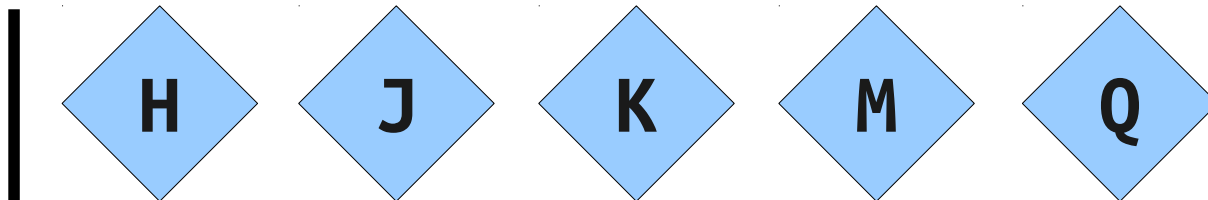
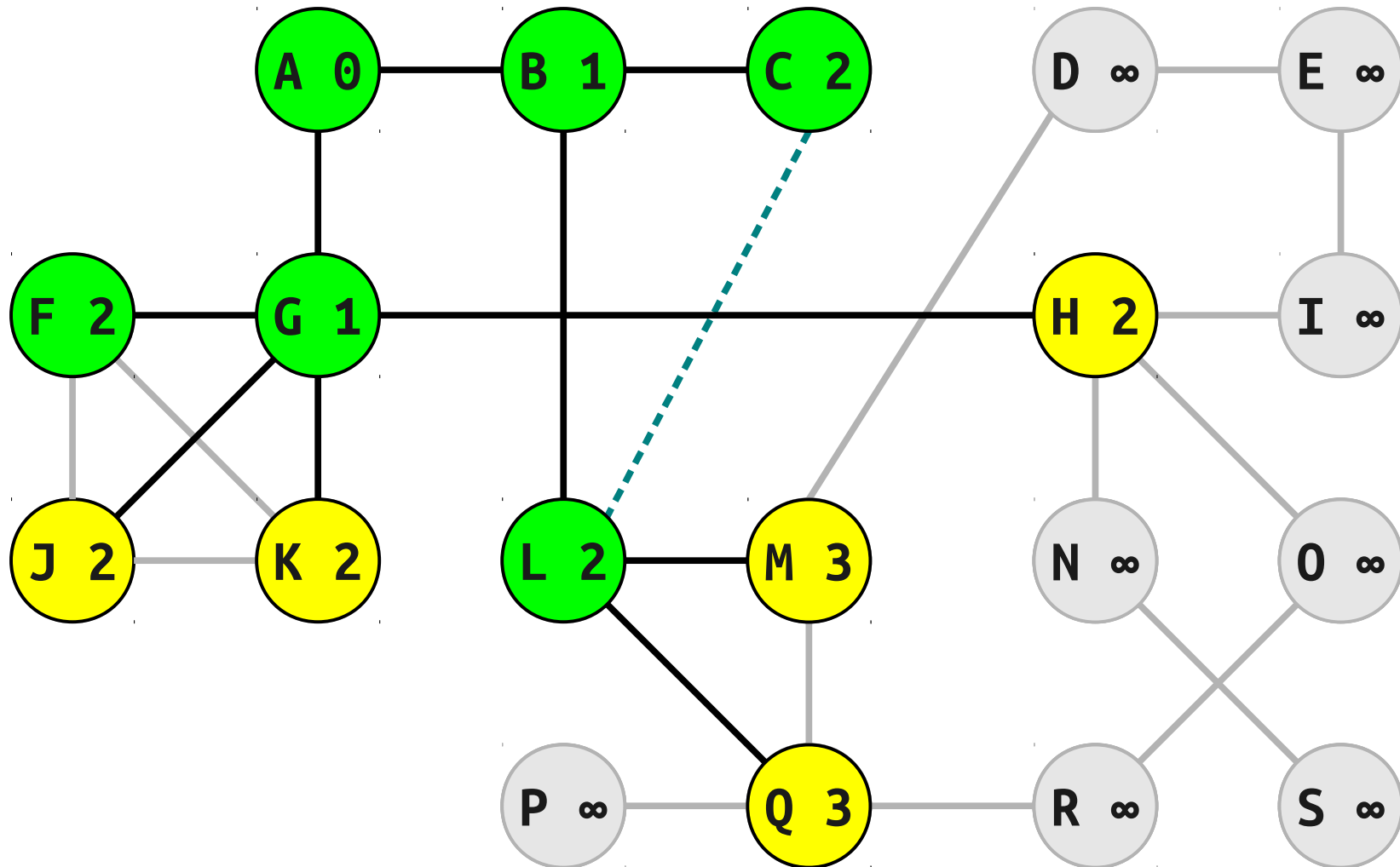
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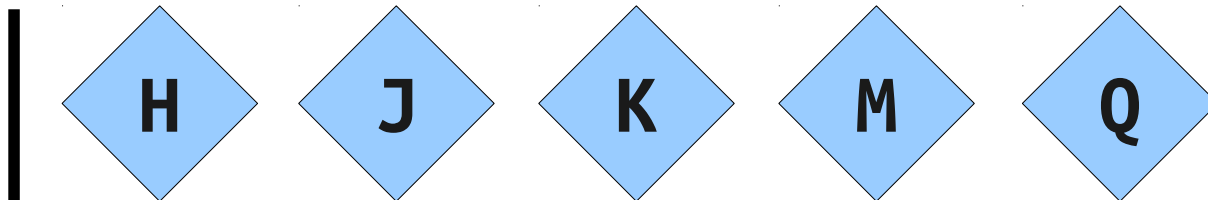
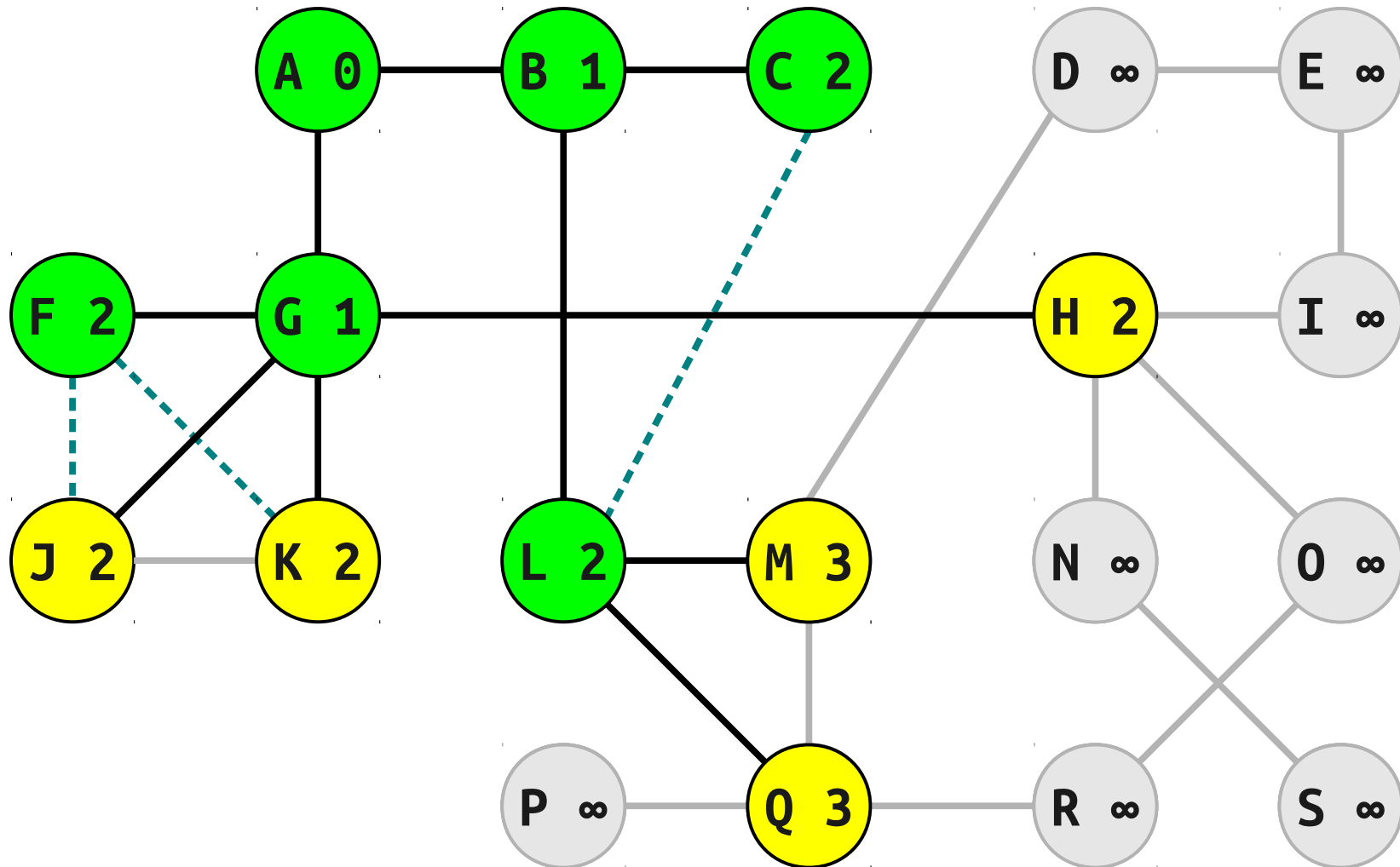
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# Breadth-First Search

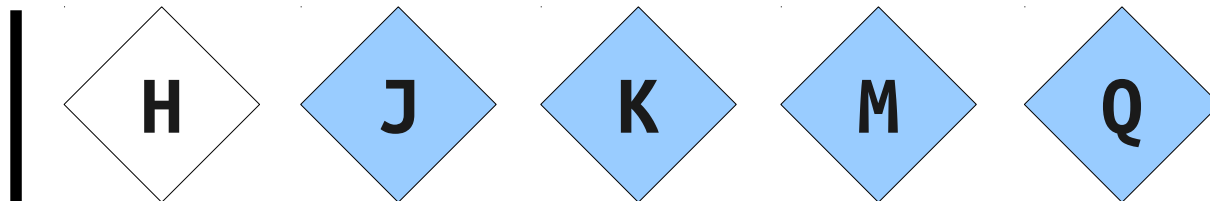
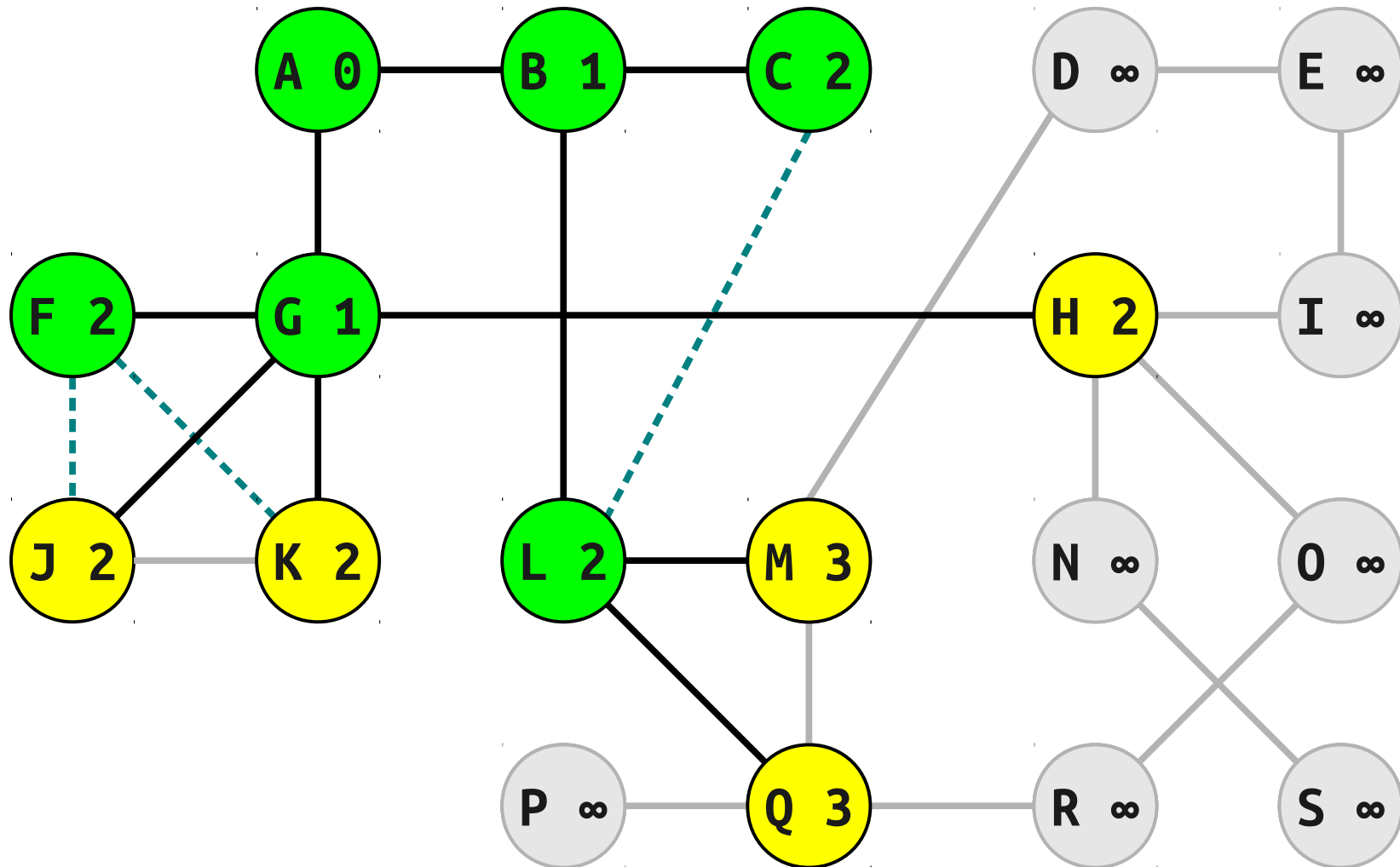


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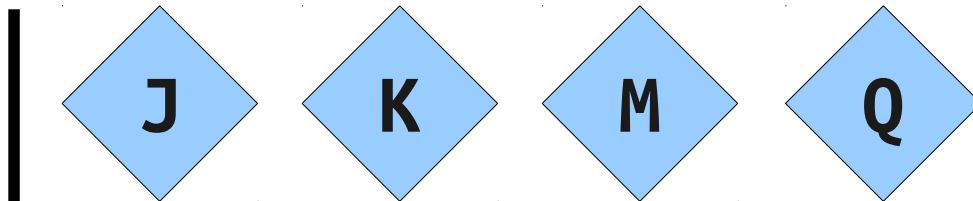
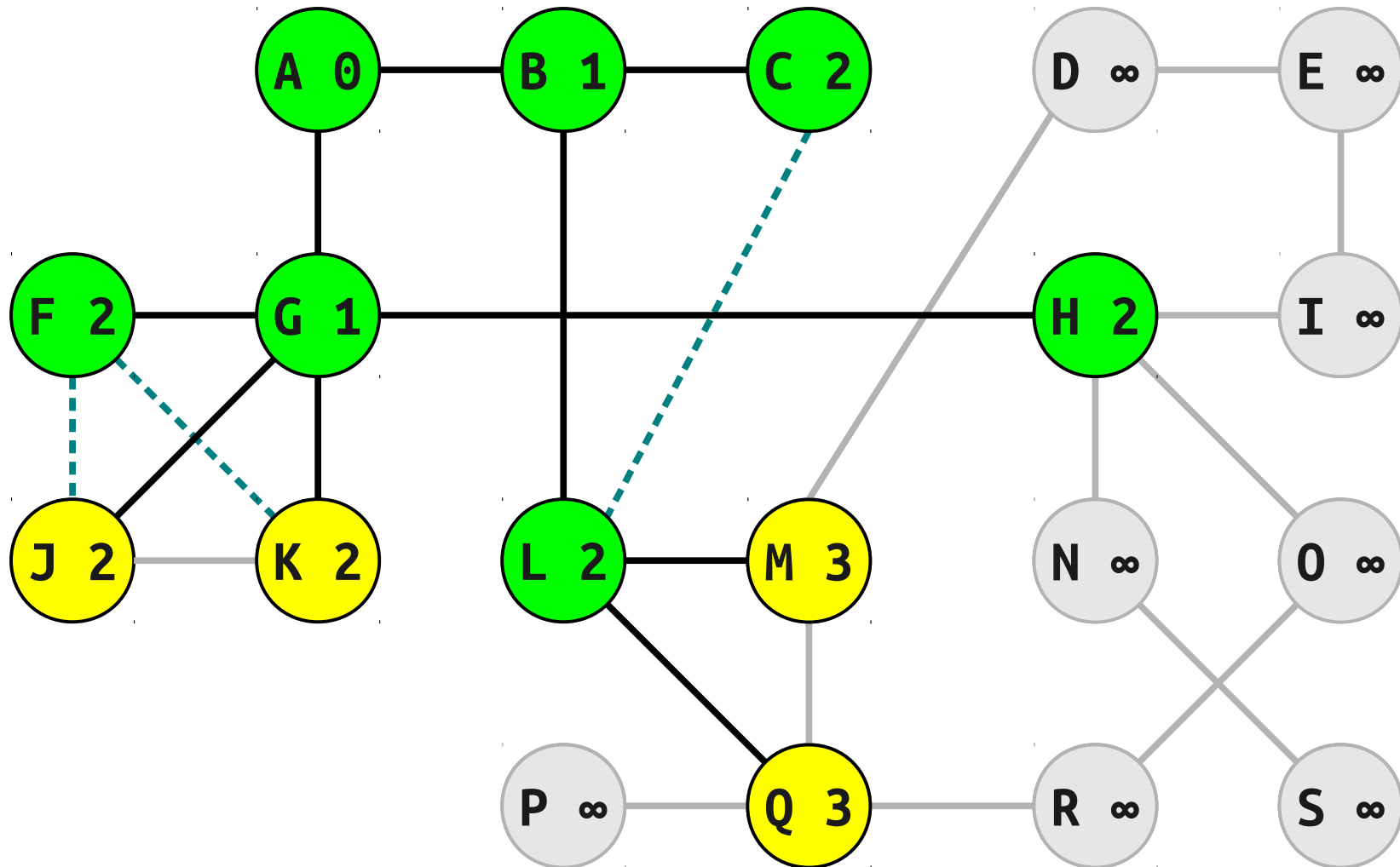




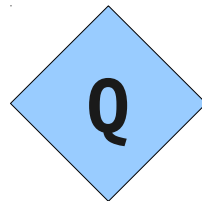
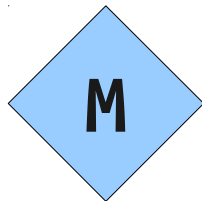
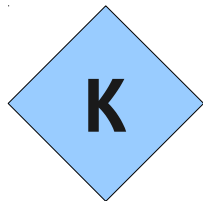
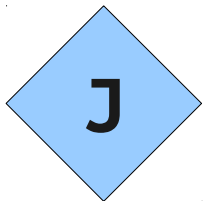
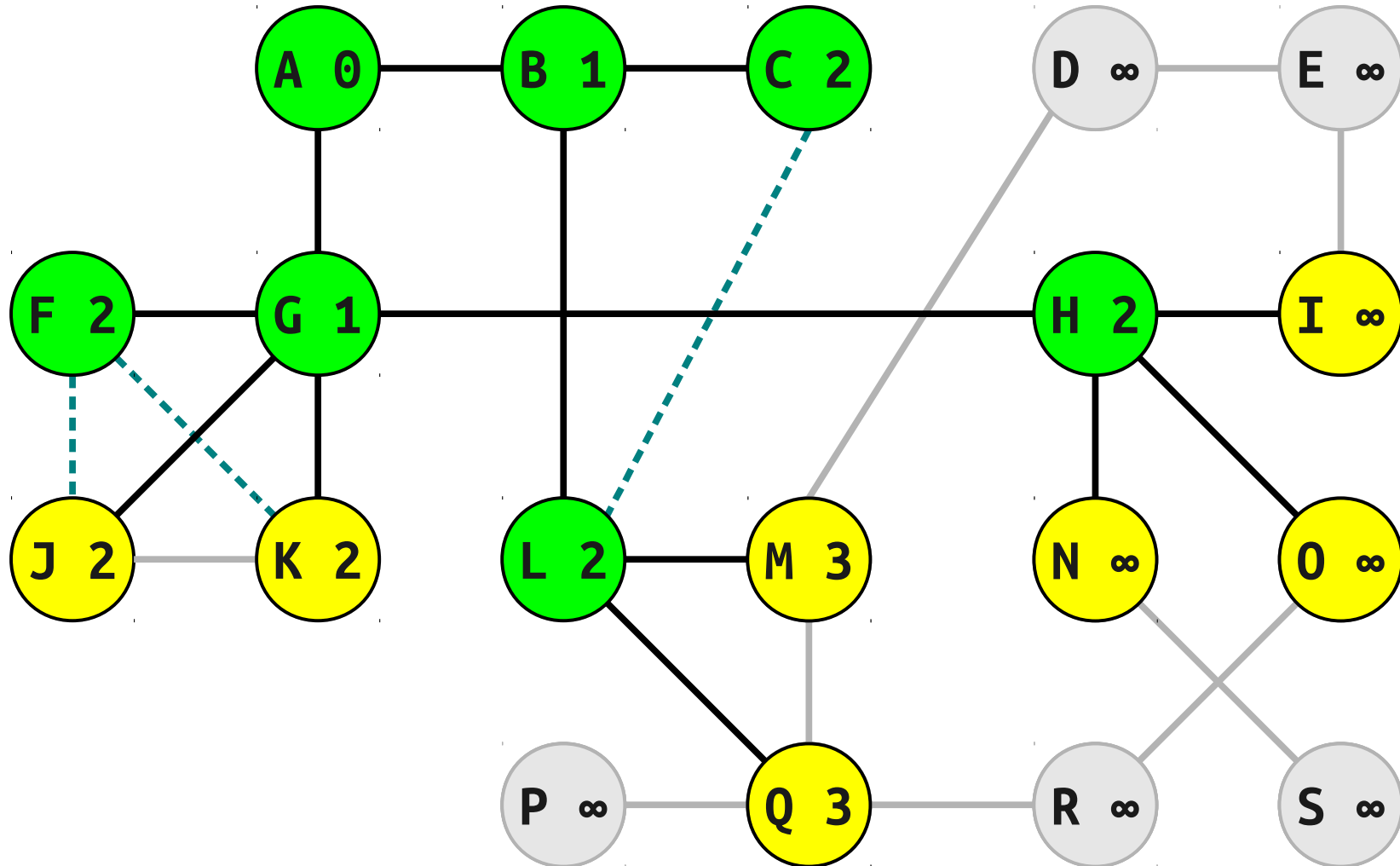
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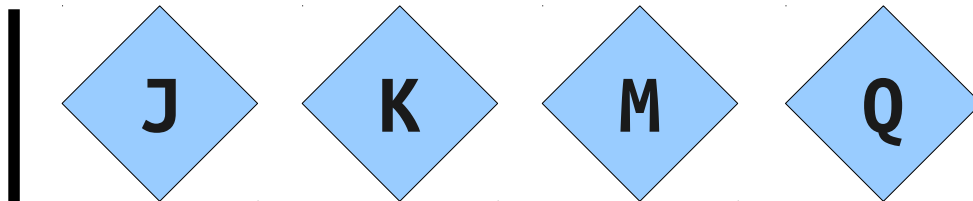
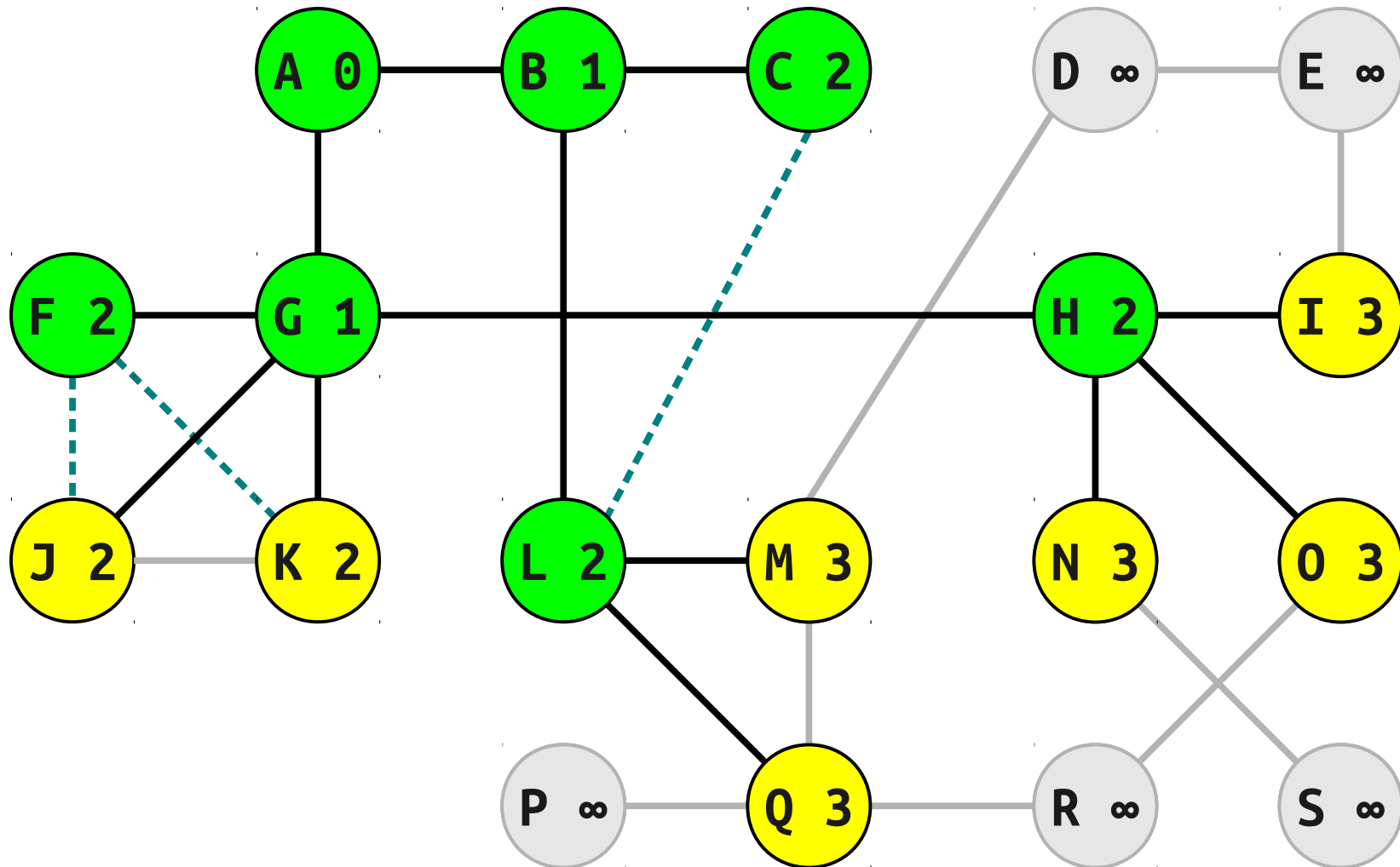
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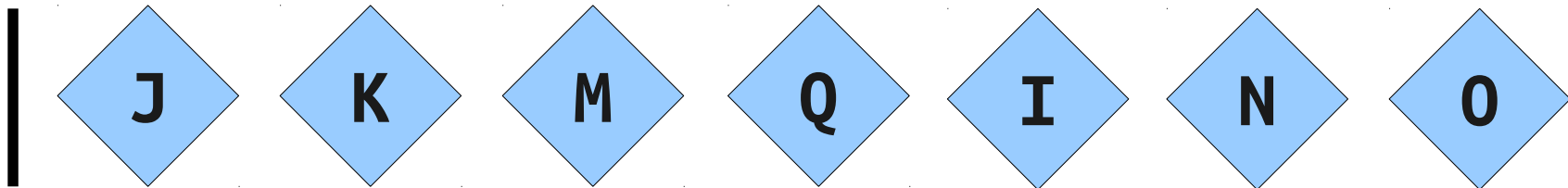
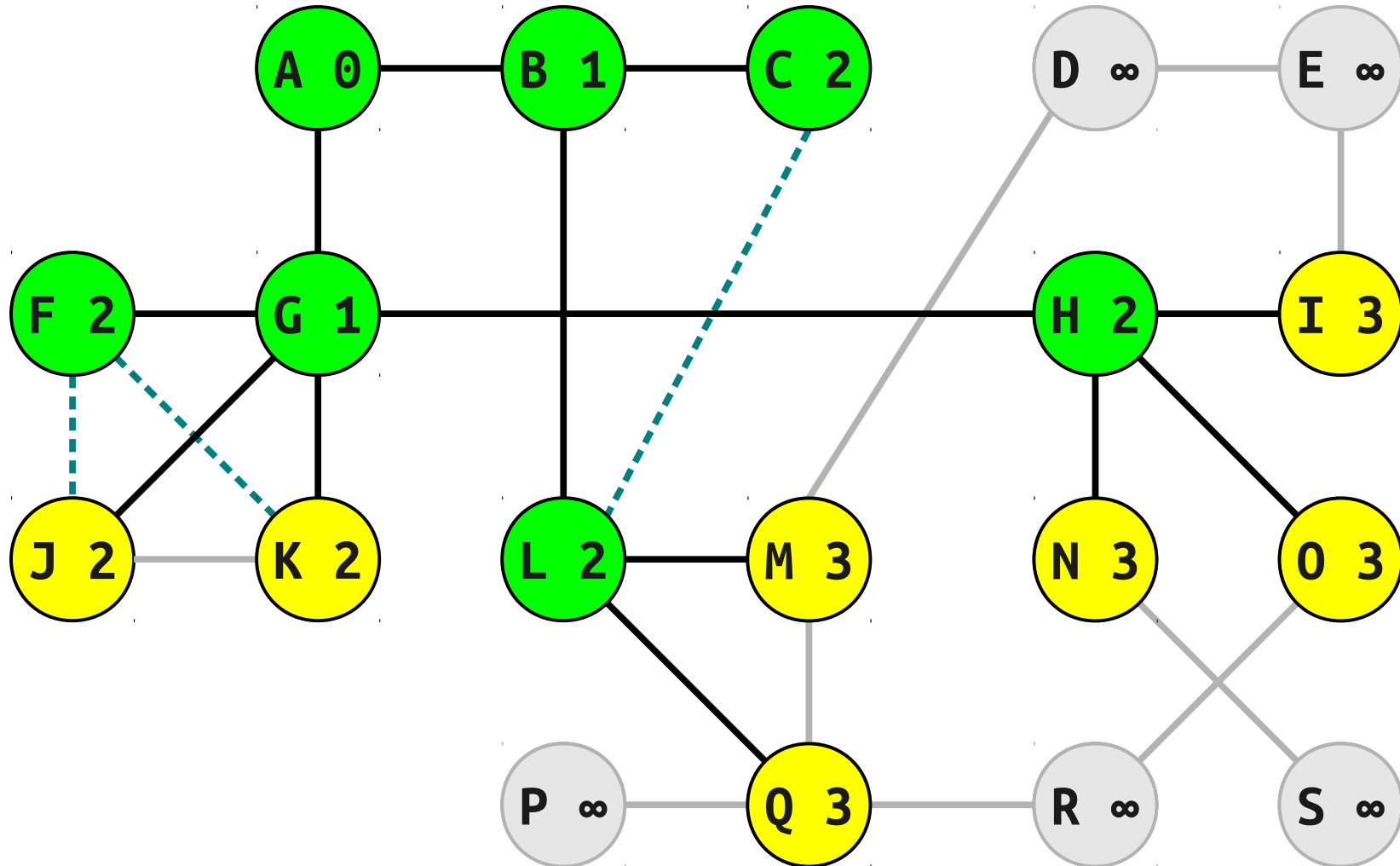
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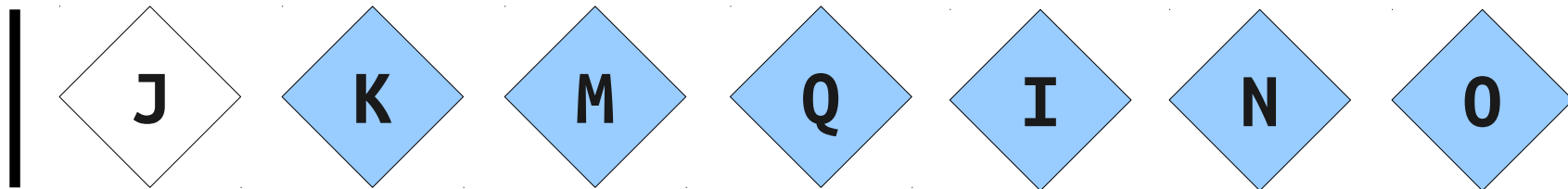
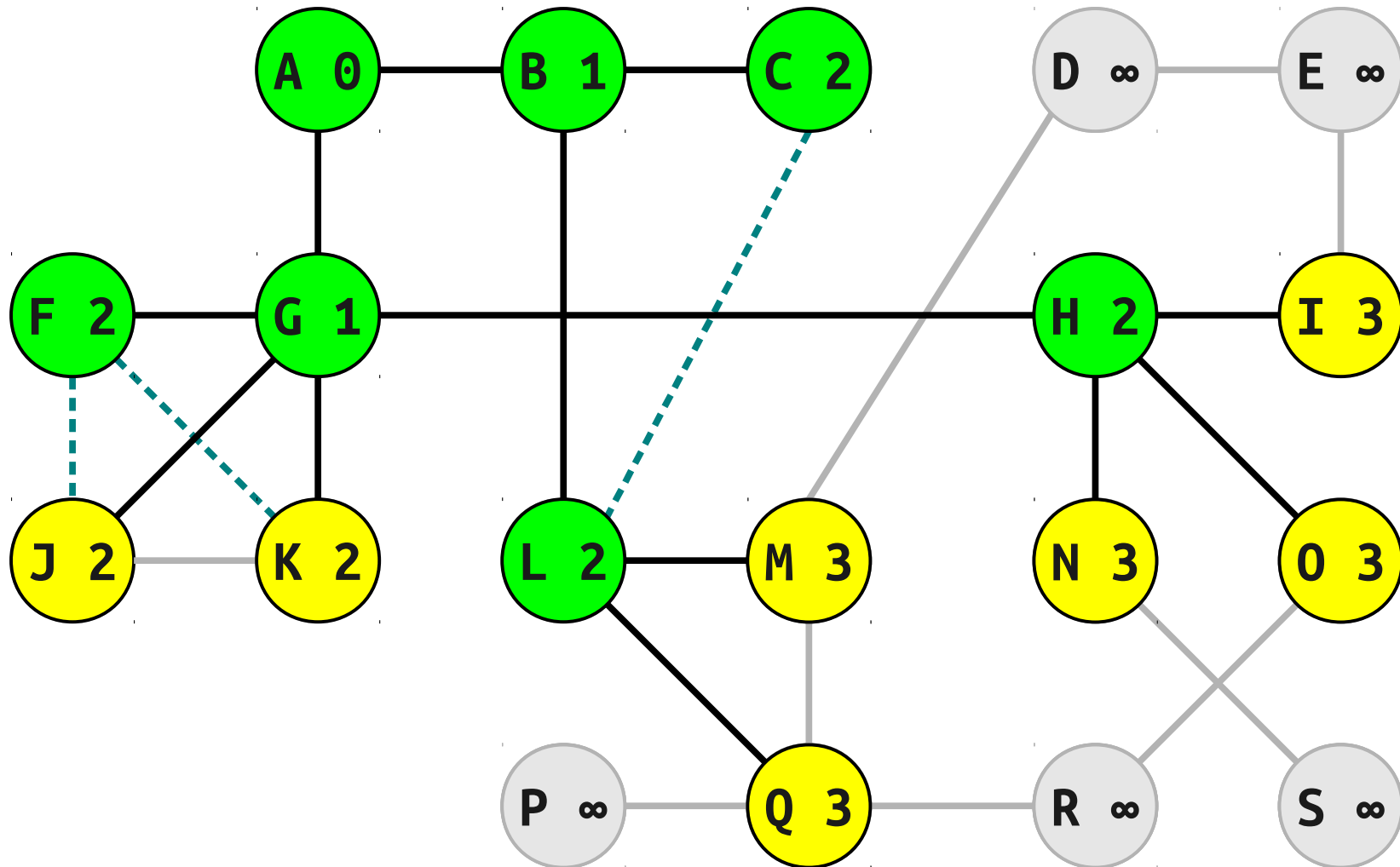
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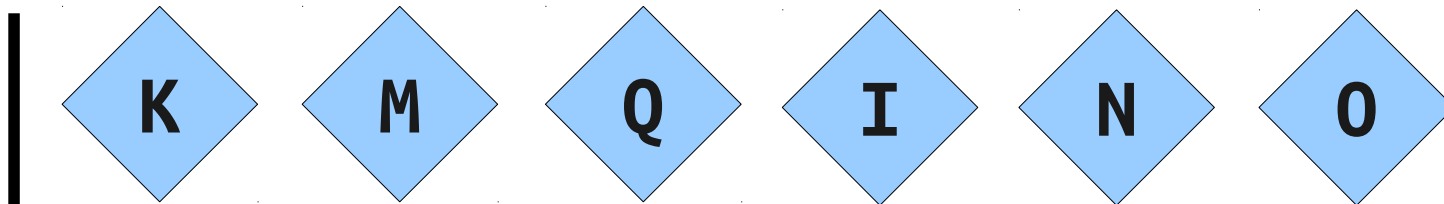
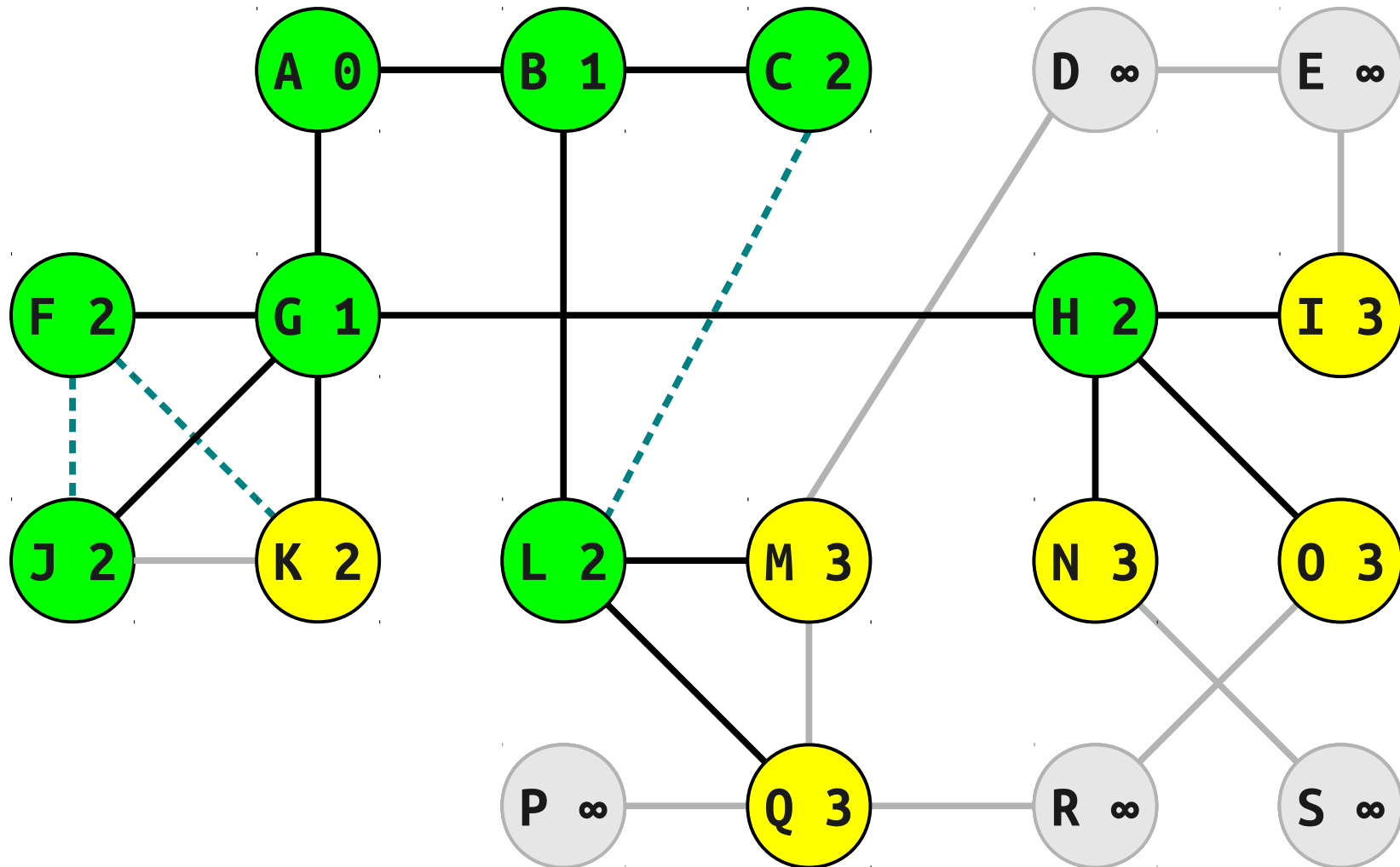
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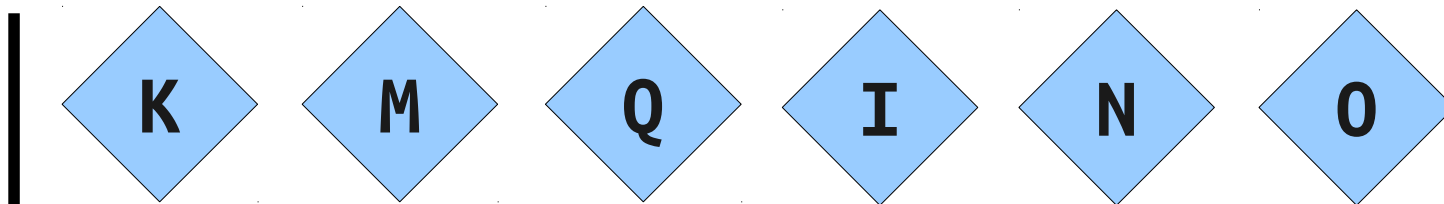
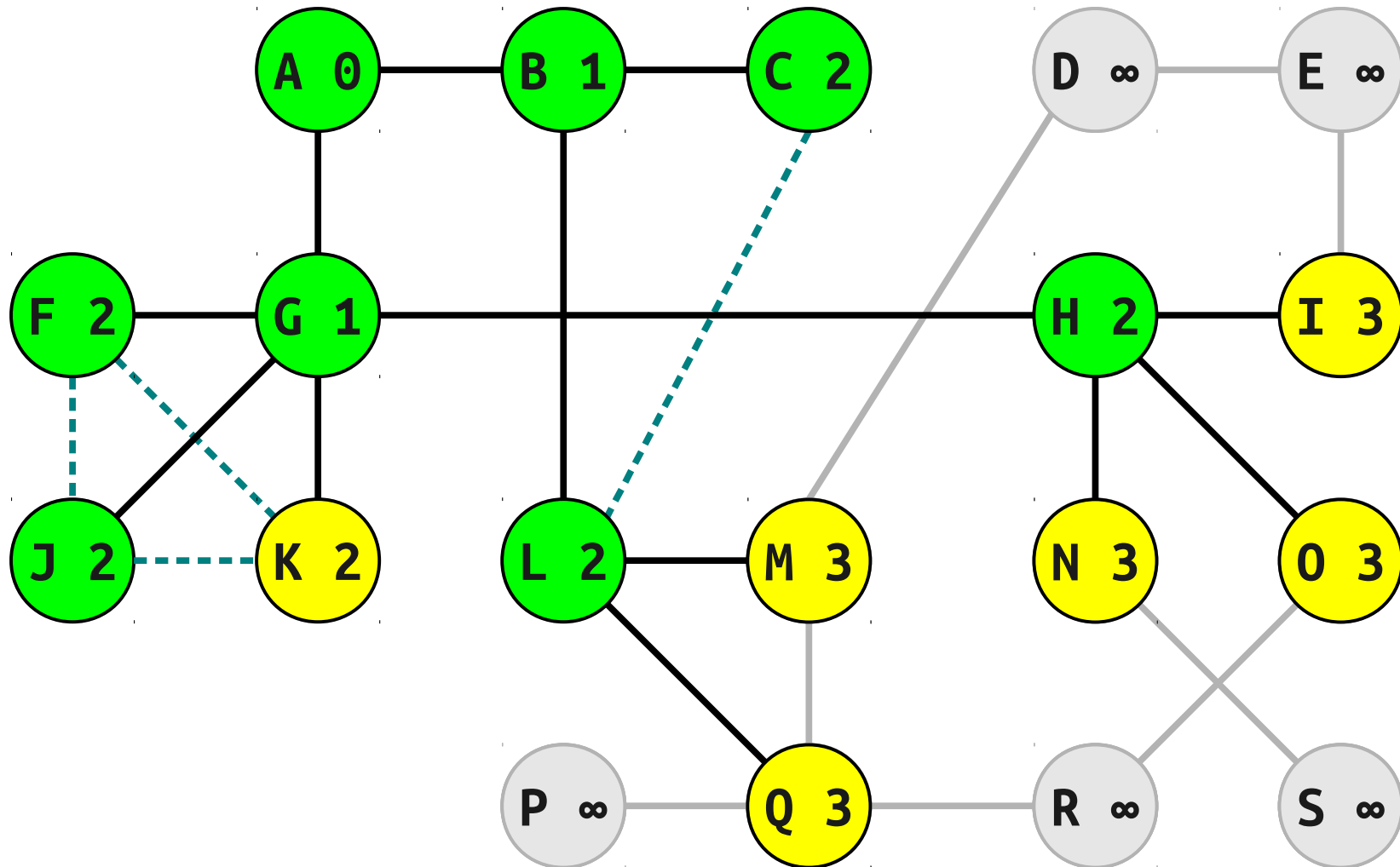
# Breadth-First Search



# Breadth-First Search

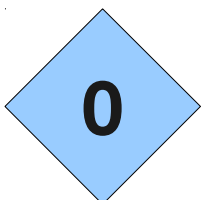
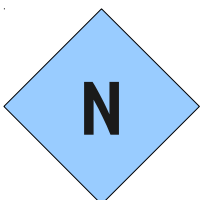
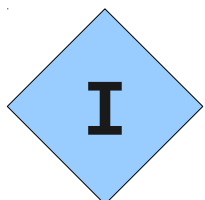
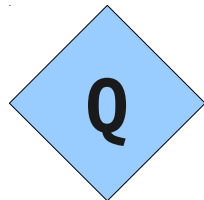
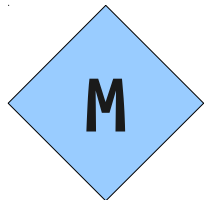
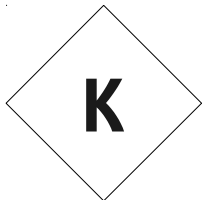
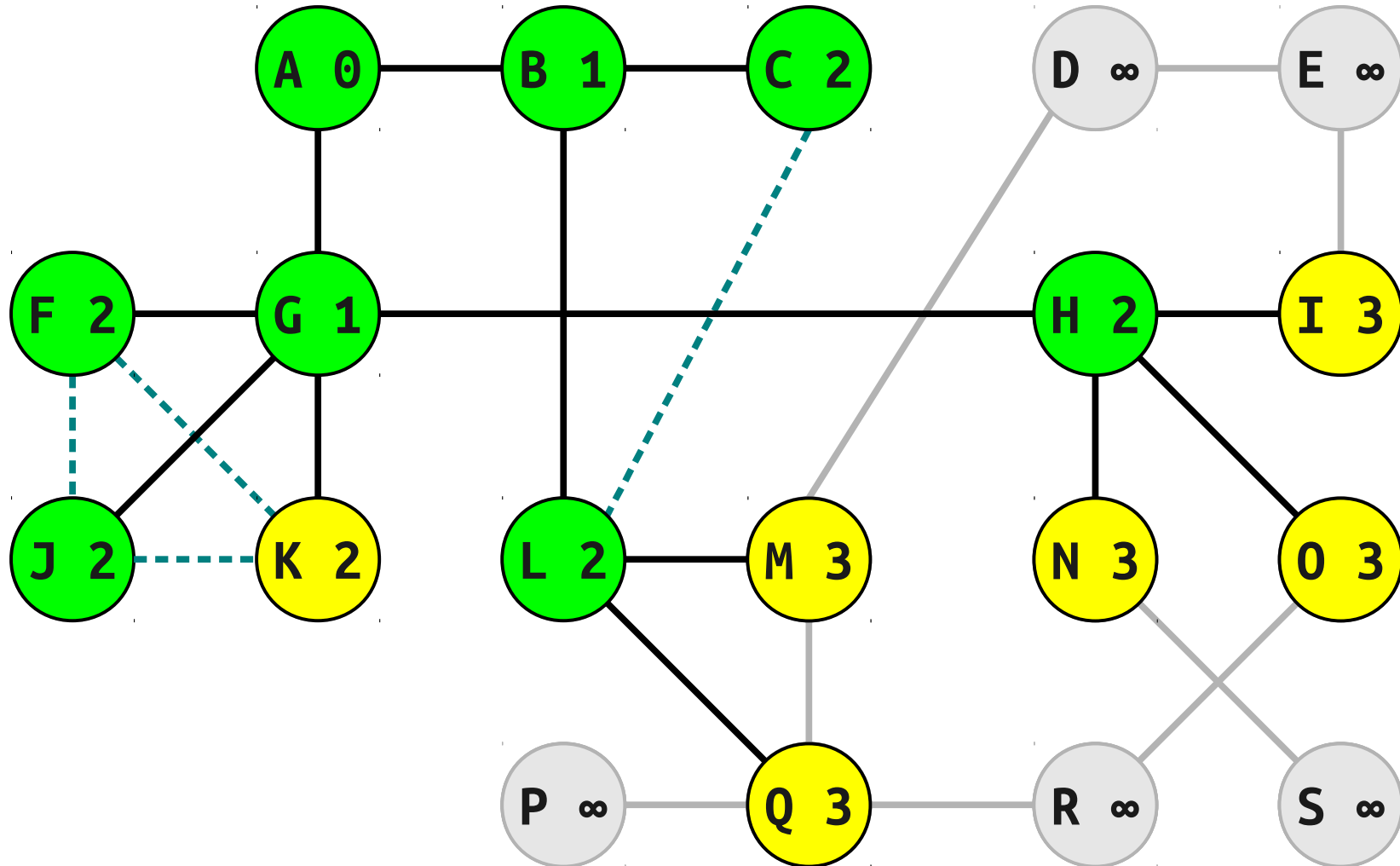


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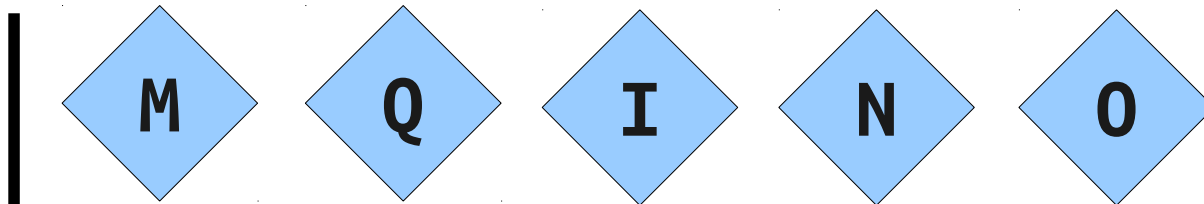
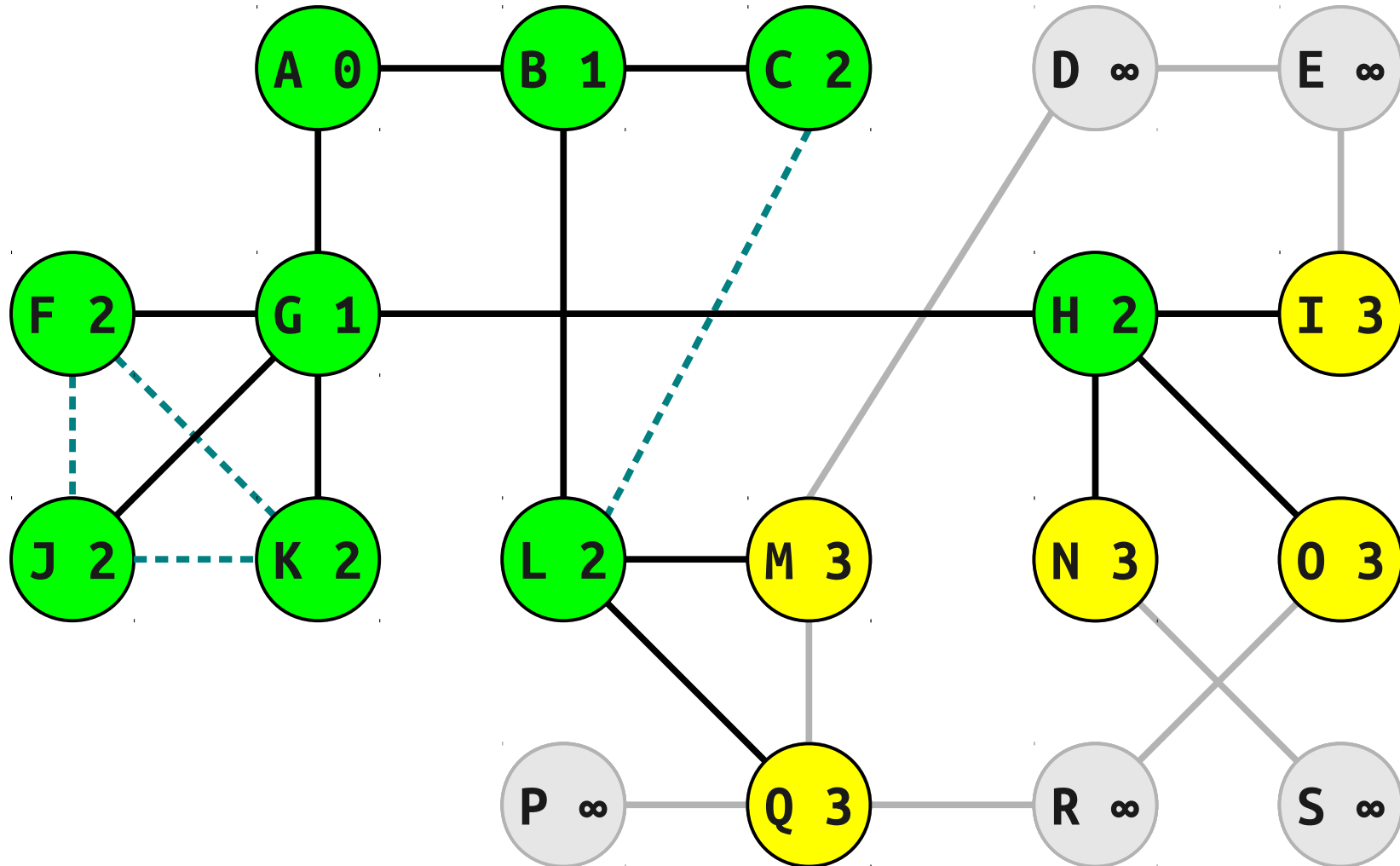




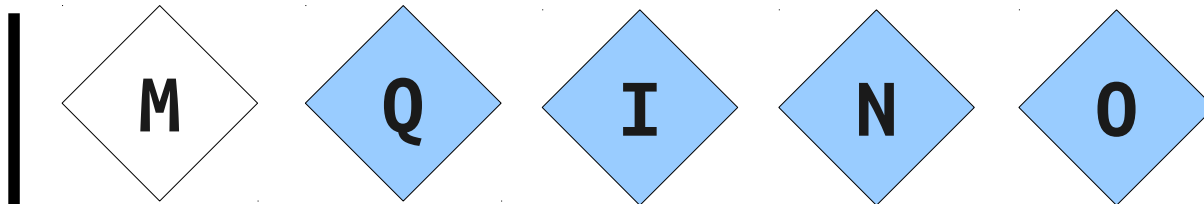
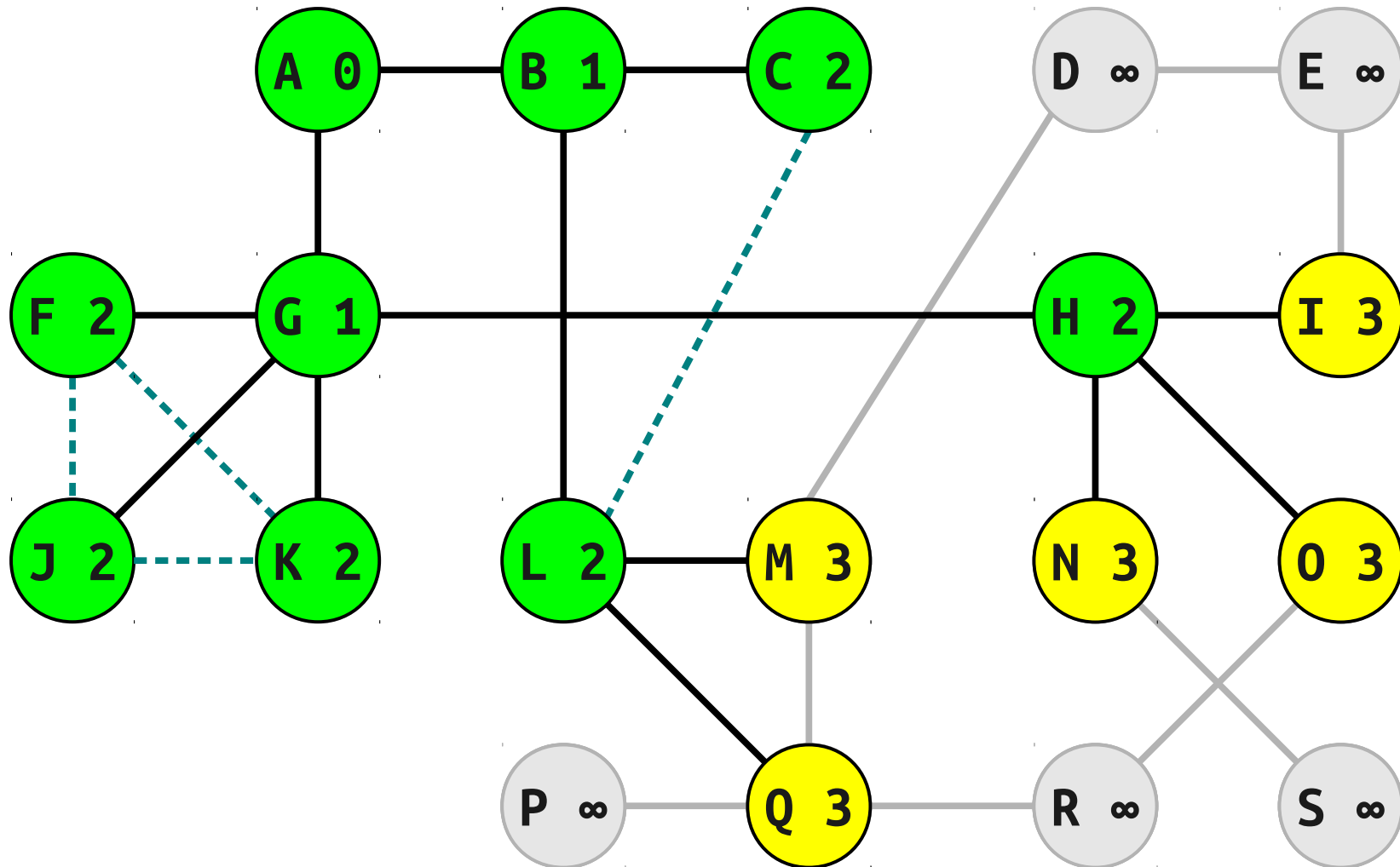
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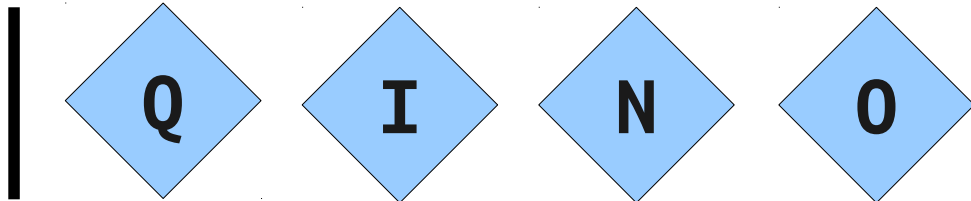
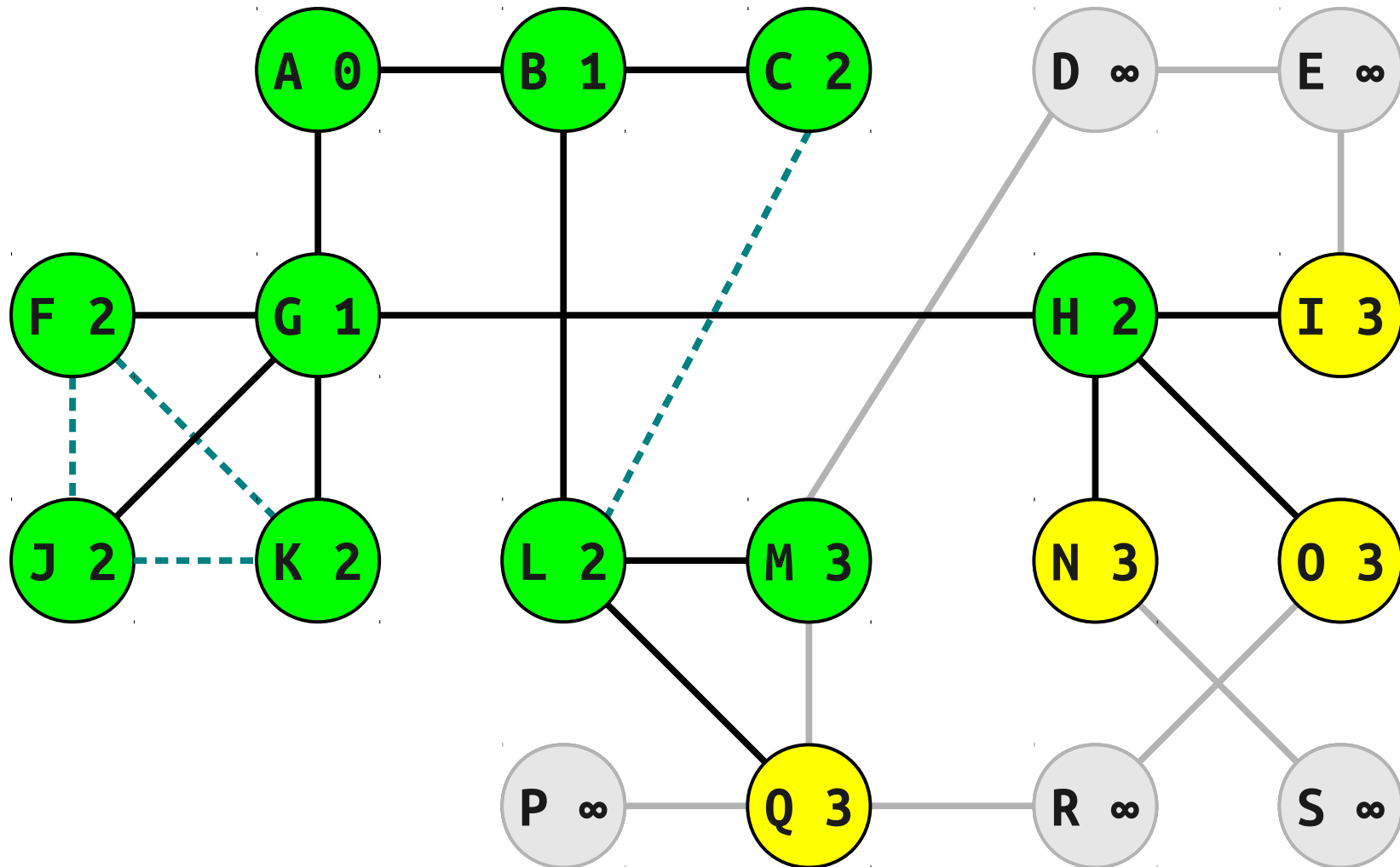
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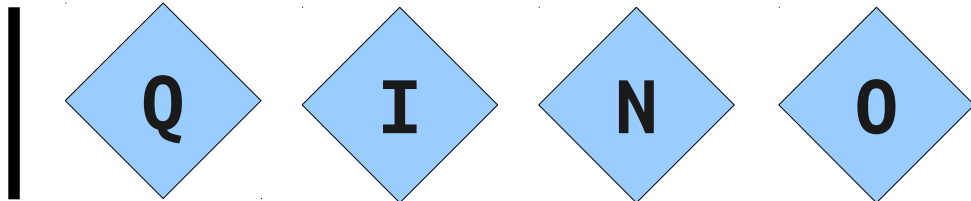
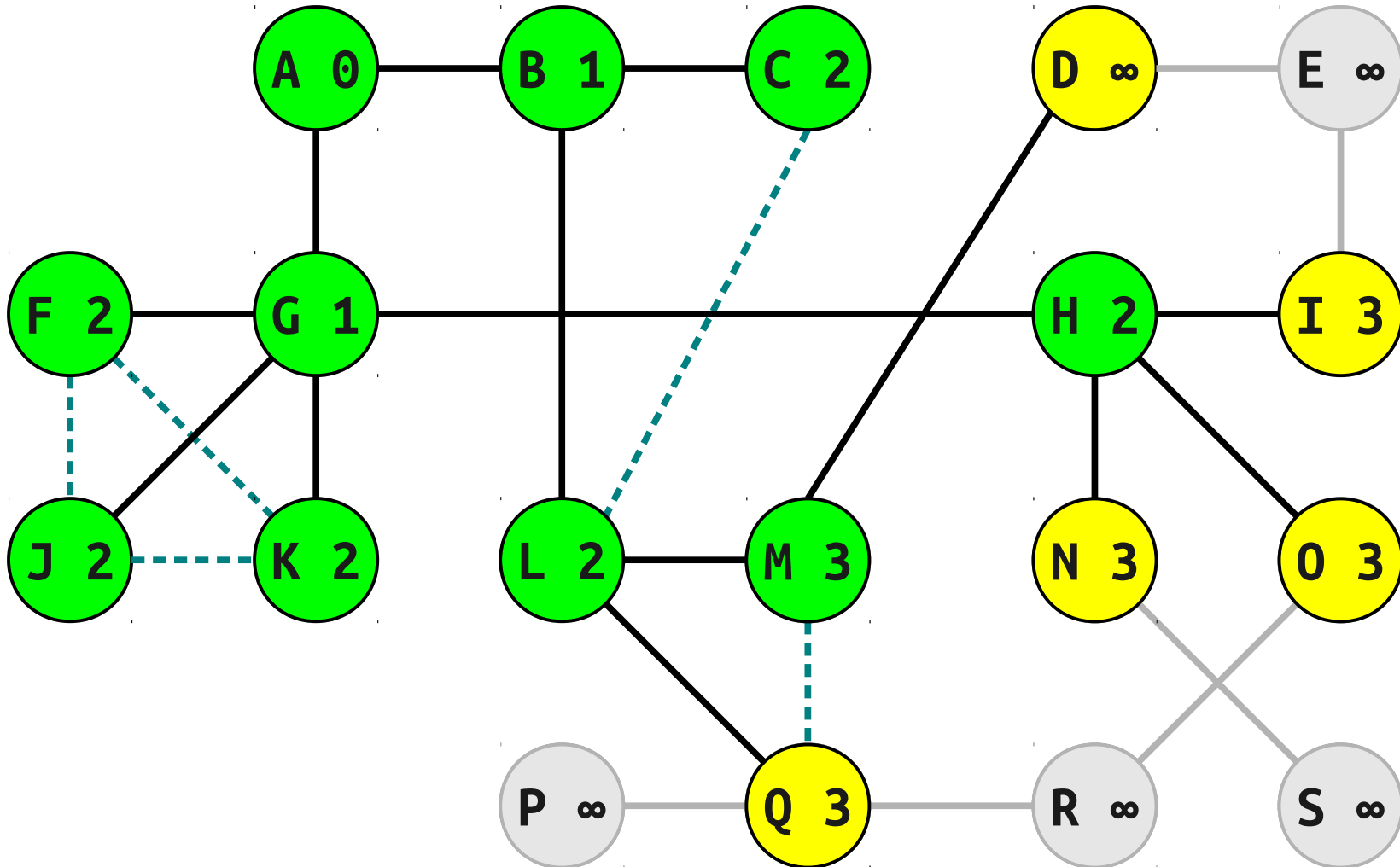
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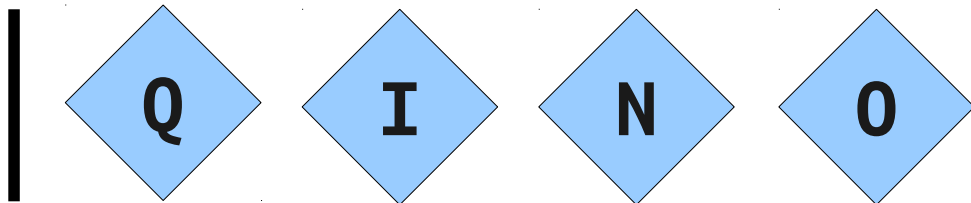
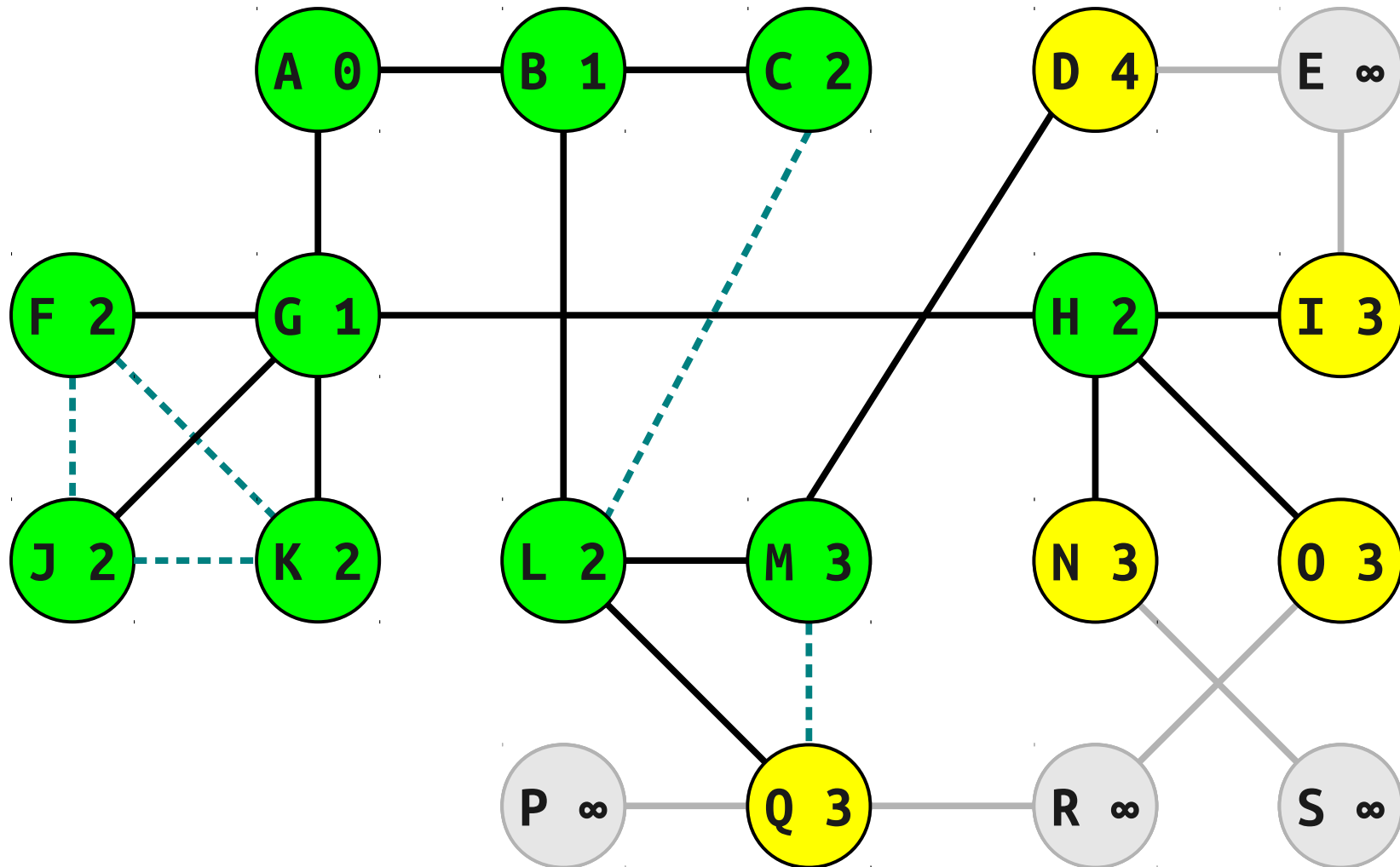
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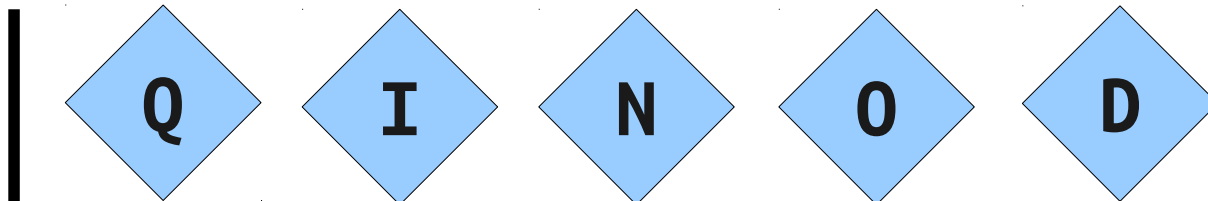
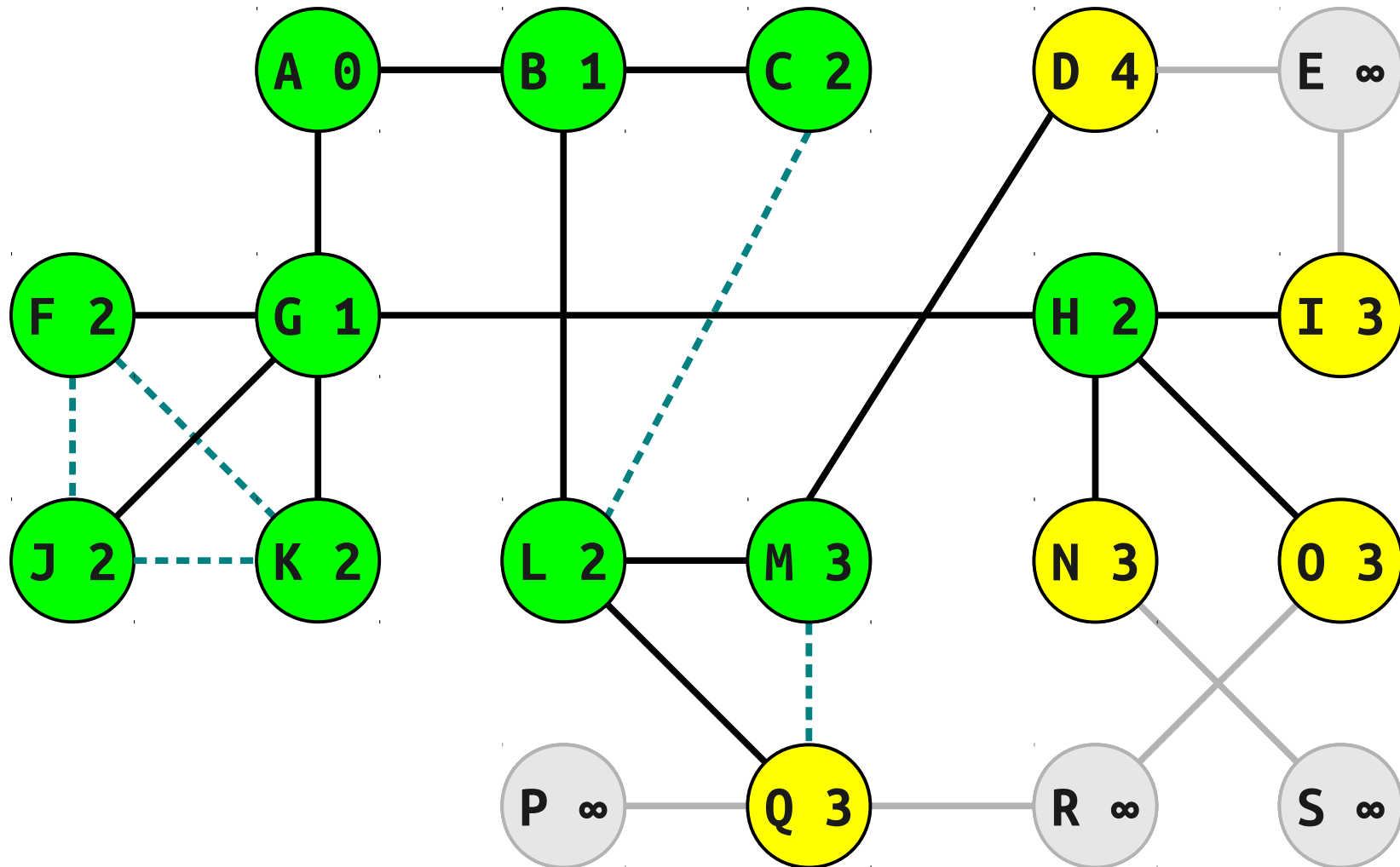
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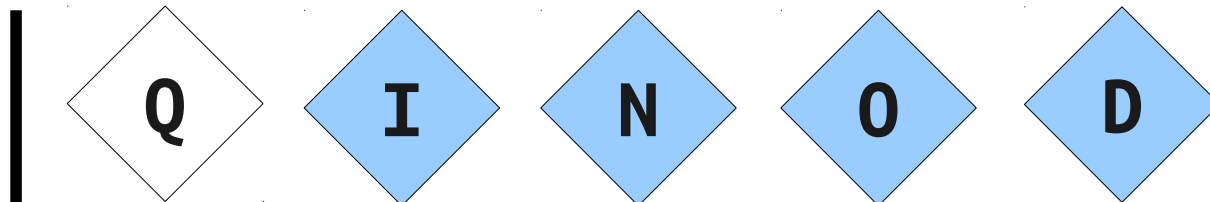
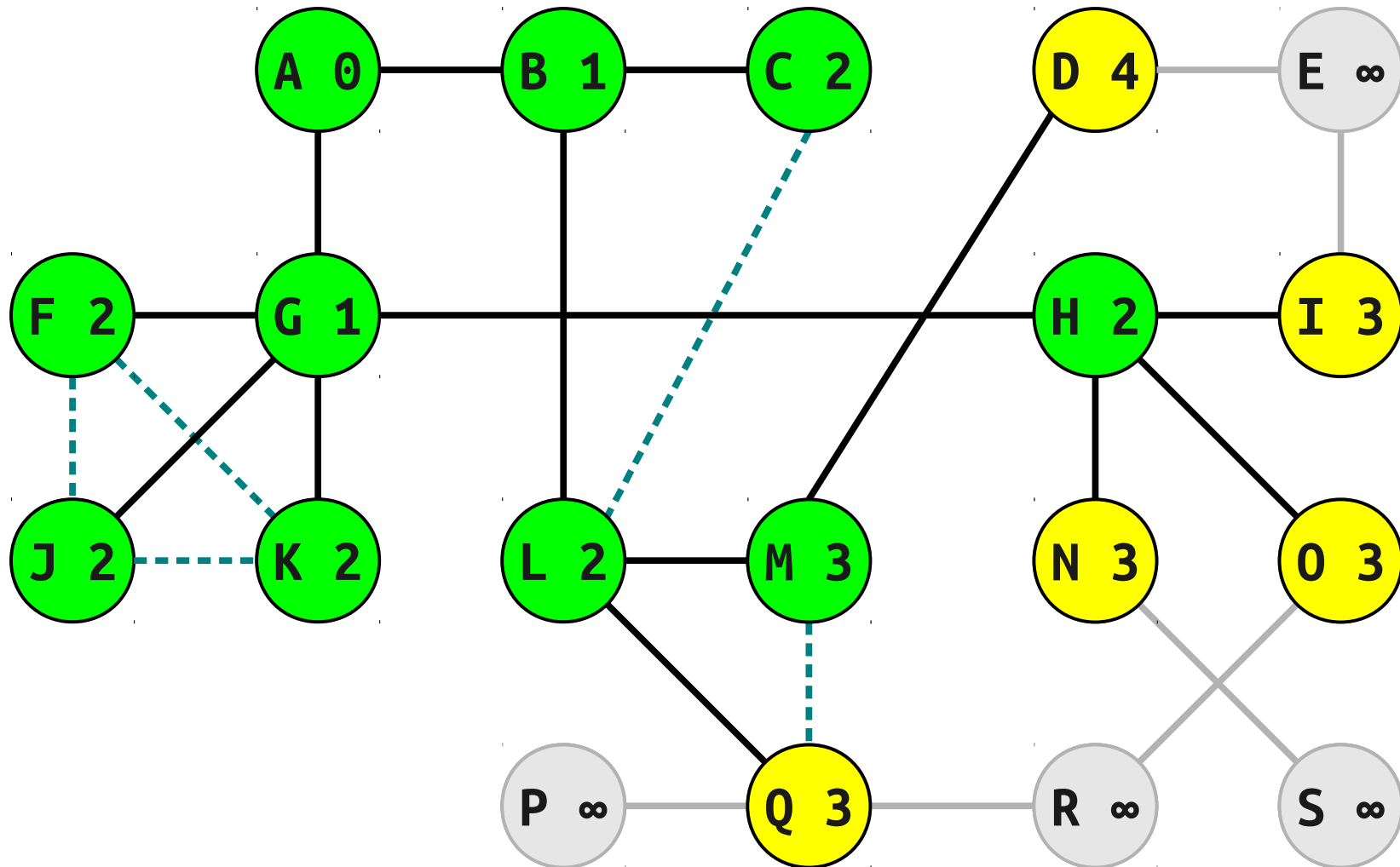
# Breadth-First Search



# Breadth-First Search

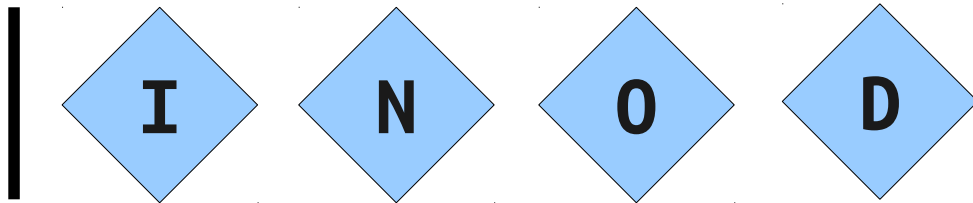
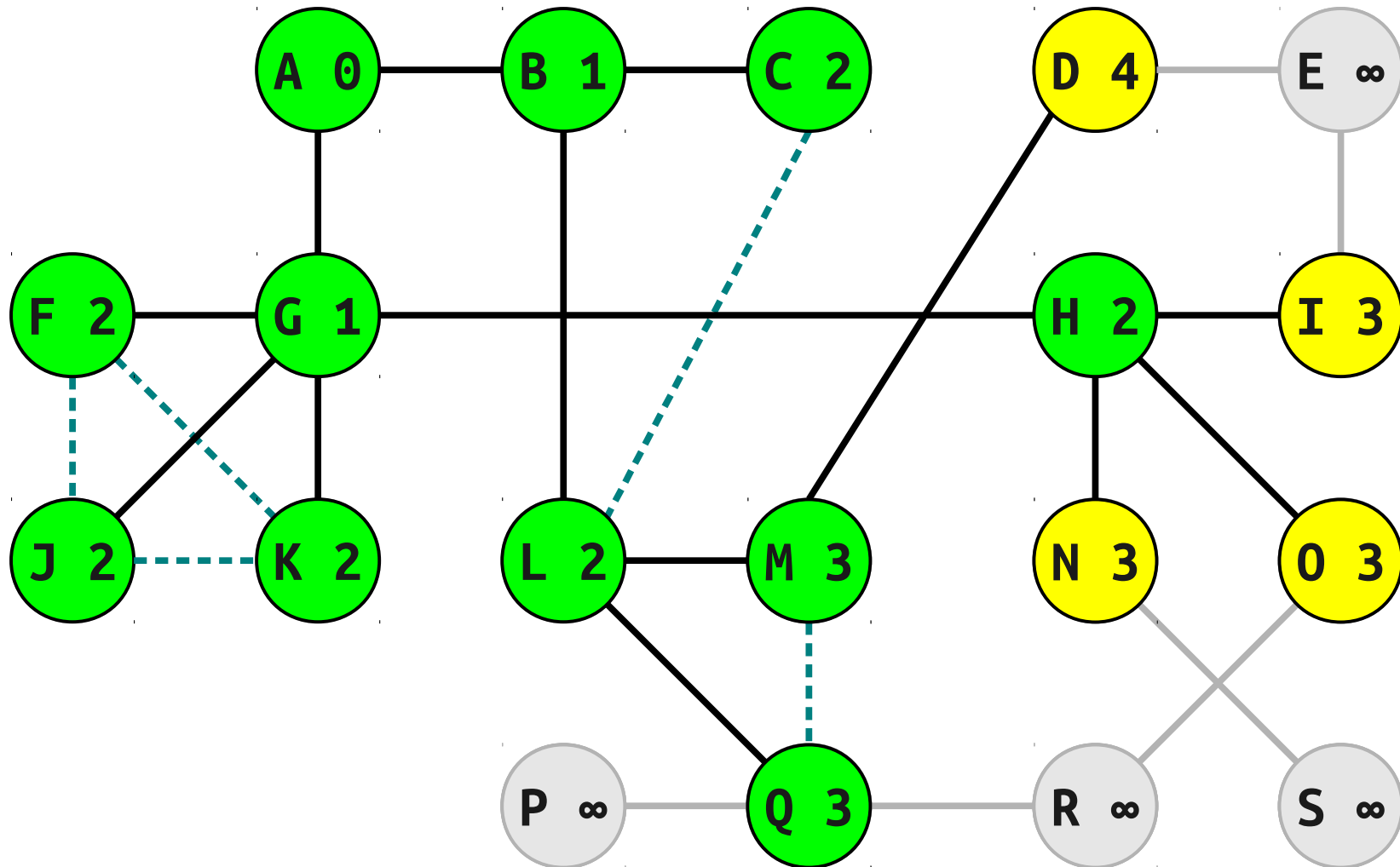


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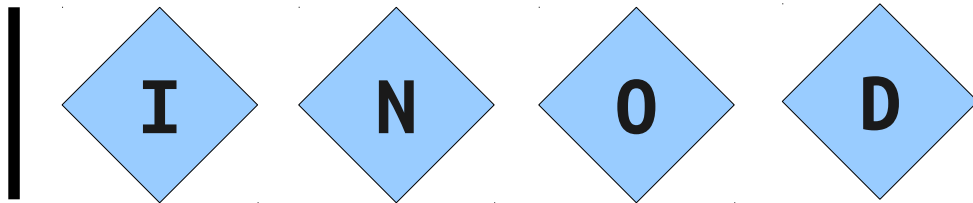
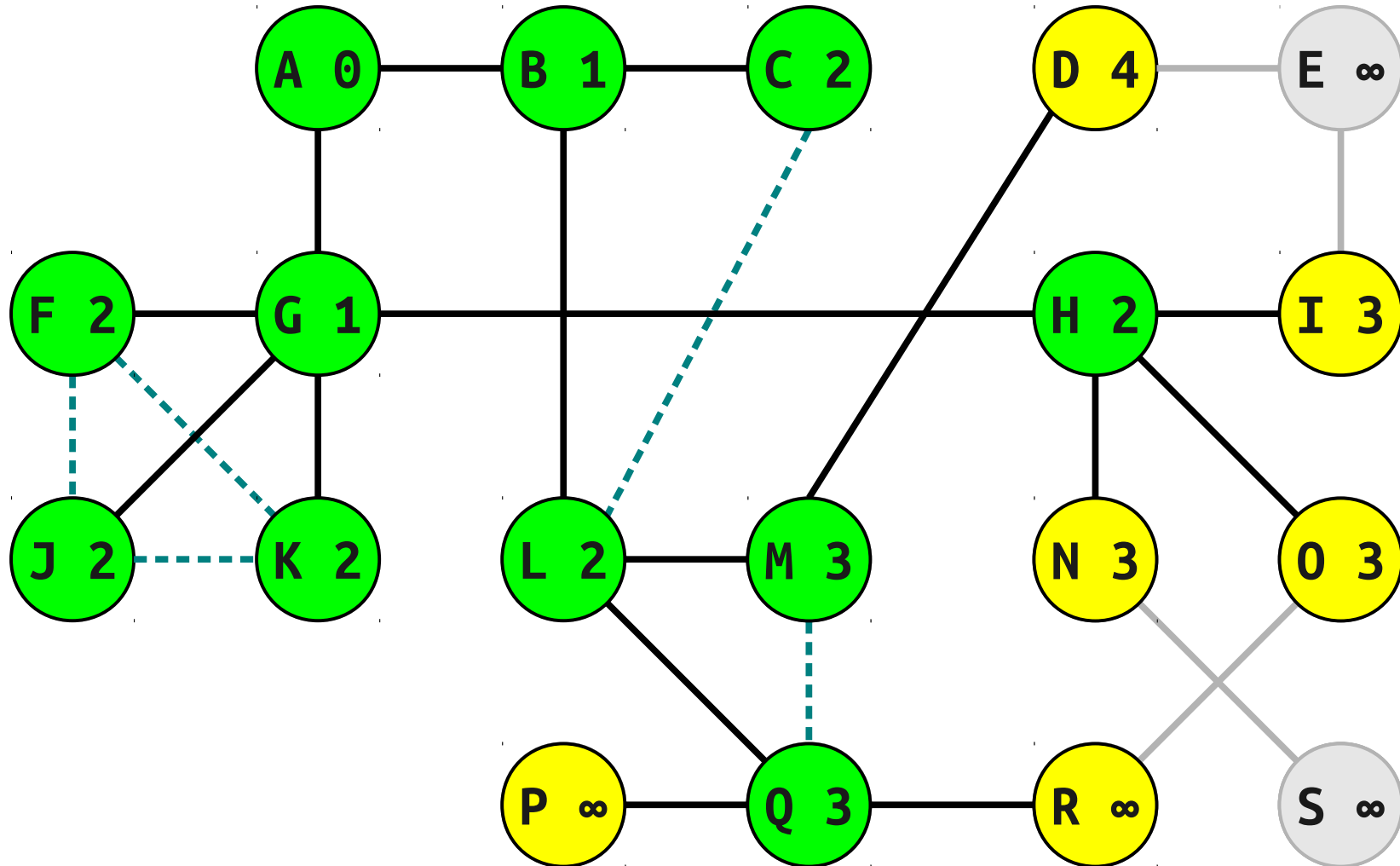




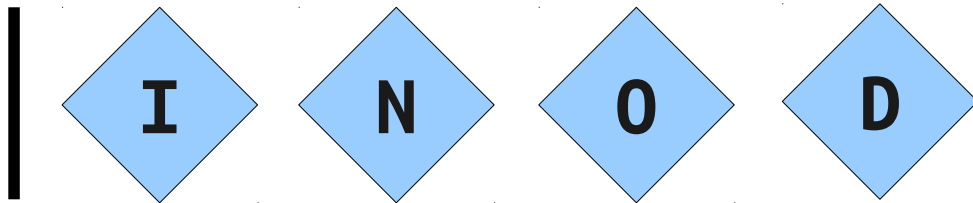
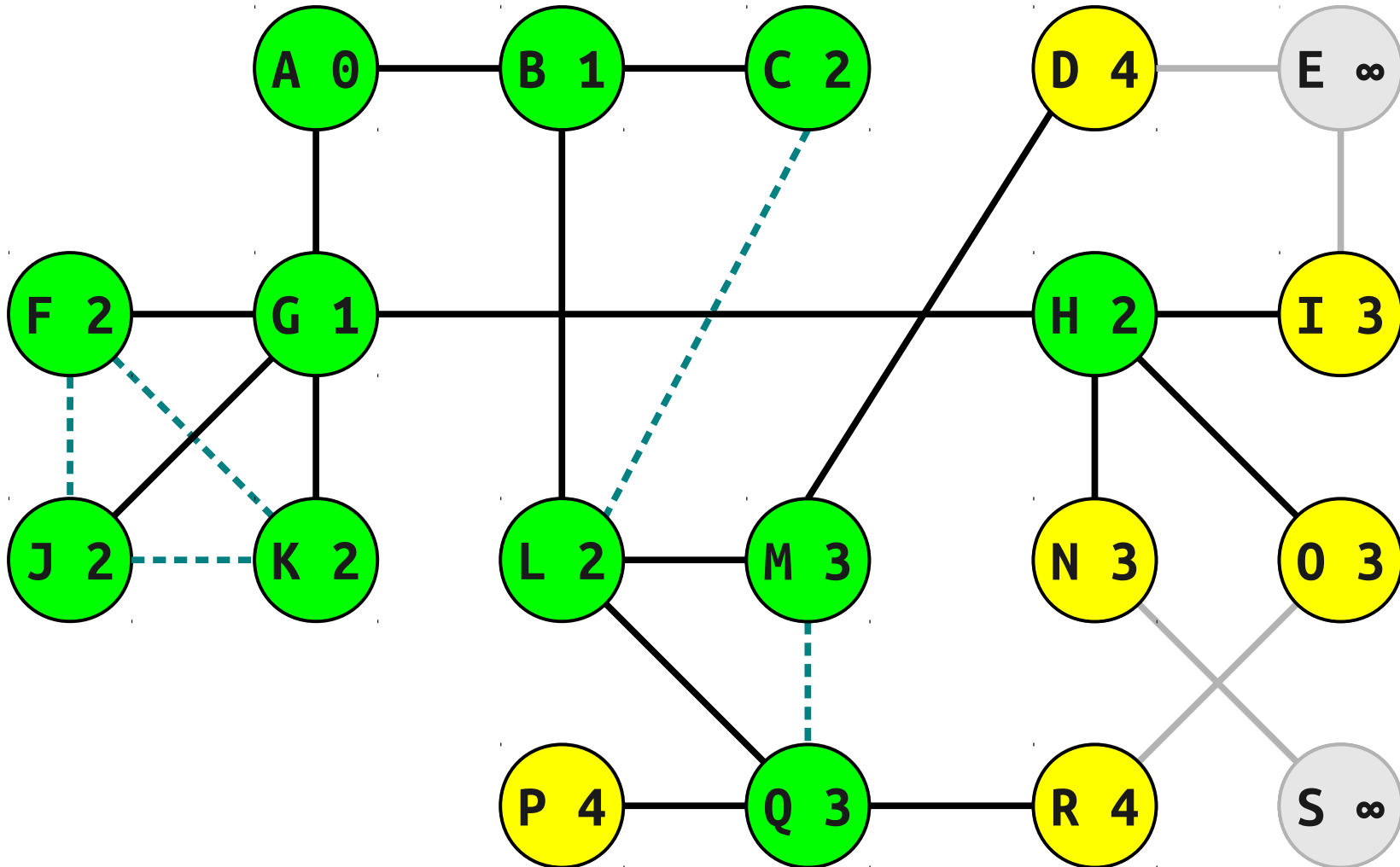
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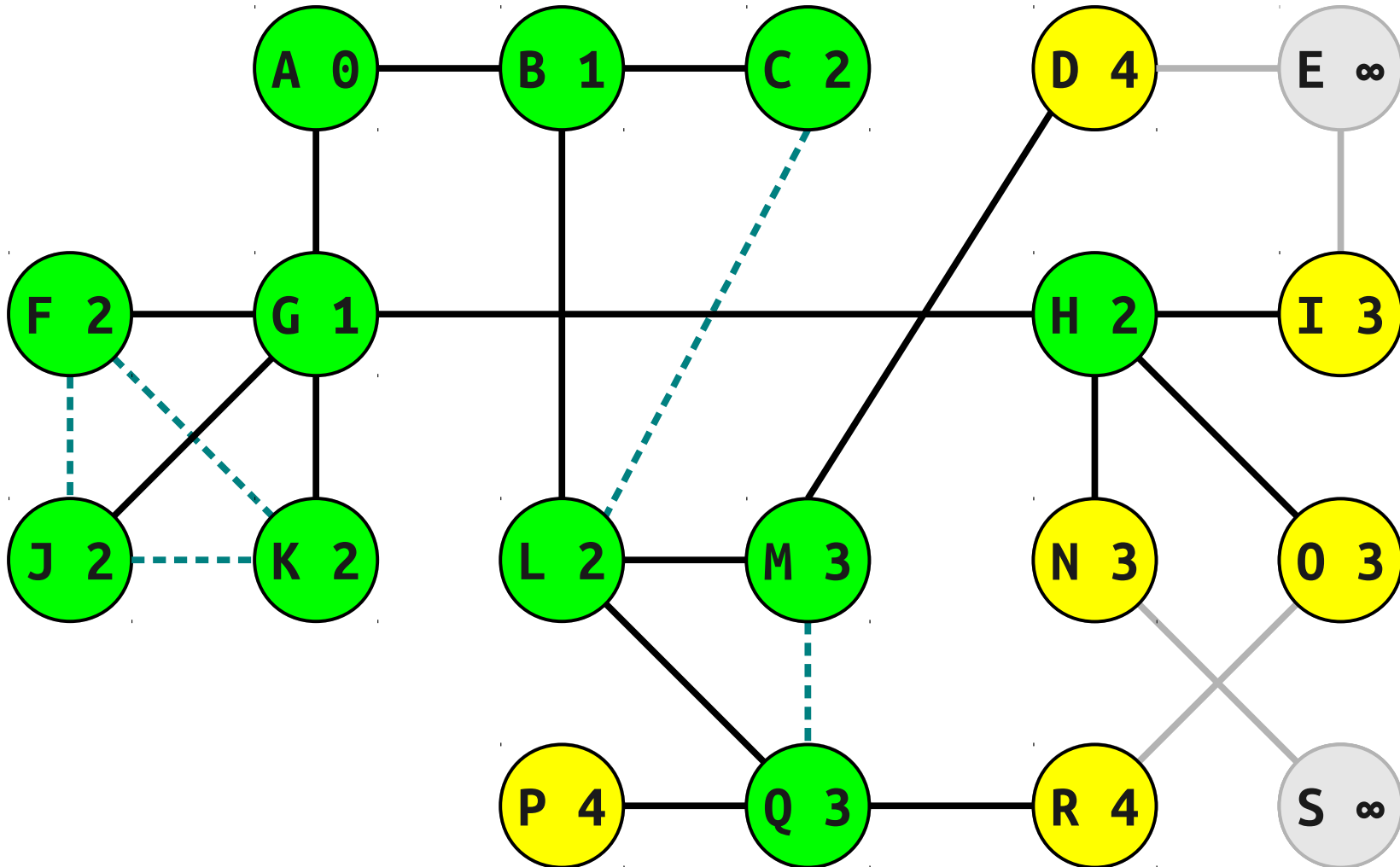
# Breadth-First Search



# Breadth-First Search



# Breadth-First Search



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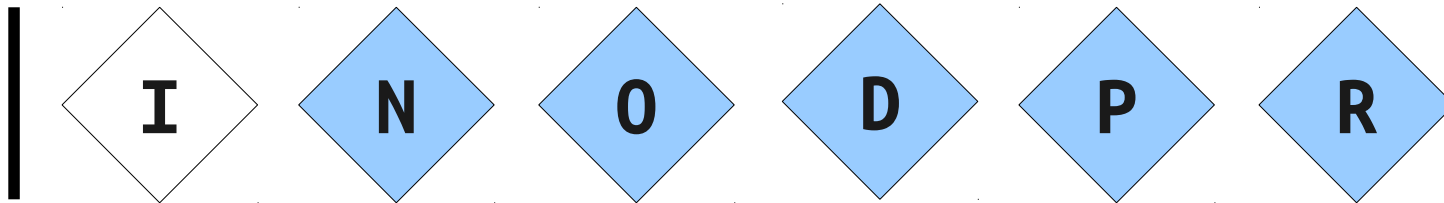
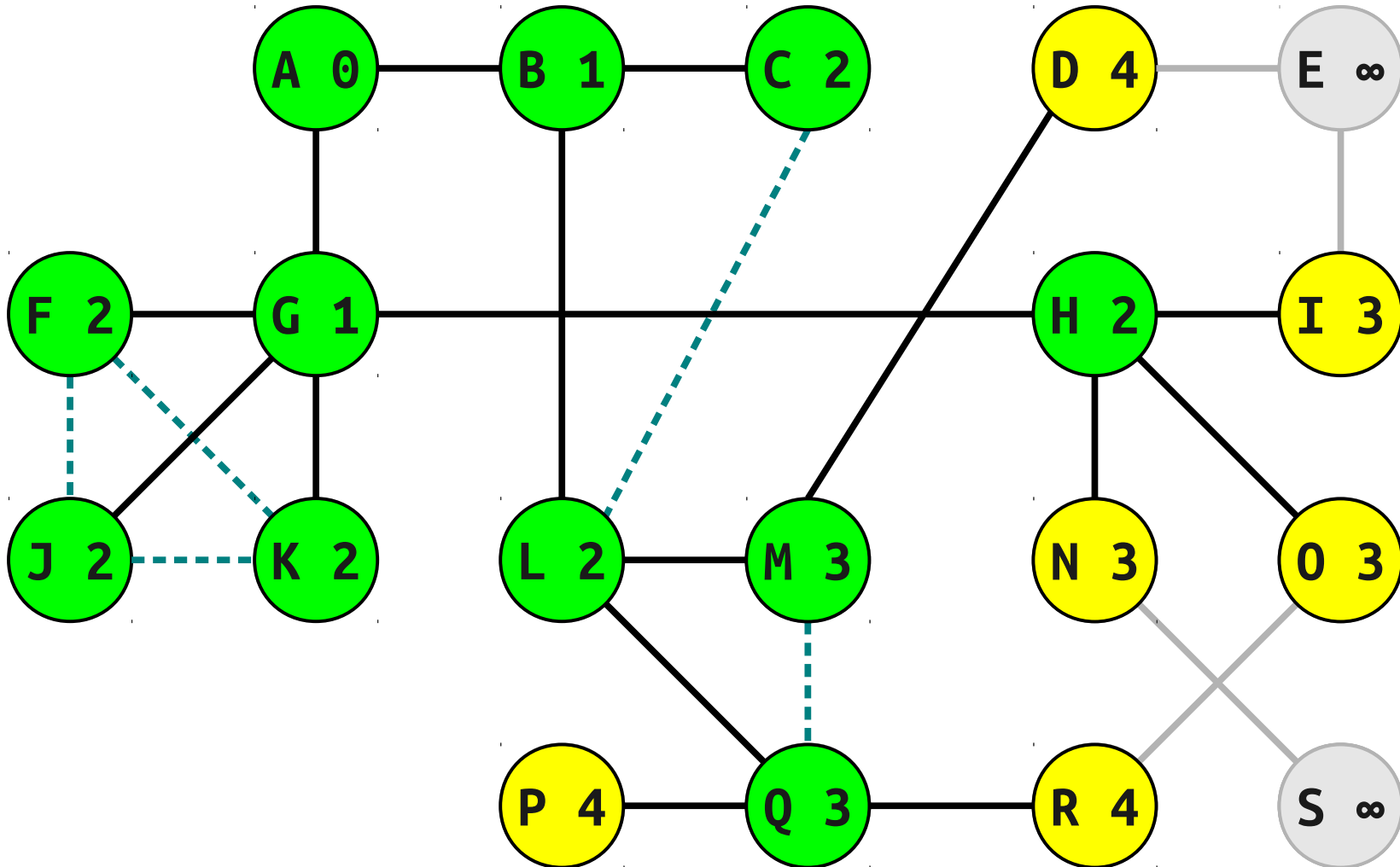
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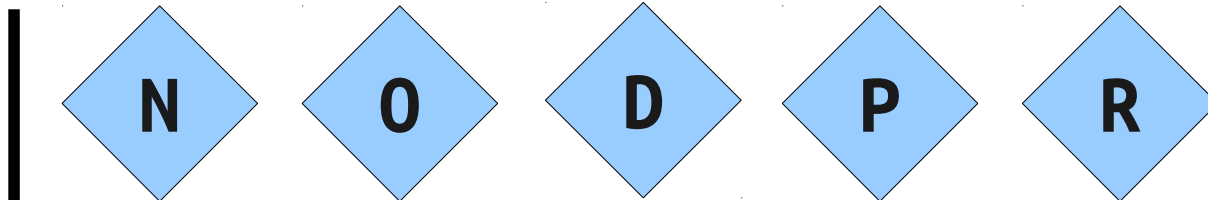
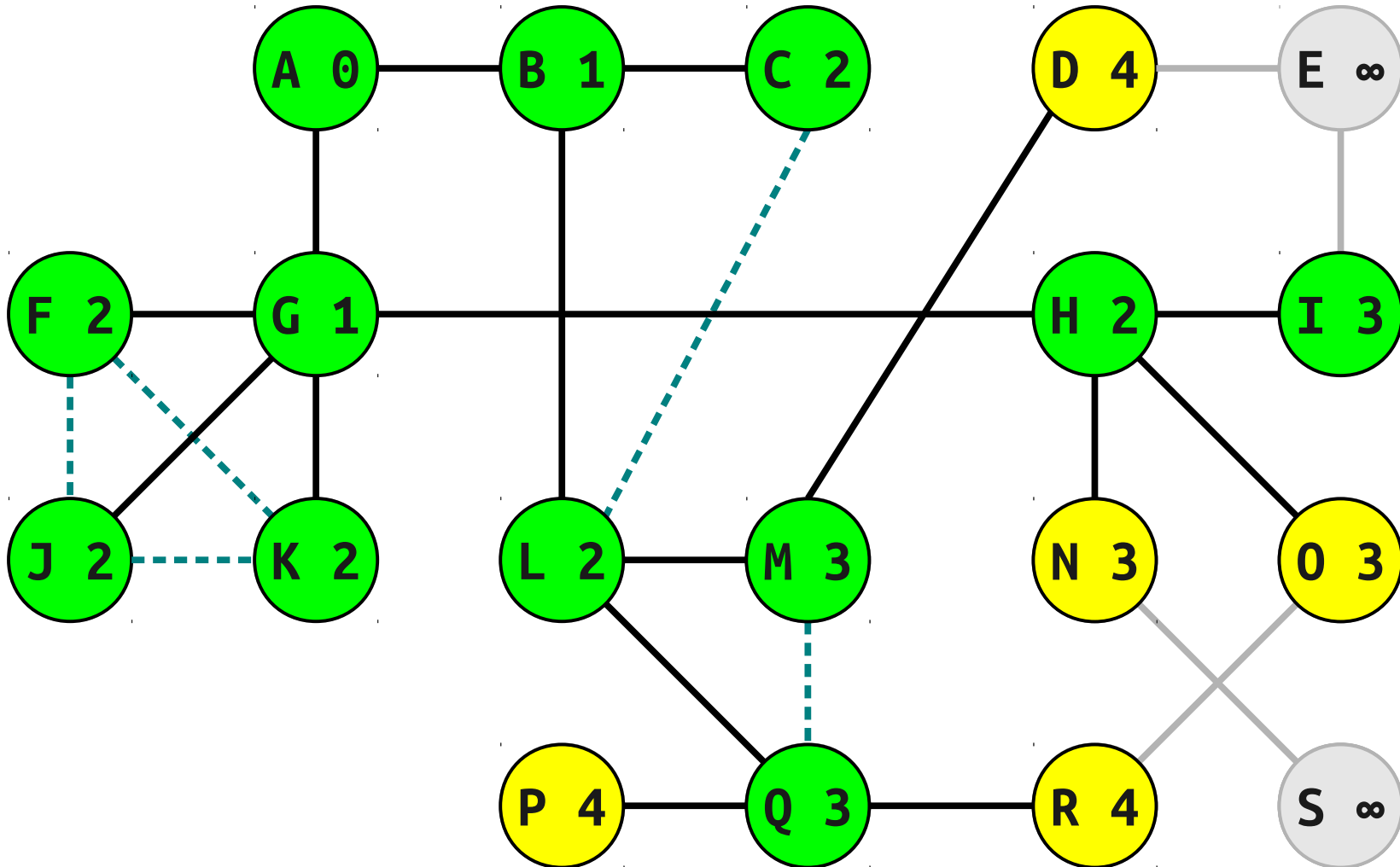
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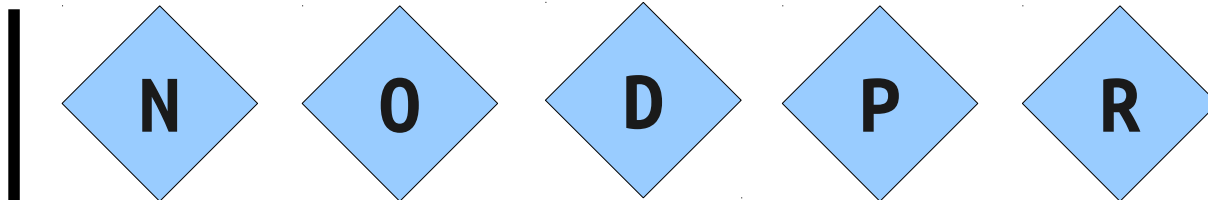
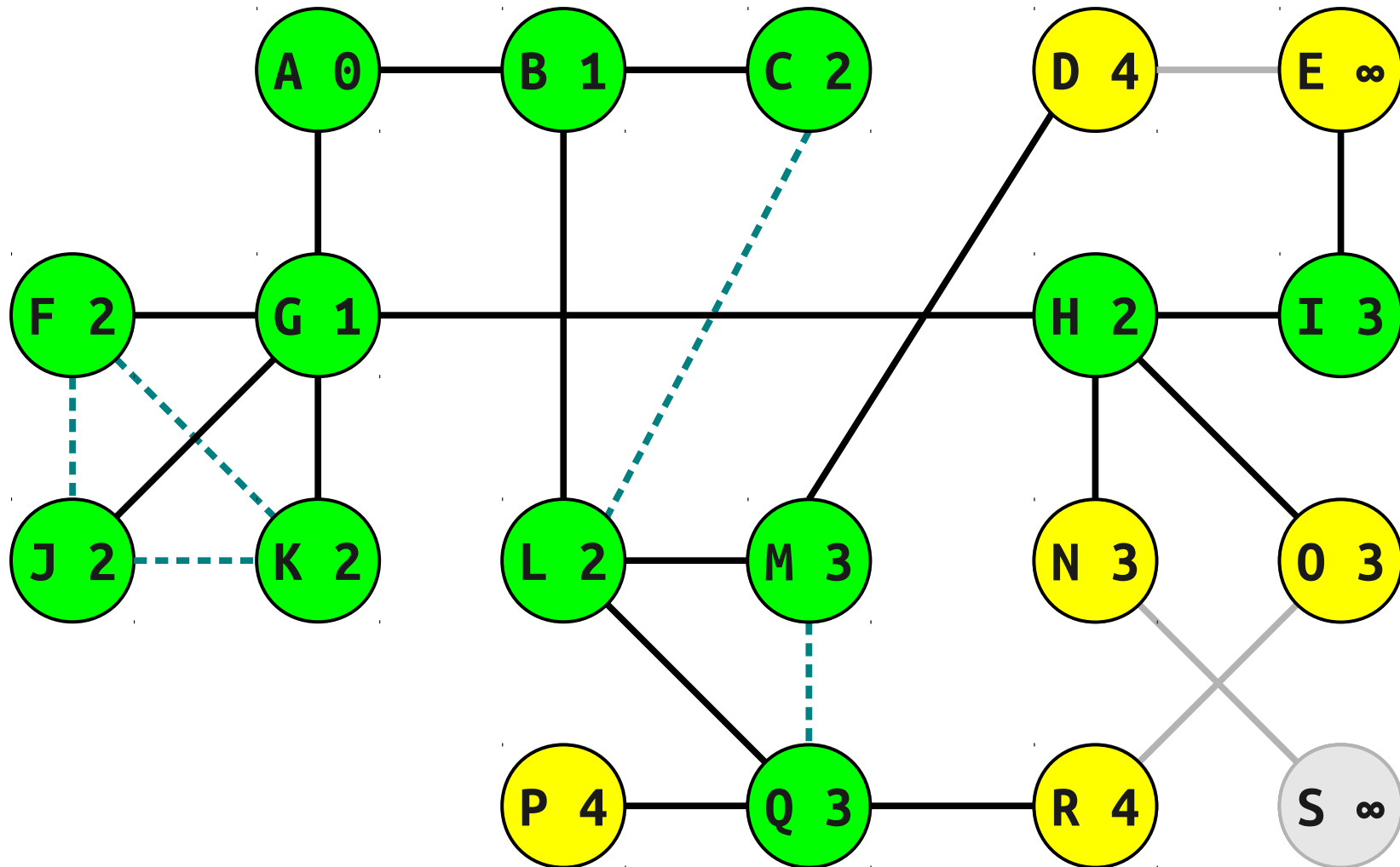
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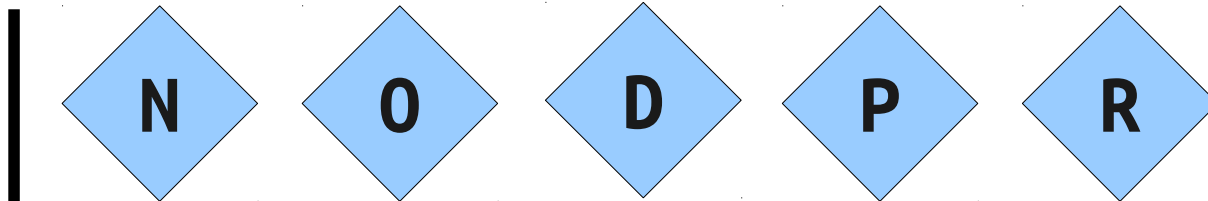
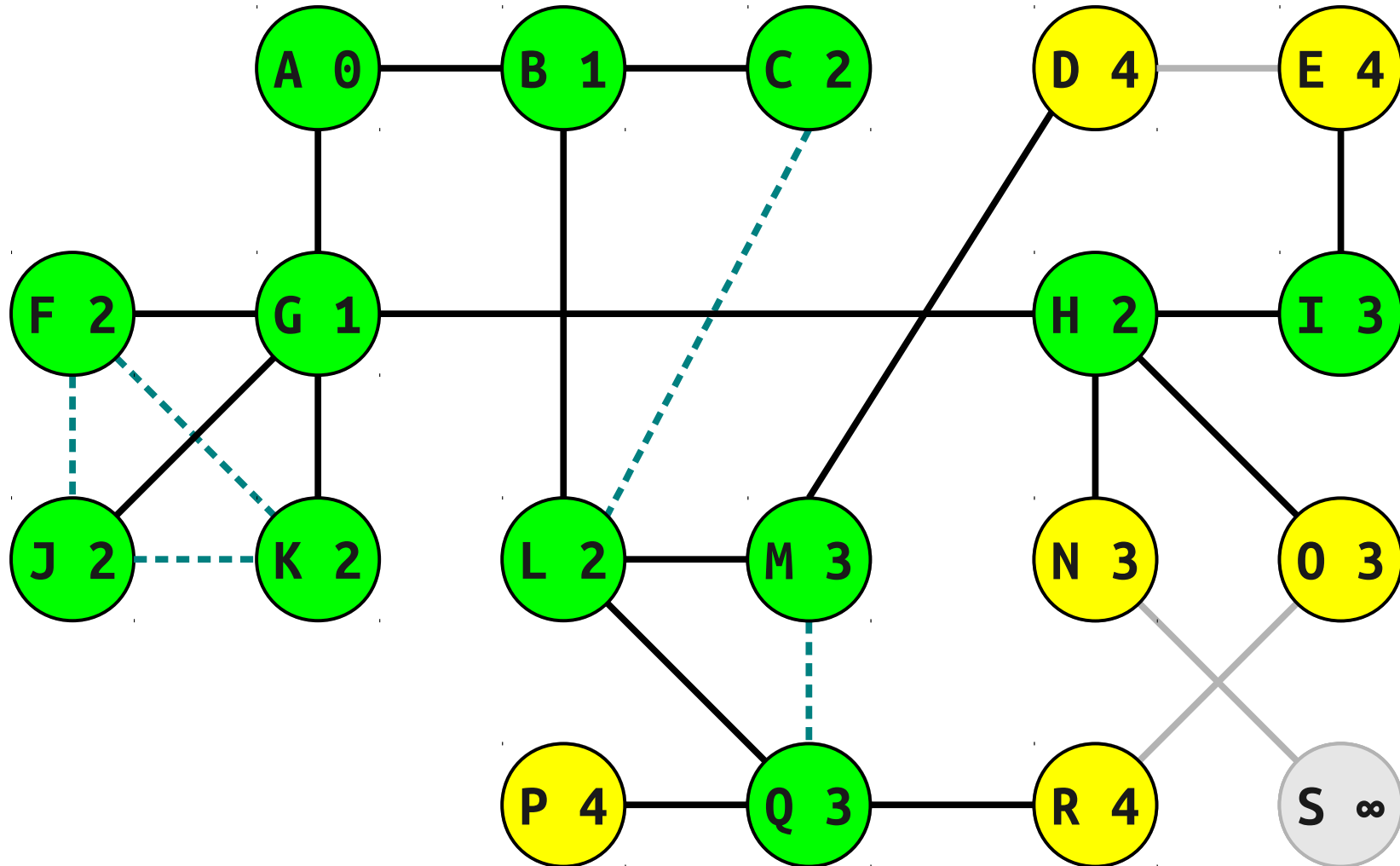
# Breadth-First Search



# Breadth-First Search

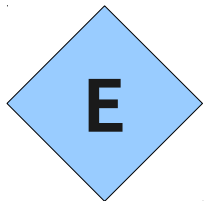
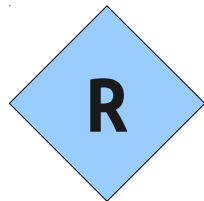
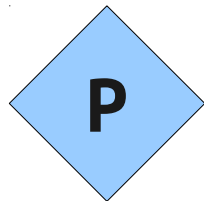
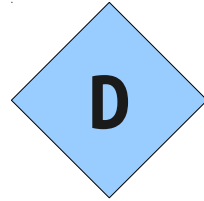
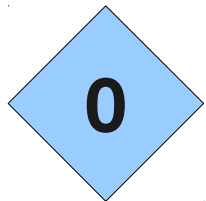
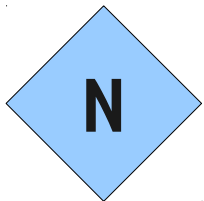
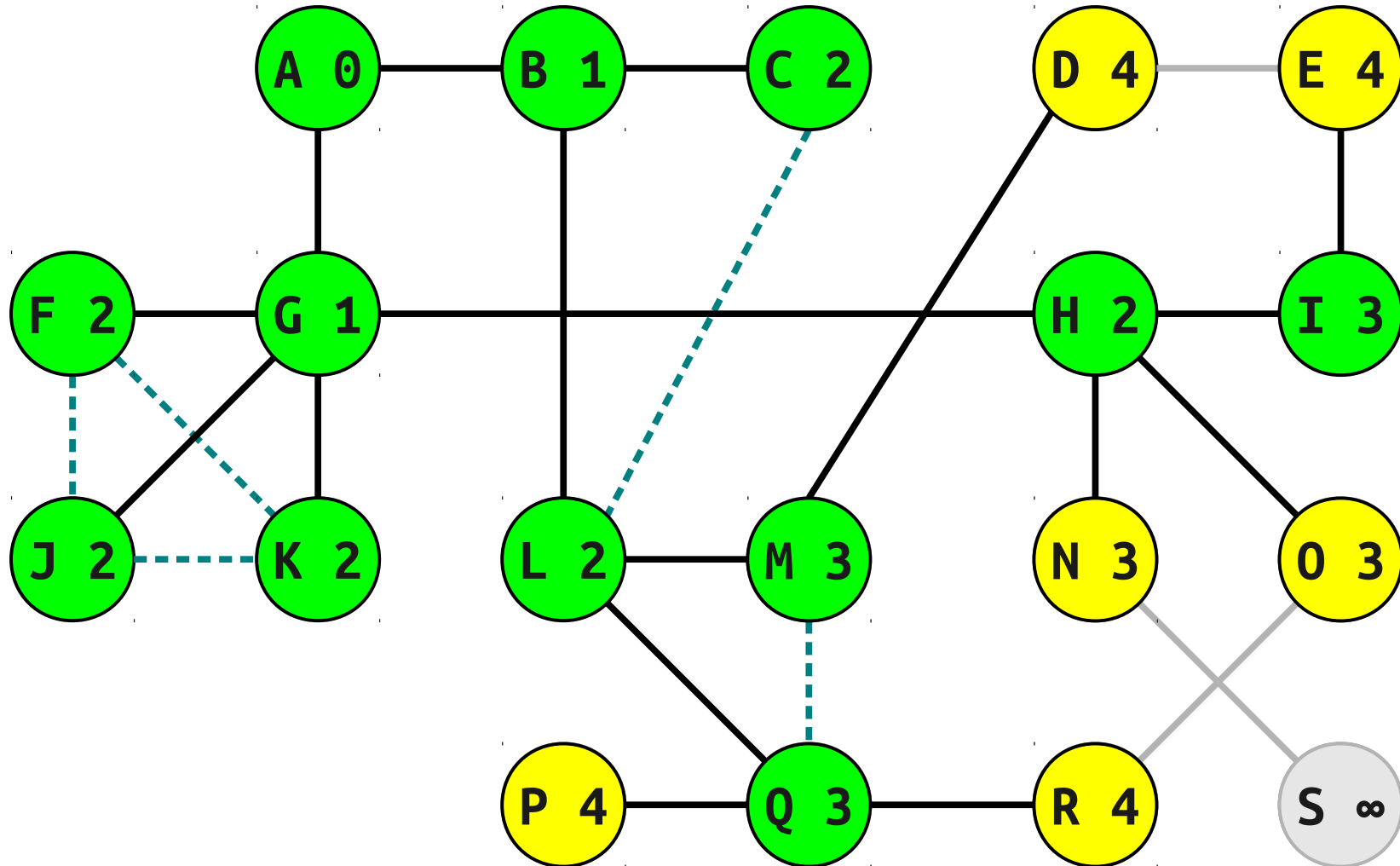


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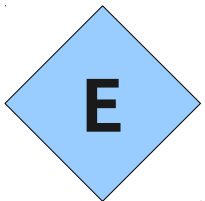
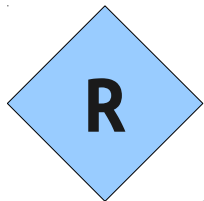
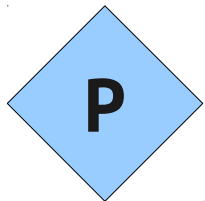
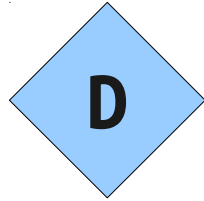
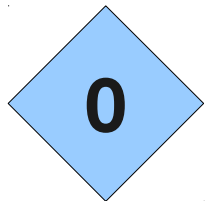
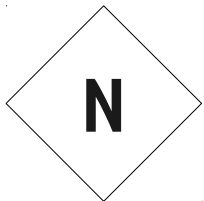
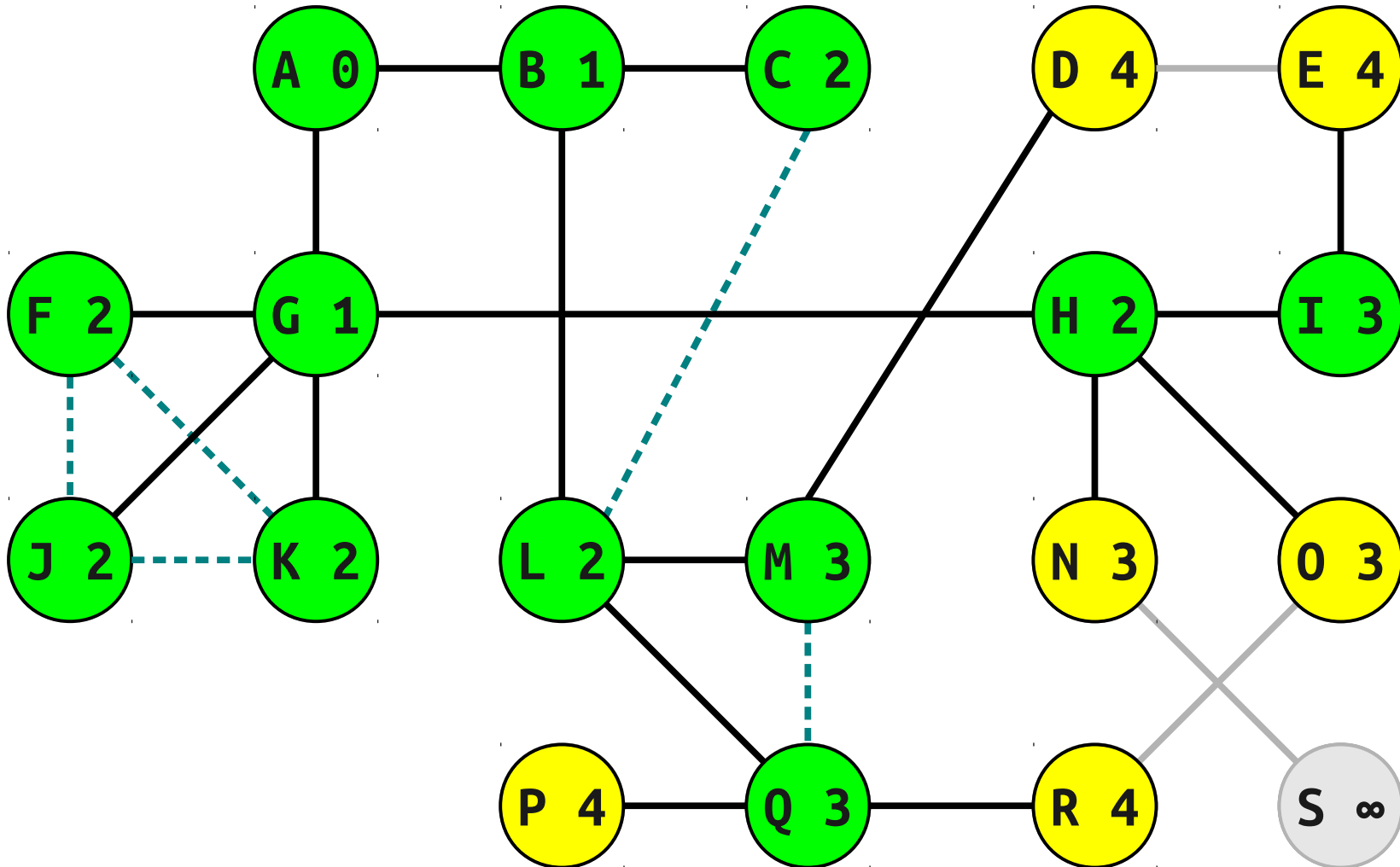




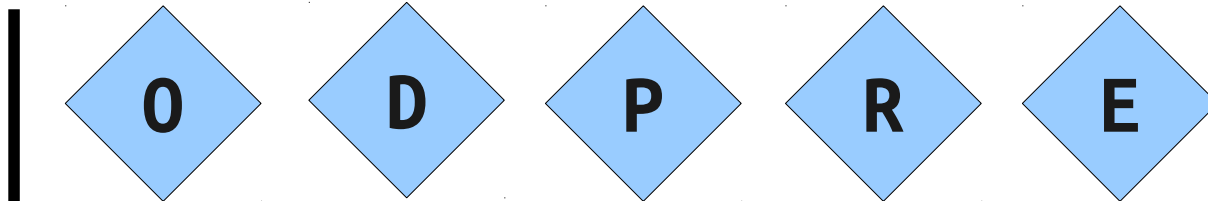
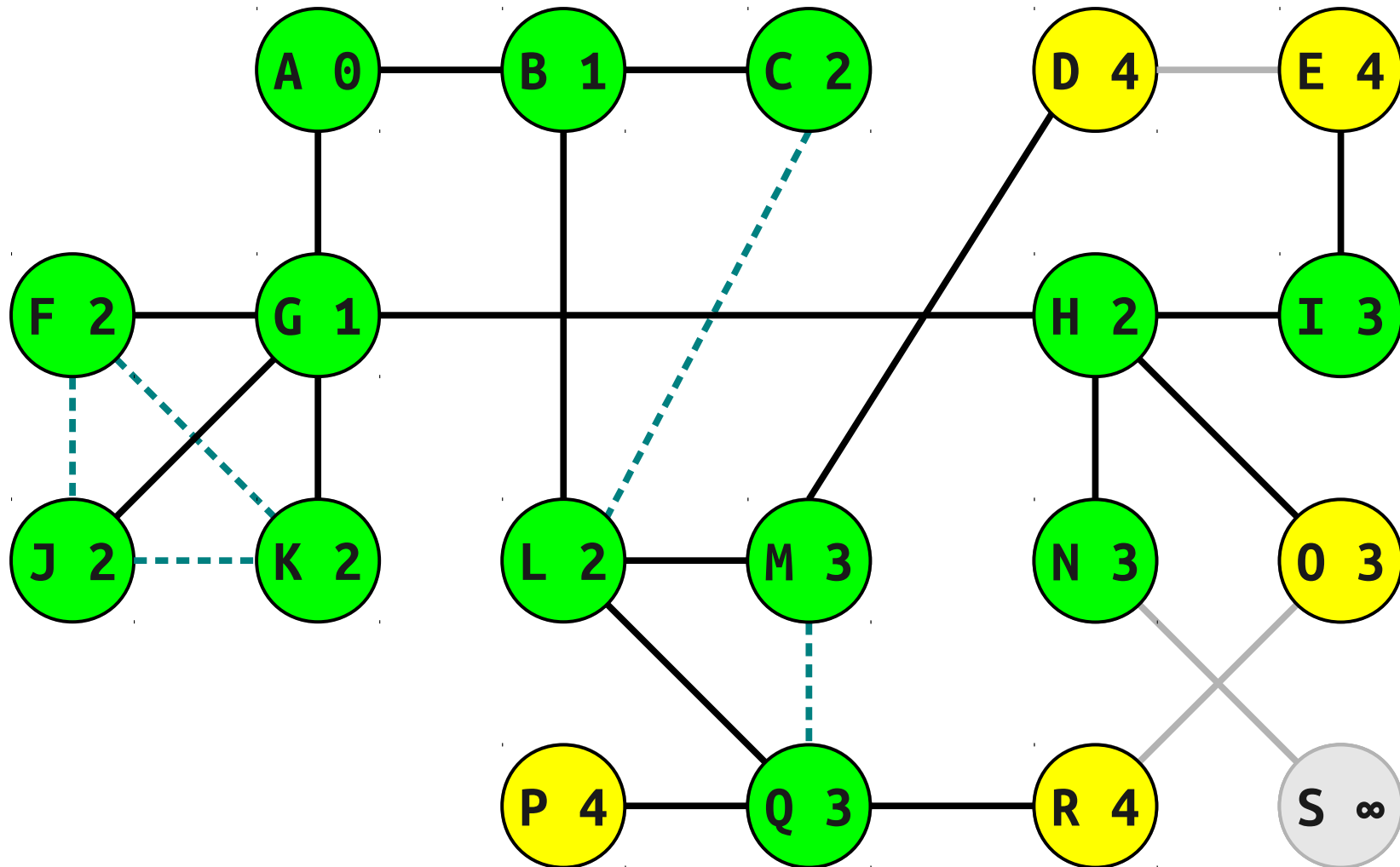
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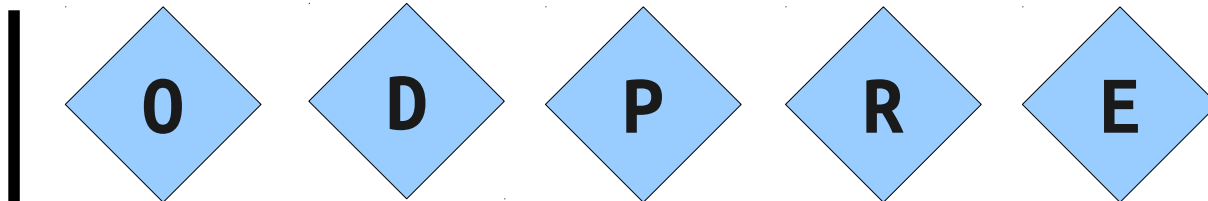
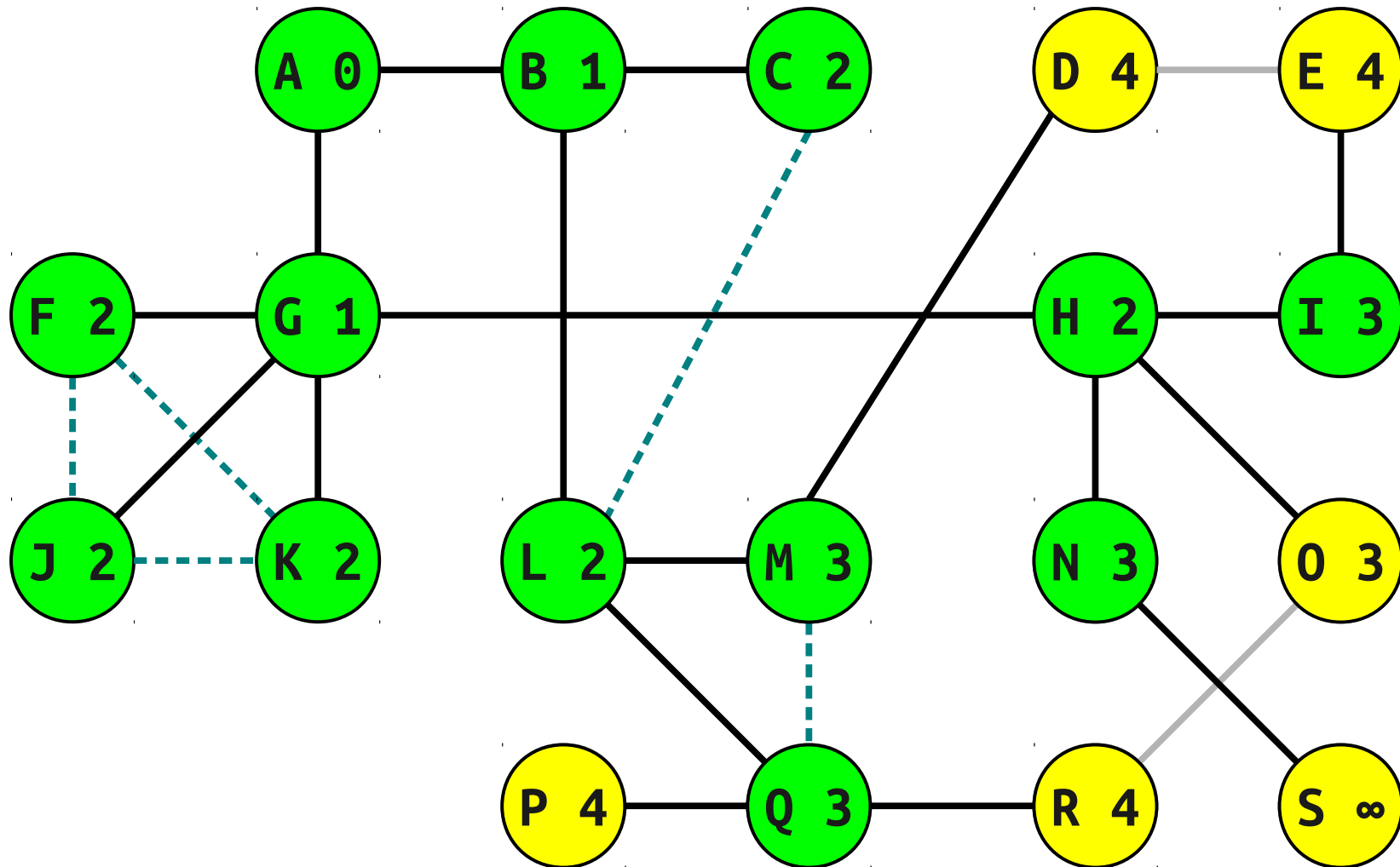
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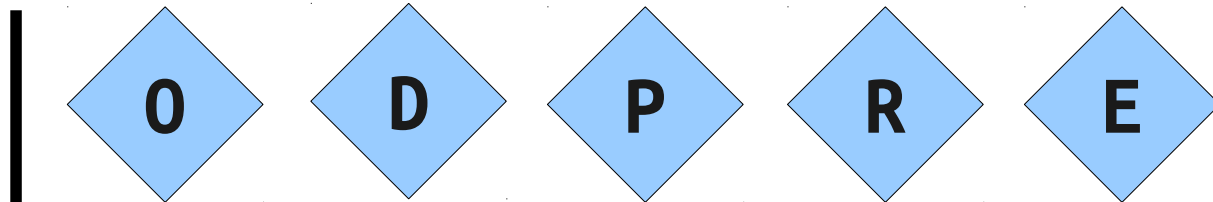
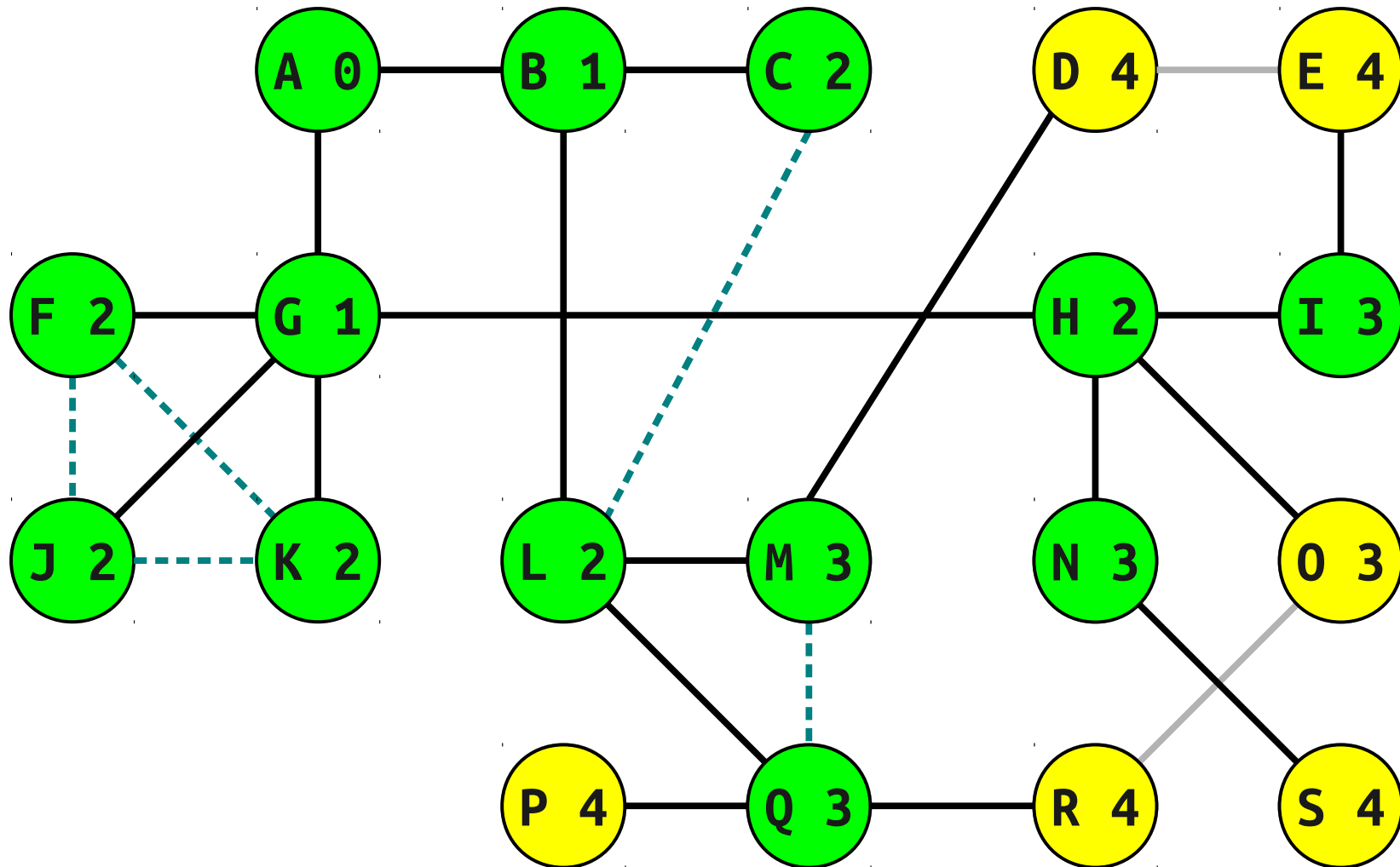
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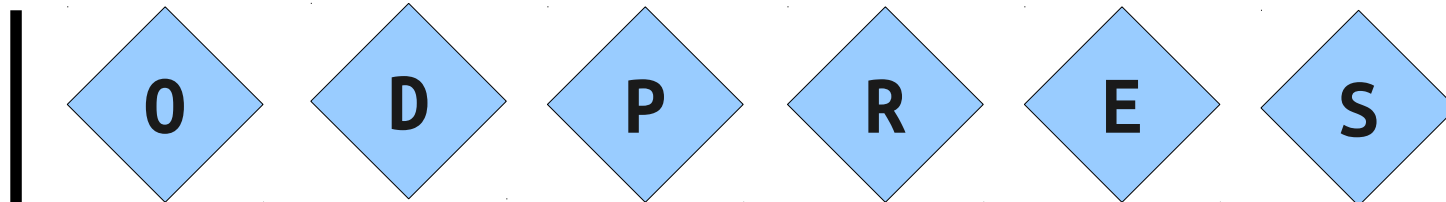
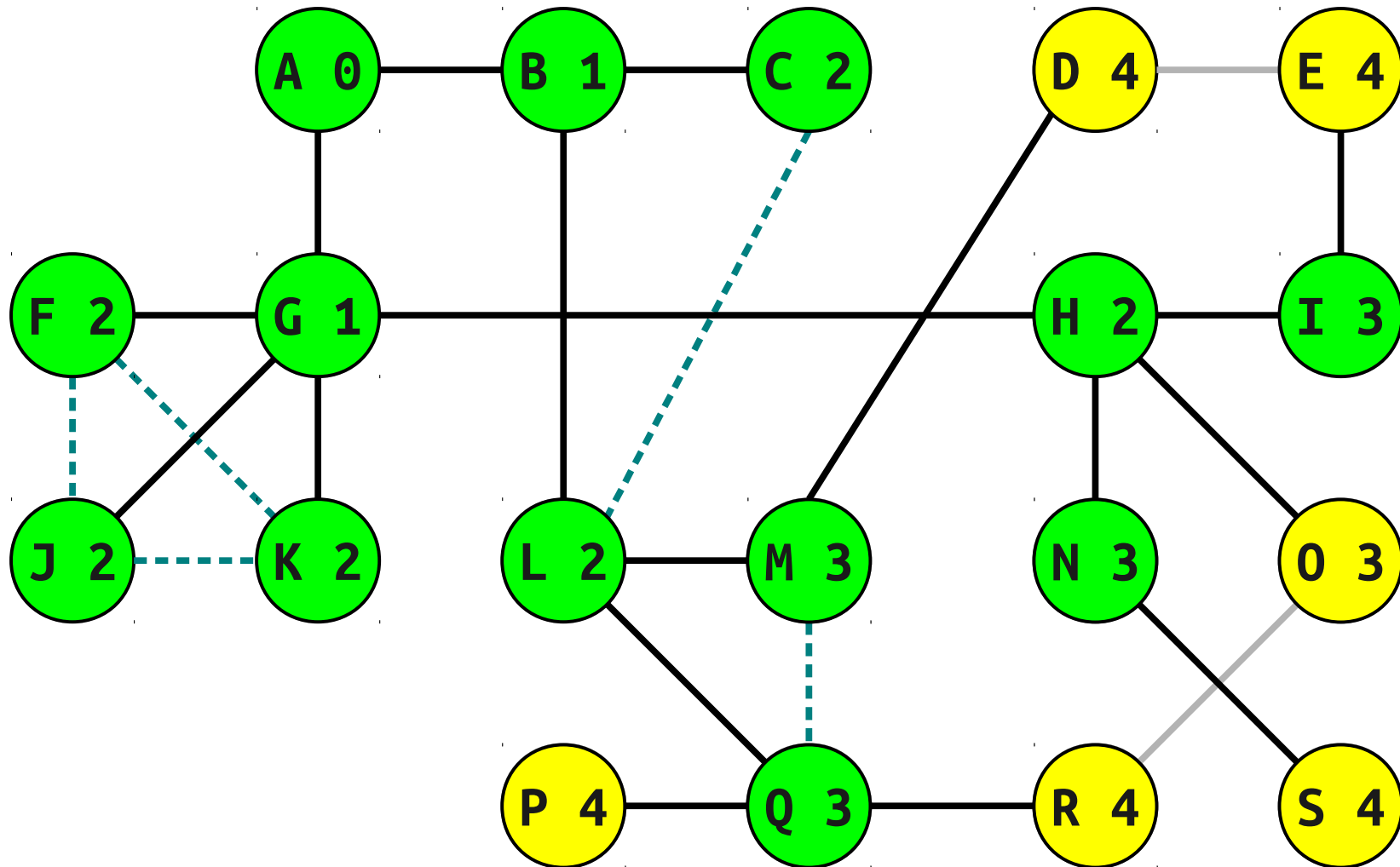
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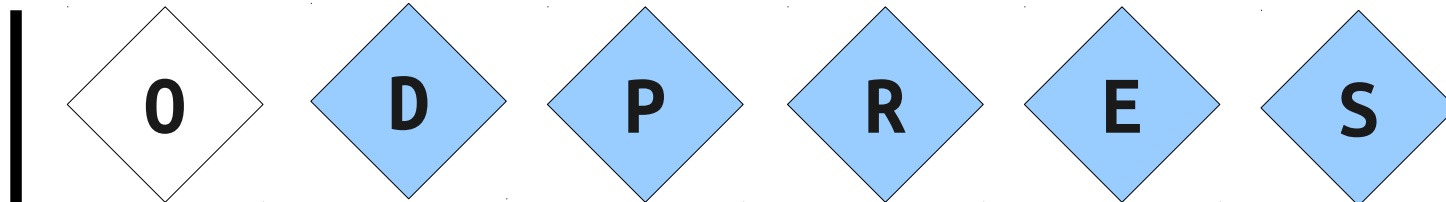
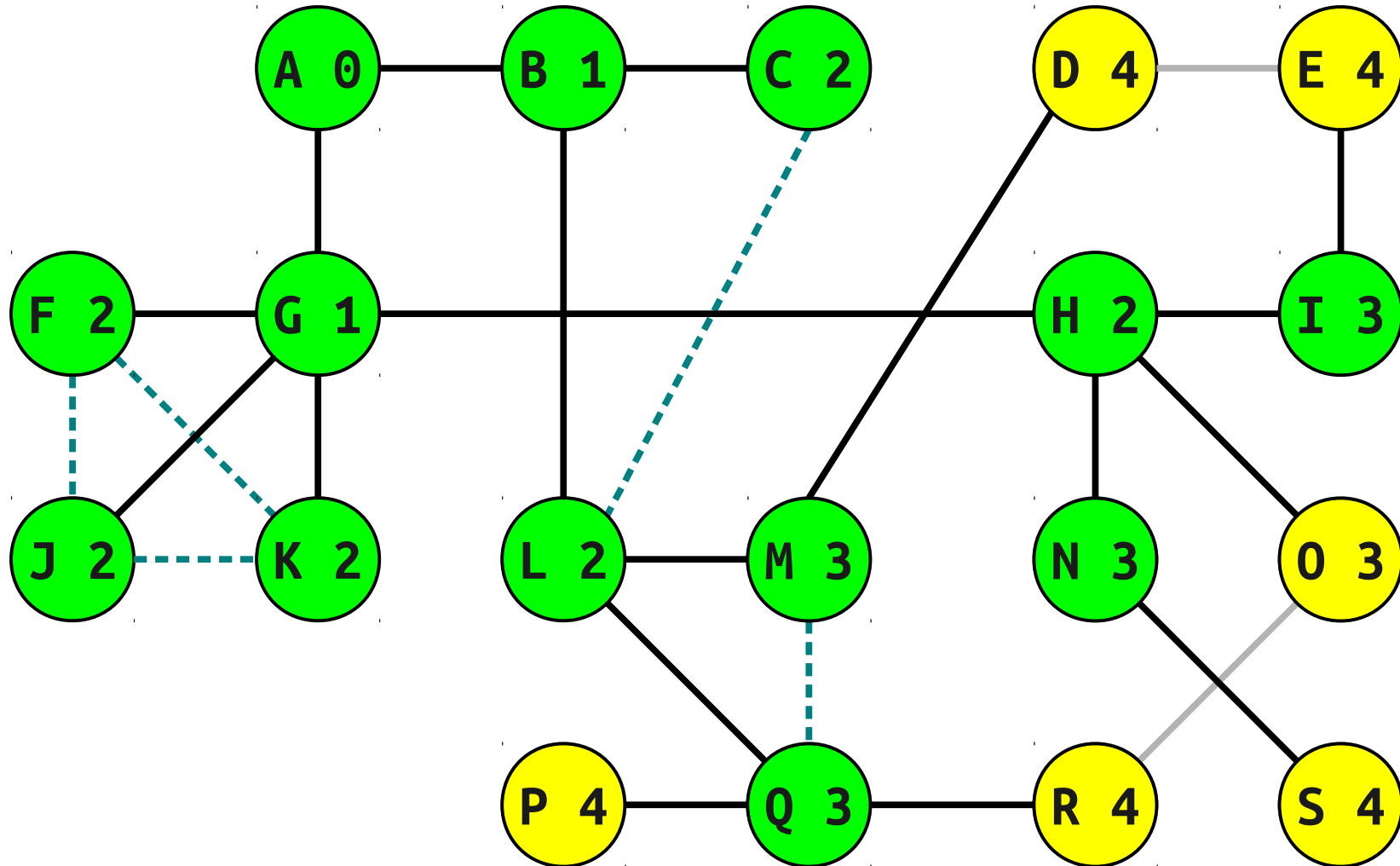
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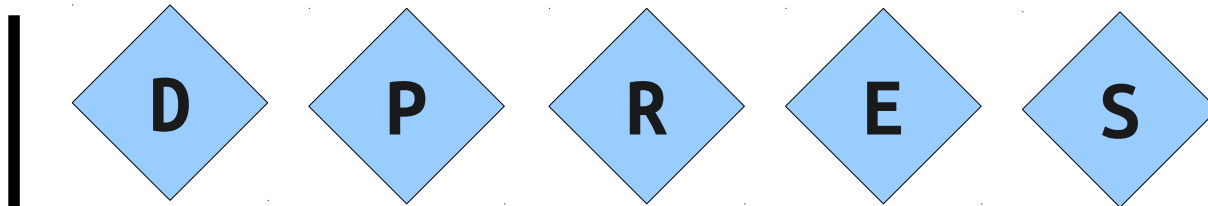
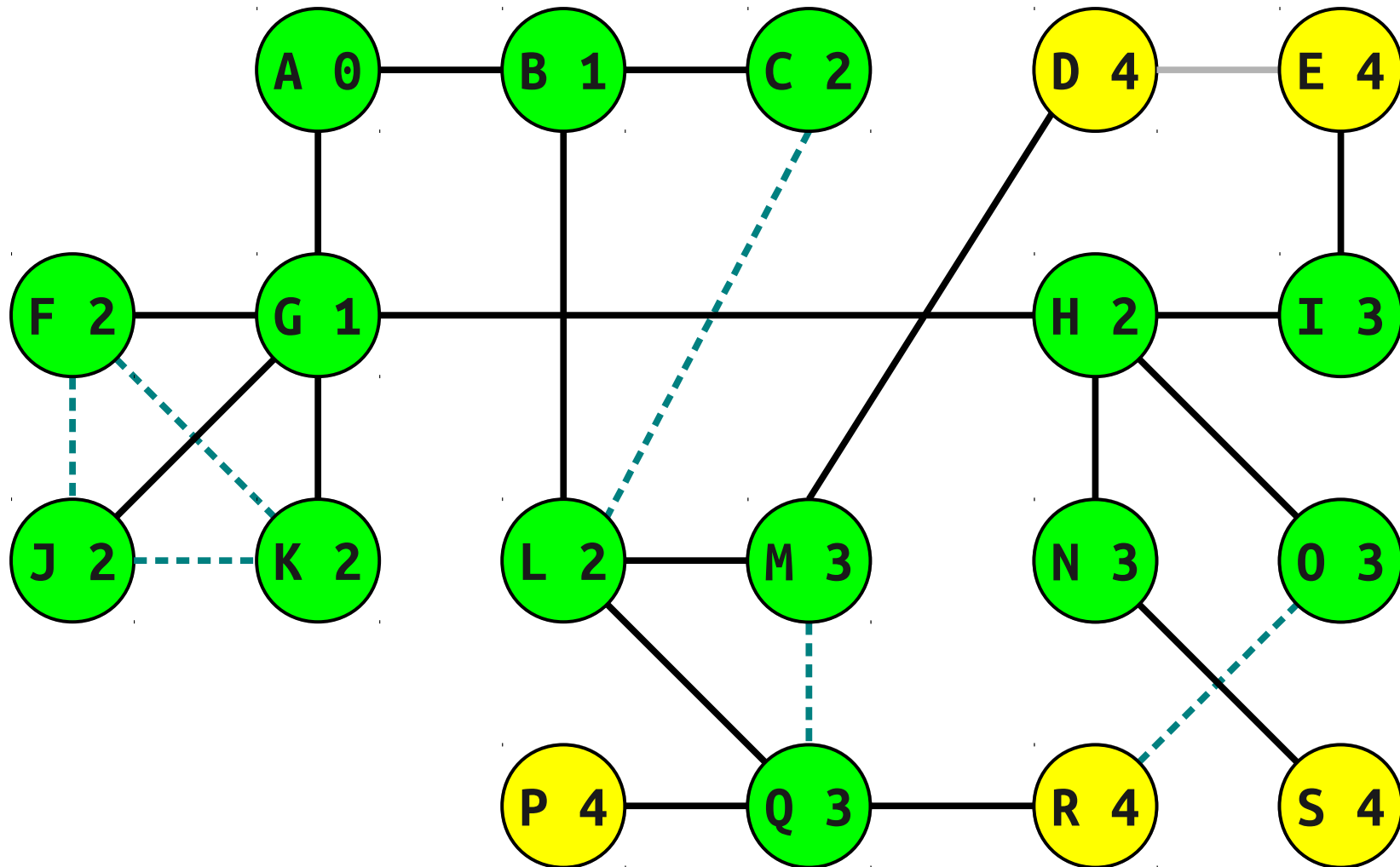
# Breadth-First Search



# Breadth-First Search

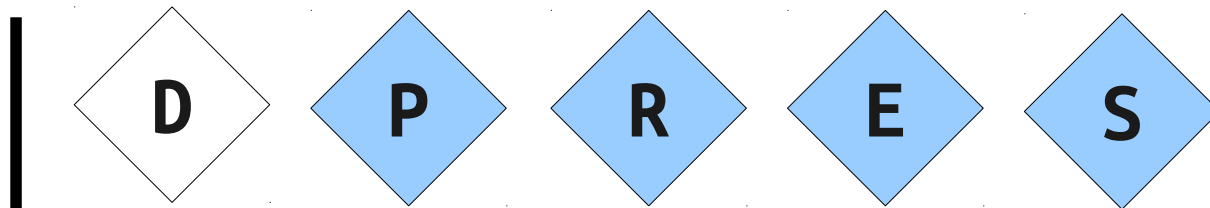
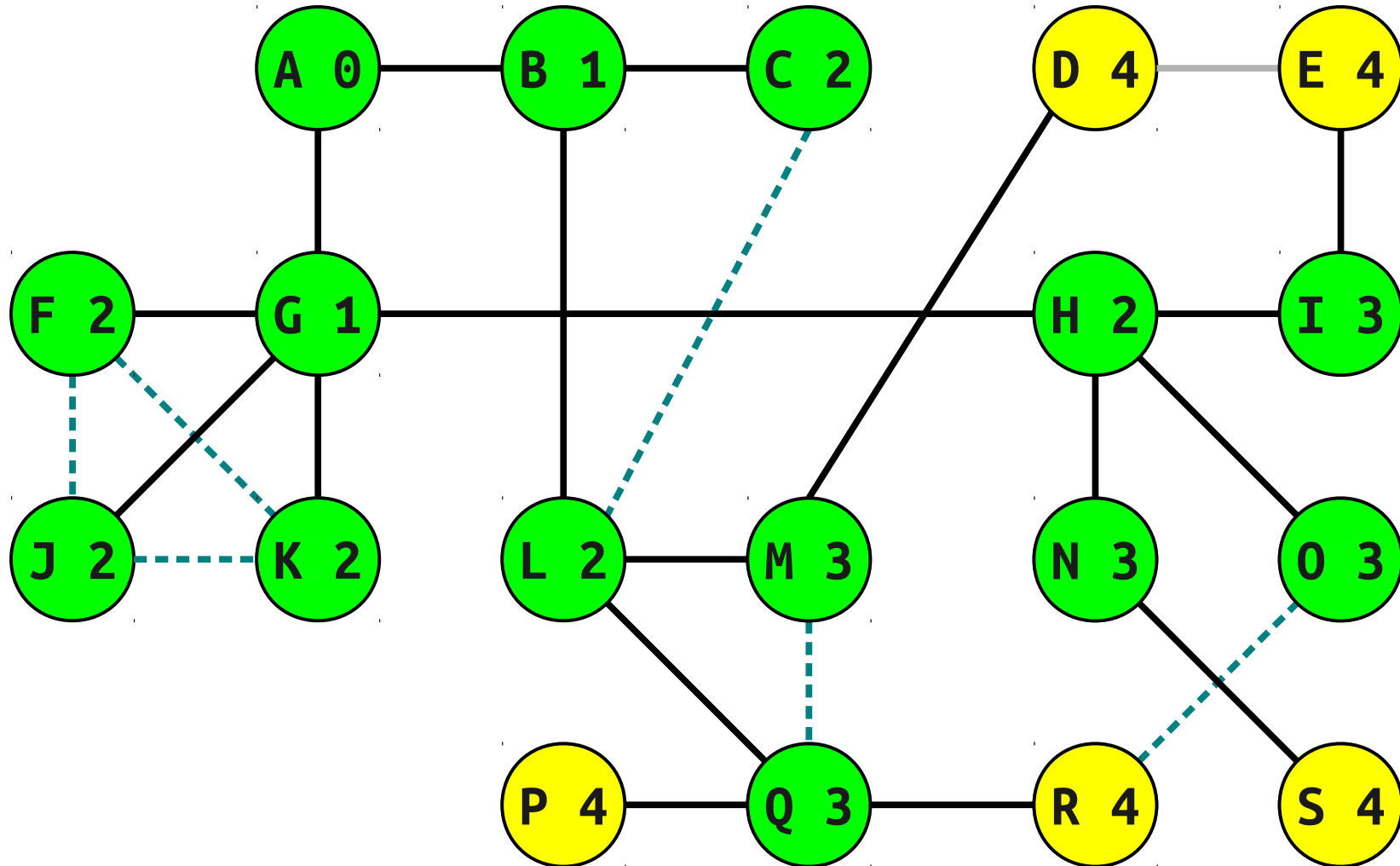


# Breadth-First Search

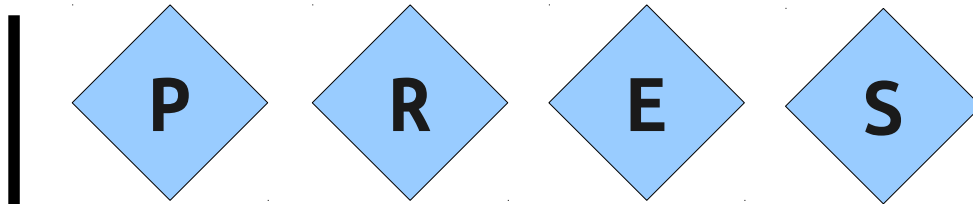
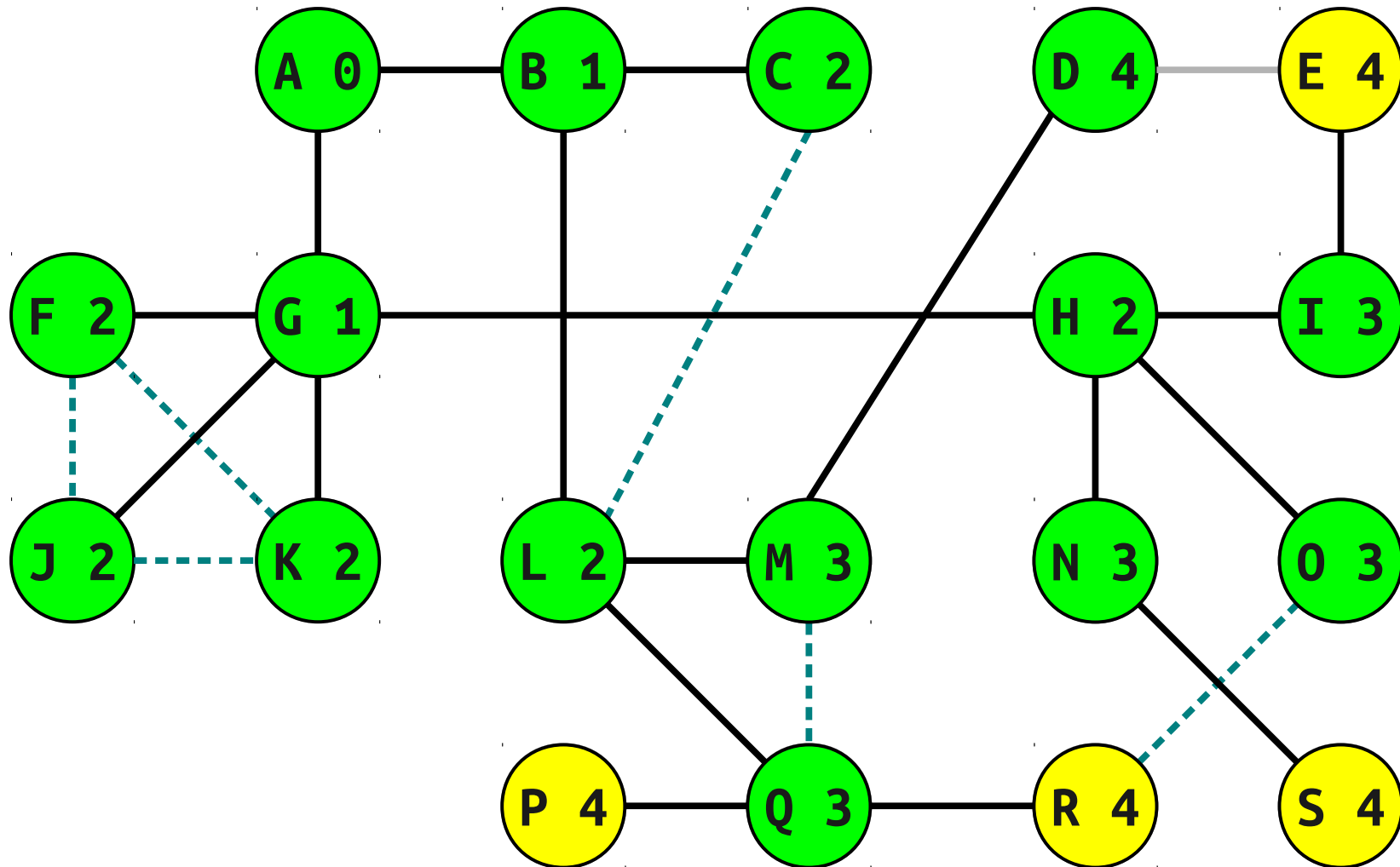




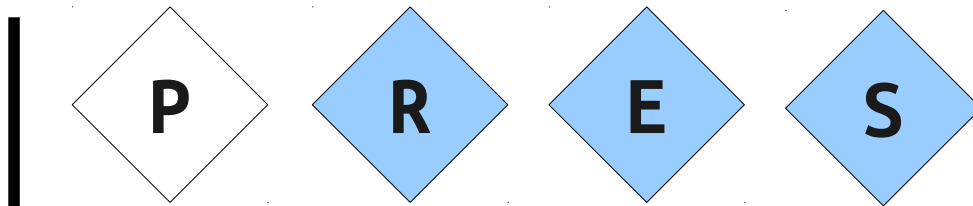
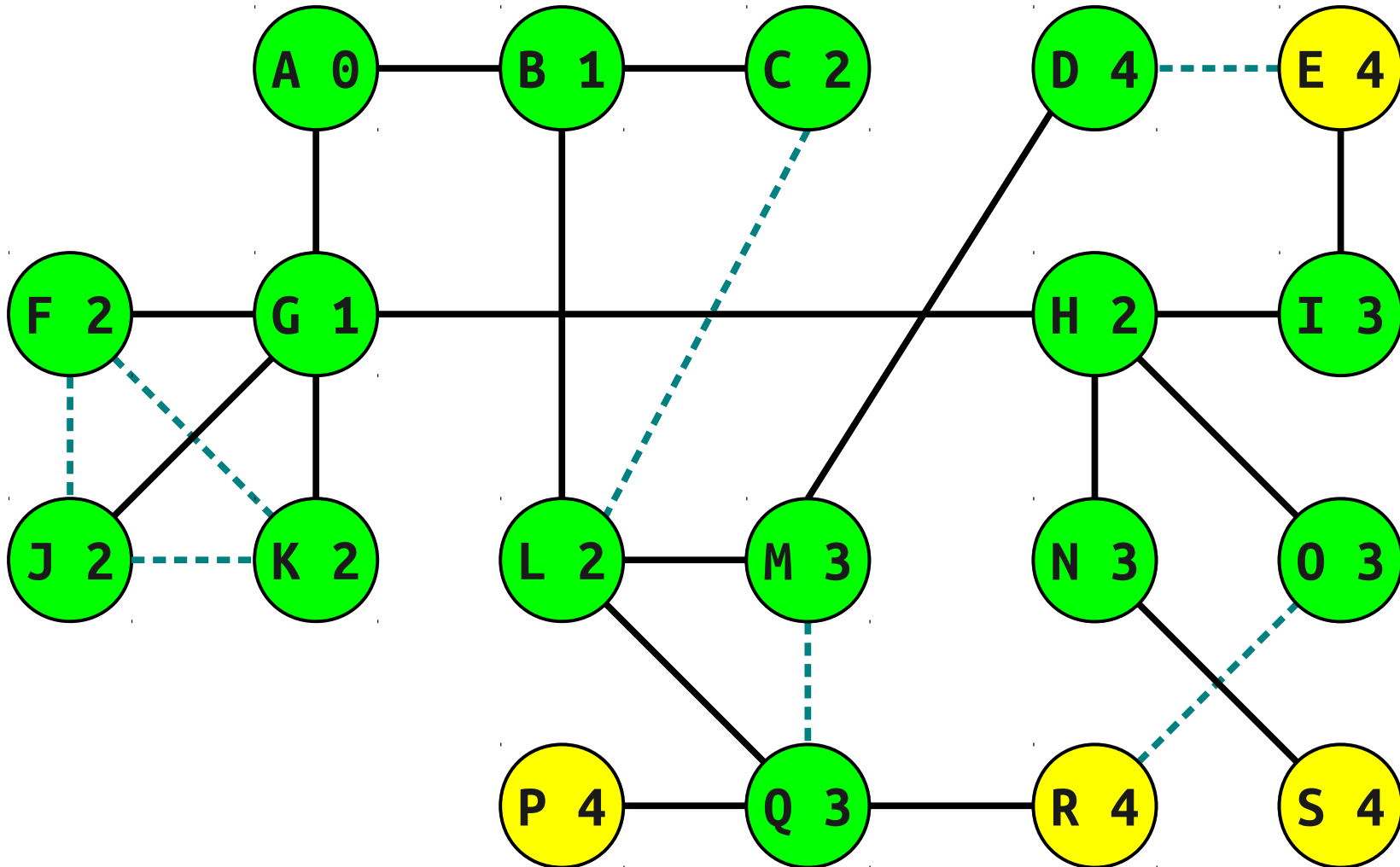
# Breadth-First Search



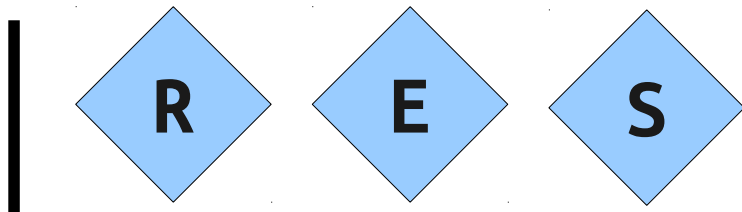
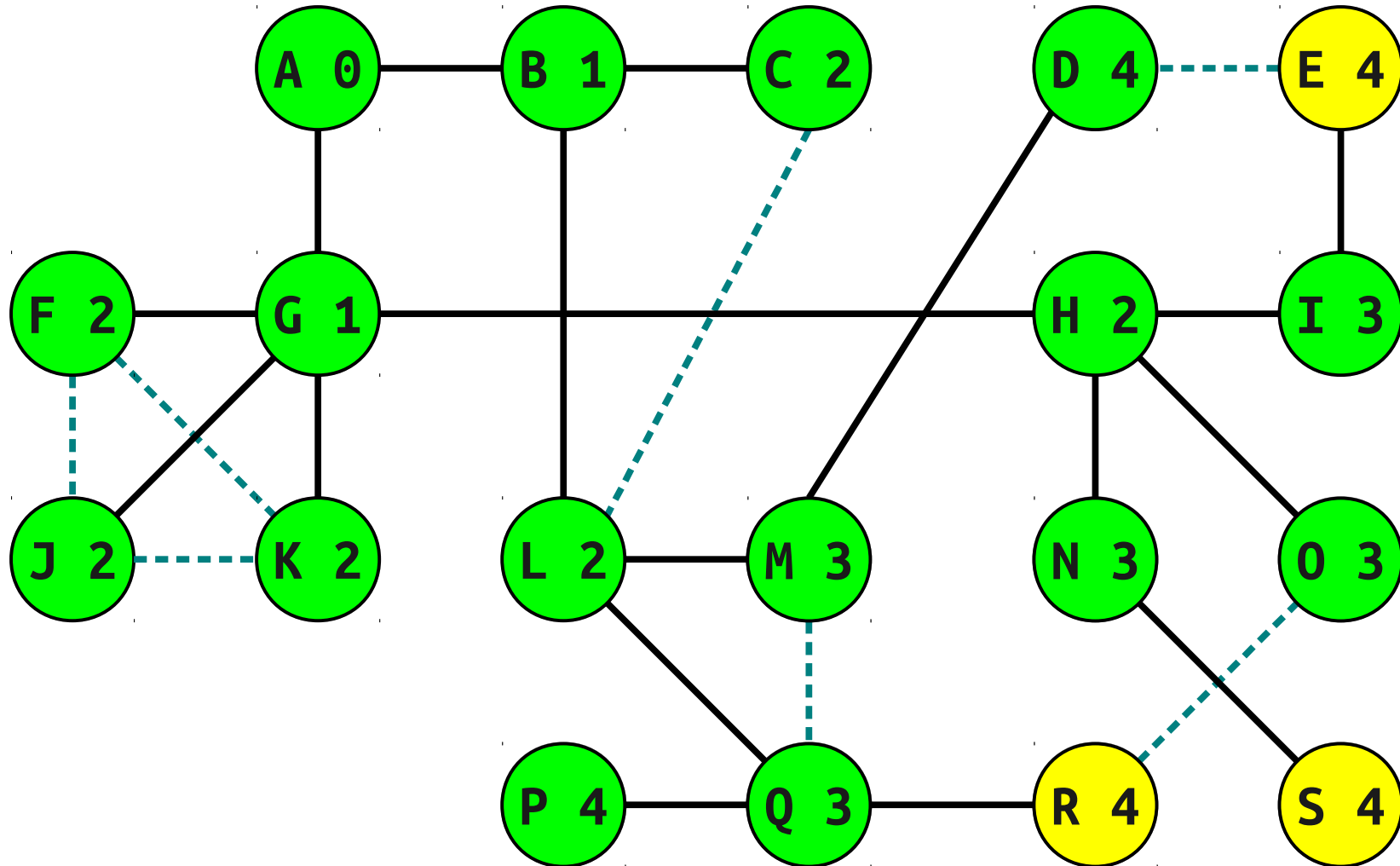
# Breadth-First Search



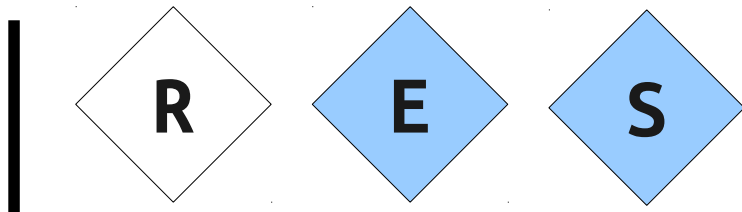
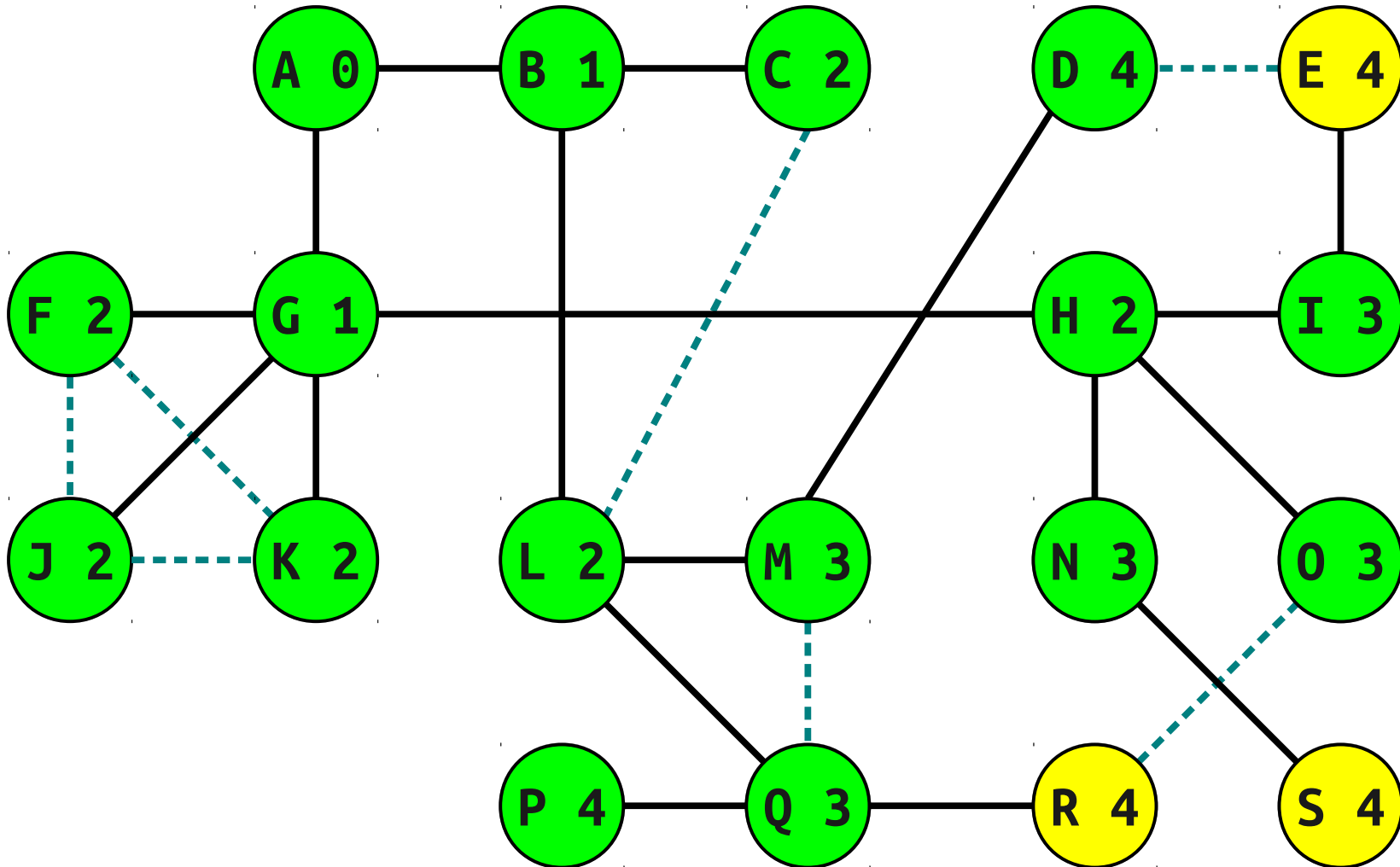
# Breadth-First Search



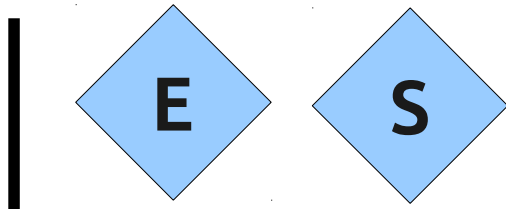
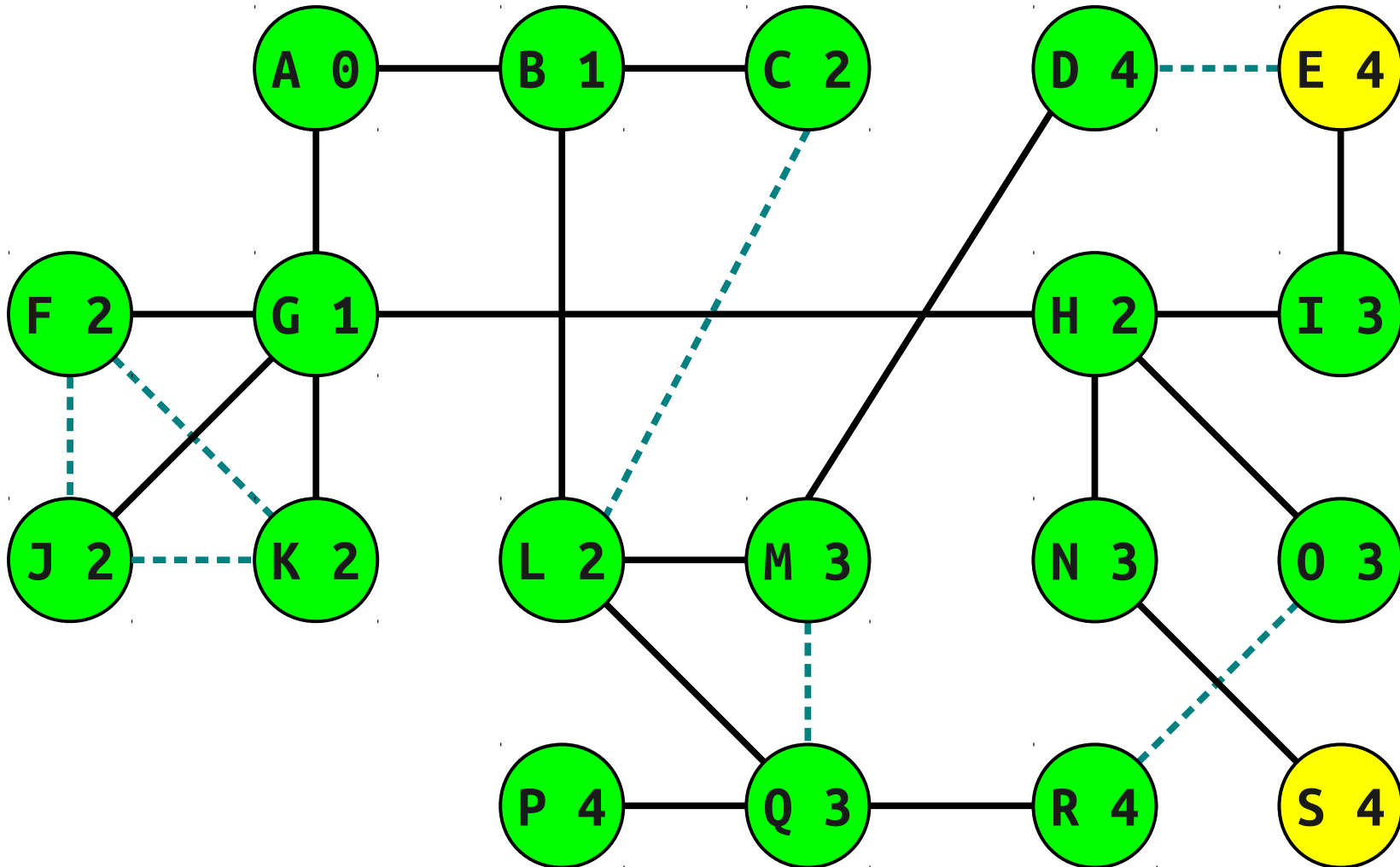
# Breadth-First Search



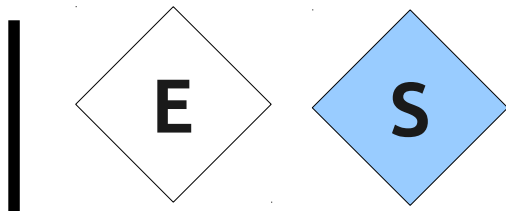
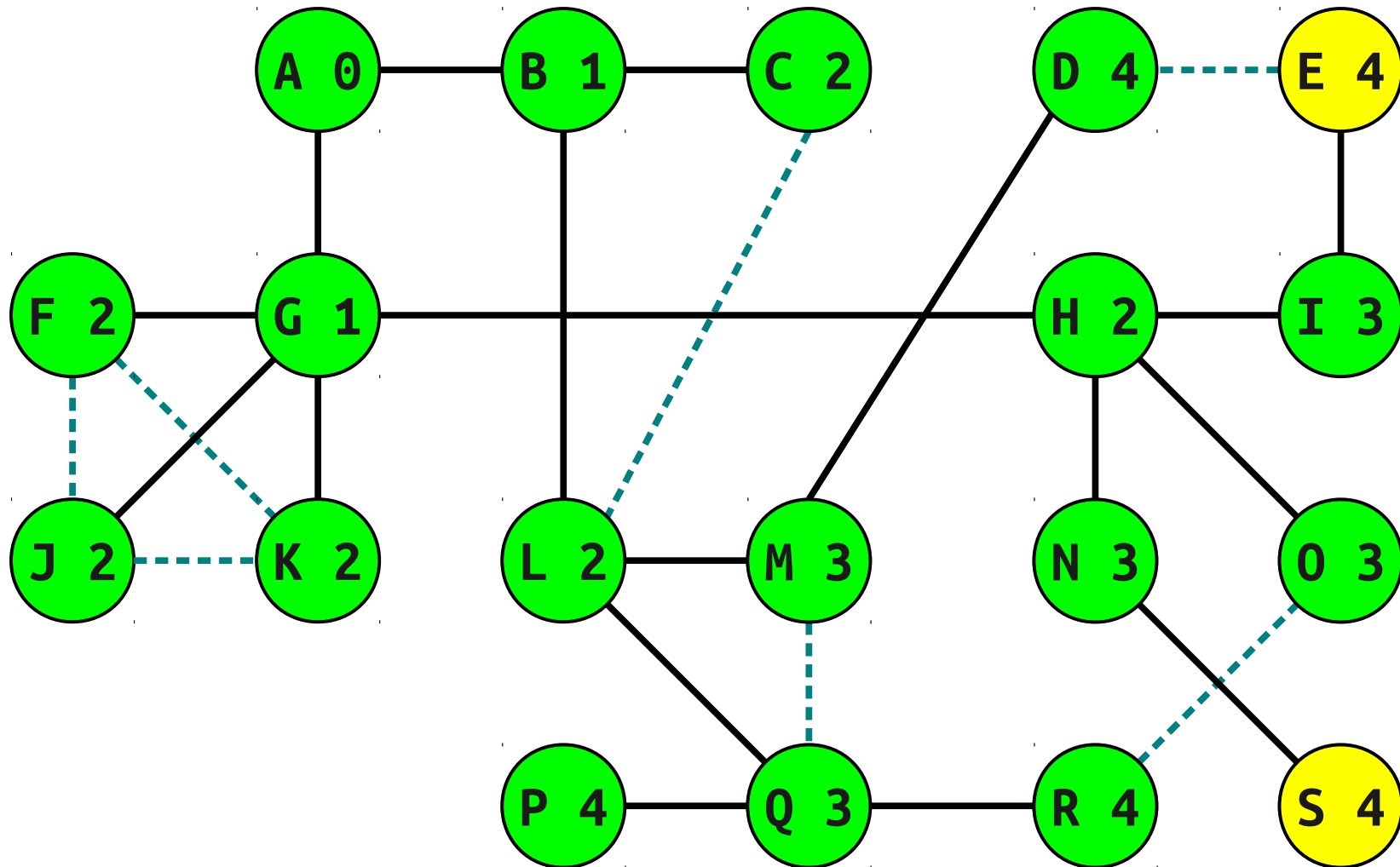
# Breadth-First Search



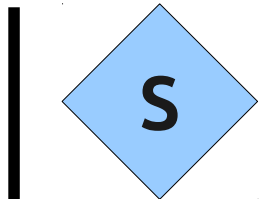
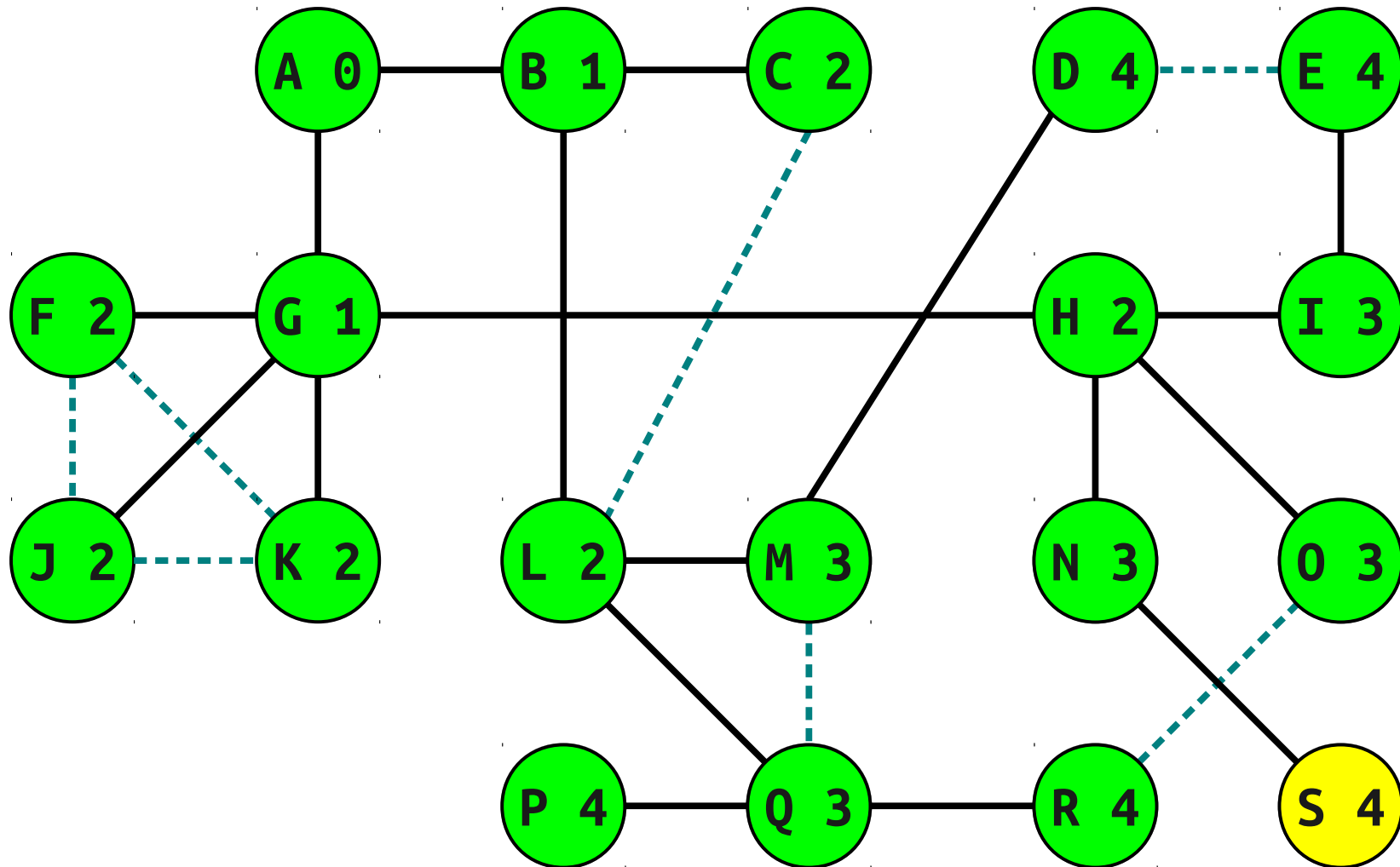
# Breadth-First Search



# Breadth-First Search

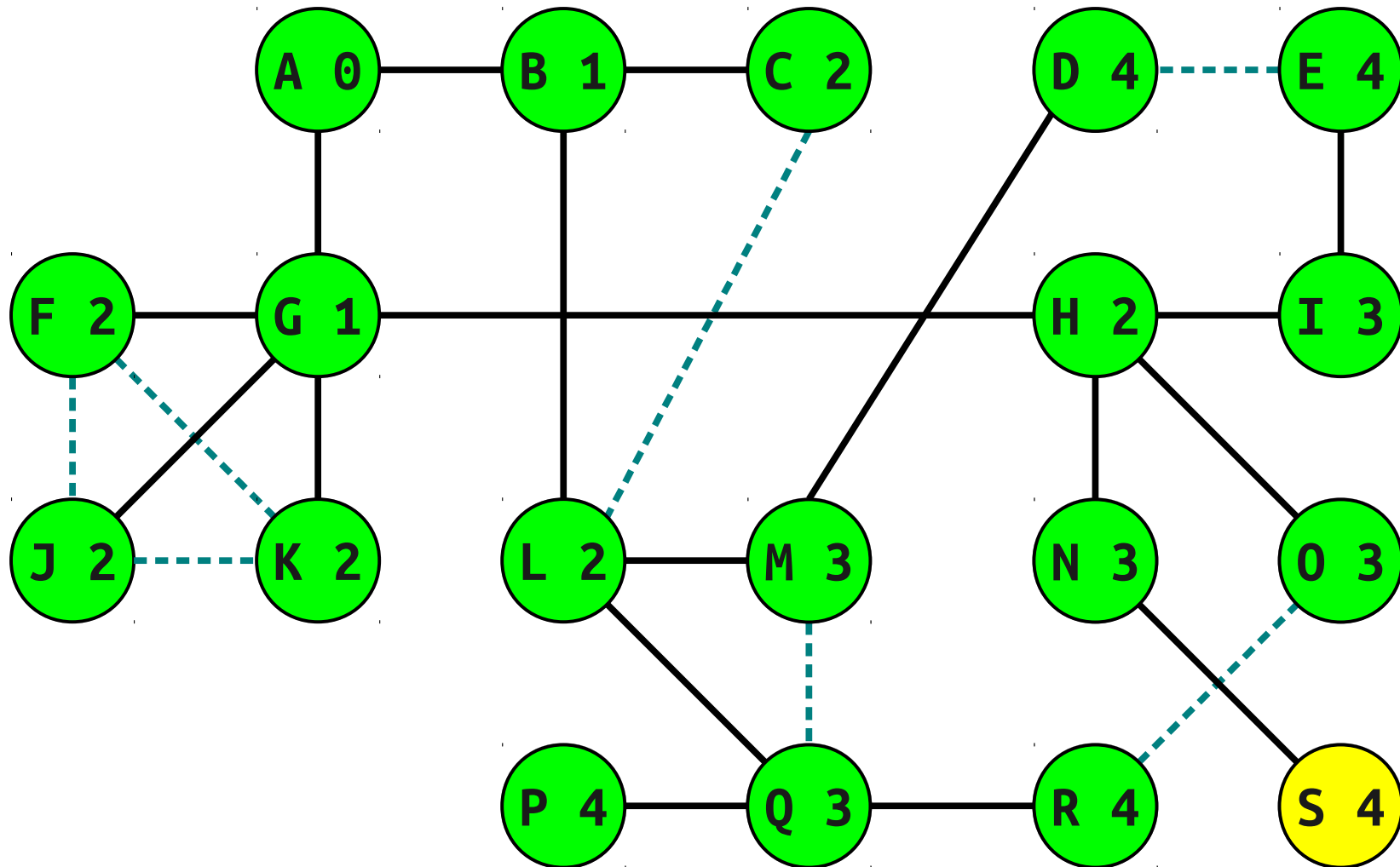


# Breadth-First Search

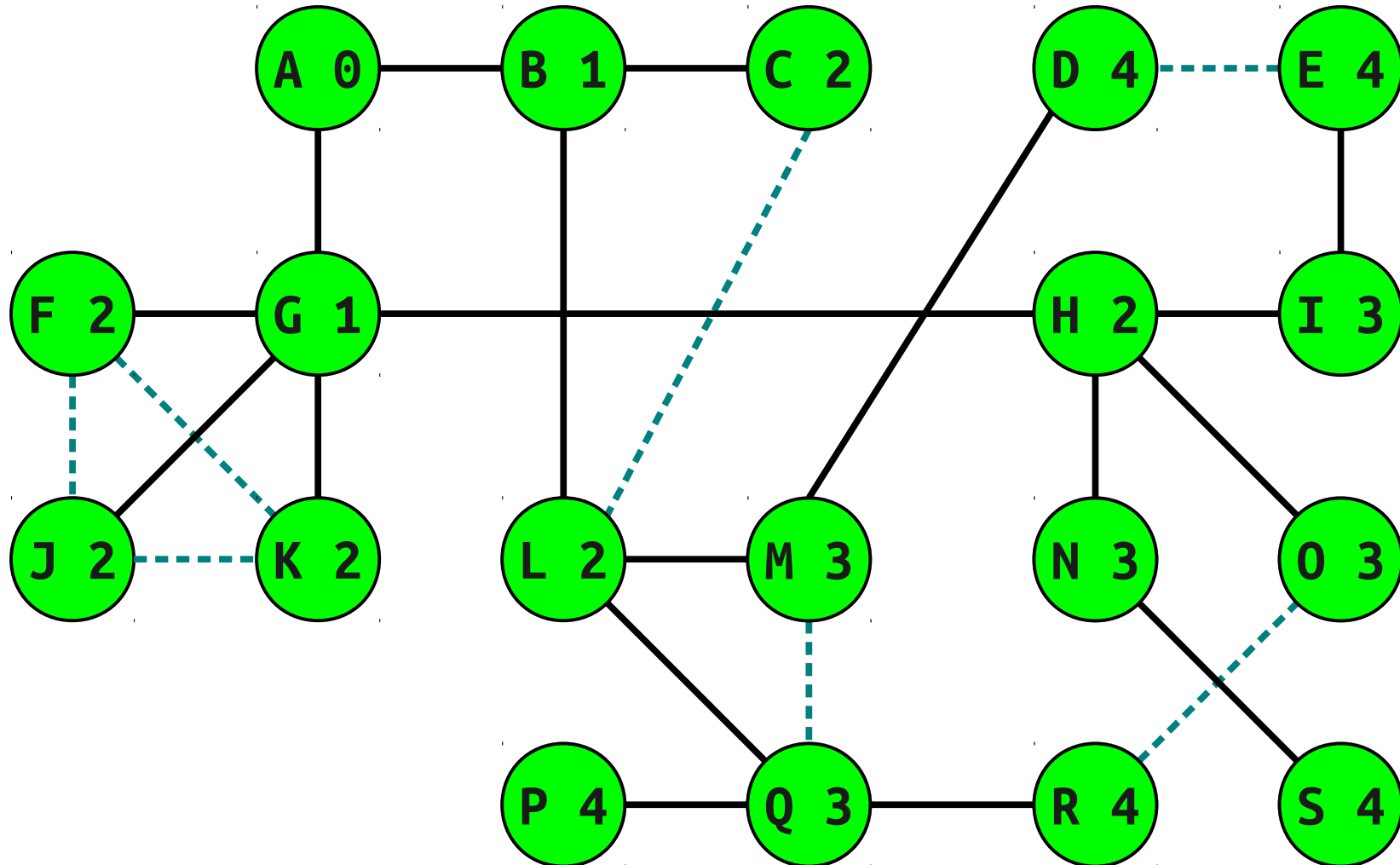




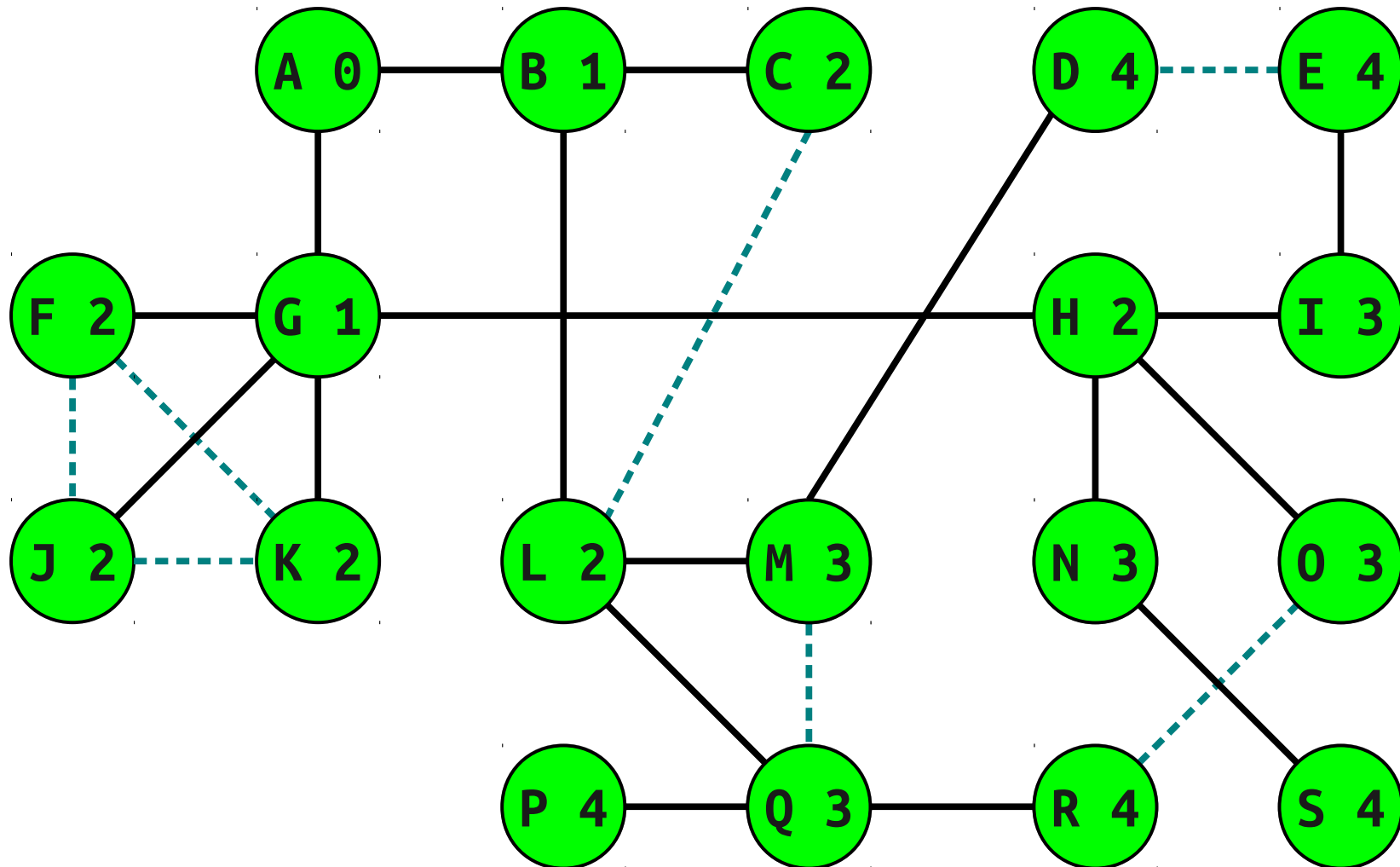
# Breadth-First Search



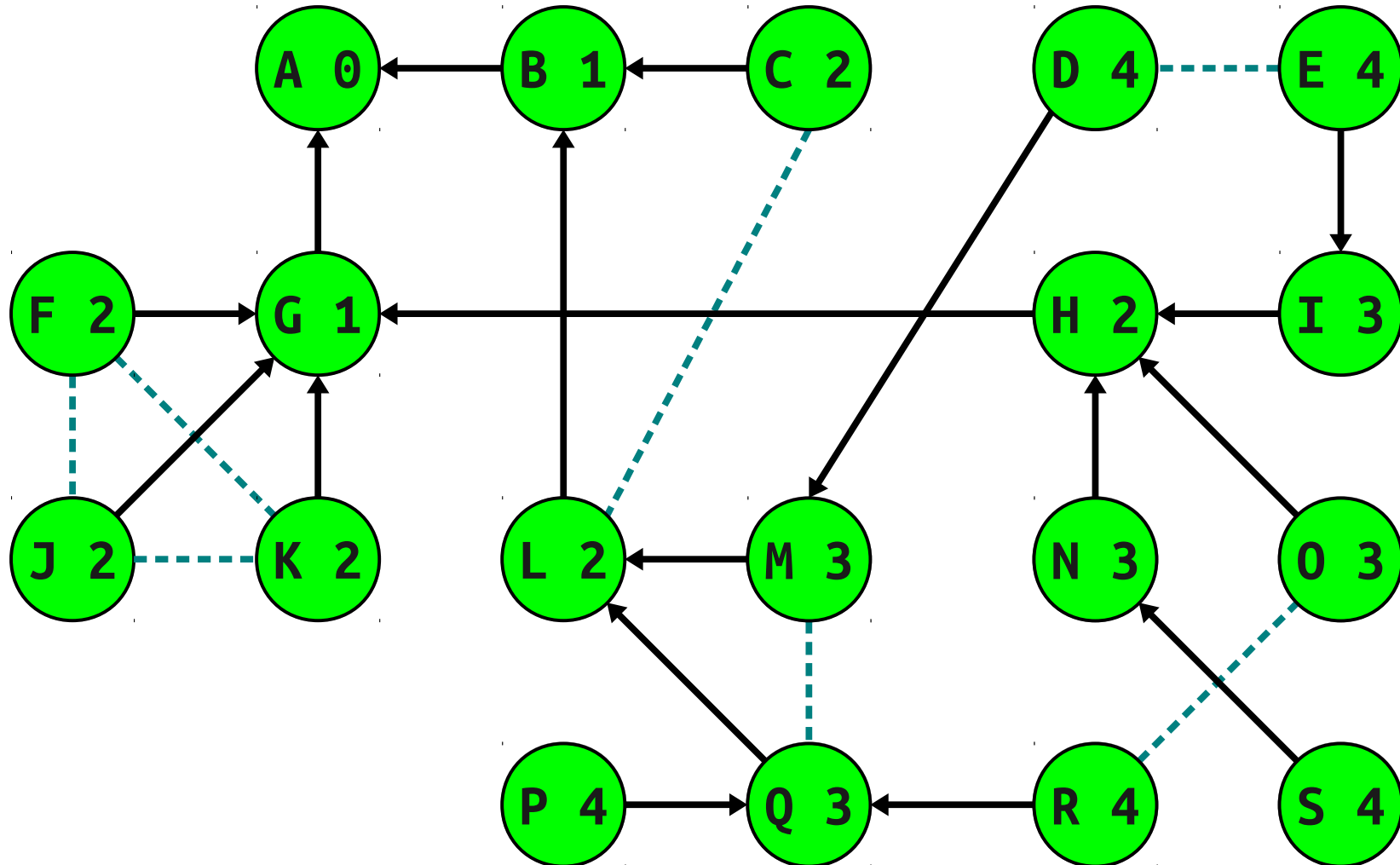
# Breadth-First Search



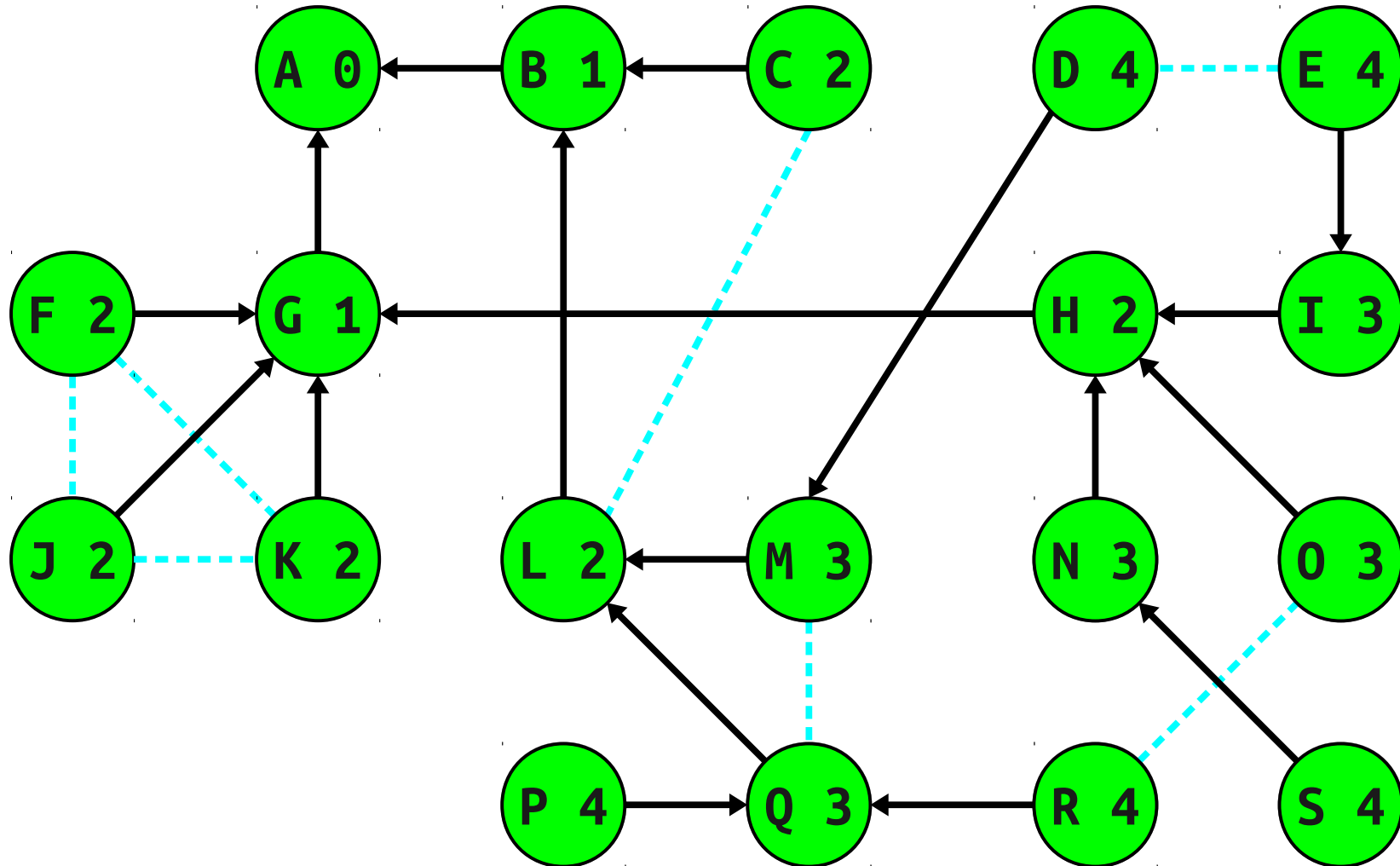
# Breadth-First Search



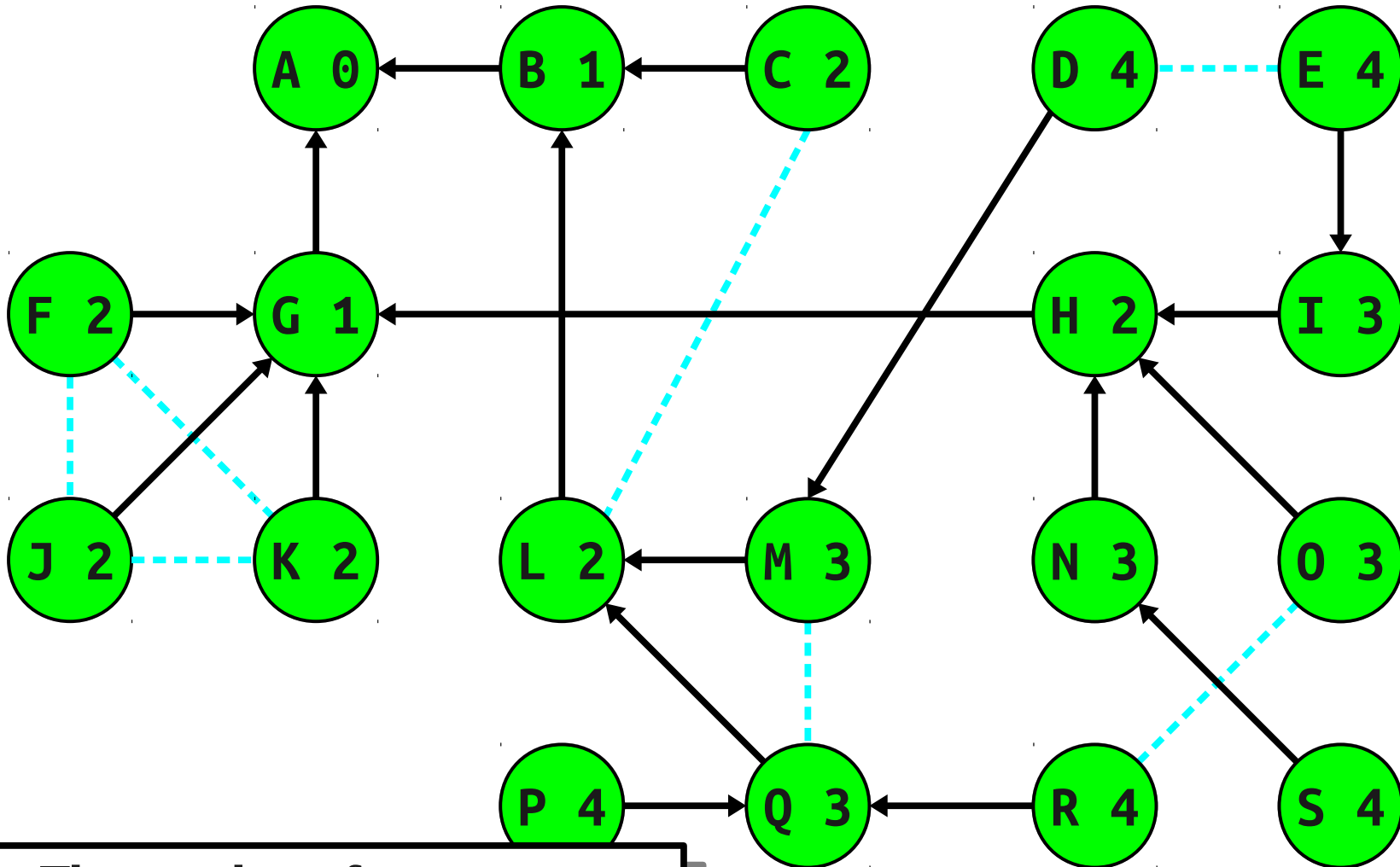
# Breadth-First Search



# Breadth-First Search



# Breadth-First Search



These edges form a **breadth-first search tree**: the path from any  $v$  to node  $A$  gives a shortest path from  $v$  to  $A$ .

```
procedure breadthFirstSearch(s, G):  
  let  $q$  be a new queue.  
  for each node  $v$  in  $G$ :  
     $\text{dist}[v] = \infty$   
  
   $\text{dist}[s] = 0$   
  enqueue( $s$ ,  $q$ )  
  
  while  $q$  is not empty:  
    let  $v = \text{dequeue}(q)$   
    for each neighbor  $u$  of  $v$ :  
      if  $\text{dist}[u] = \infty$ :  
         $\text{dist}[u] = \text{dist}[v] + 1$   
        enqueue( $u$ ,  $q$ )
```

Question 1: How do we prove this always finds the right distances?

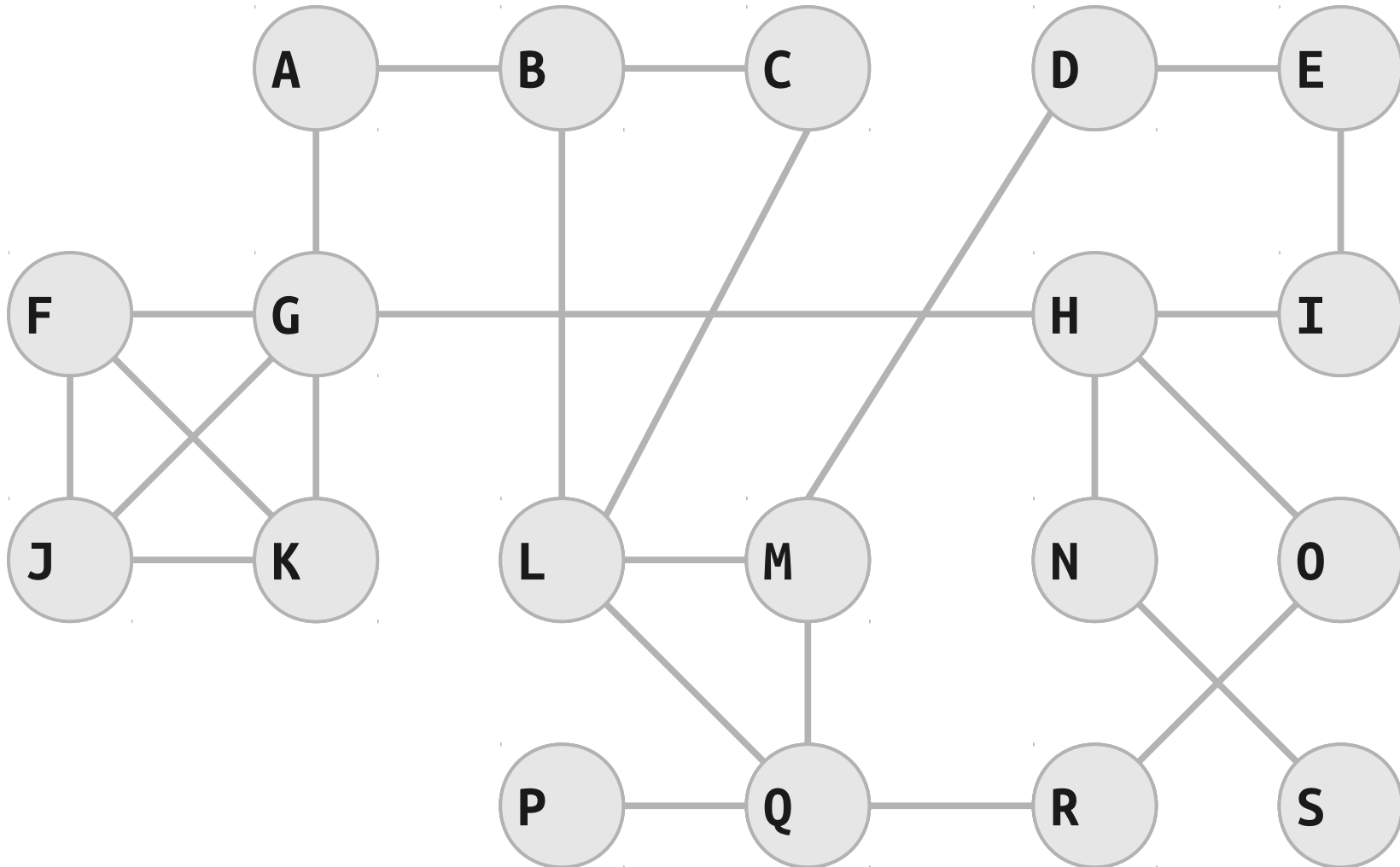
Question 2: How *efficiently* does this find the right distances?



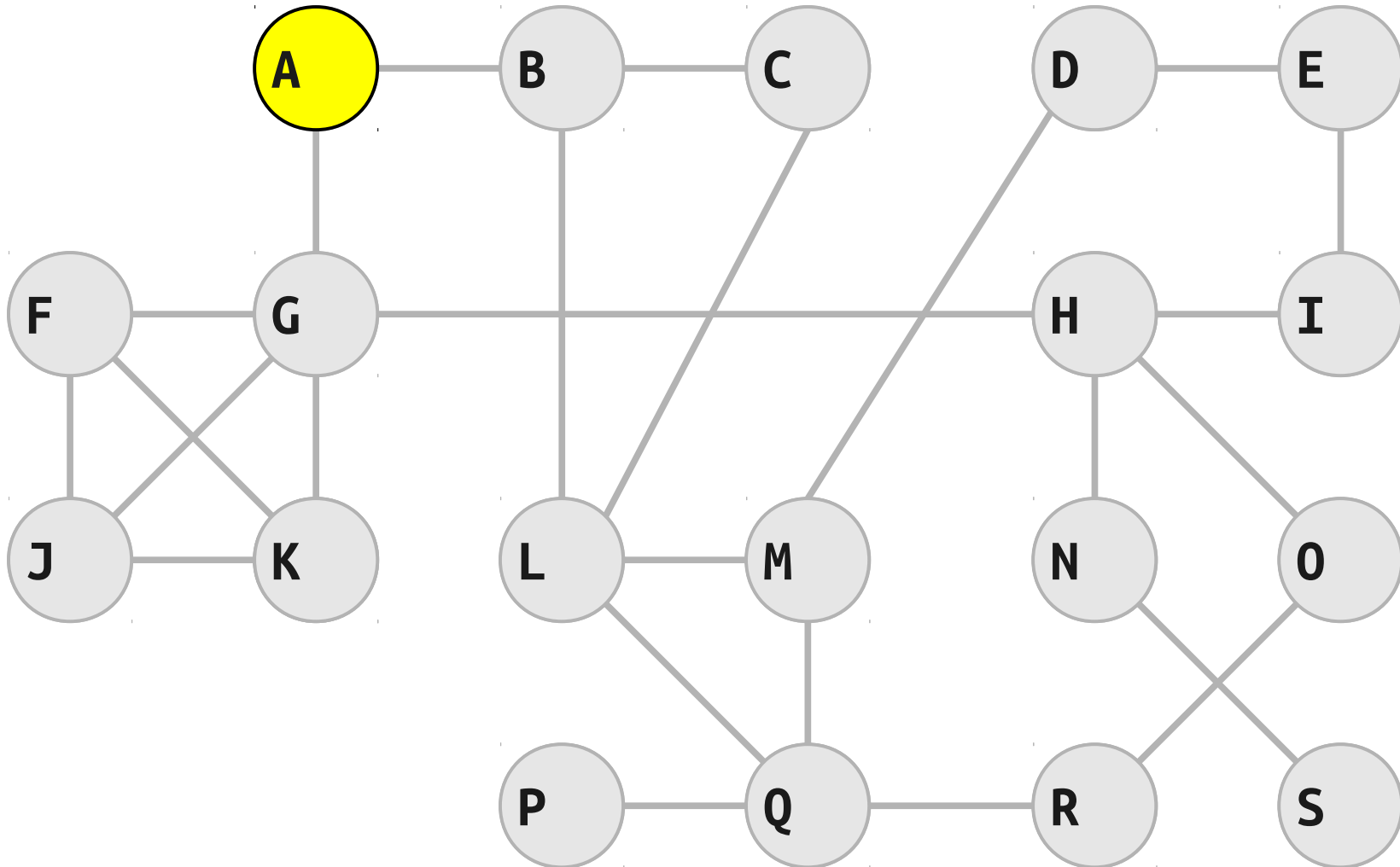
Question 1: How do we prove this always finds the right distances?

Question 2: How *efficiently* does this find the right distances?

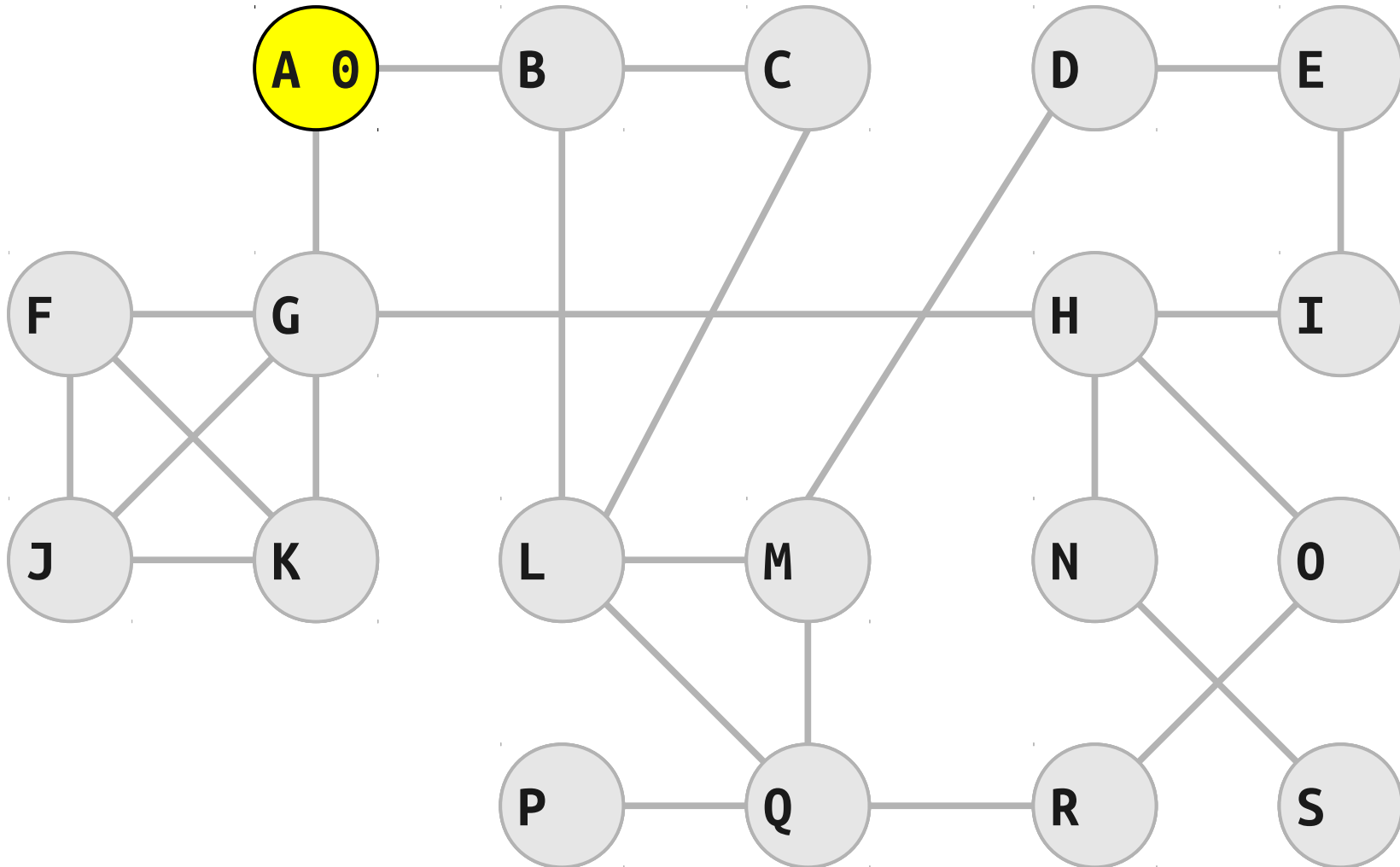
# Breadth-First Search



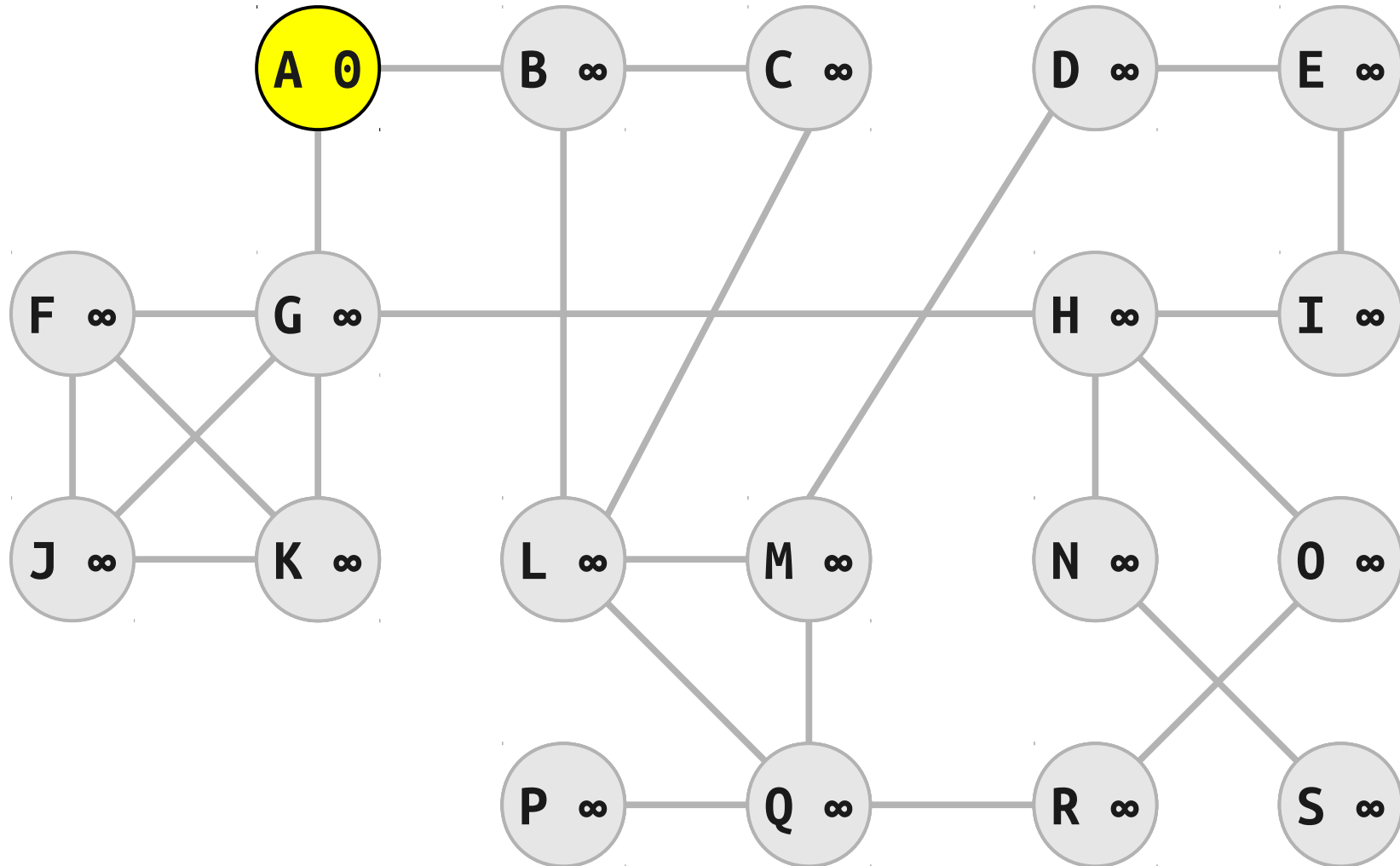
# Breadth-First Search



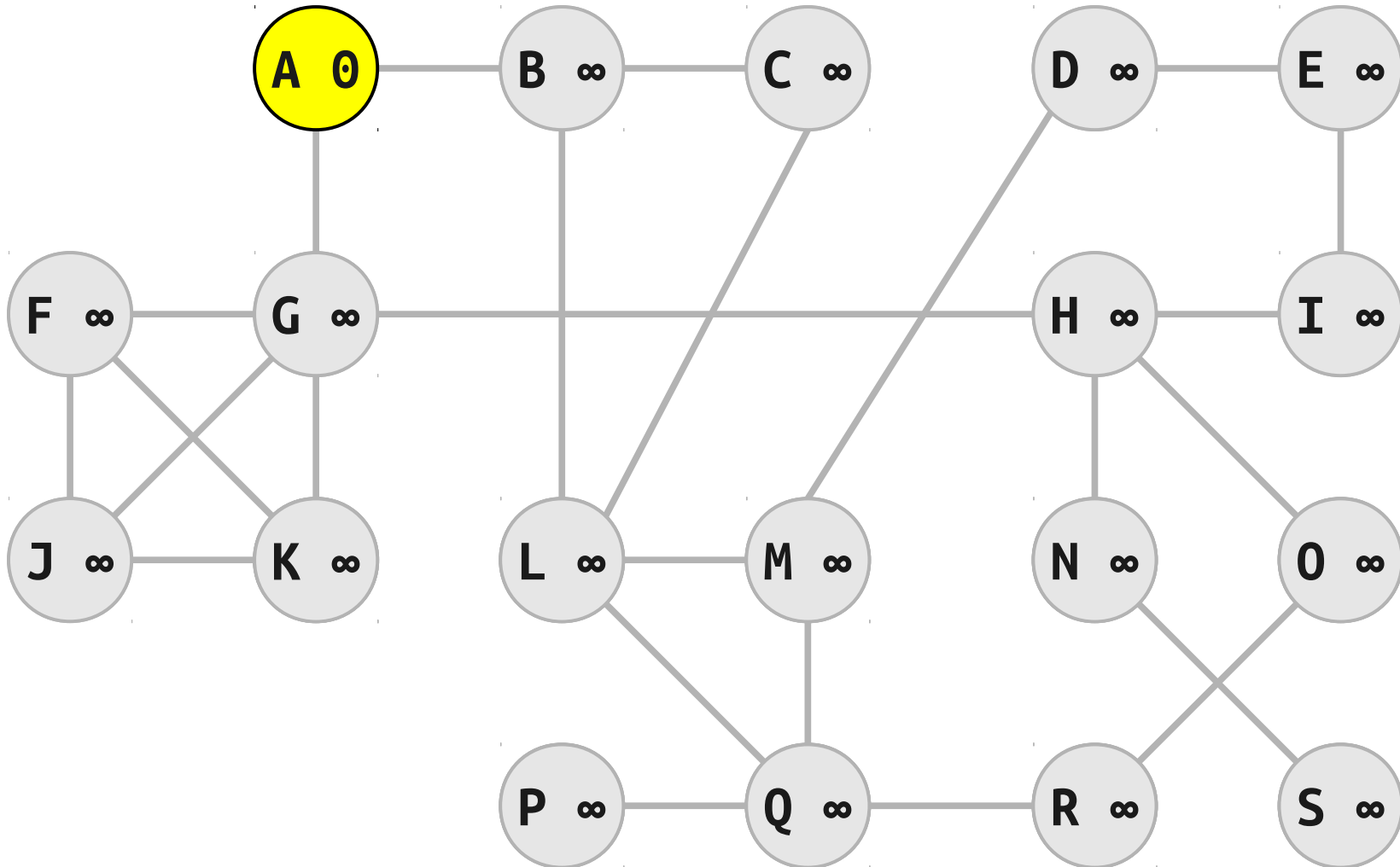
# Breadth-First Search



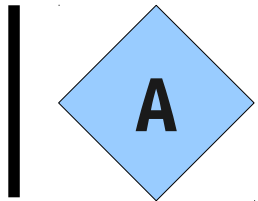
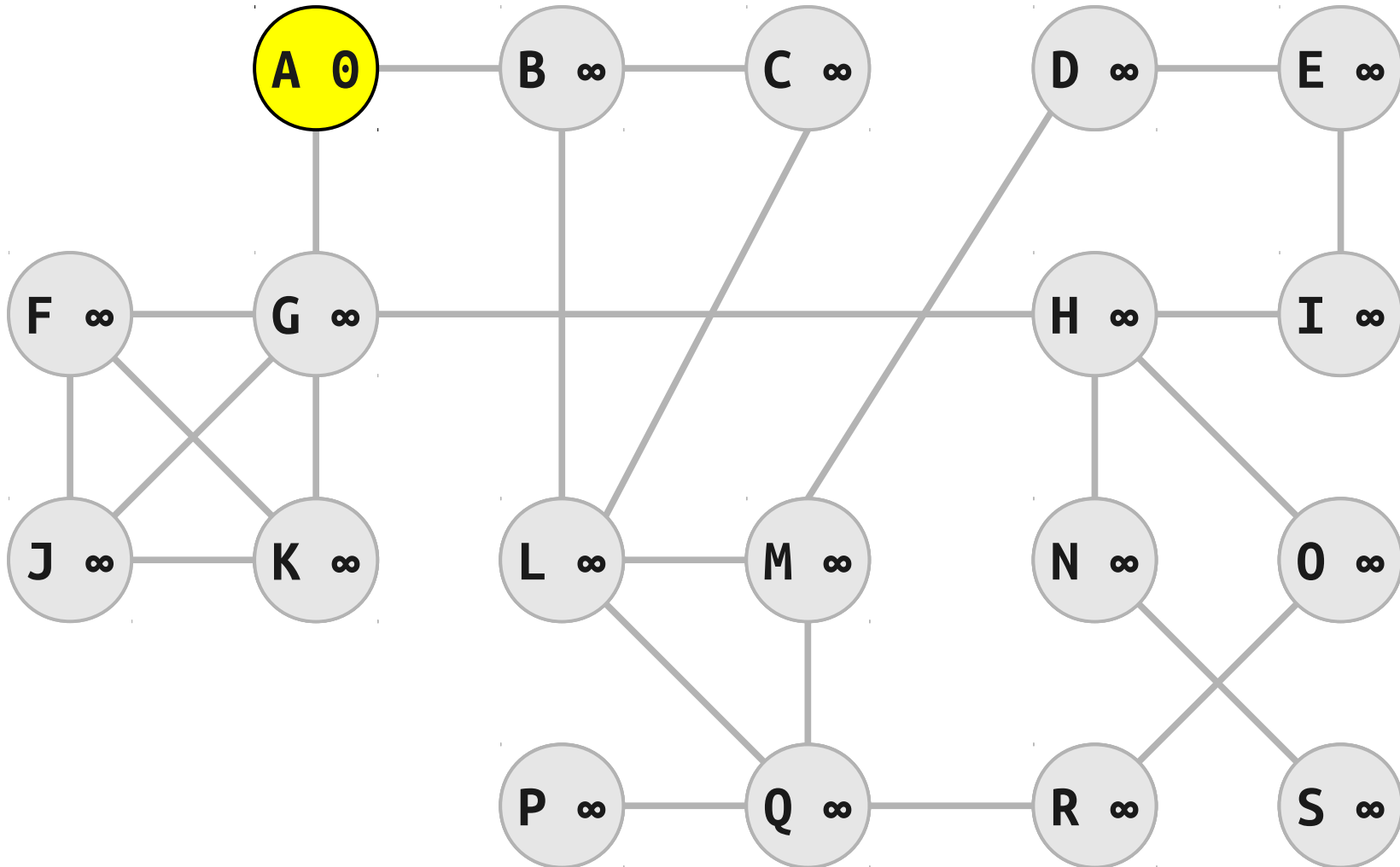
# Breadth-First Search



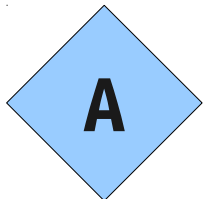
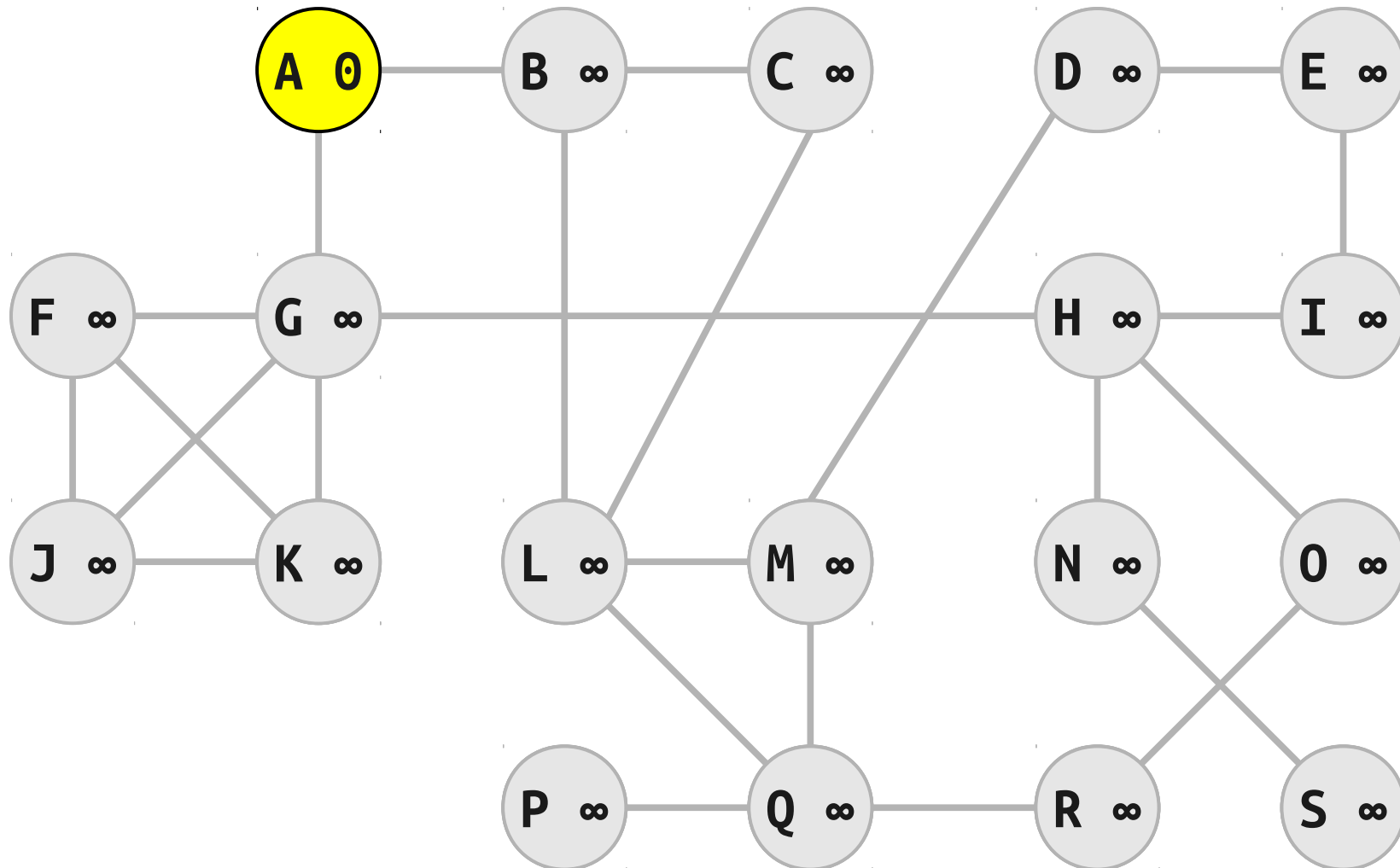
# Breadth-First Search



# Breadth-First Search



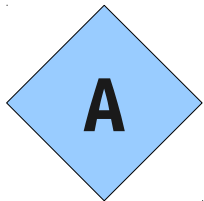
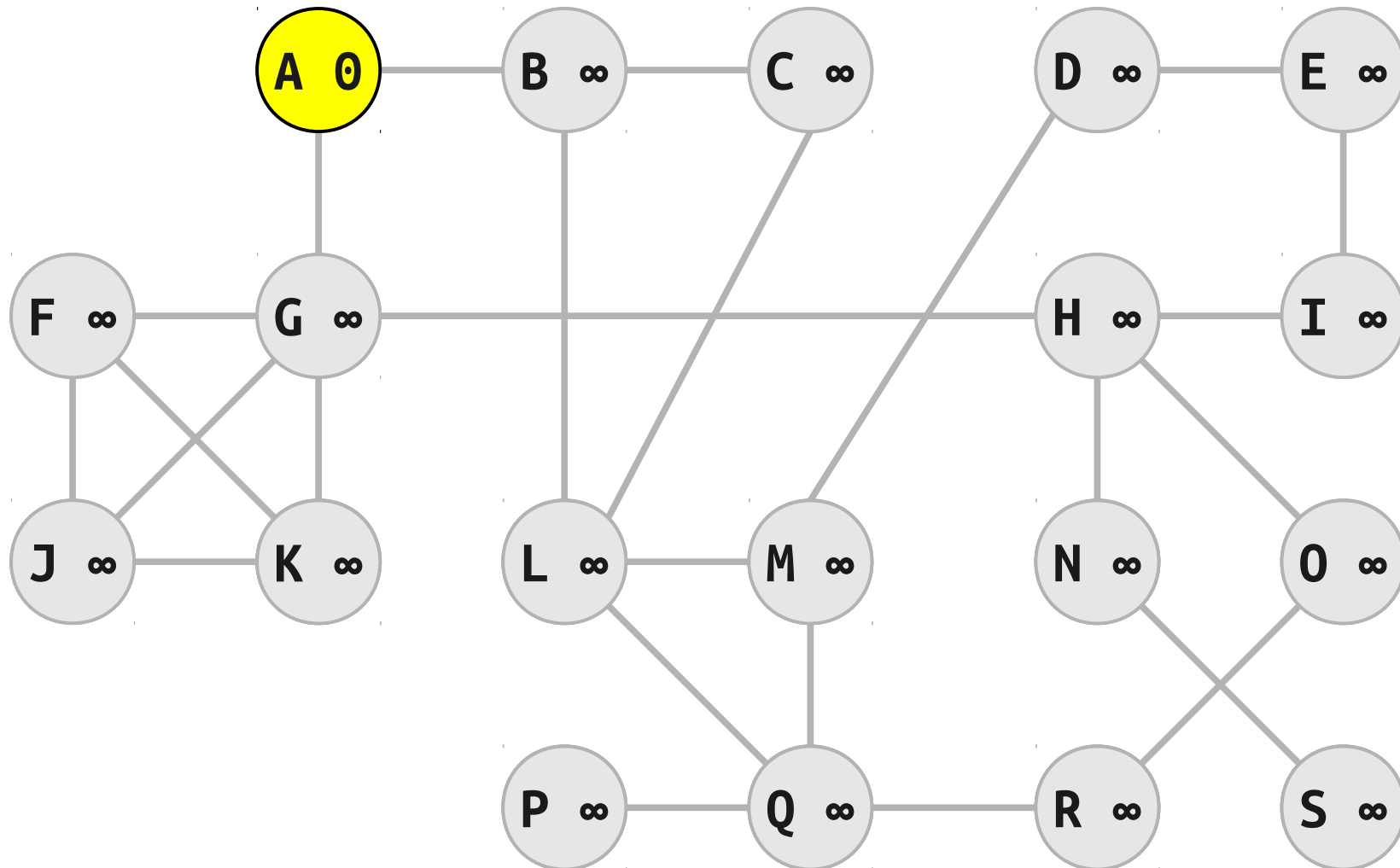
# Breadth-First Search



All nodes in the queue are at distance **0** from A.

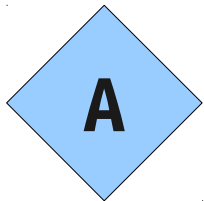
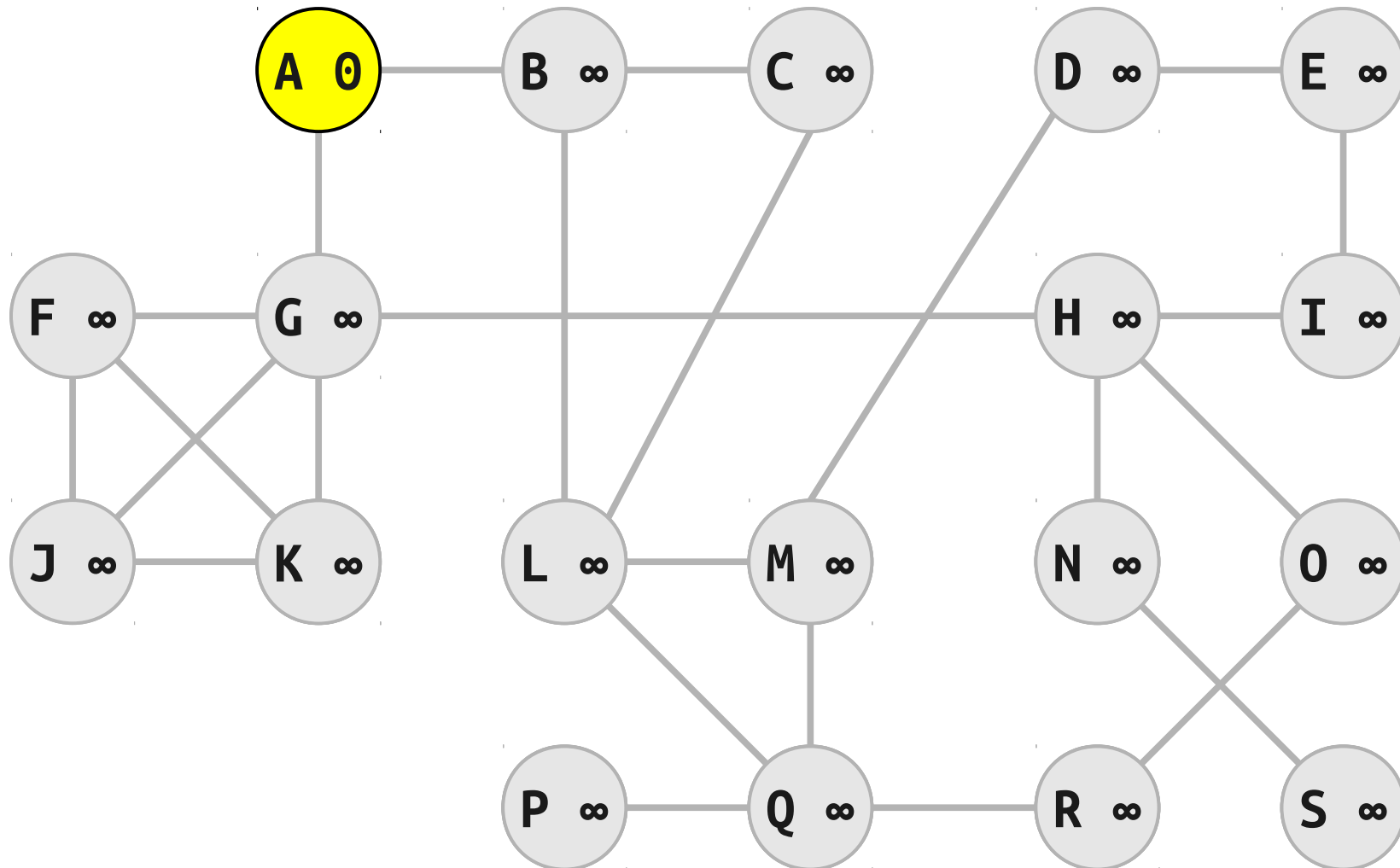


# Breadth-First Search



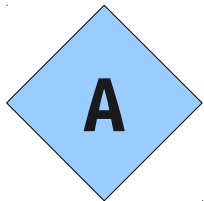
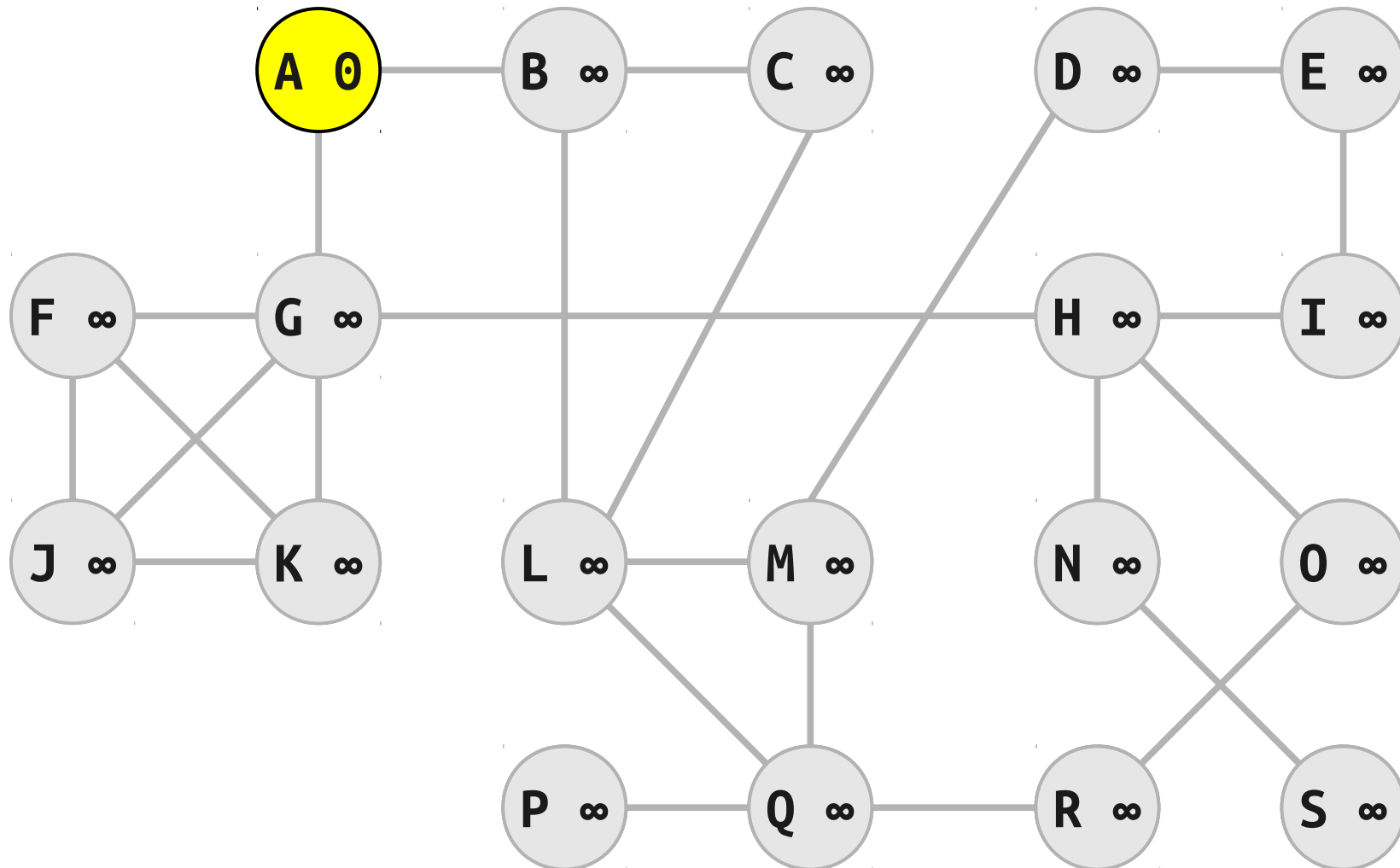
All nodes at distance **0** from A are in the queue.

# Breadth-First Search



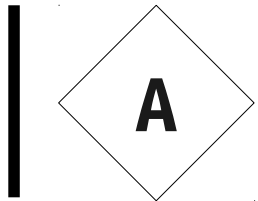
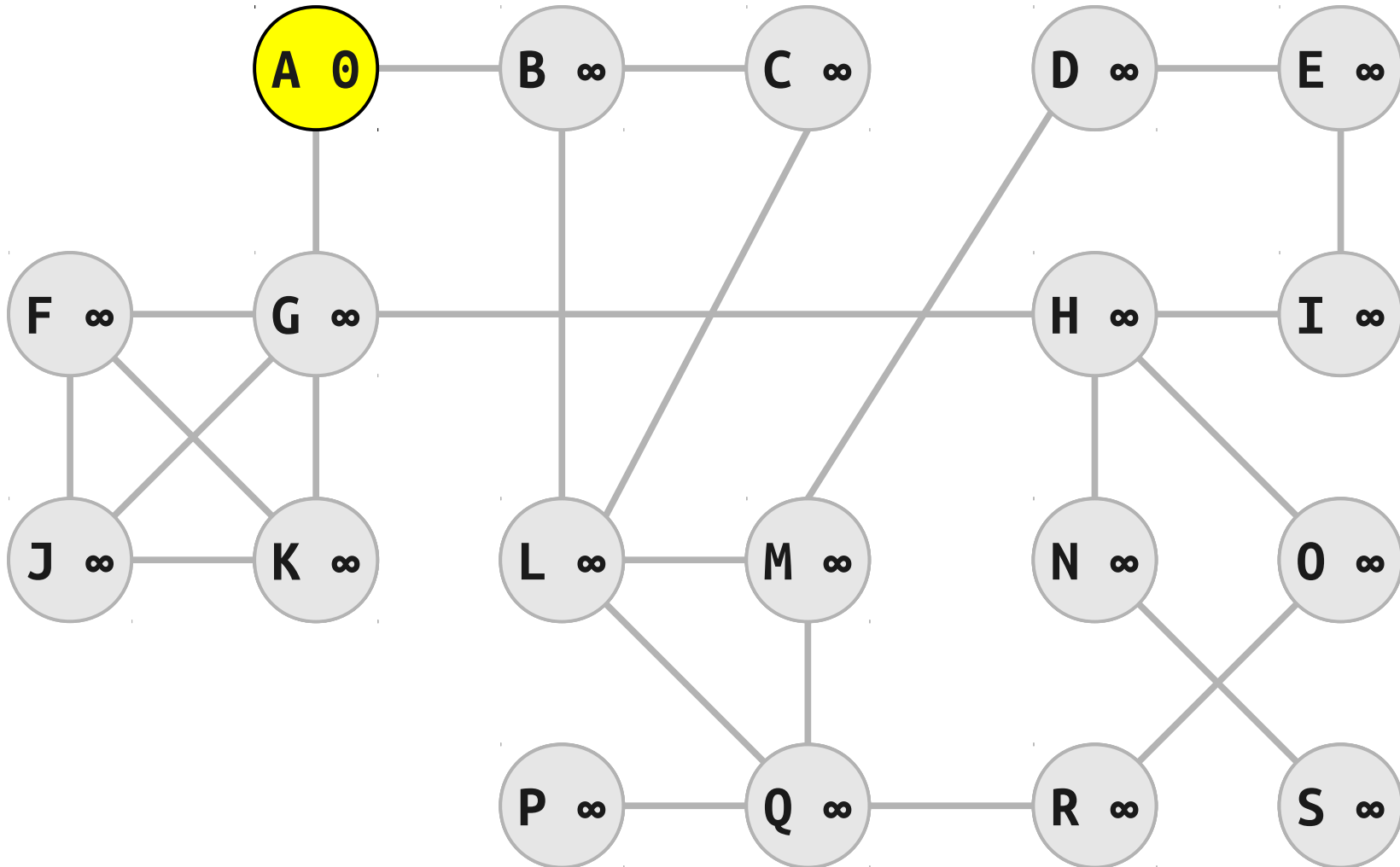
All nodes at distance  $\leq 0$  from A have the right distance set.

# Breadth-First Search

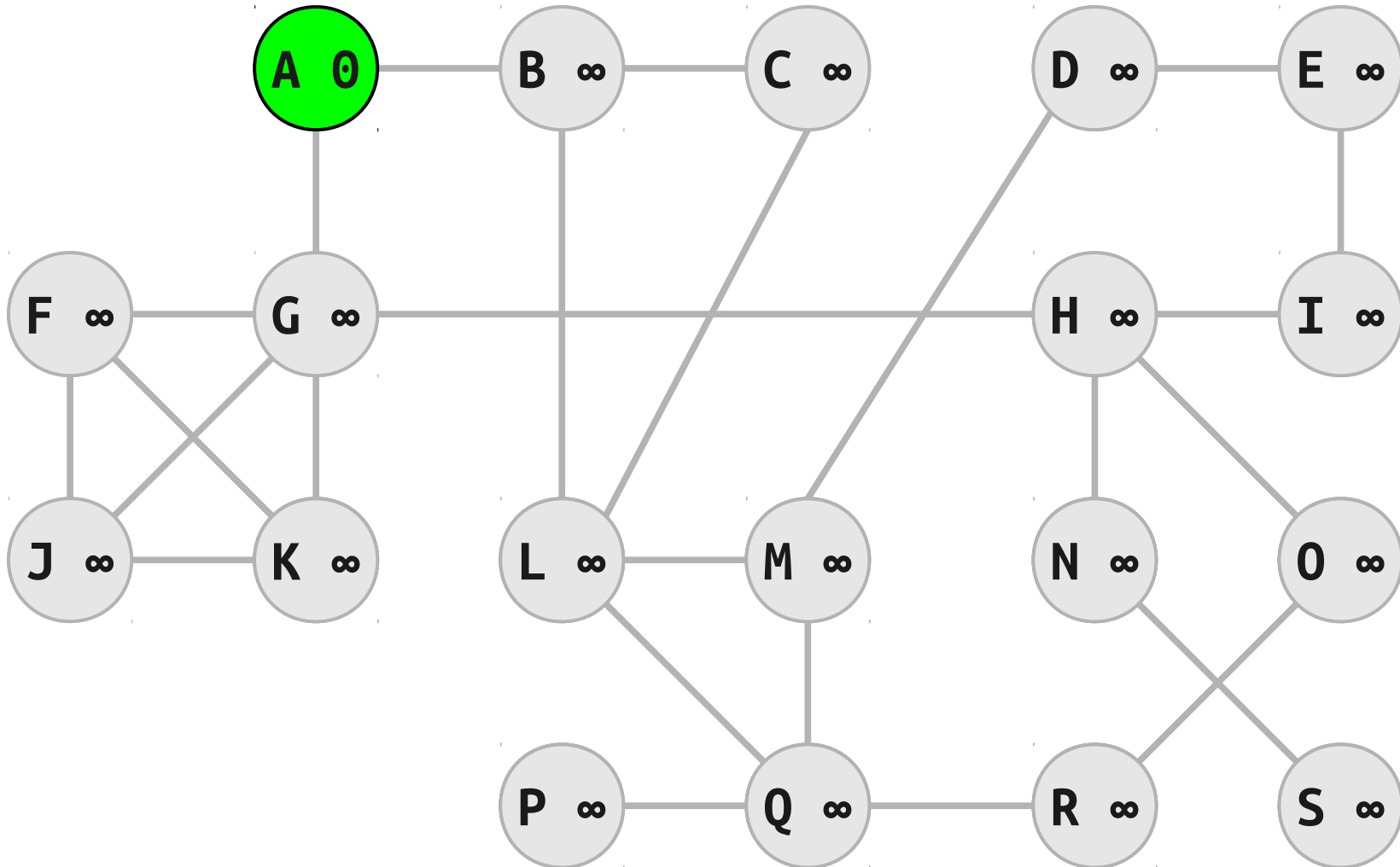


All nodes at distance  
> 0 from A have  
distance set to ∞

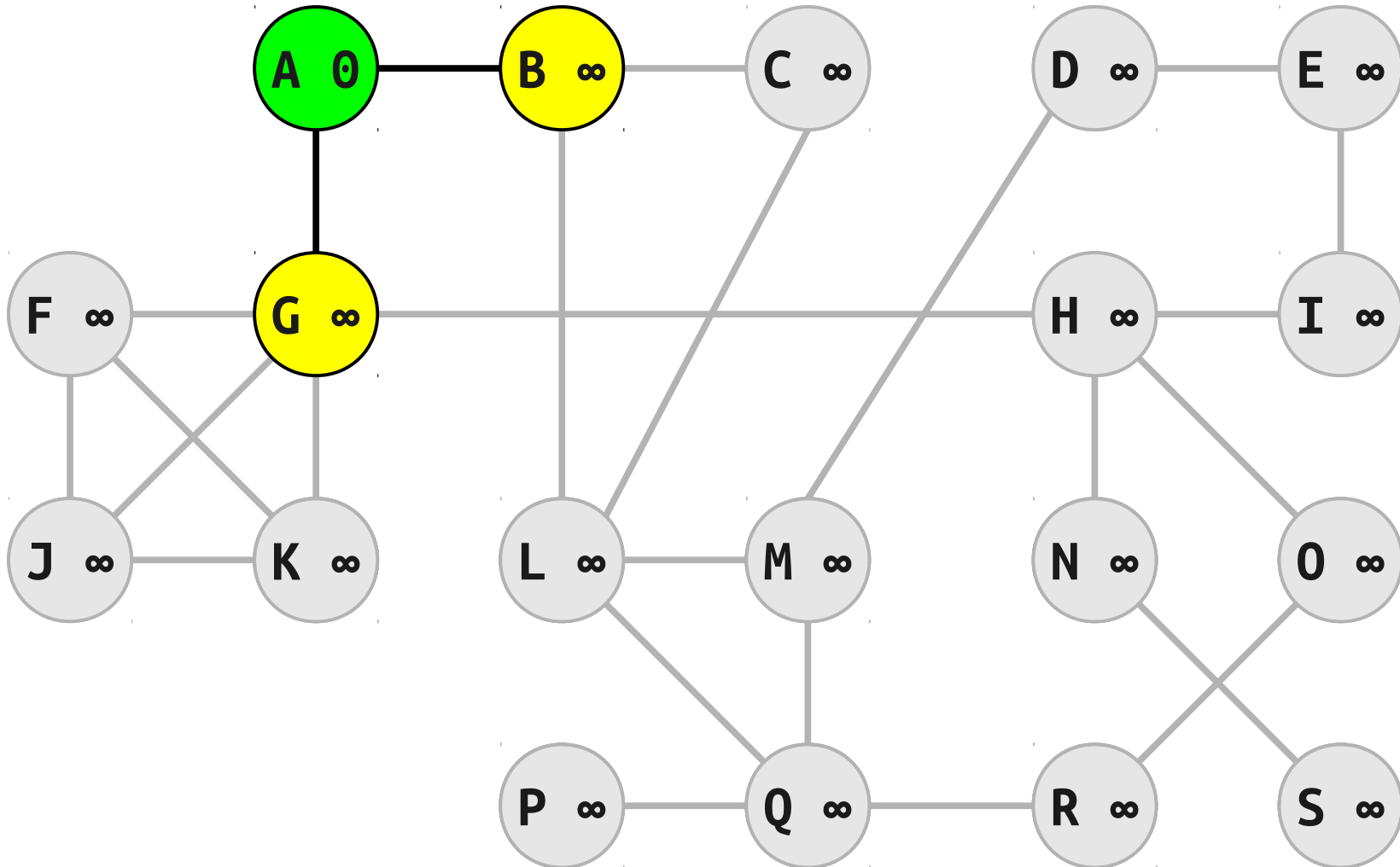
# Breadth-First Search



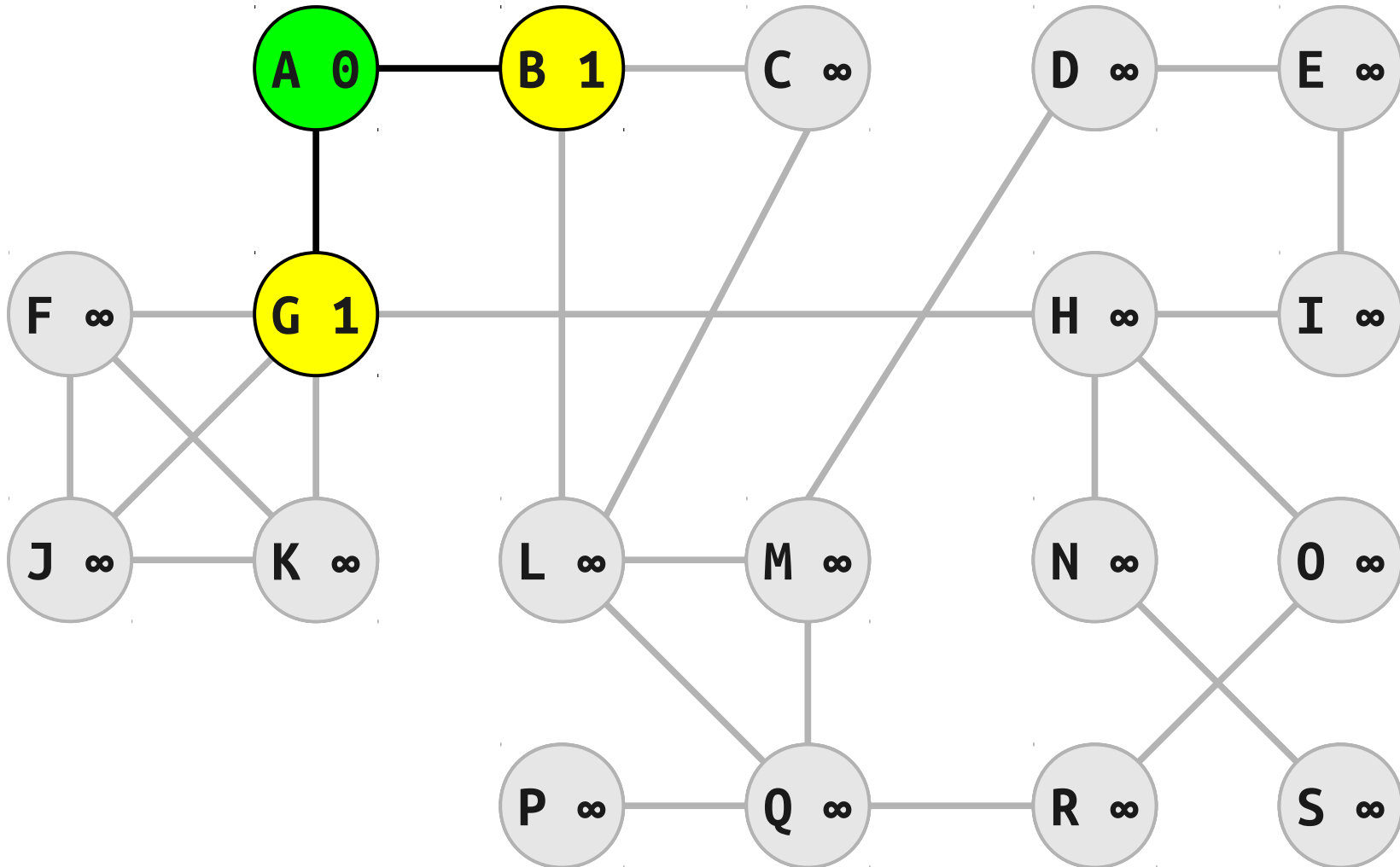
# Breadth-First Search



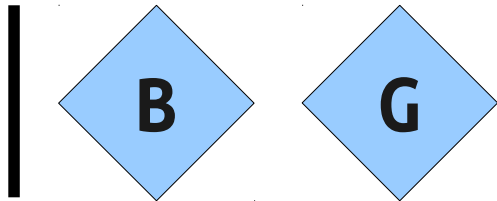
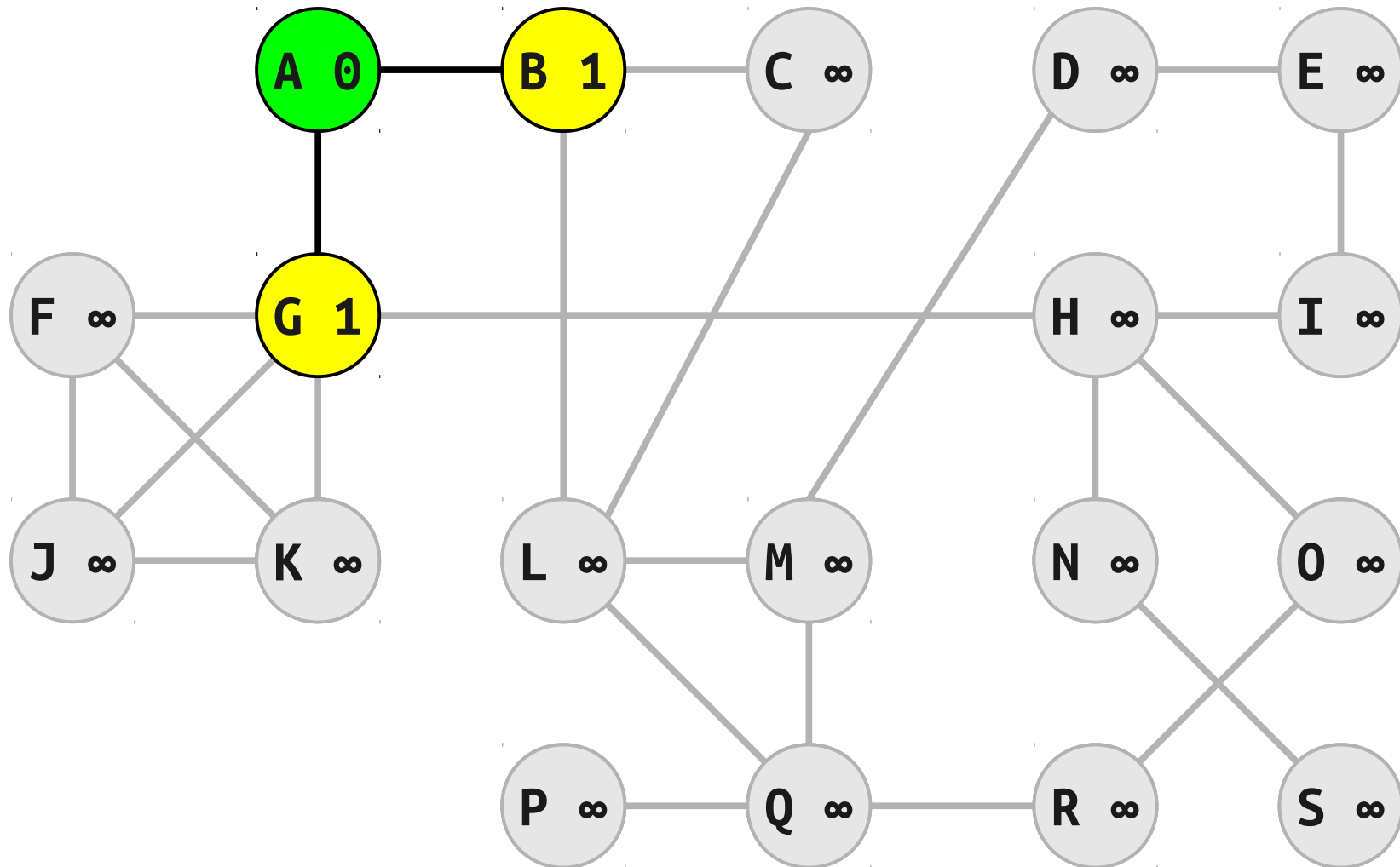
# Breadth-First Search



# Breadth-First Search

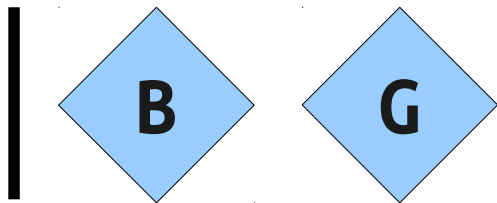
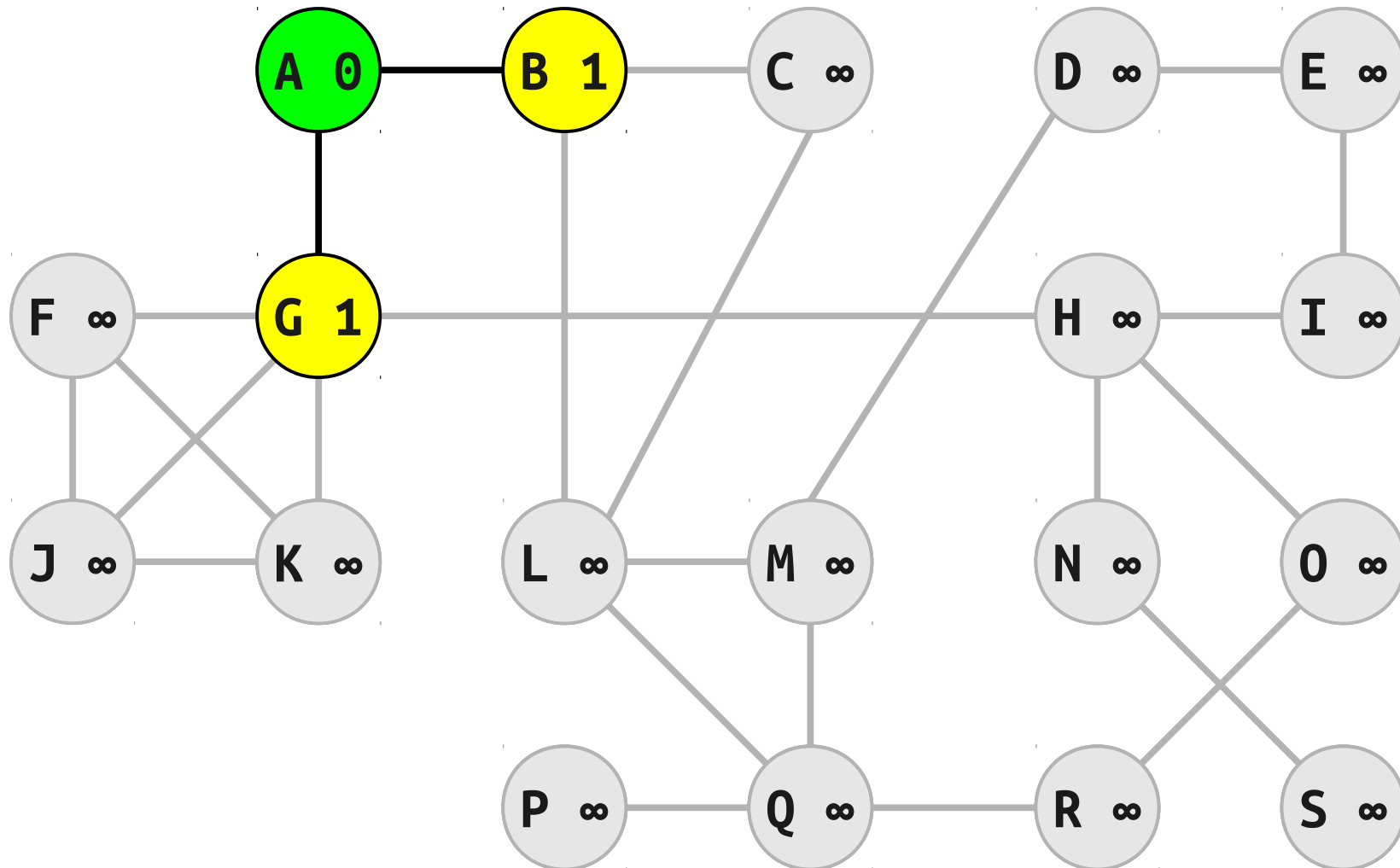


# Breadth-First Search



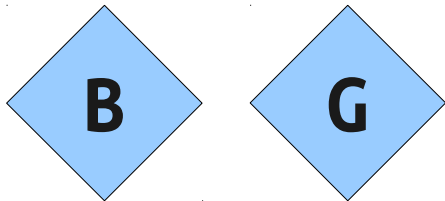
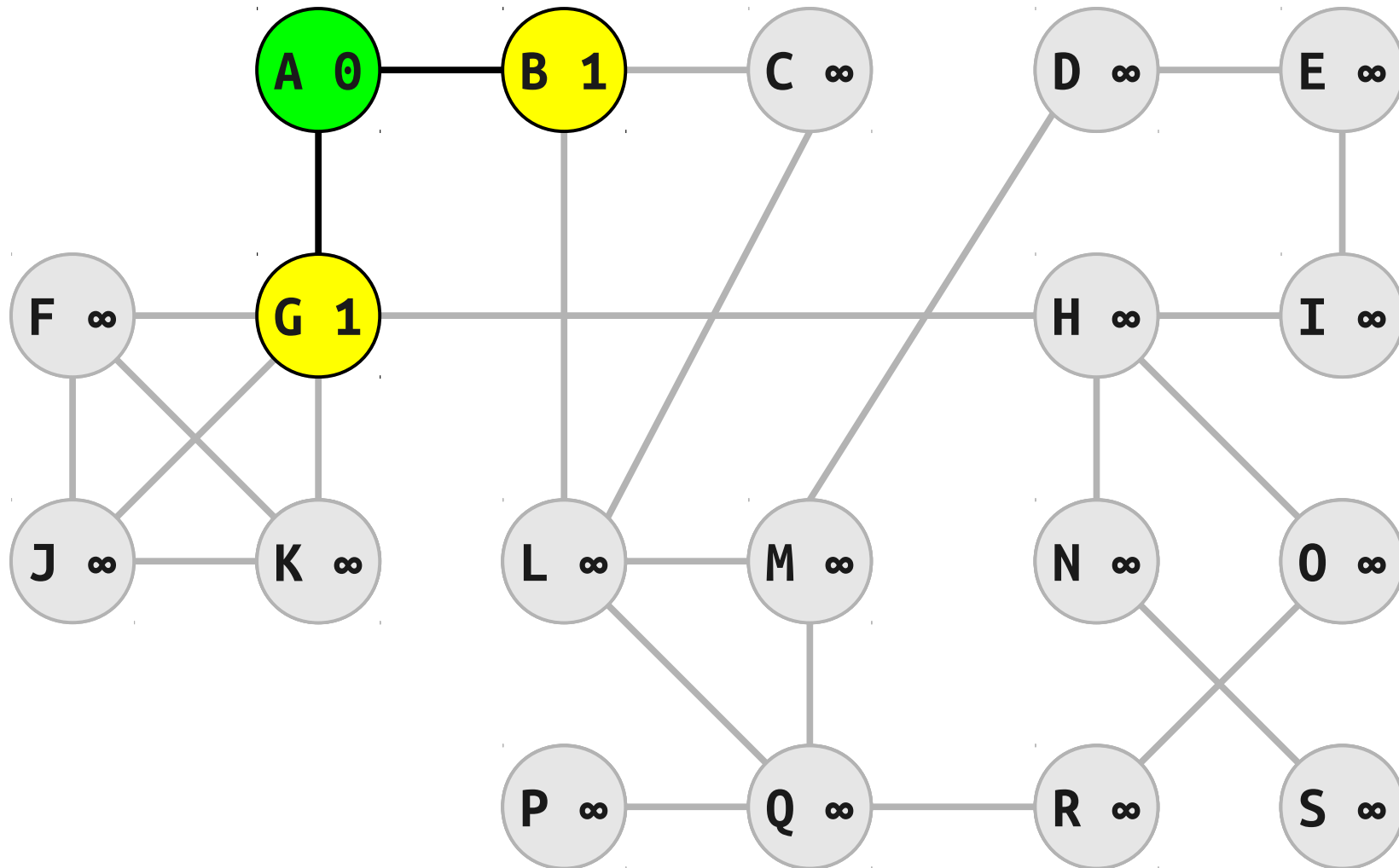


# Breadth-First Search



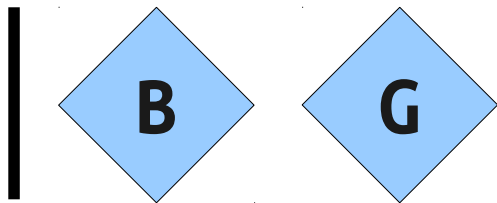
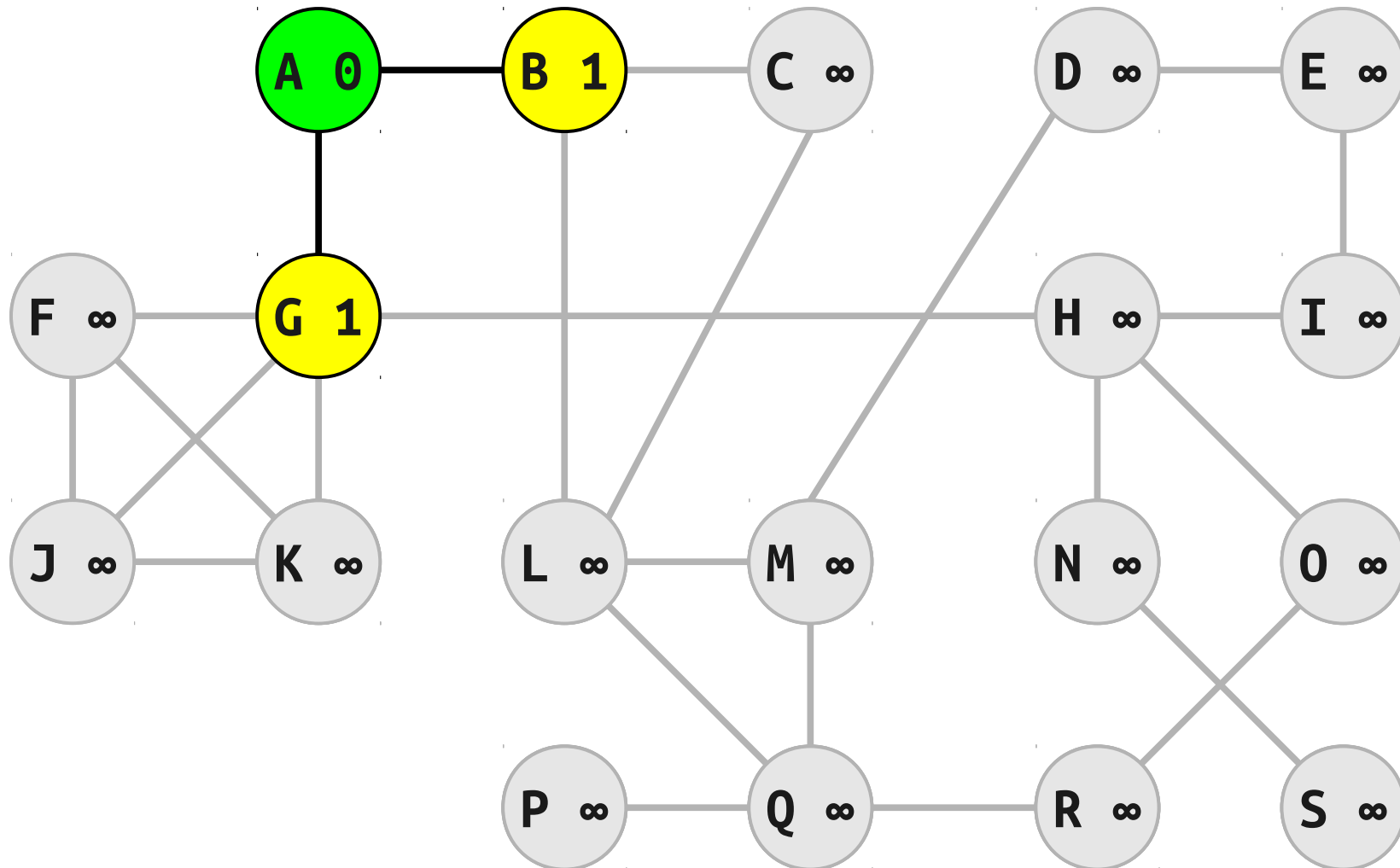
All nodes in the queue are at distance **1** from A.

# Breadth-First Search



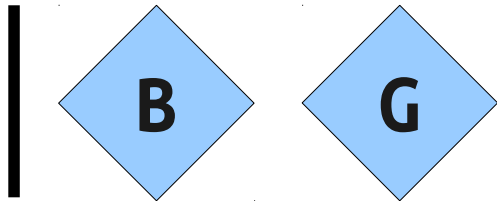
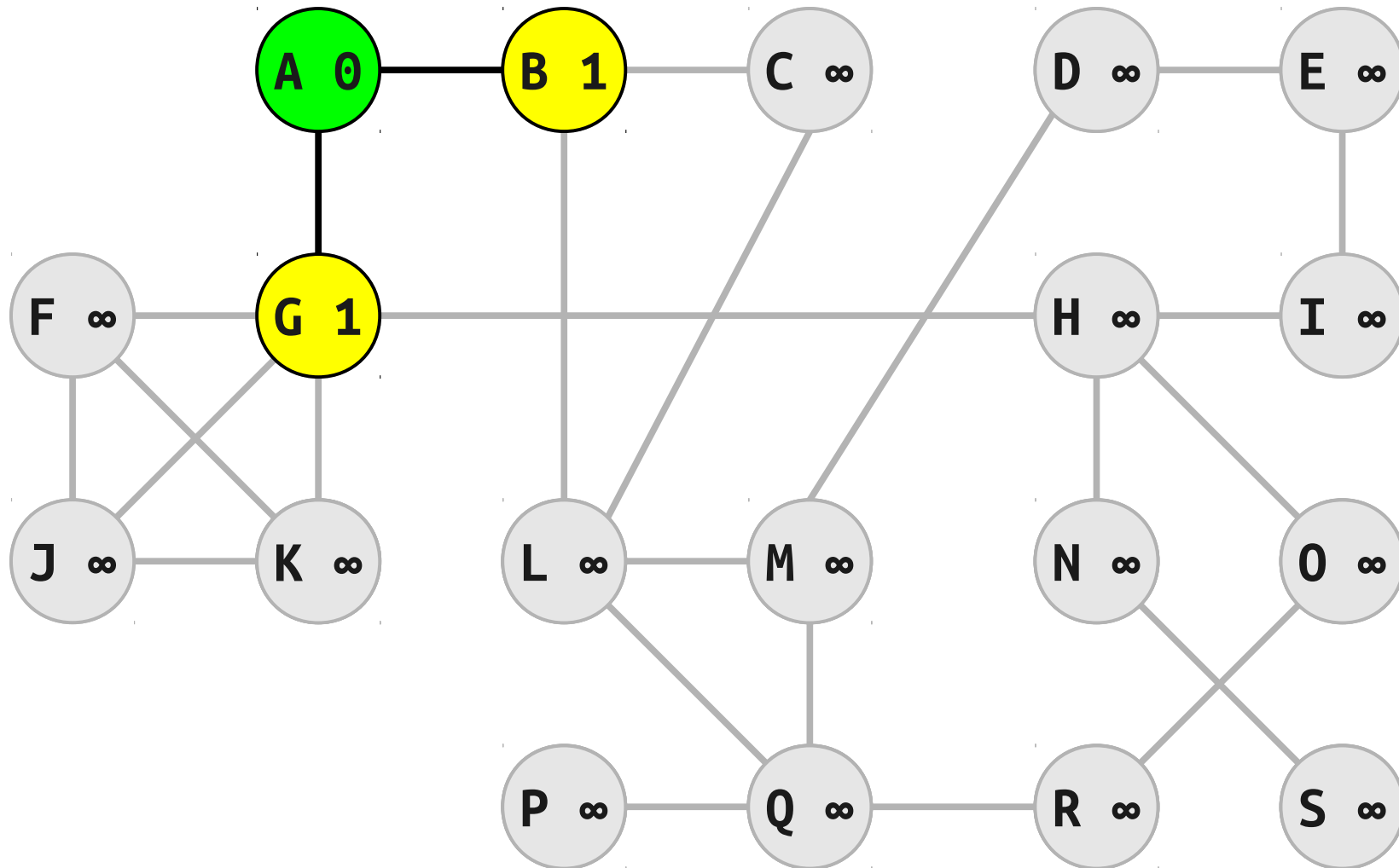
All nodes at distance **1** from A are in the queue.

# Breadth-First Search



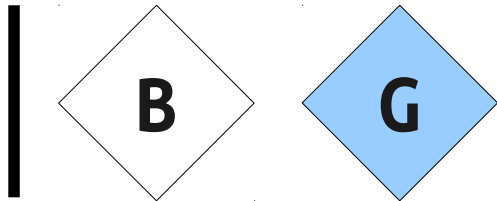
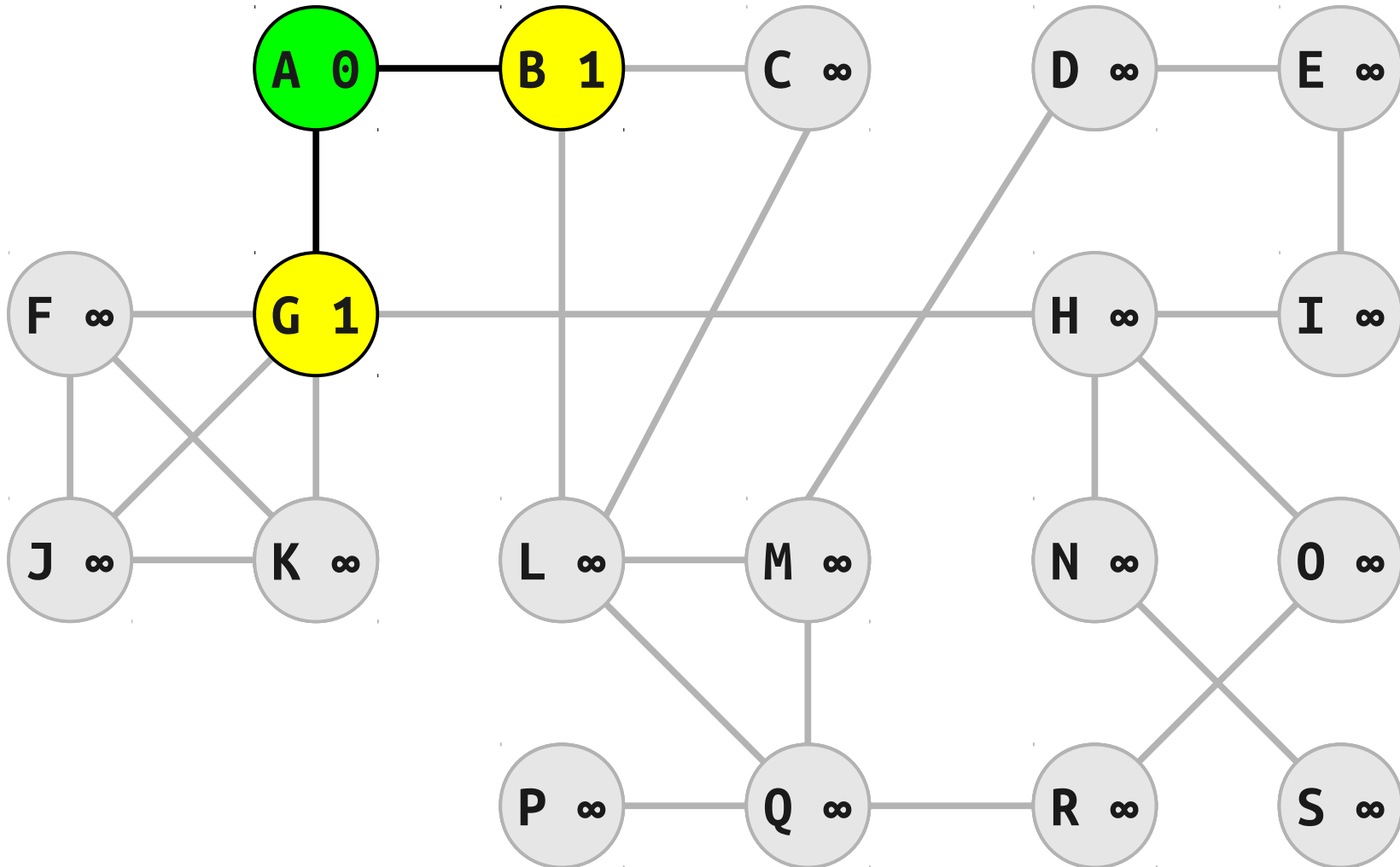
All nodes at distance  $\leq 1$  from A have the right distance set.

# Breadth-First Search

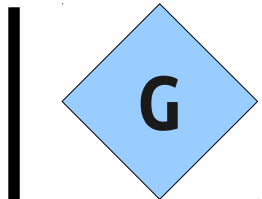
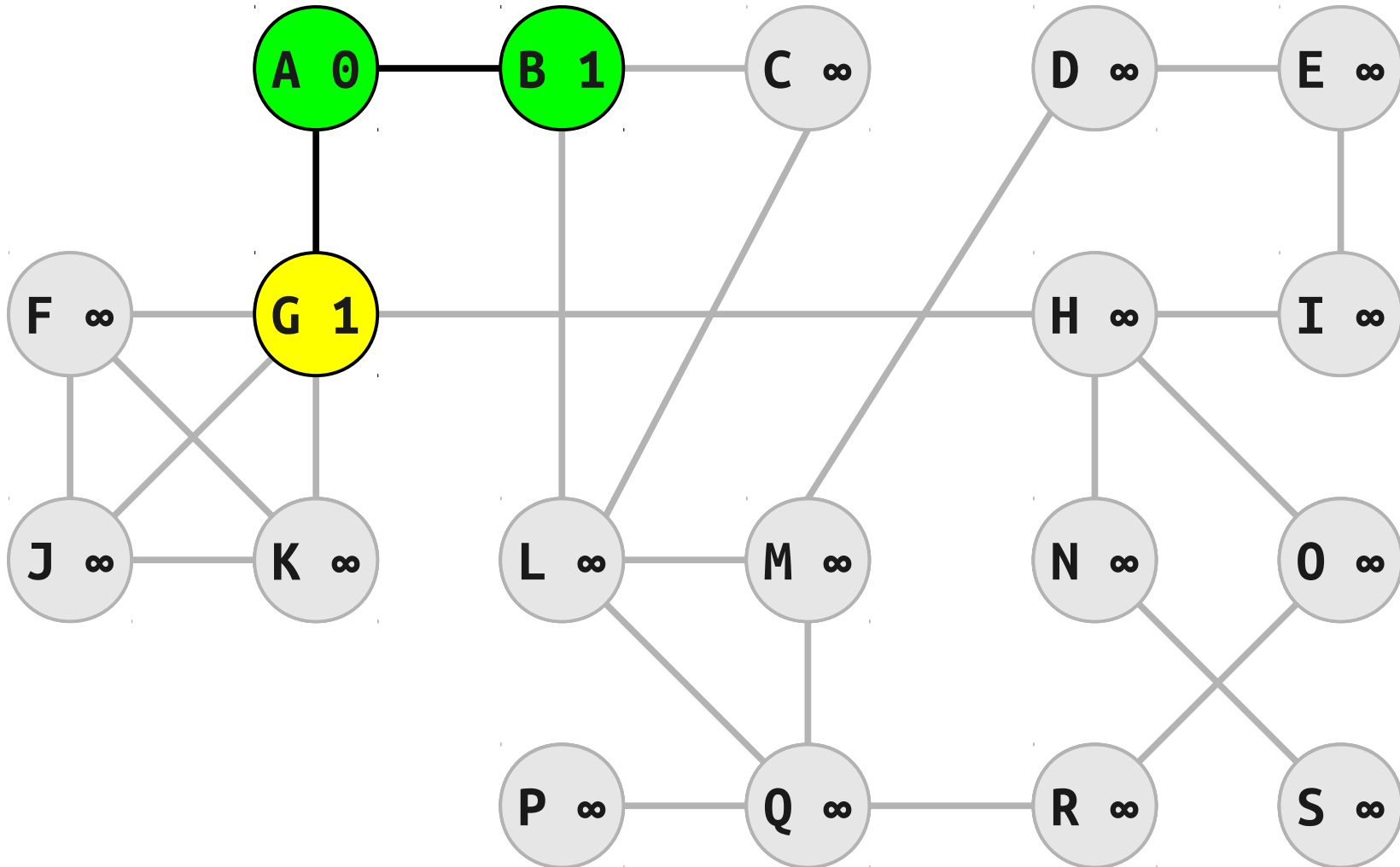


All nodes at distance  
> 1 from A have  
distance set to ∞

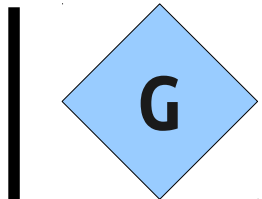
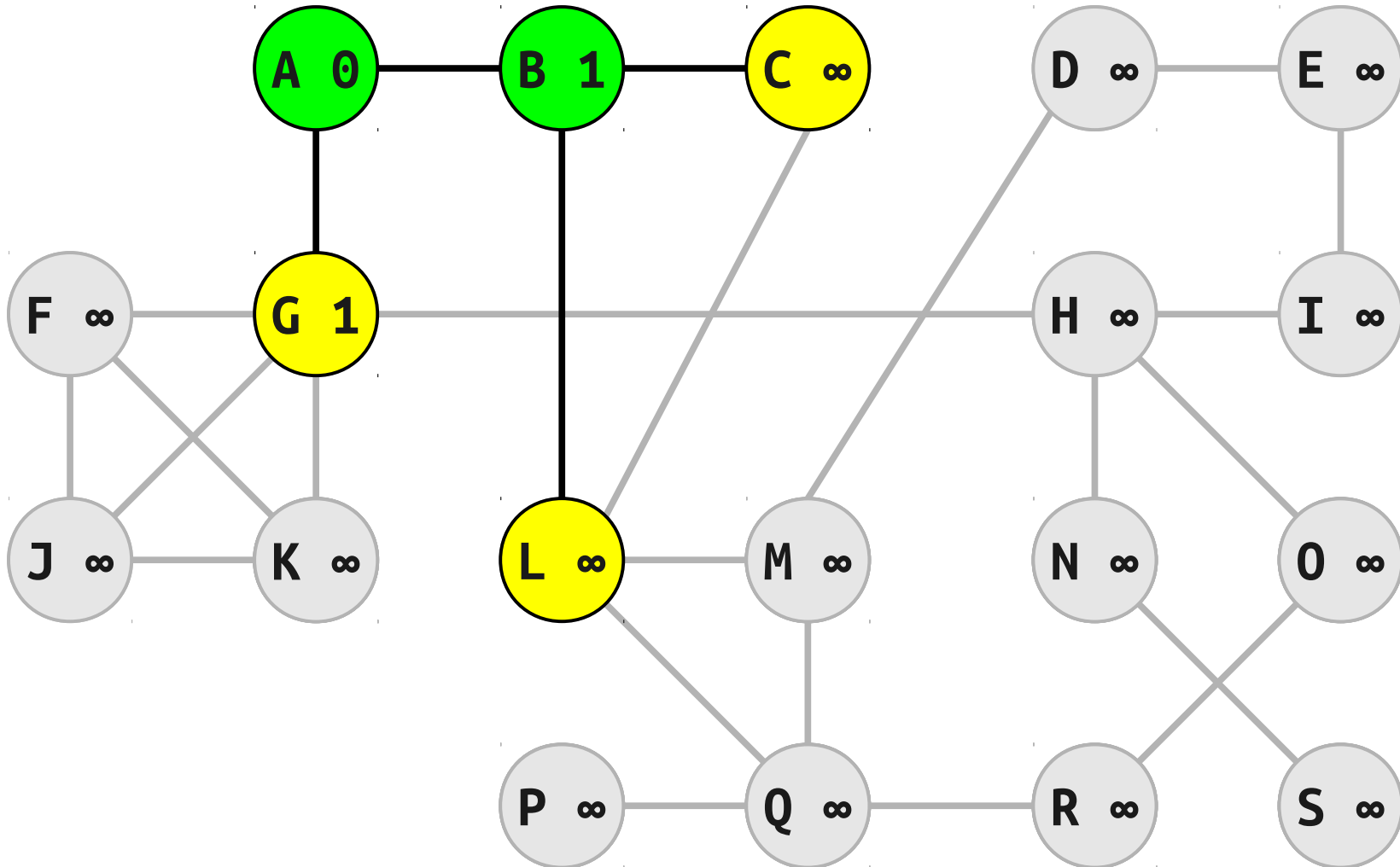
# Breadth-First Search



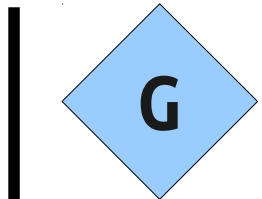
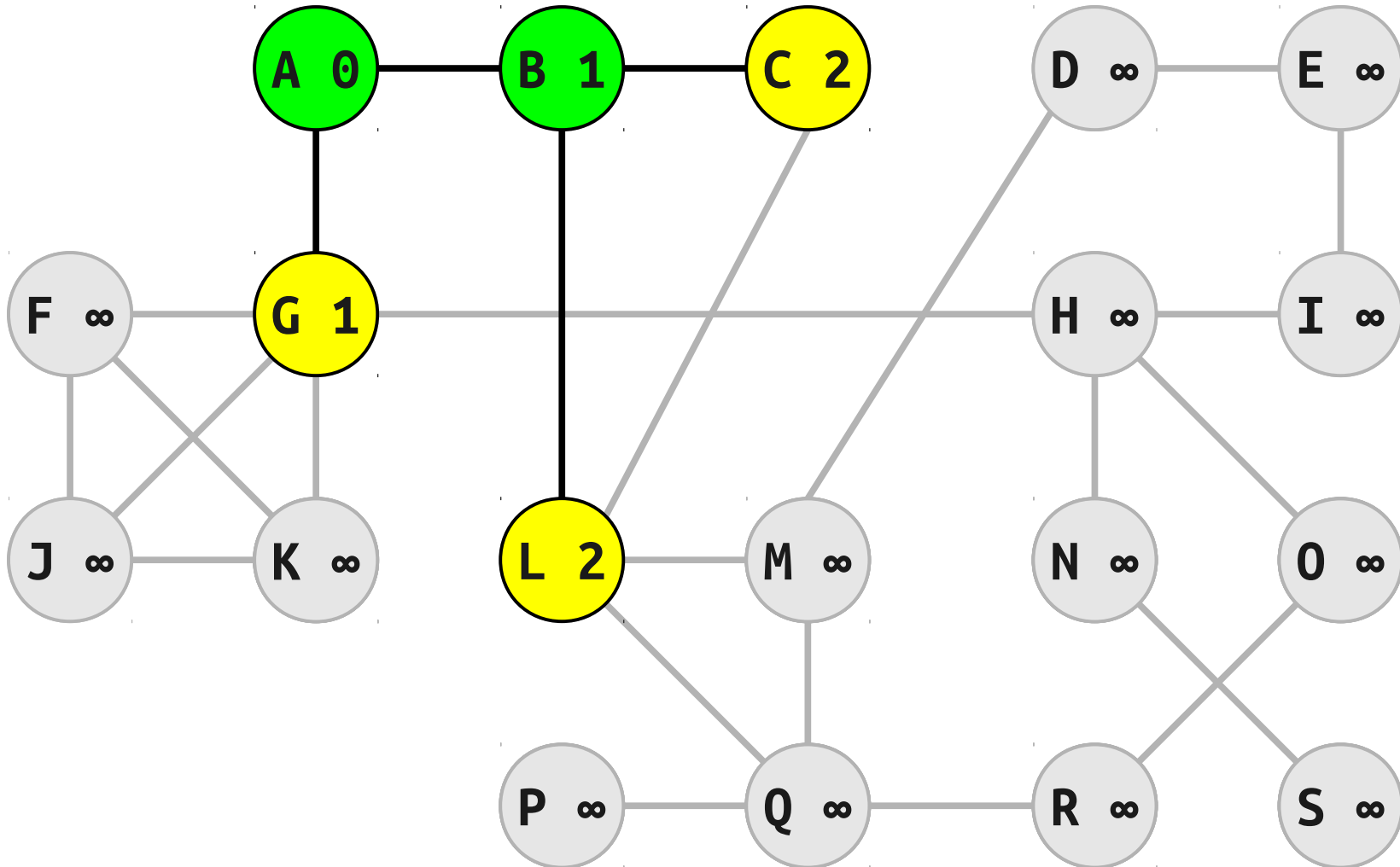
# Breadth-First Search



# Breadth-First Search

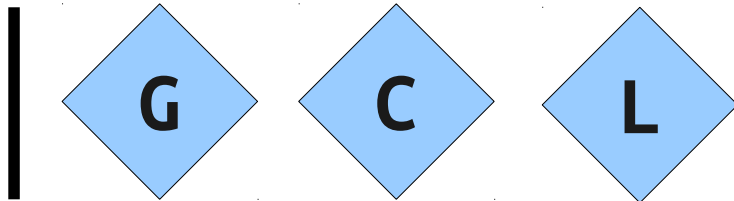
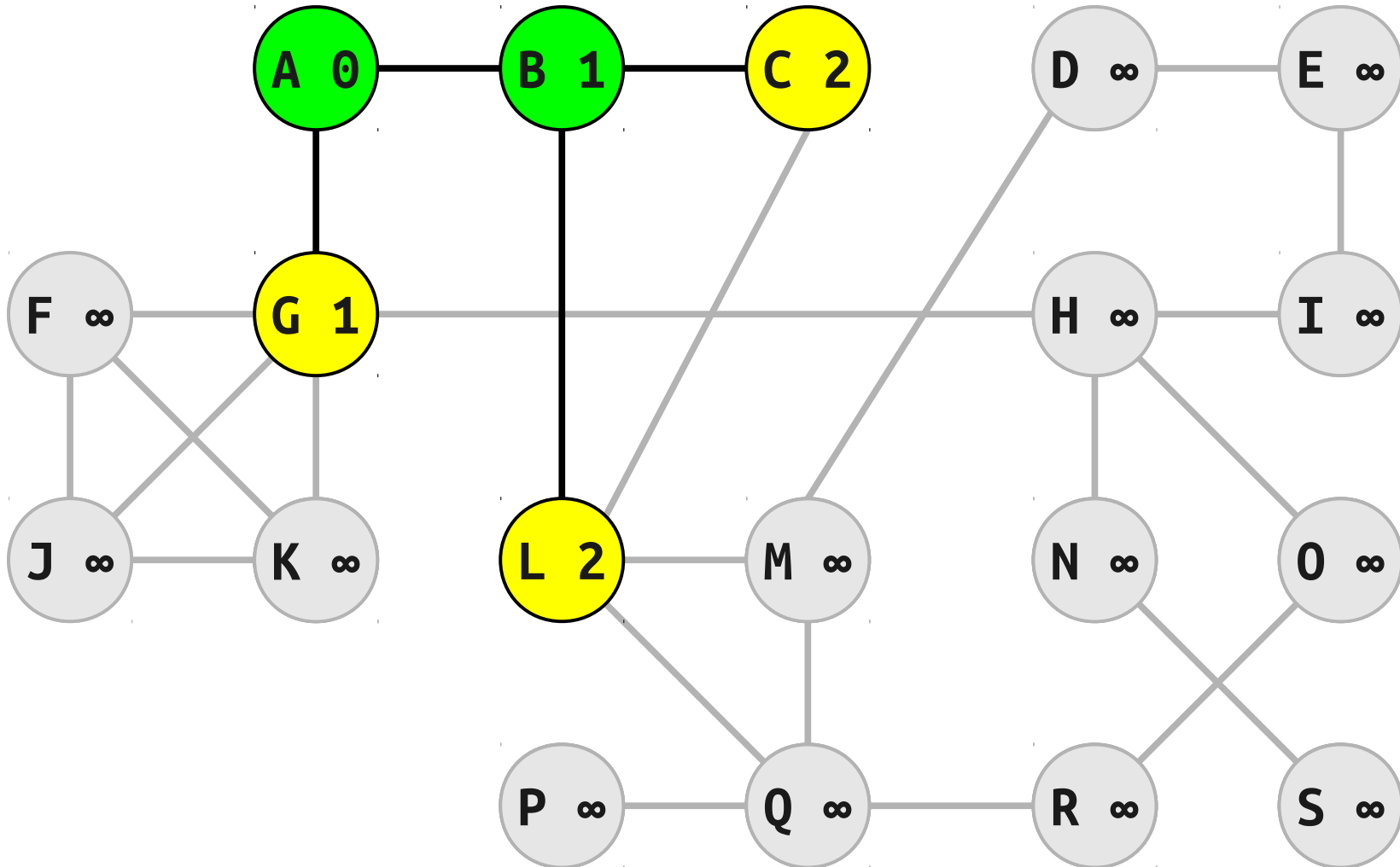


# Breadth-First Search

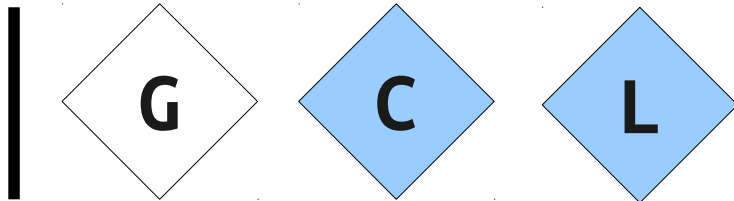
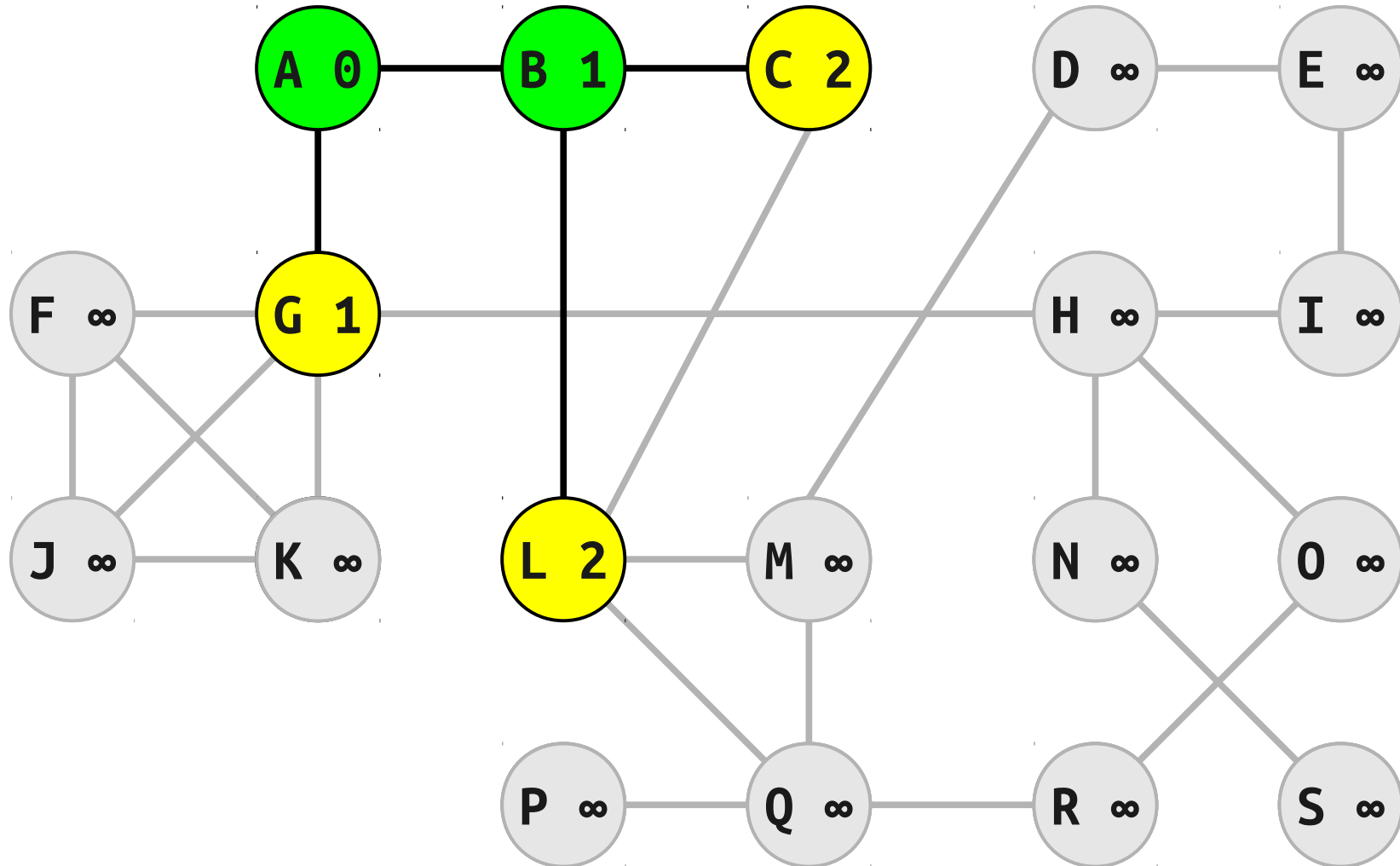




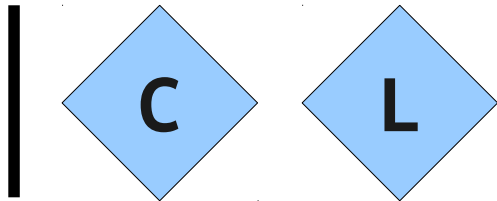
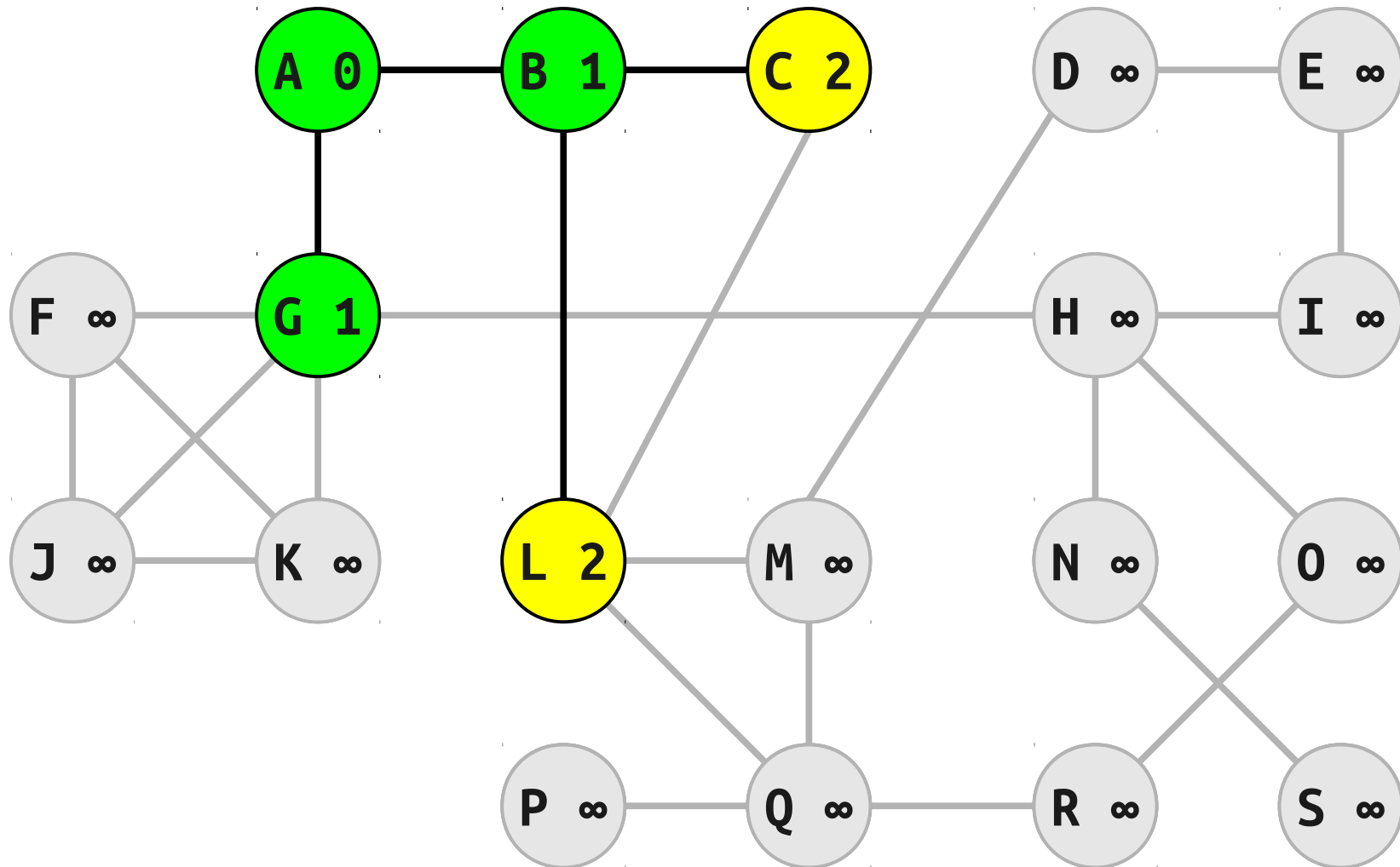
# Breadth-First Search



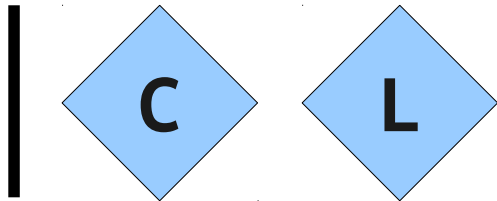
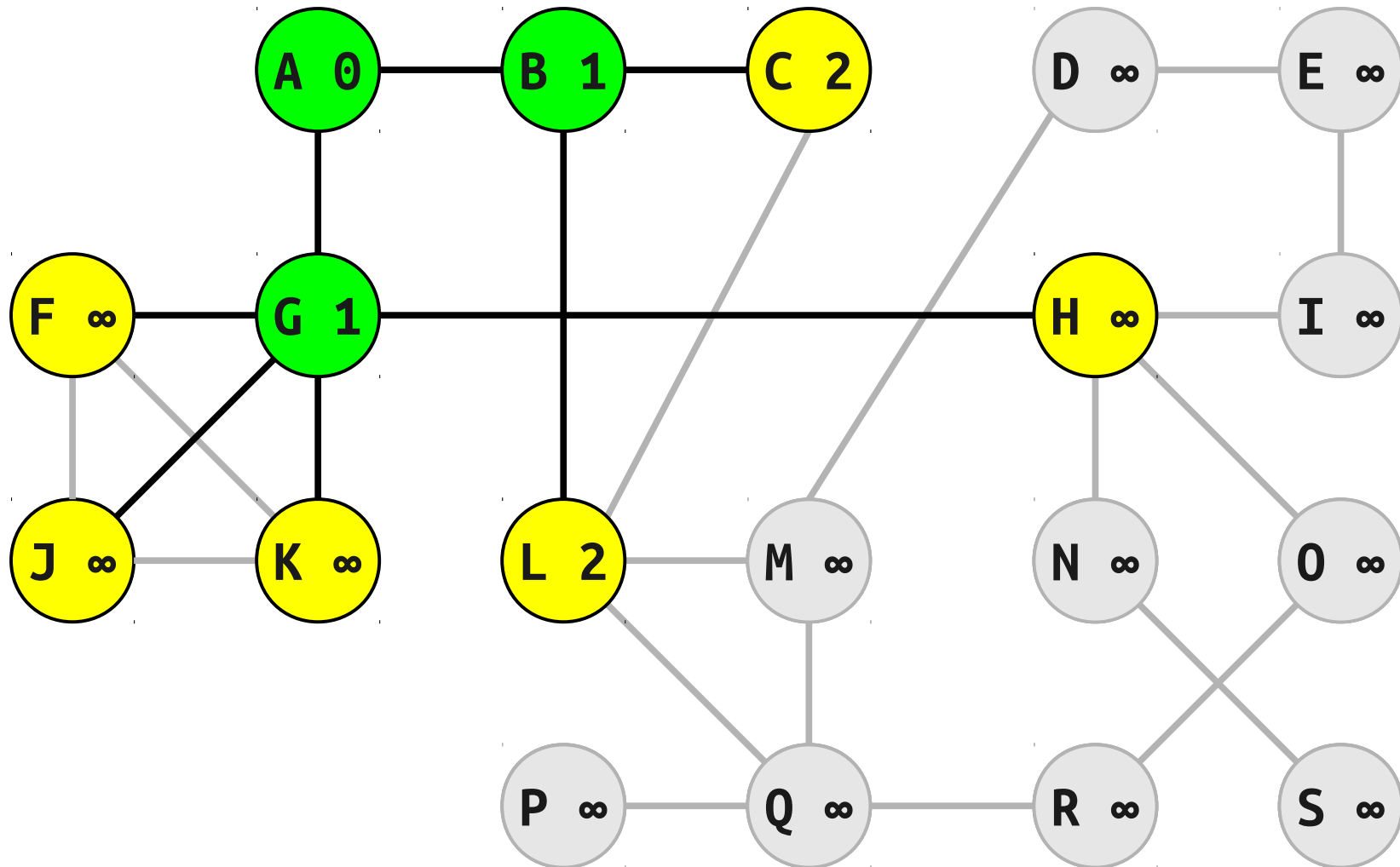
# Breadth-First Search



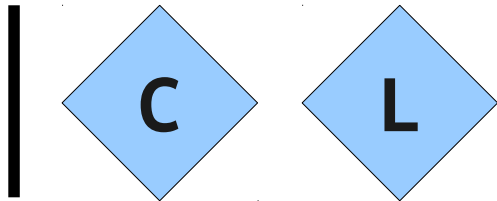
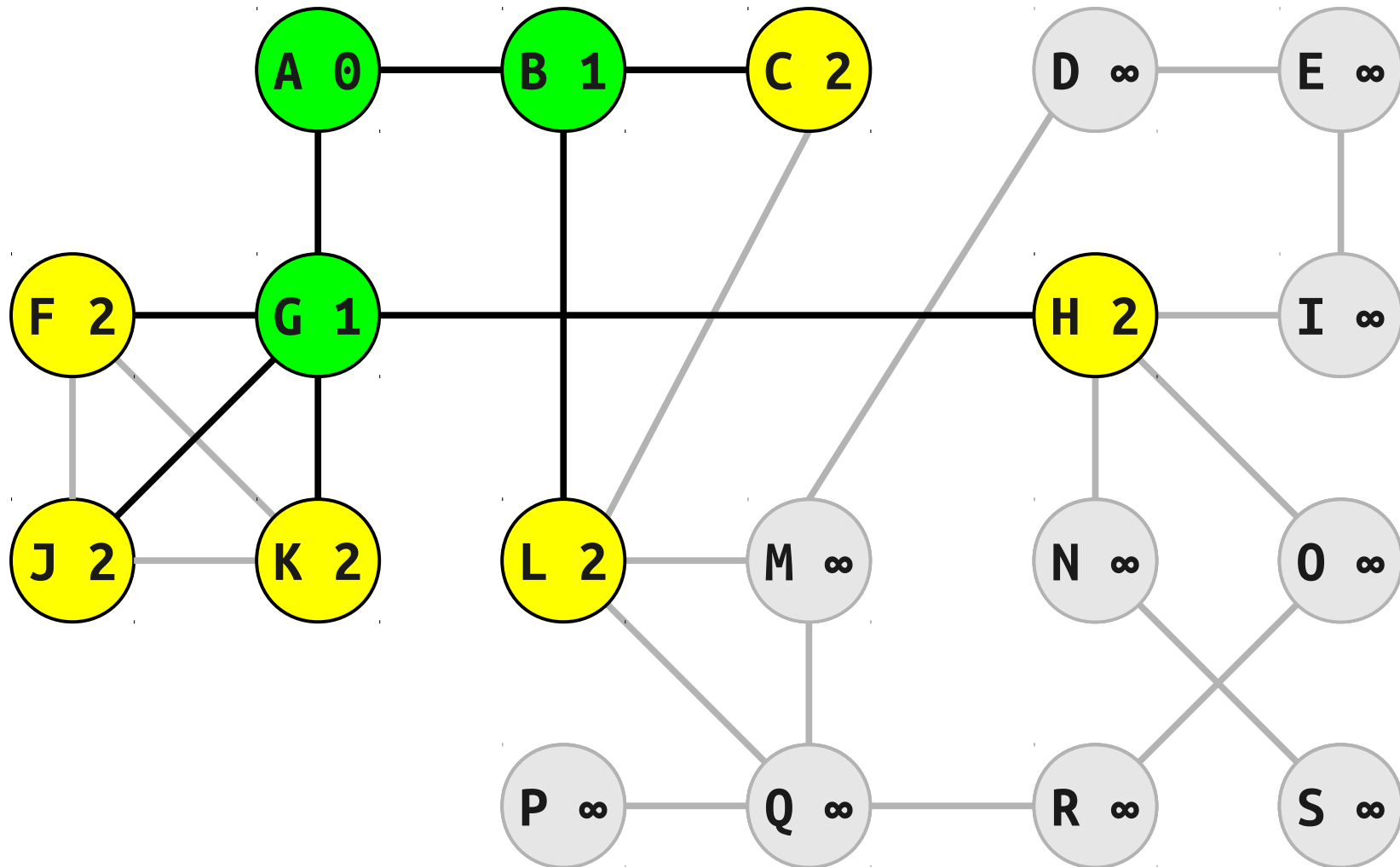
# Breadth-First Search



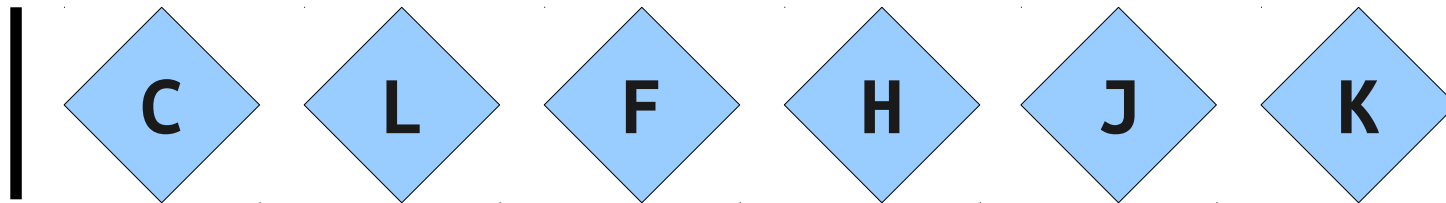
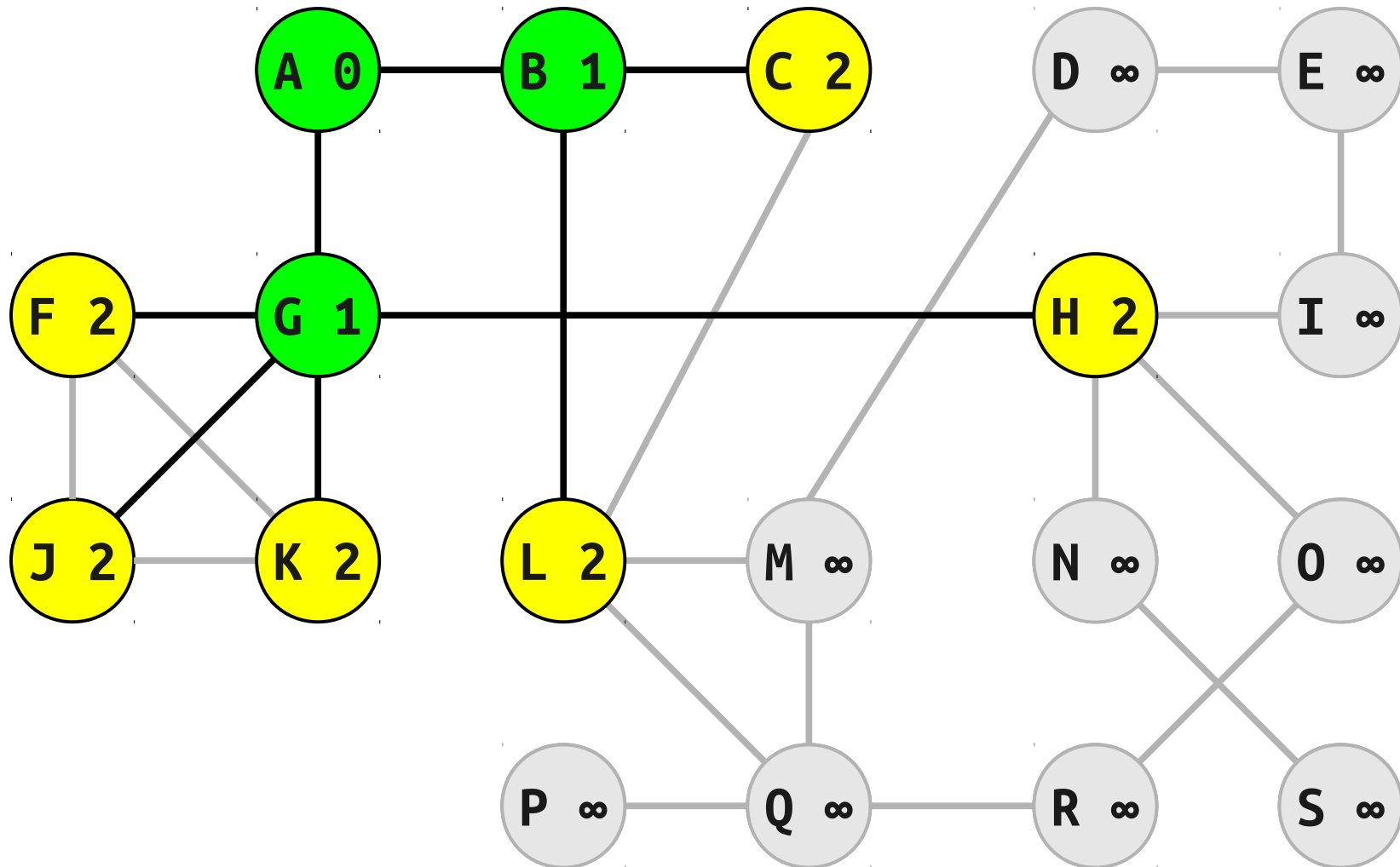
# Breadth-First Search



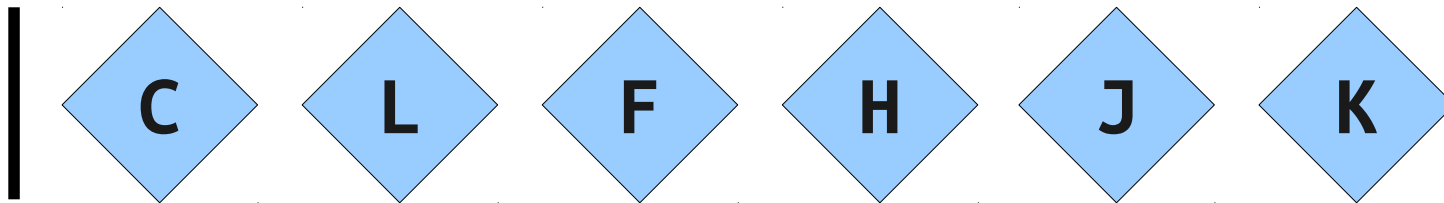
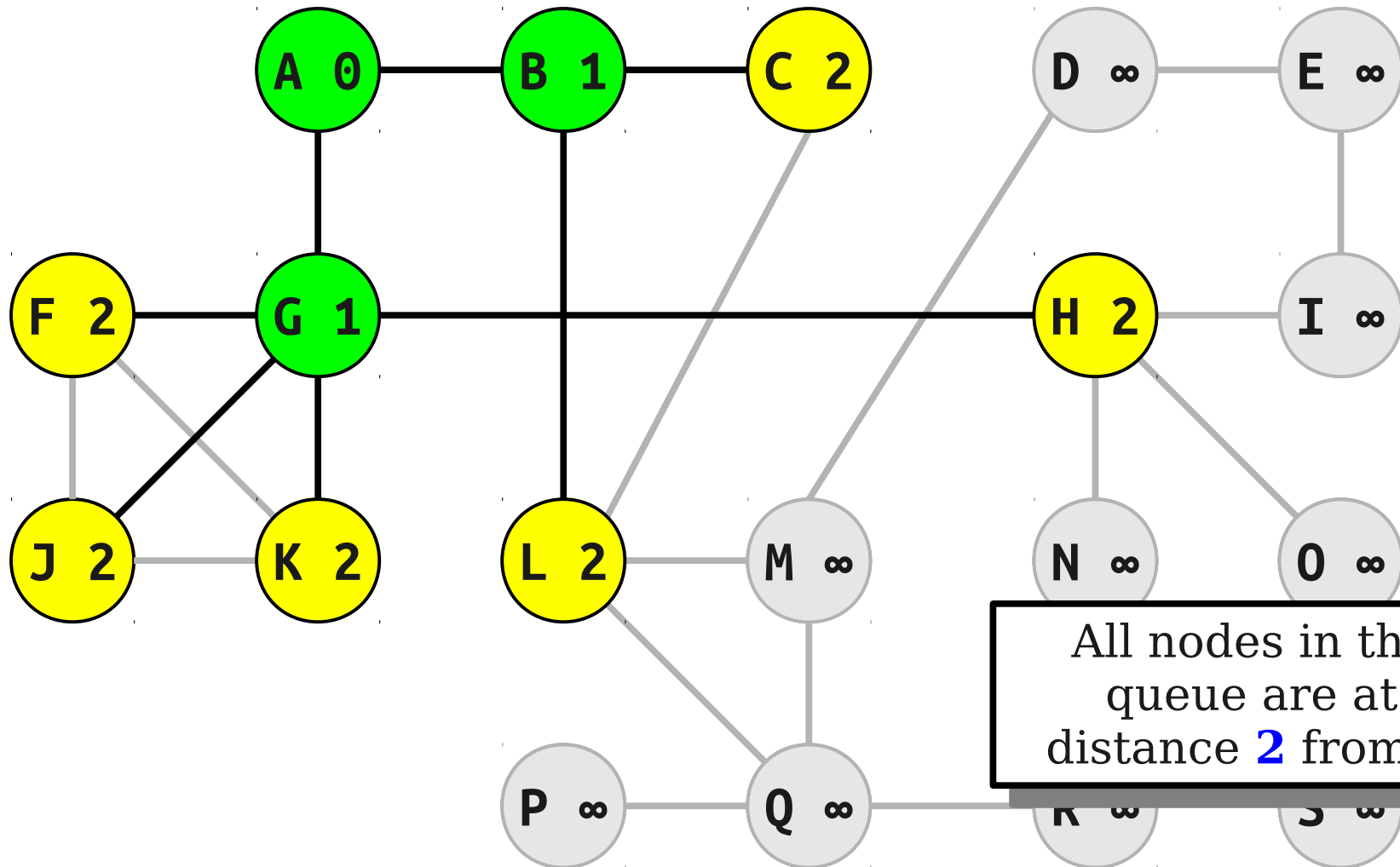
# Breadth-First Search



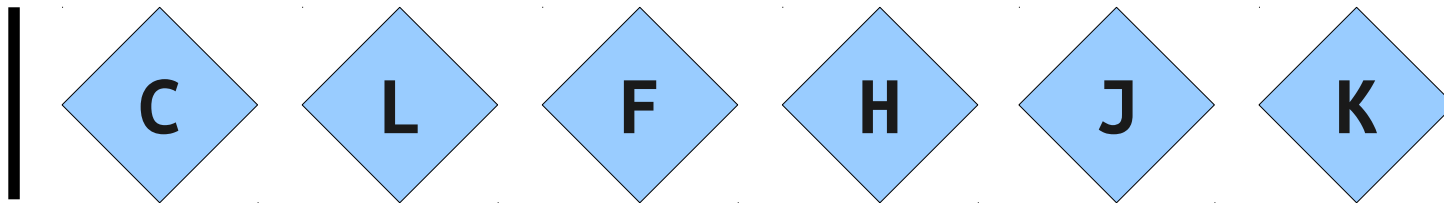
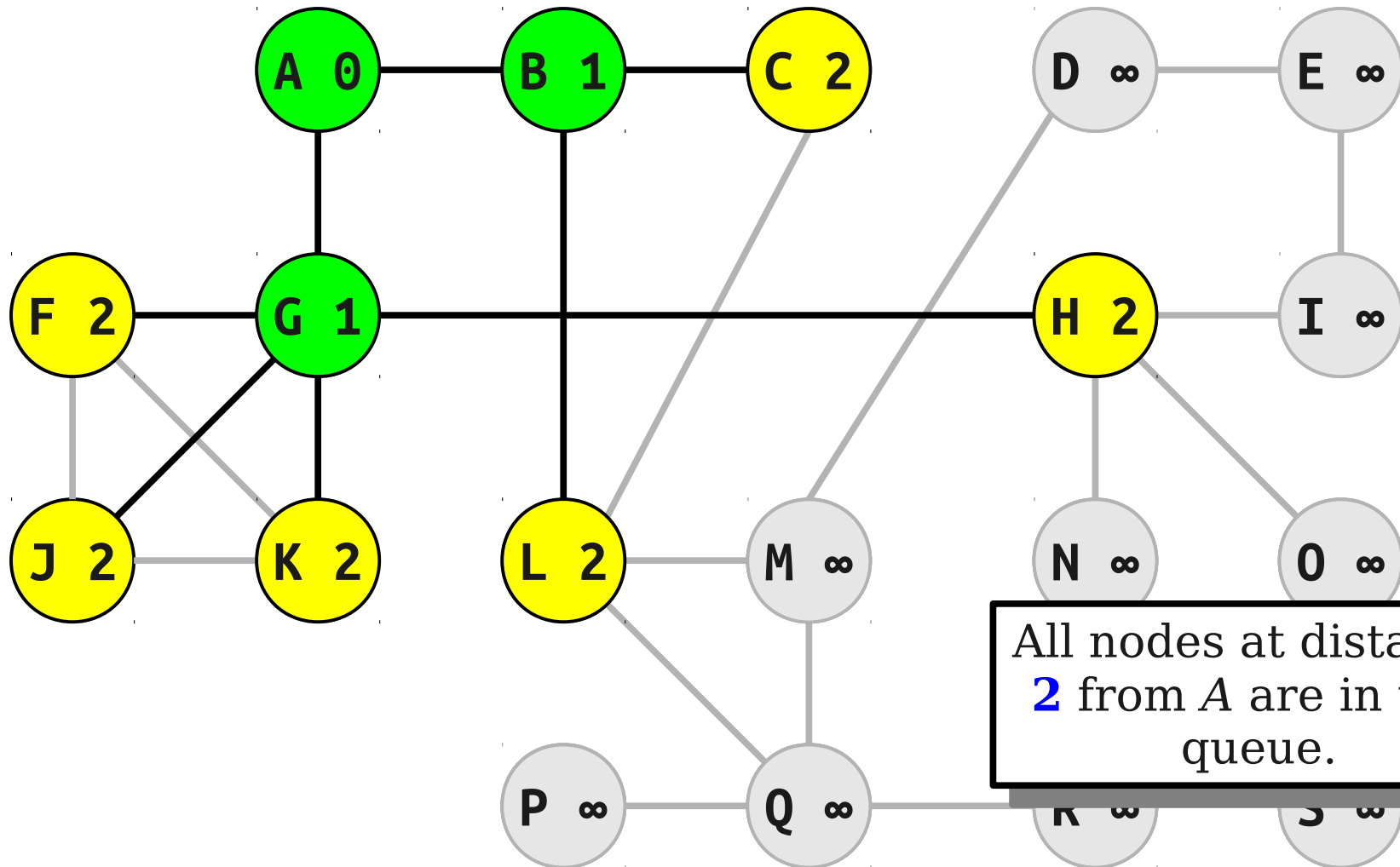
# Breadth-First Search



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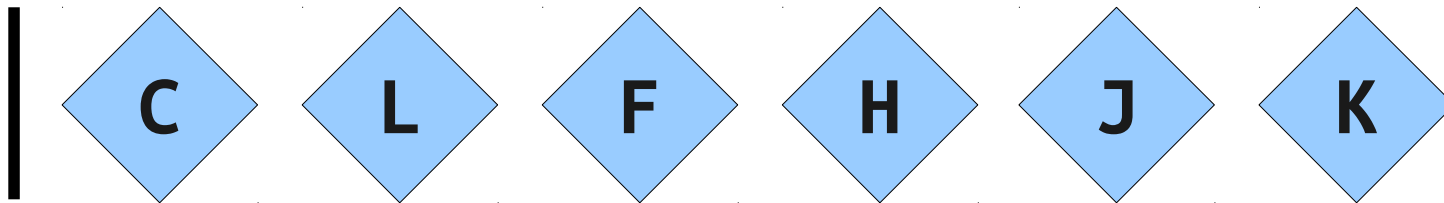
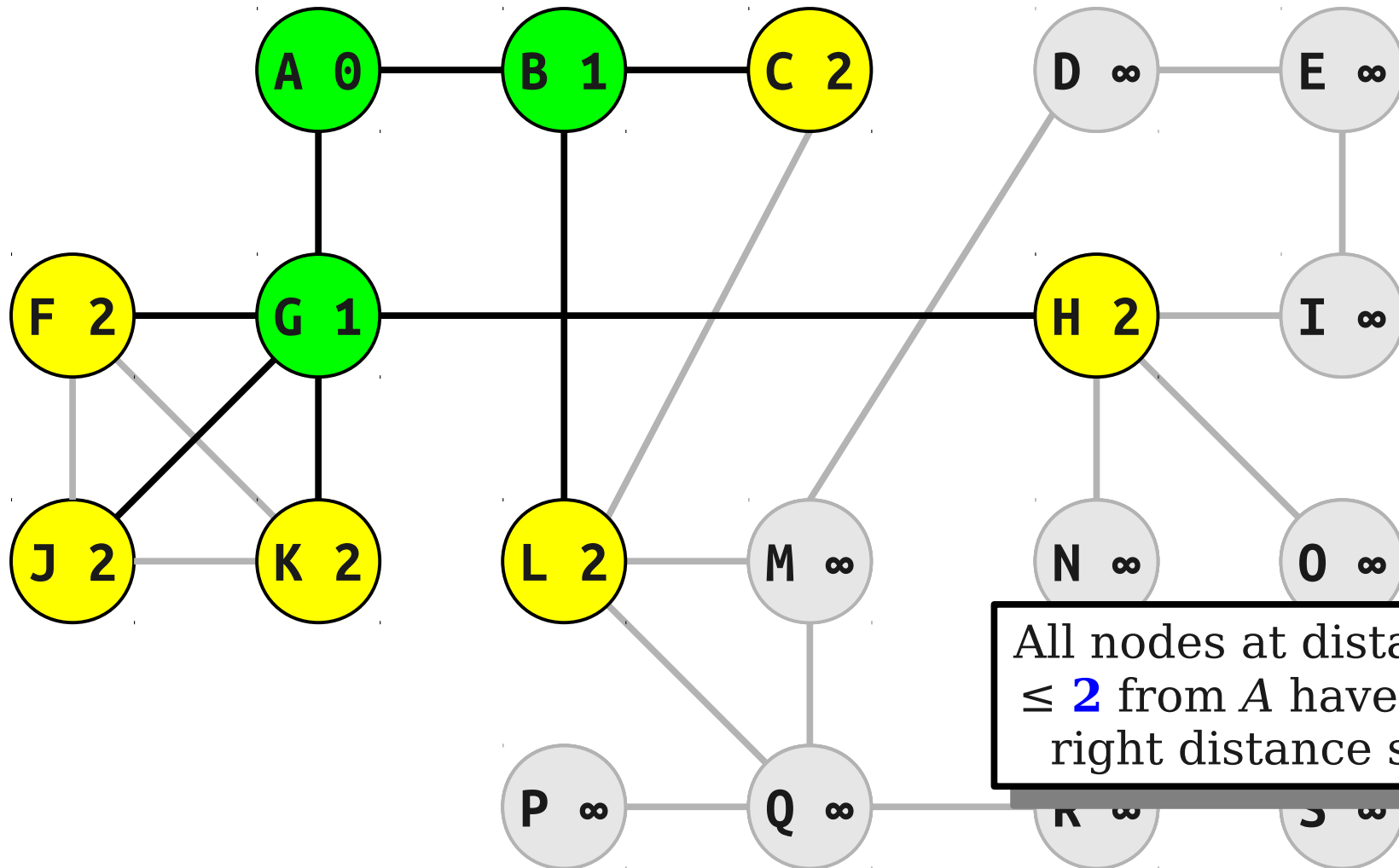


# Breadth-First Search

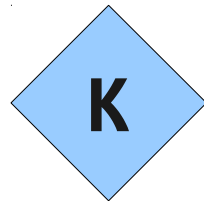
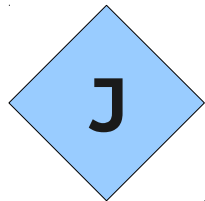
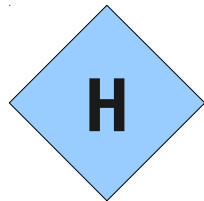
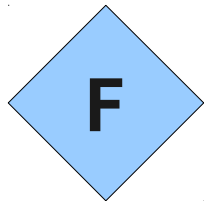
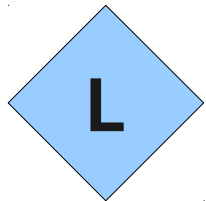
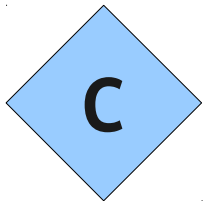
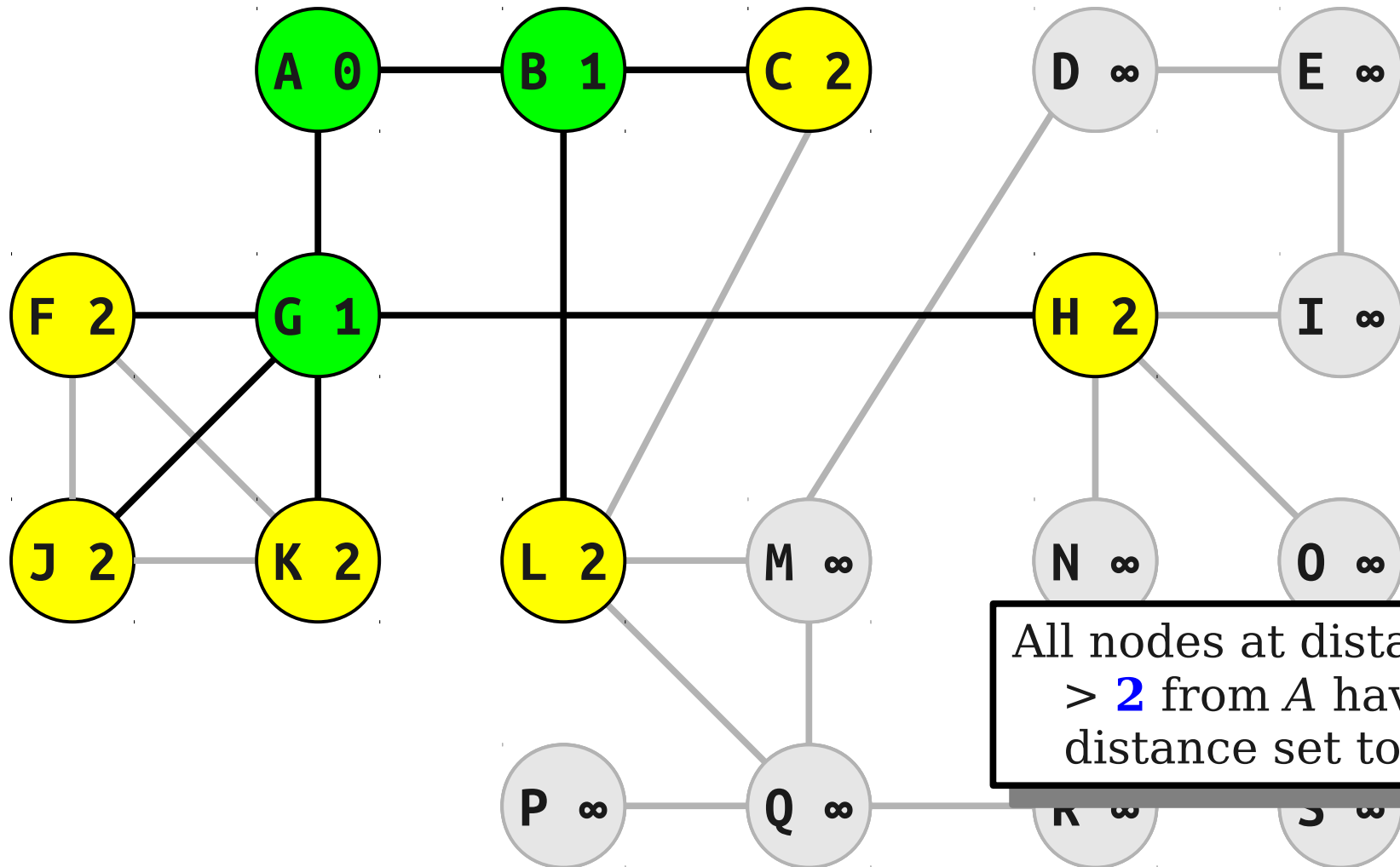




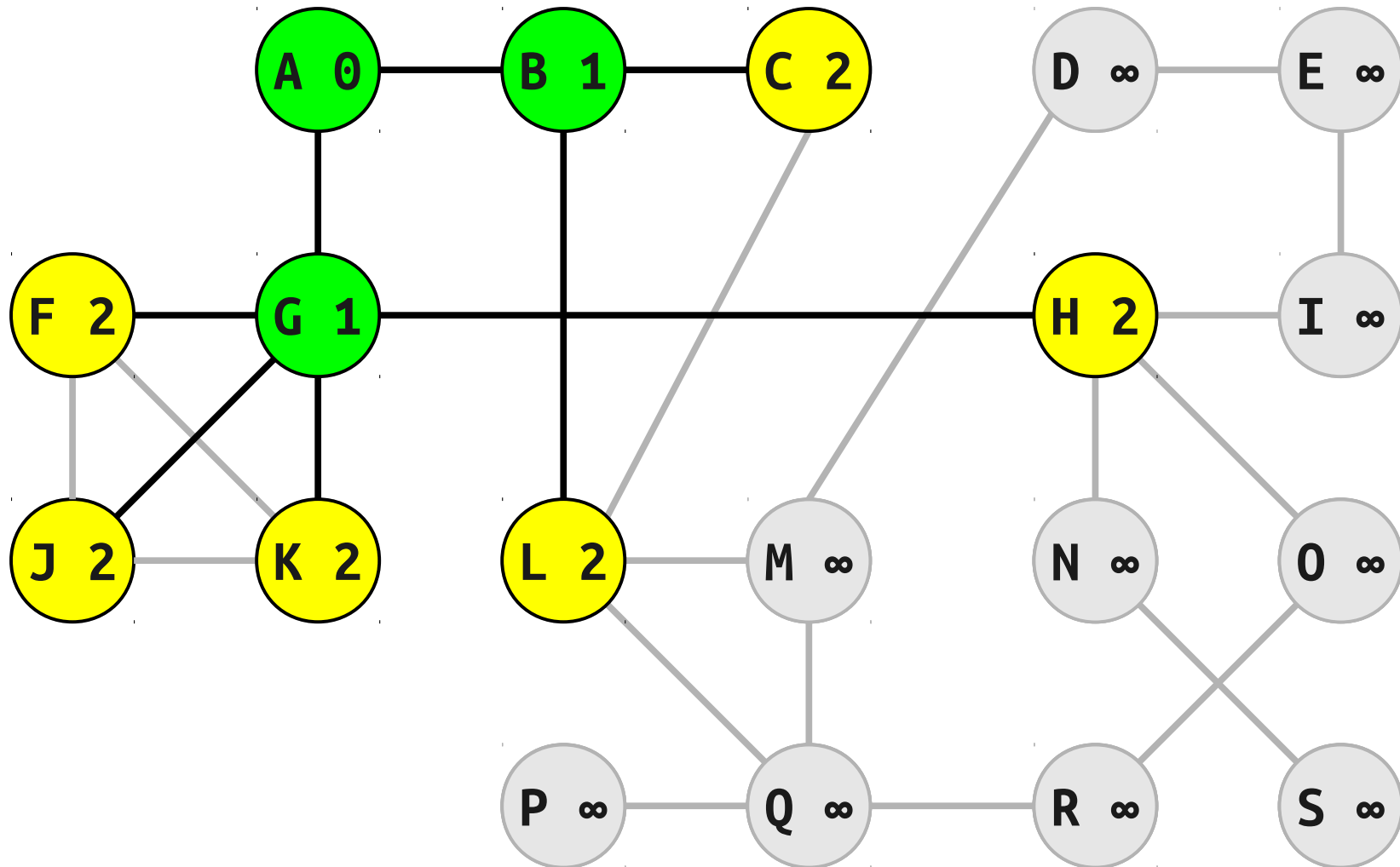
# Breadth-First Search



# Breadth-First Search



# Breadth-First Search



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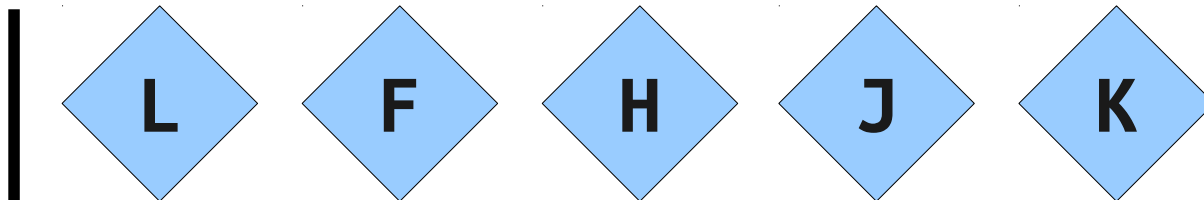
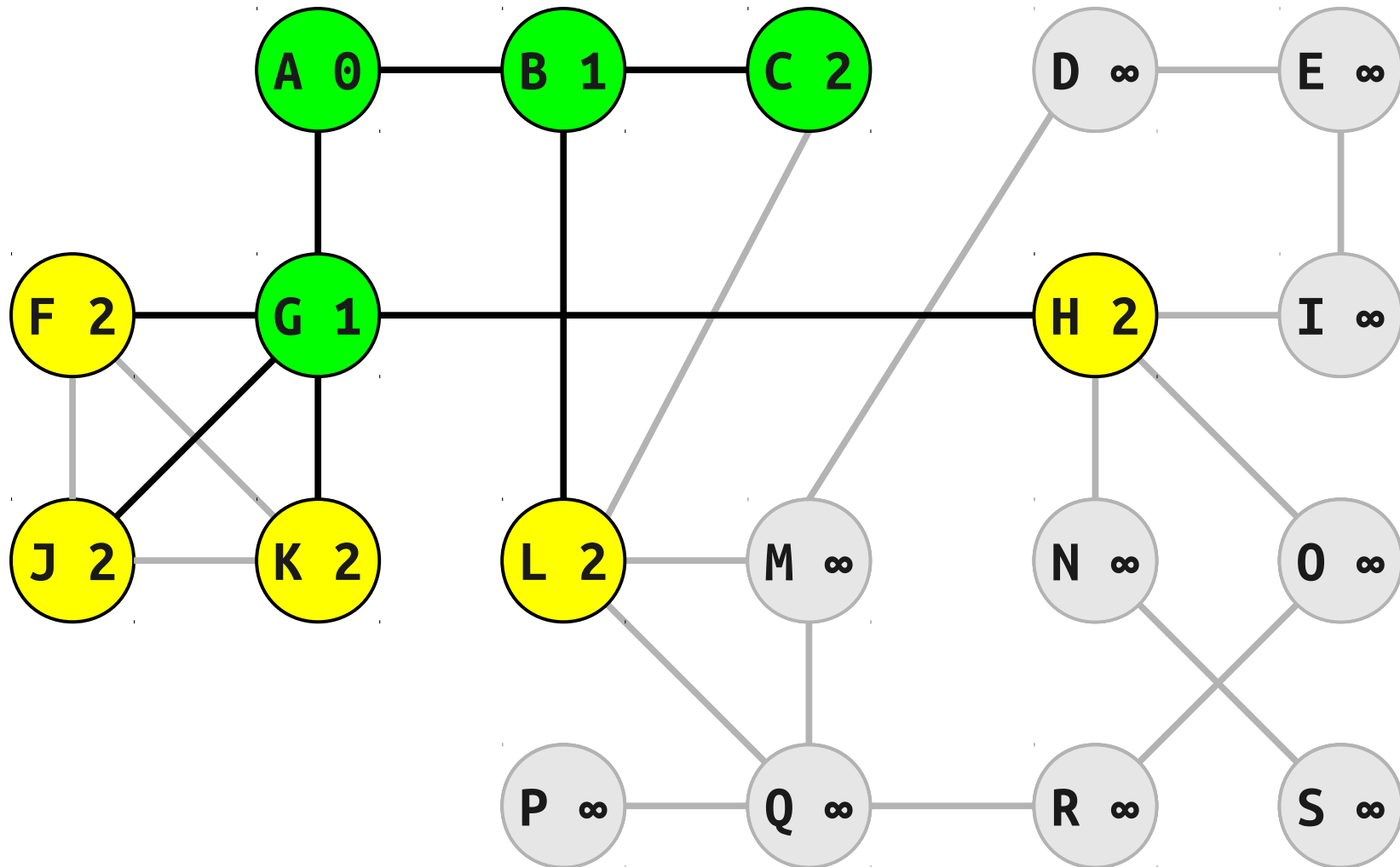
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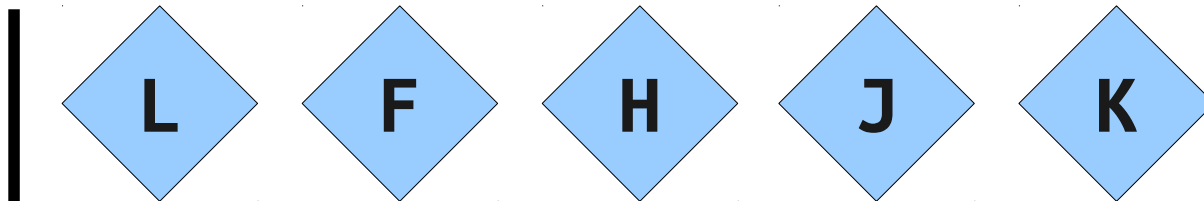
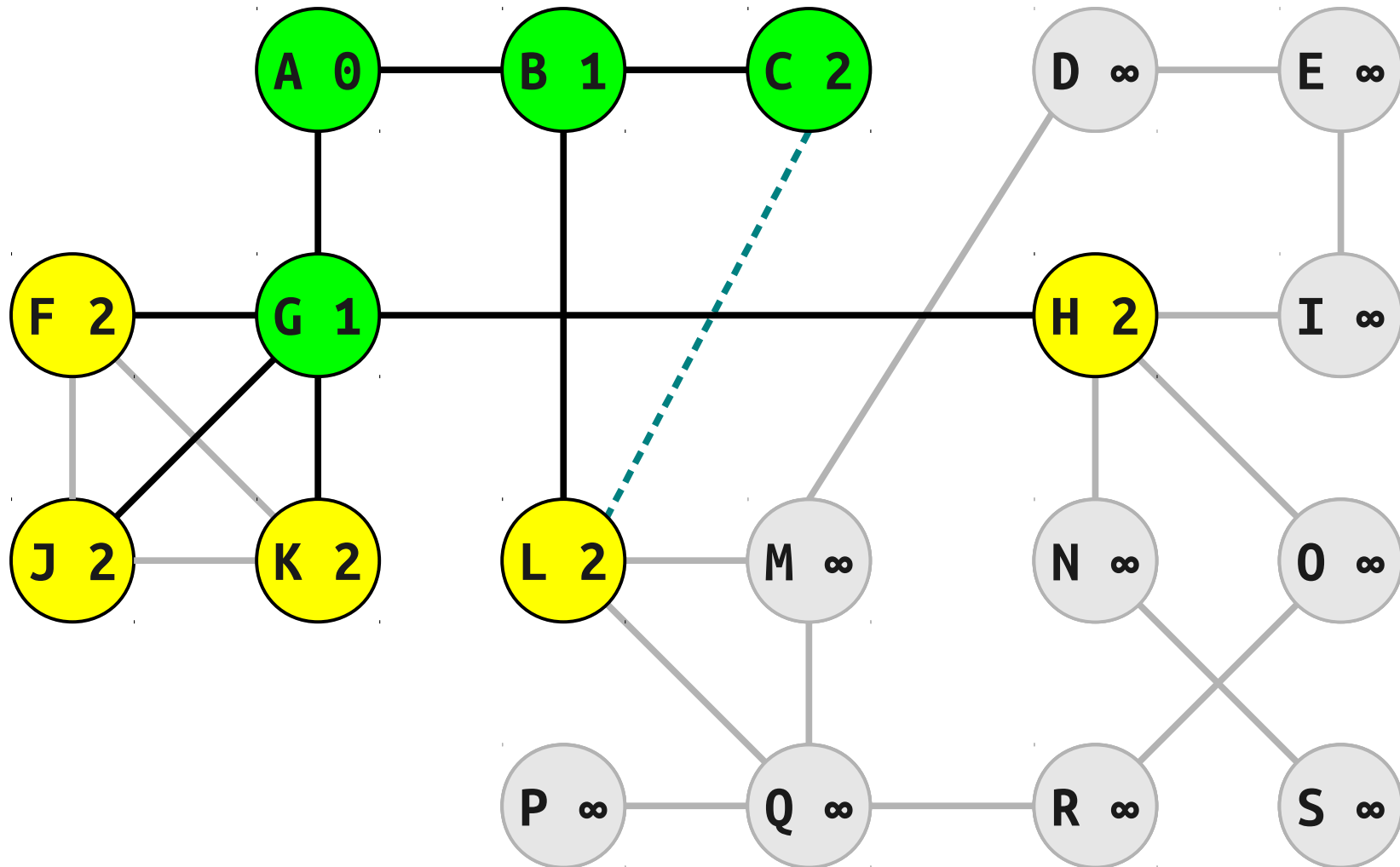
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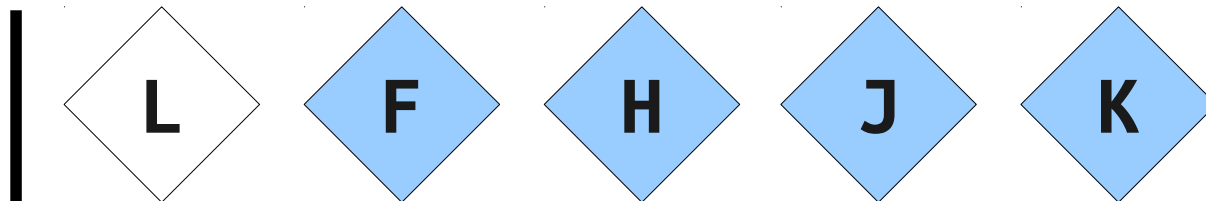
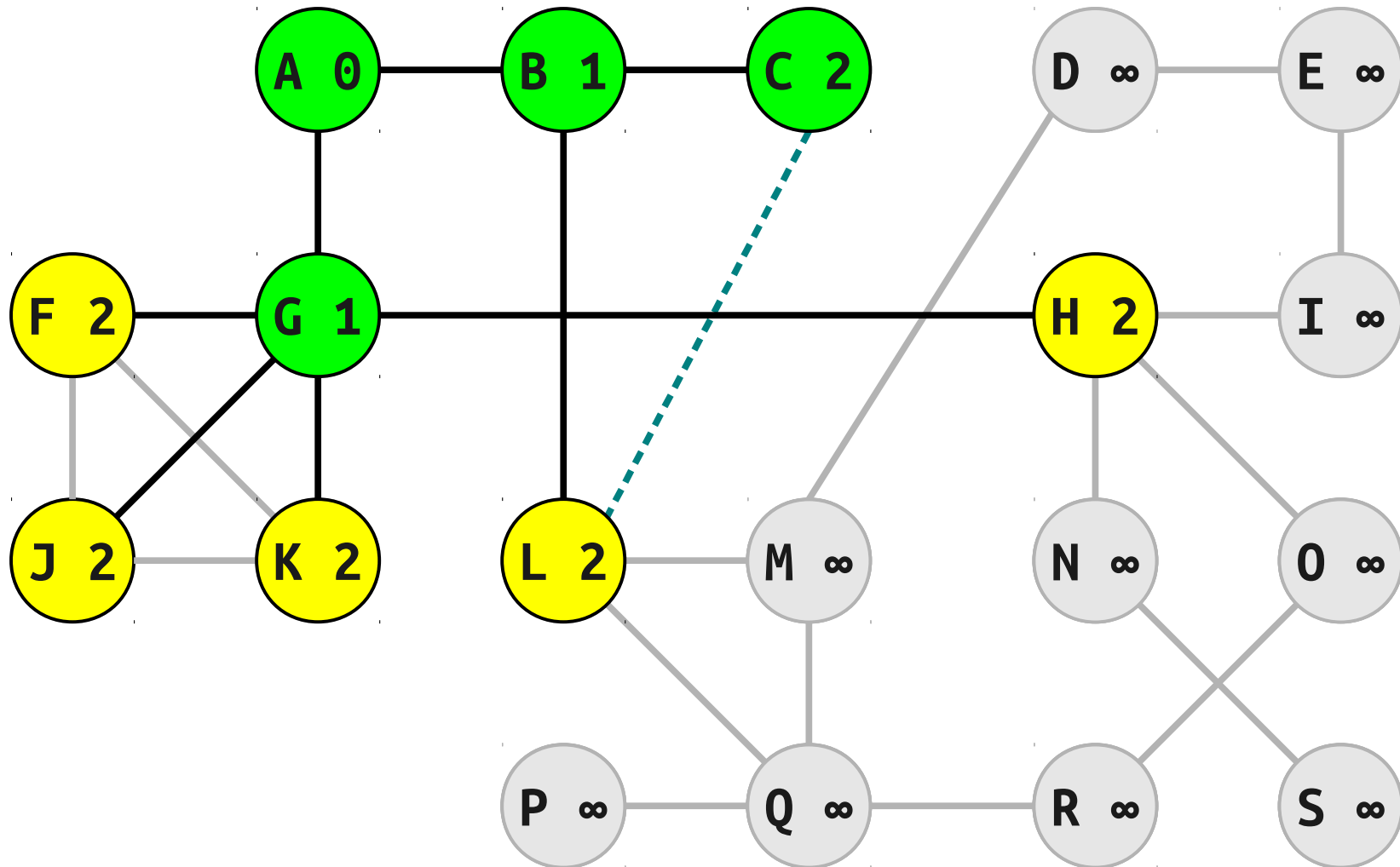
# Breadth-First Search



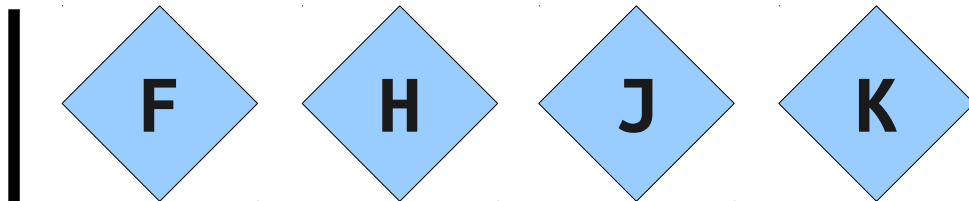
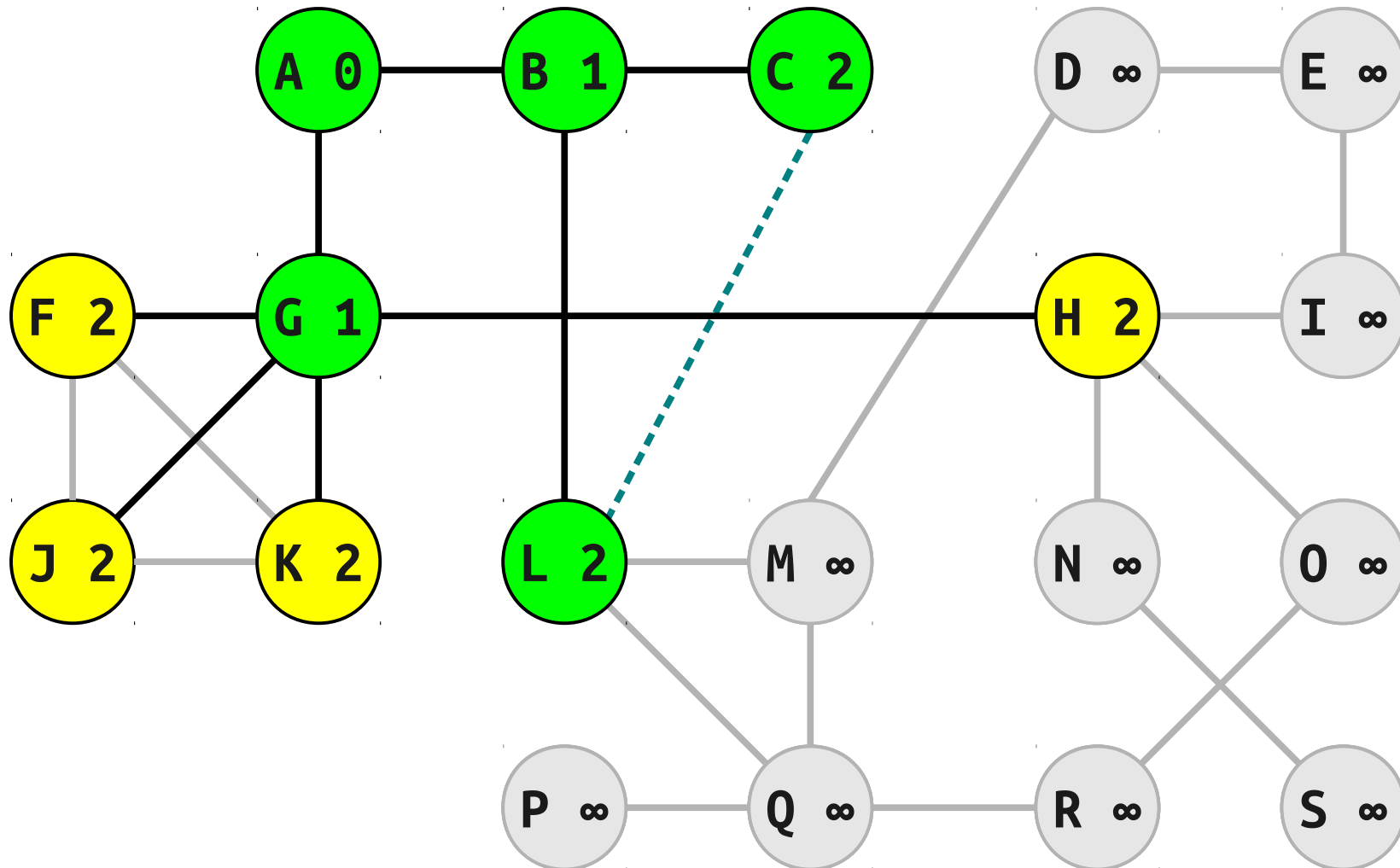
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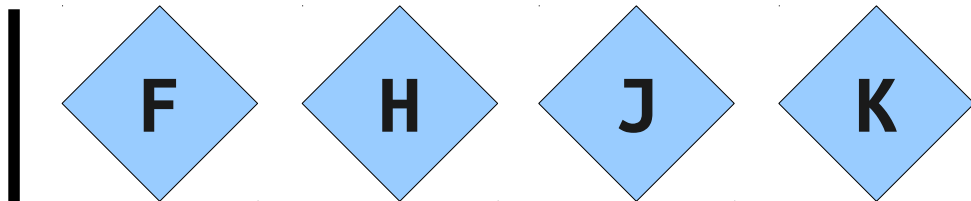
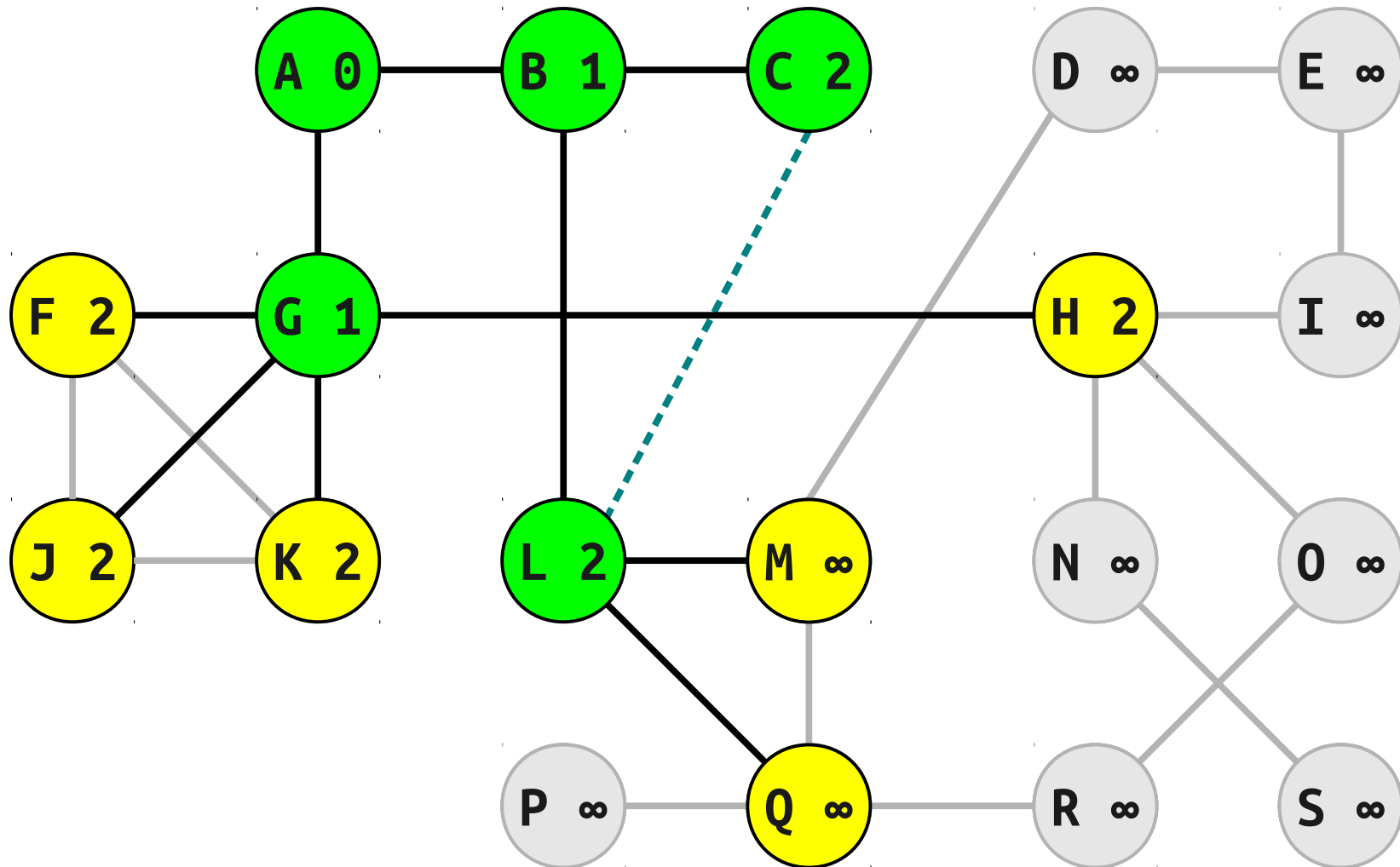
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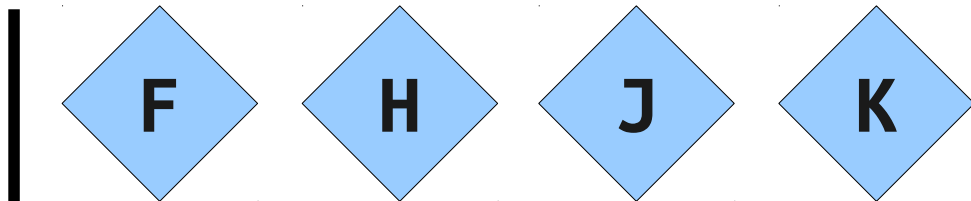
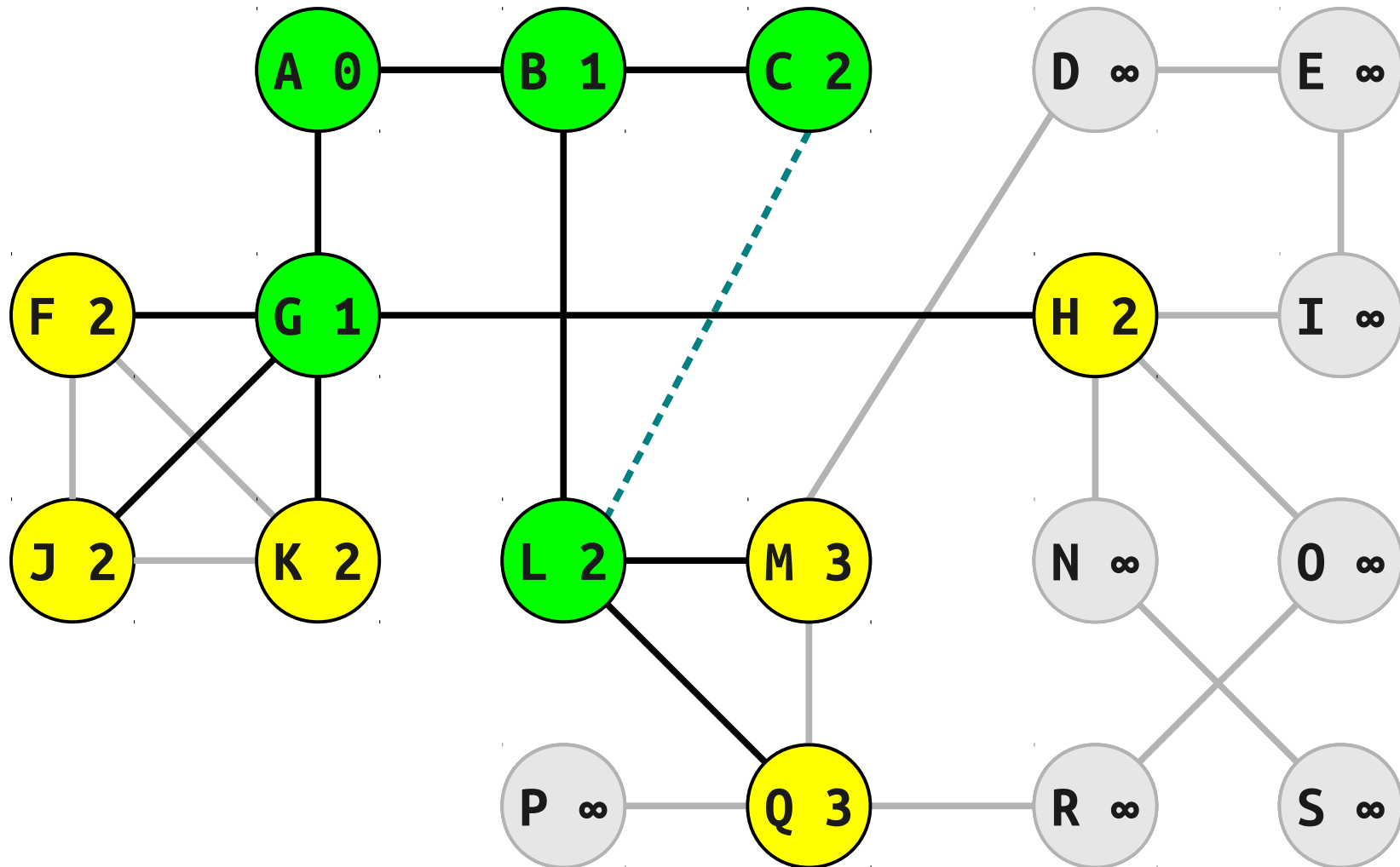


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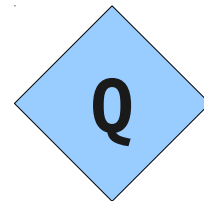
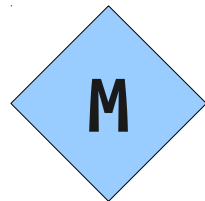
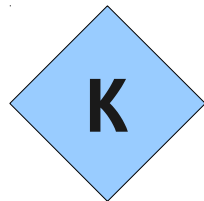
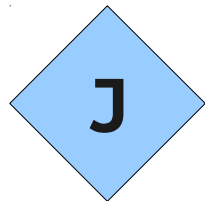
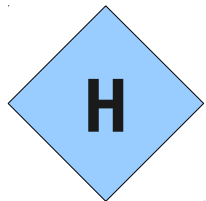
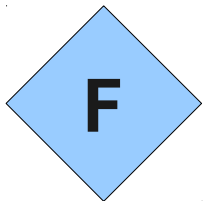
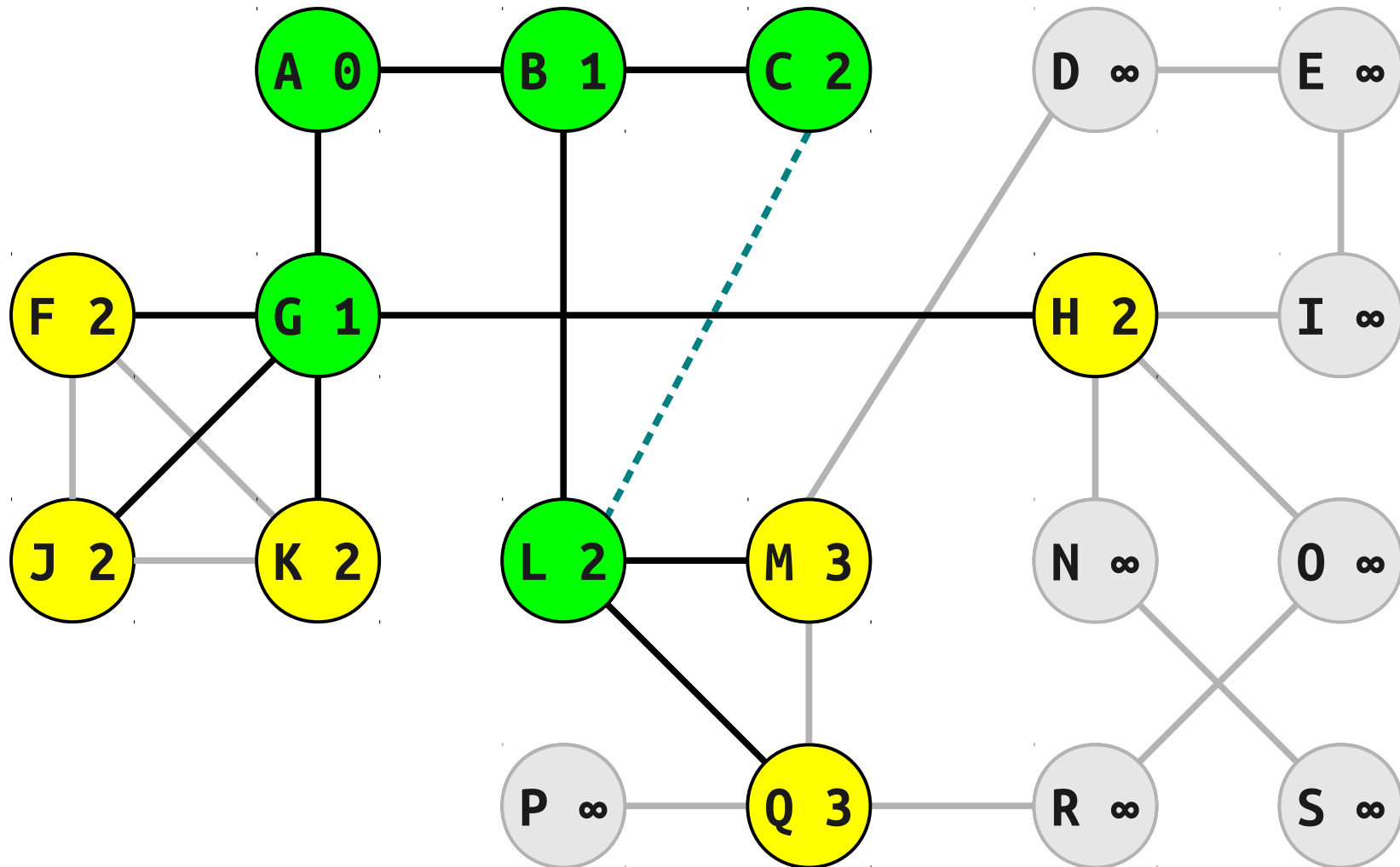




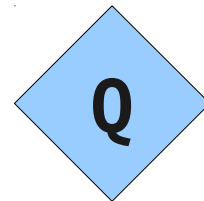
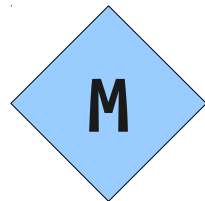
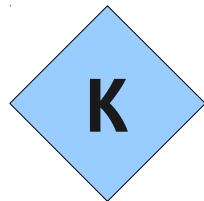
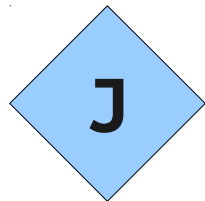
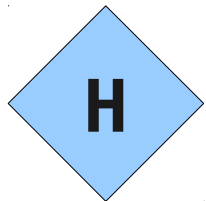
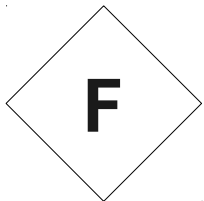
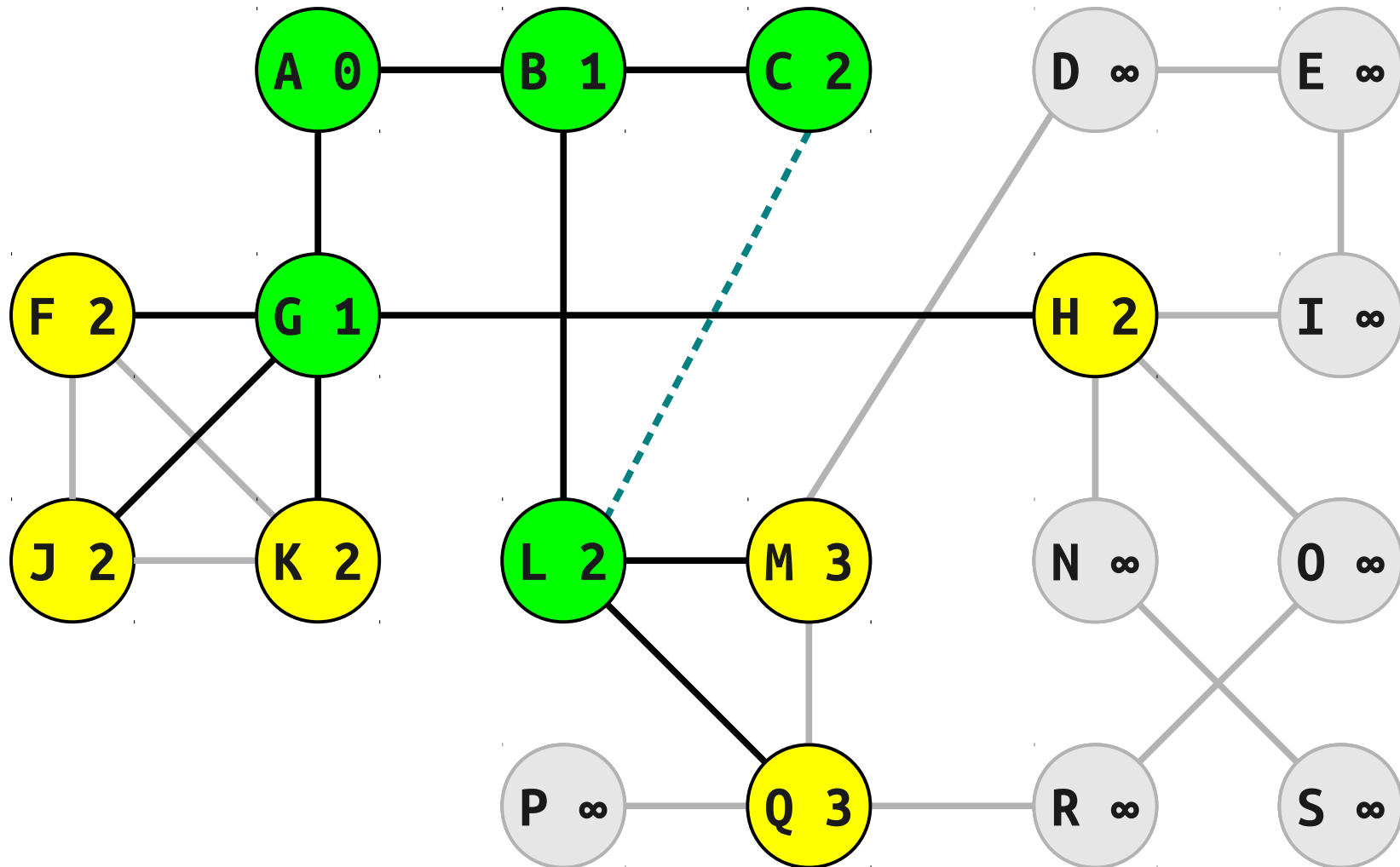
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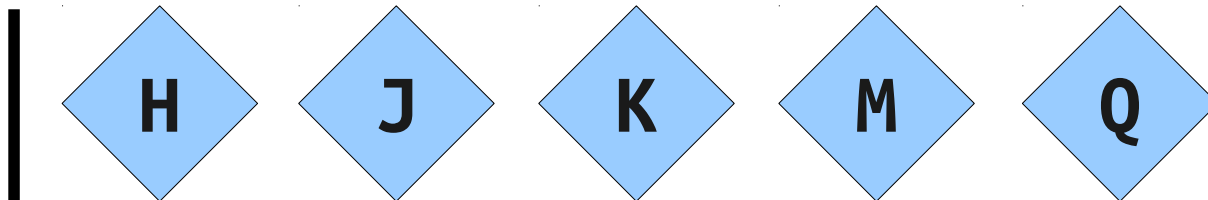
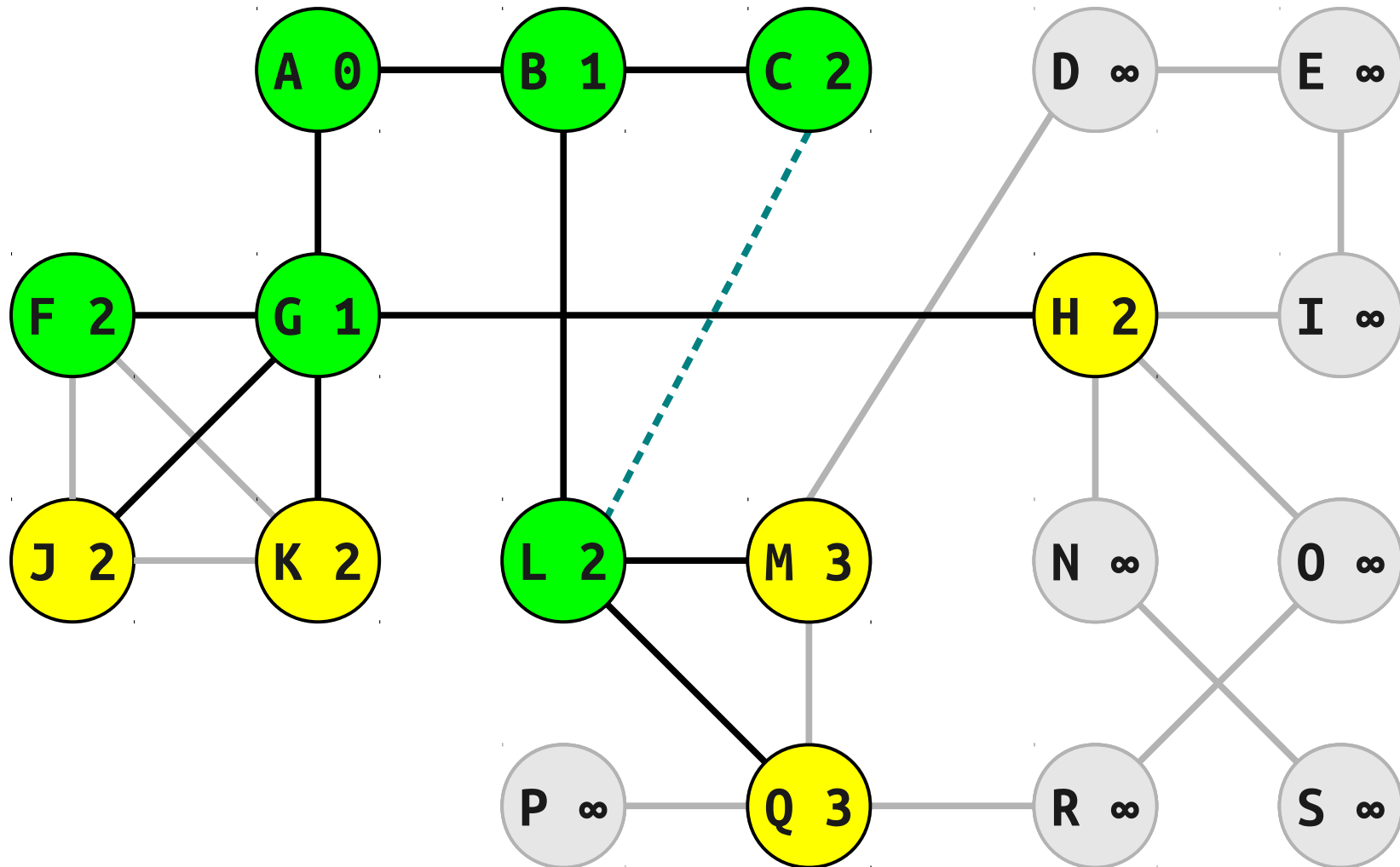
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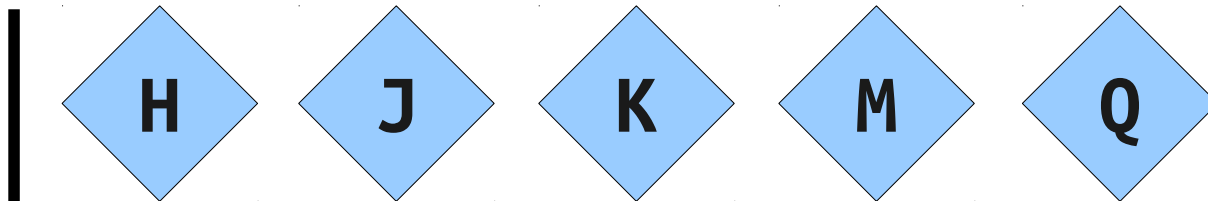
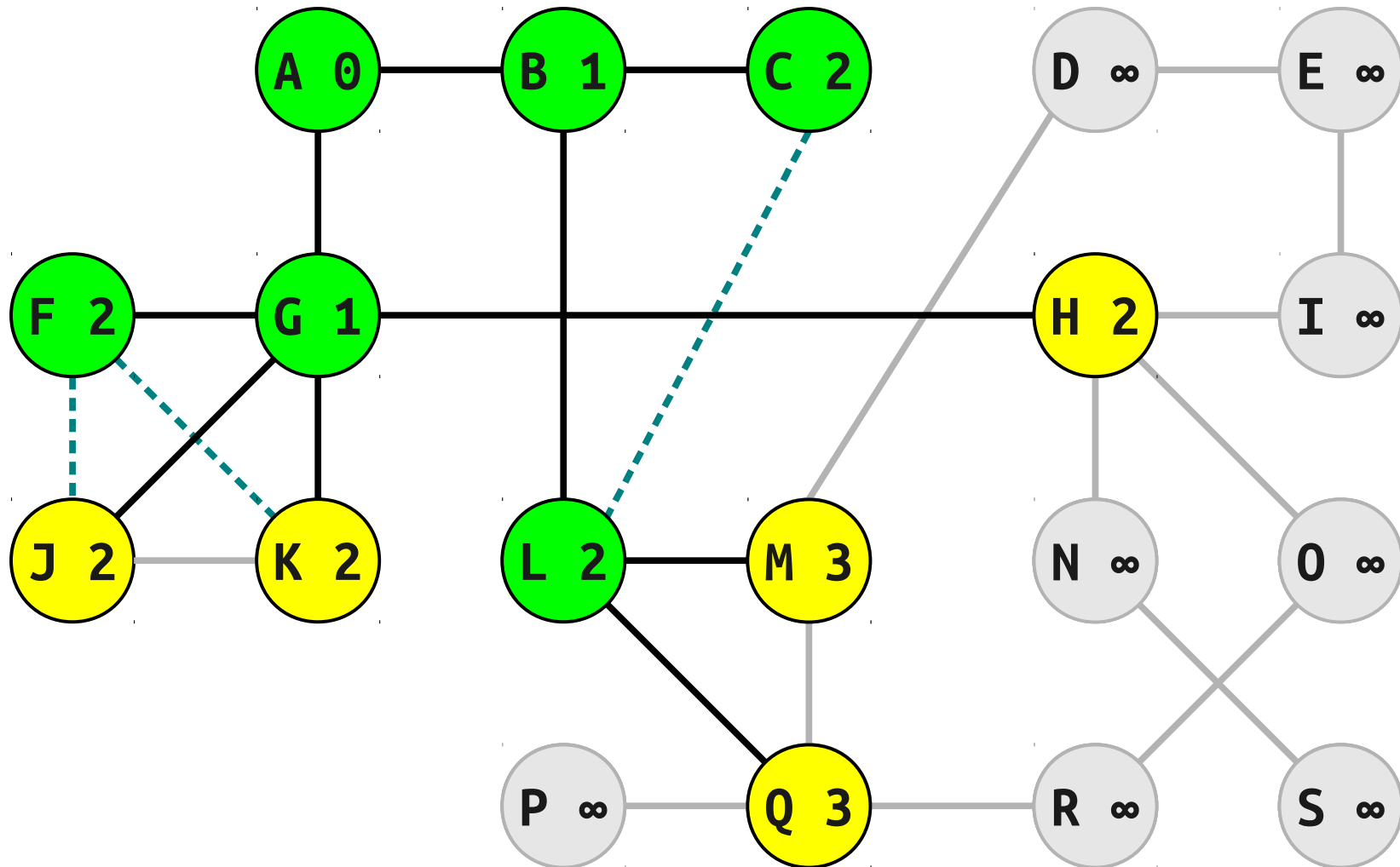
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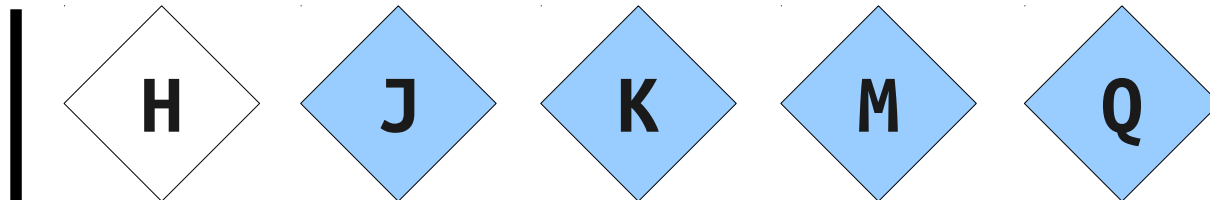
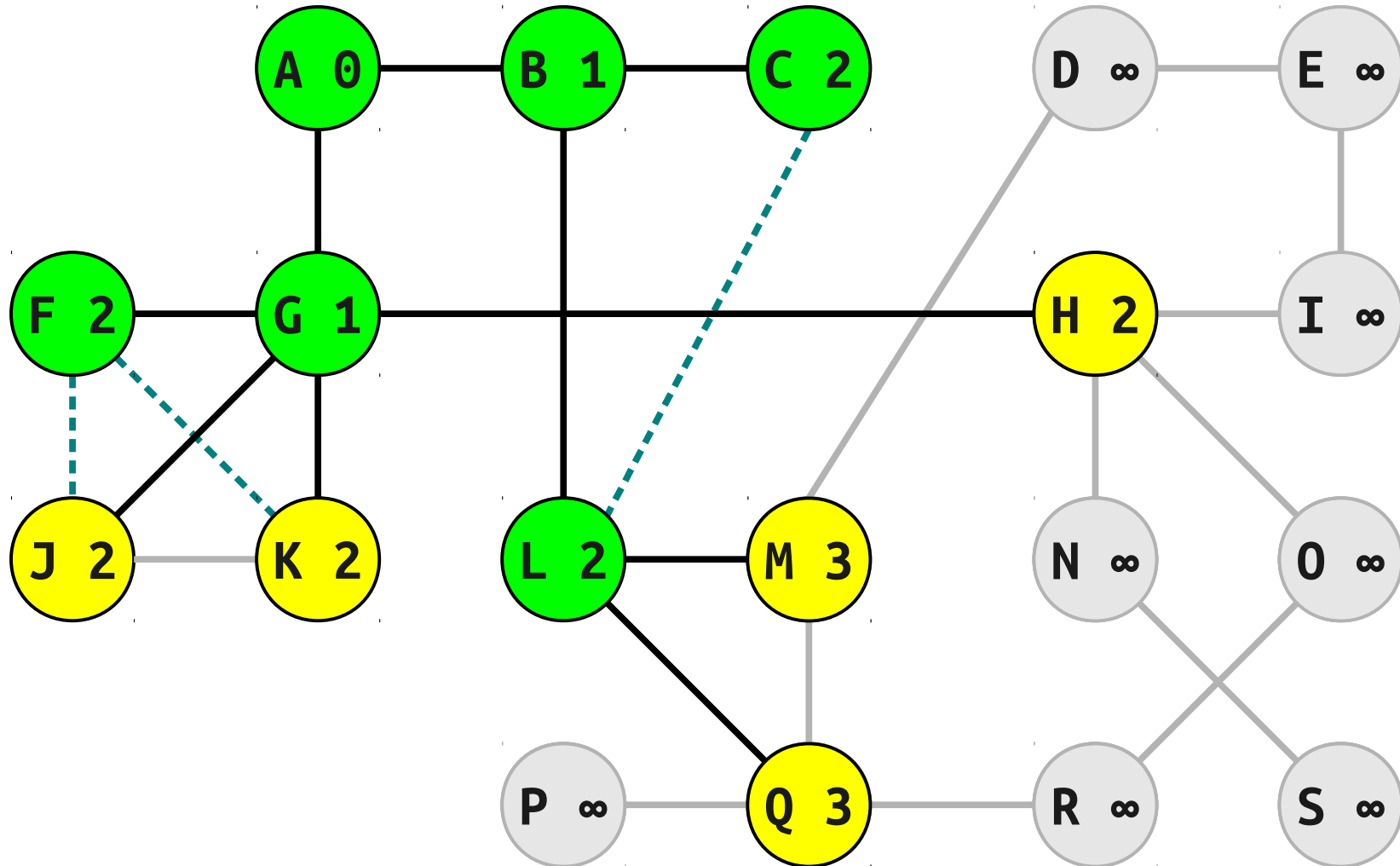
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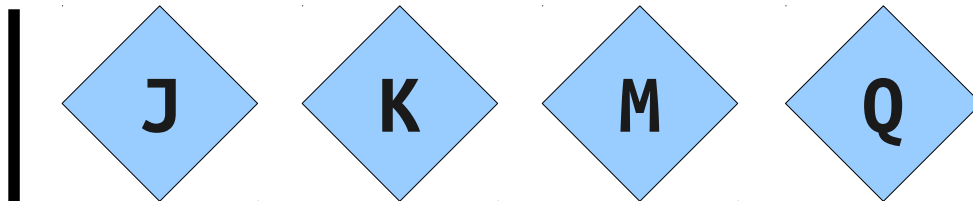
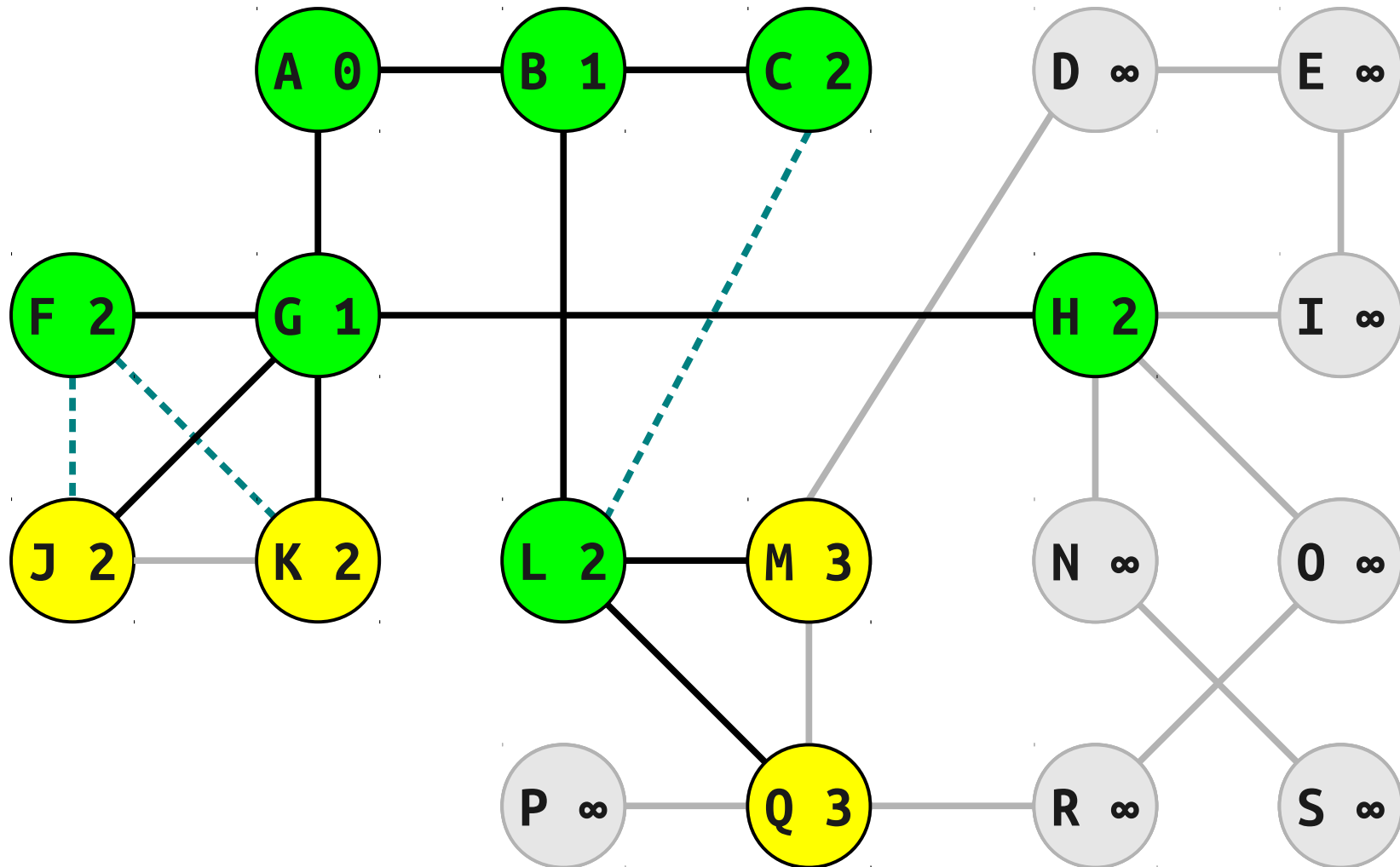
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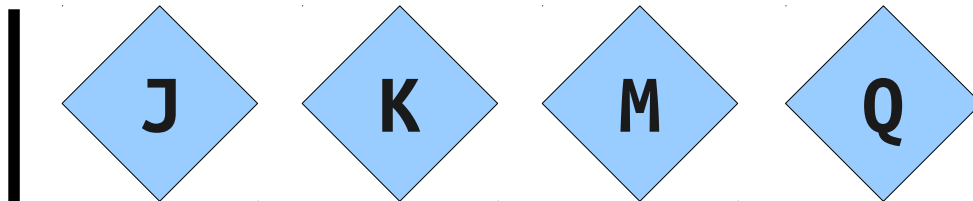
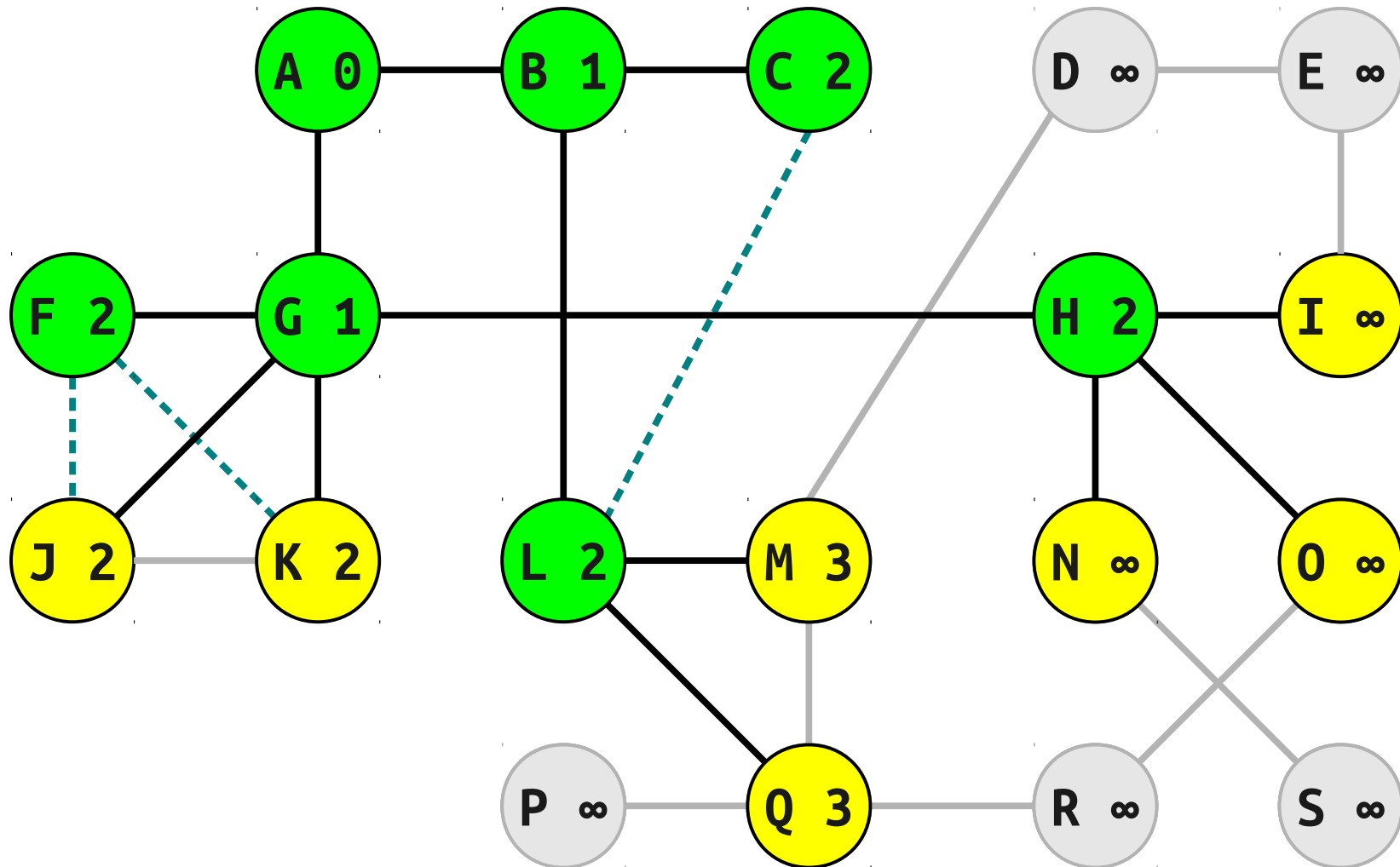
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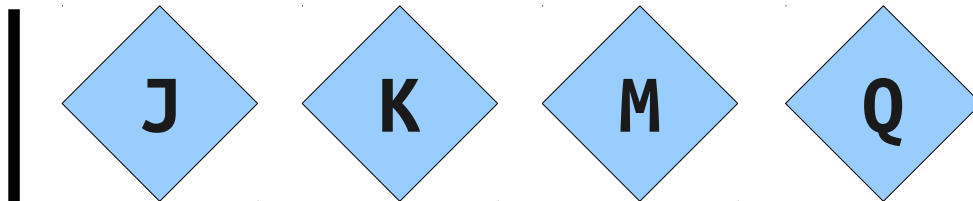
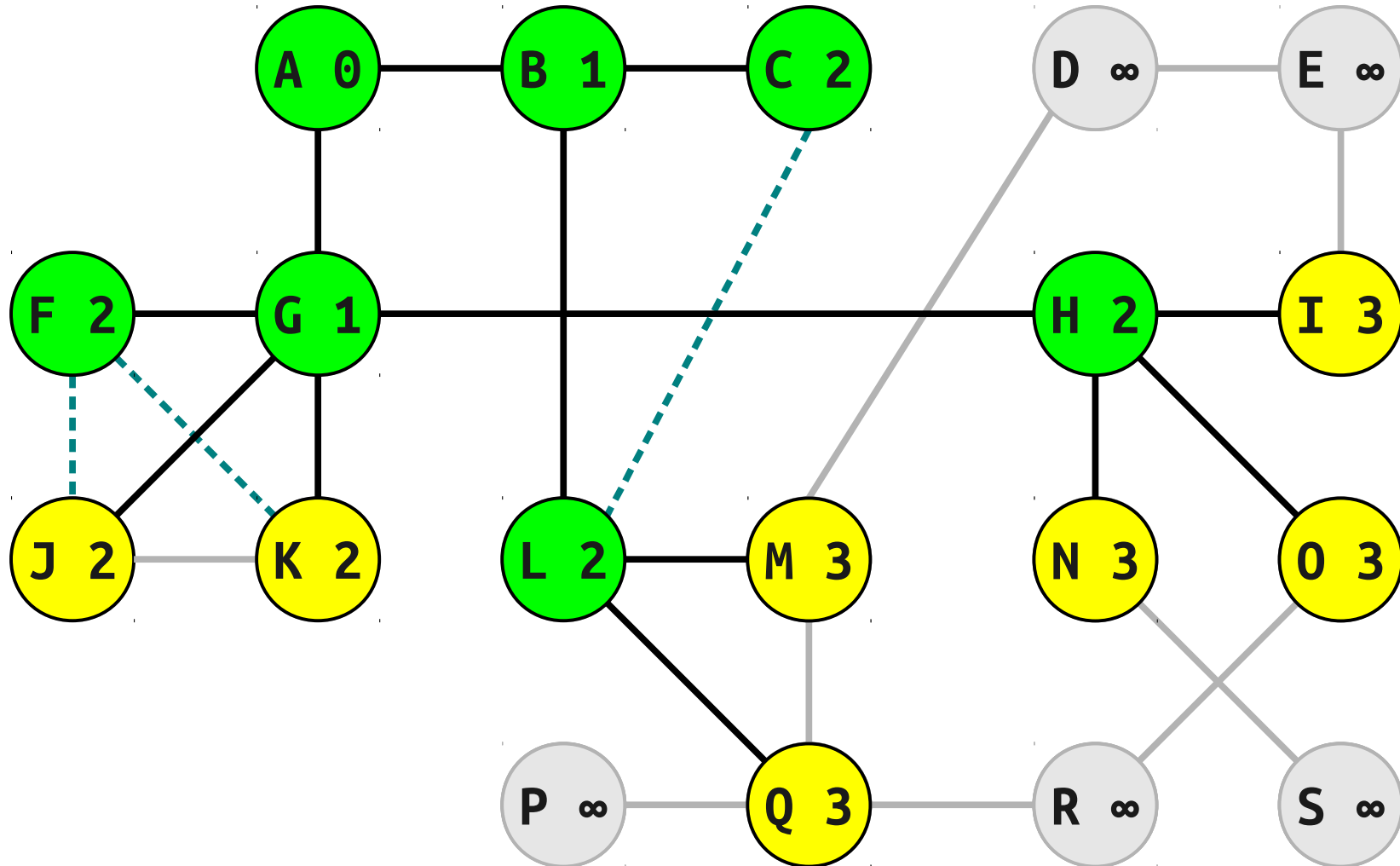


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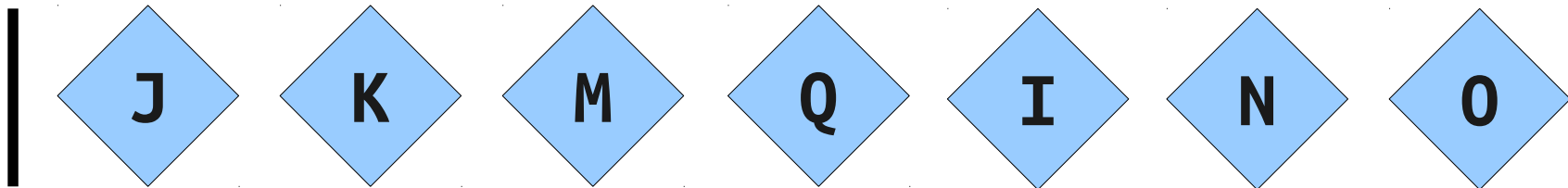
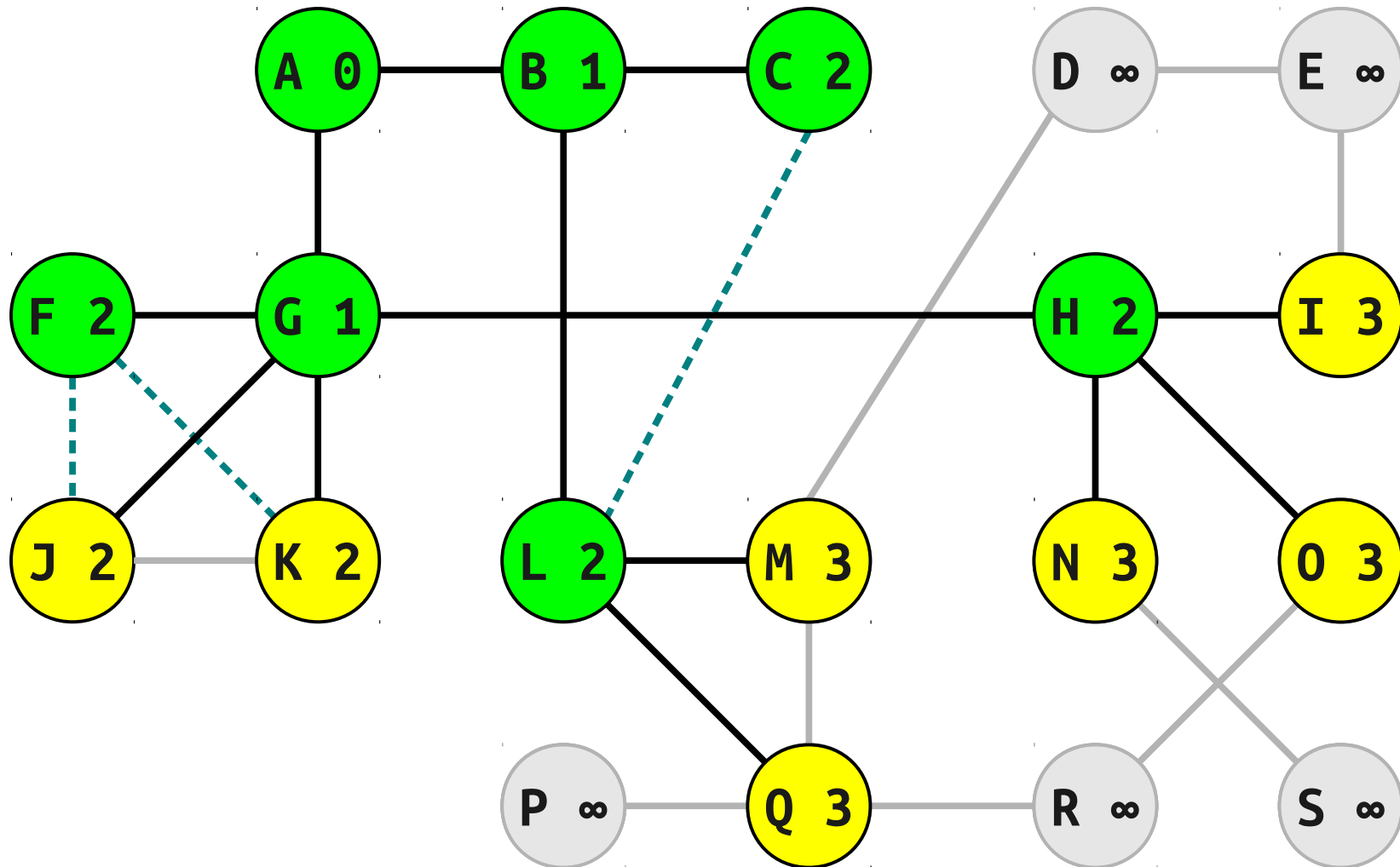




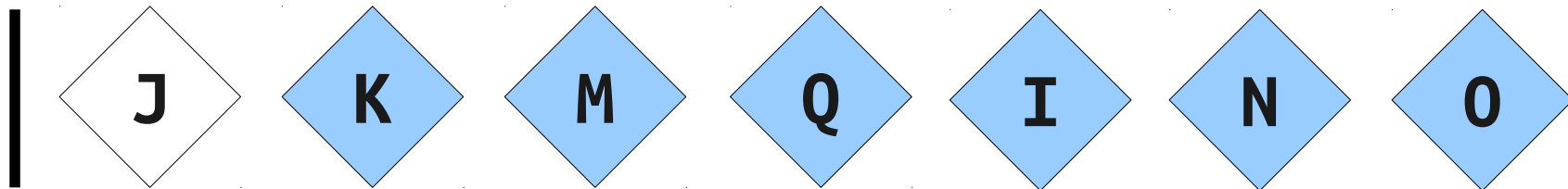
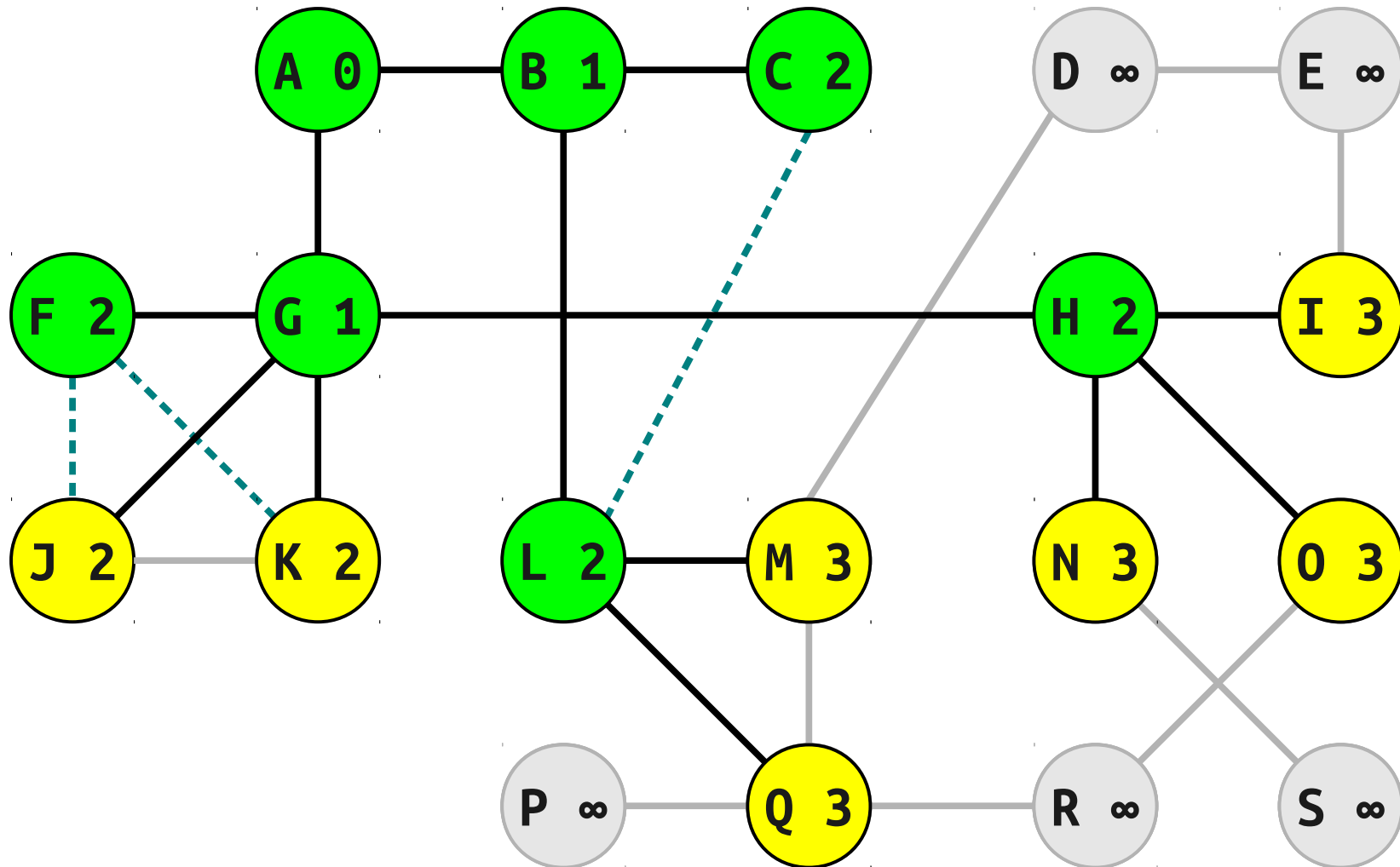
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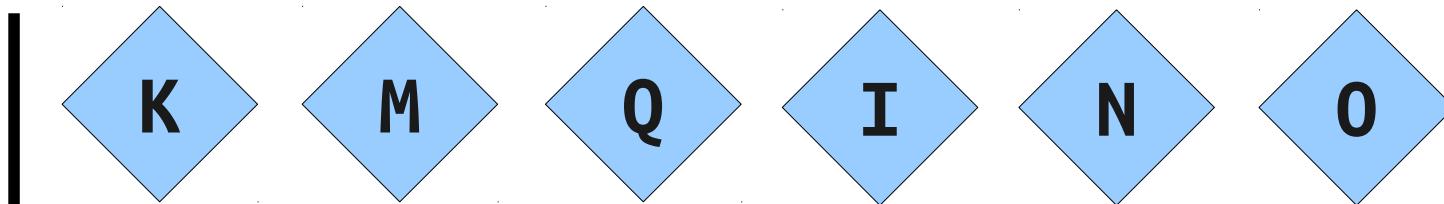
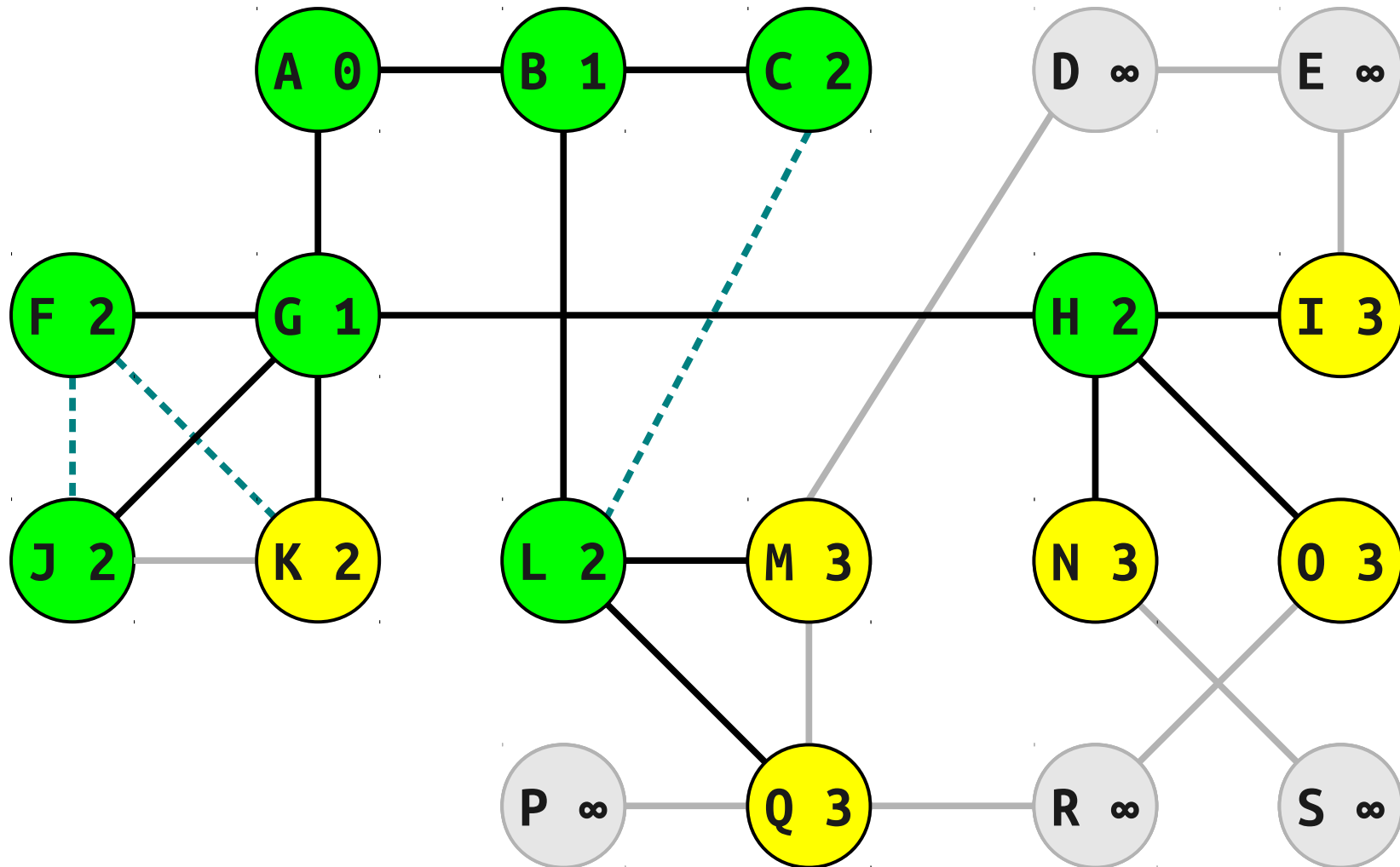
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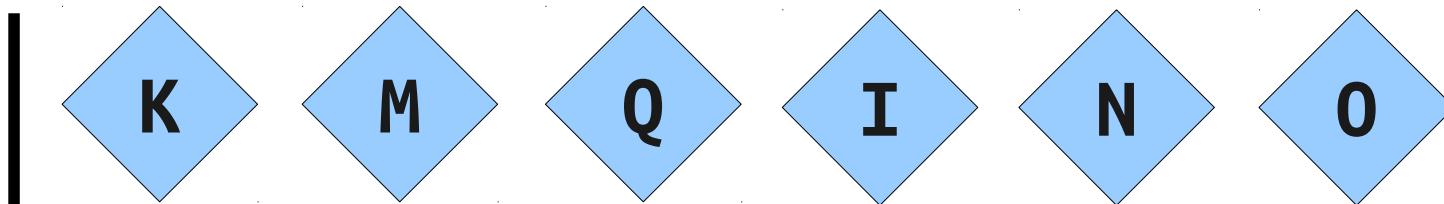
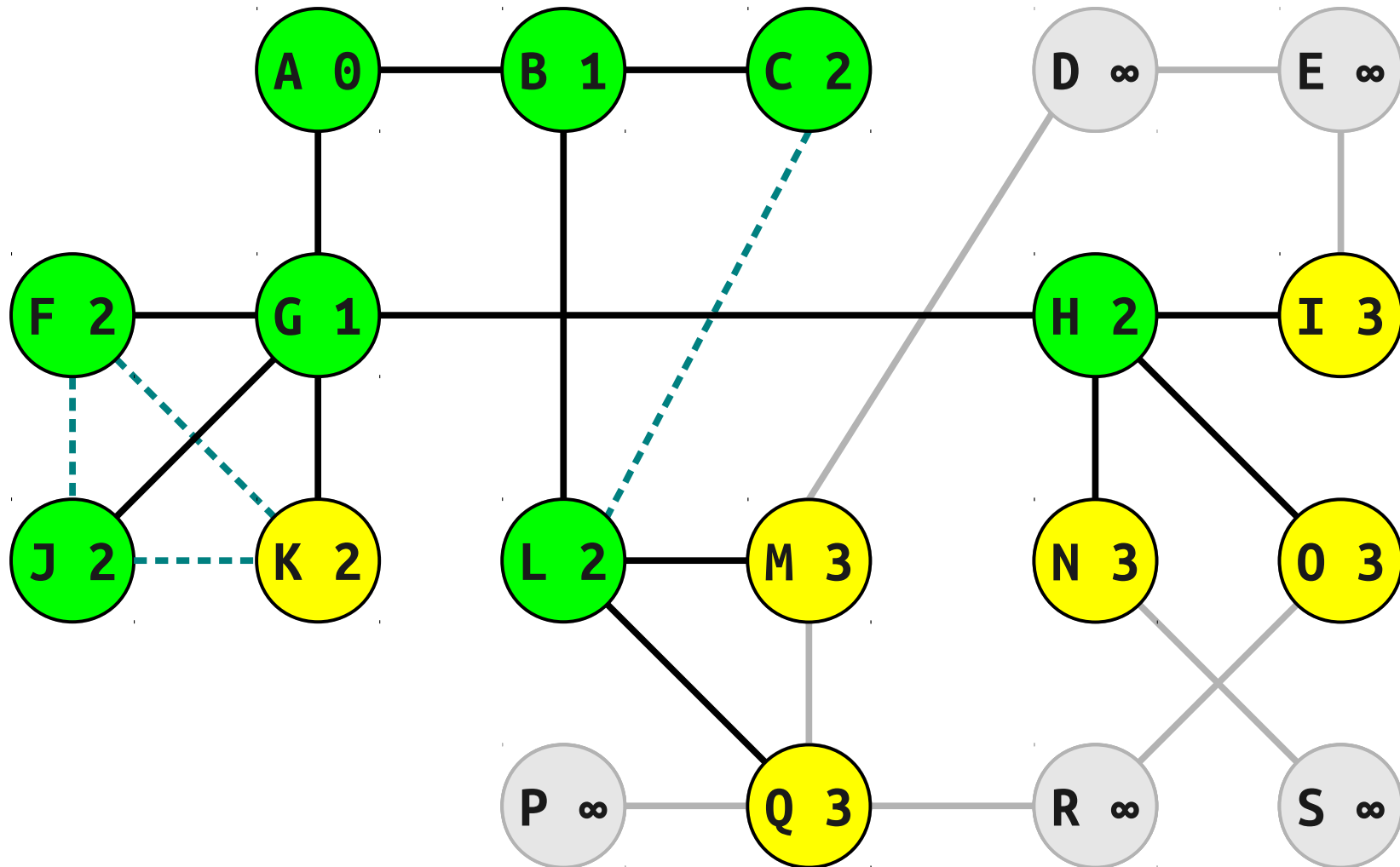
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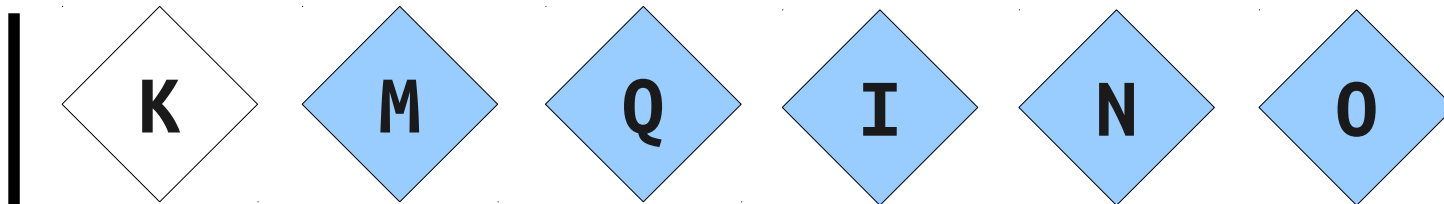
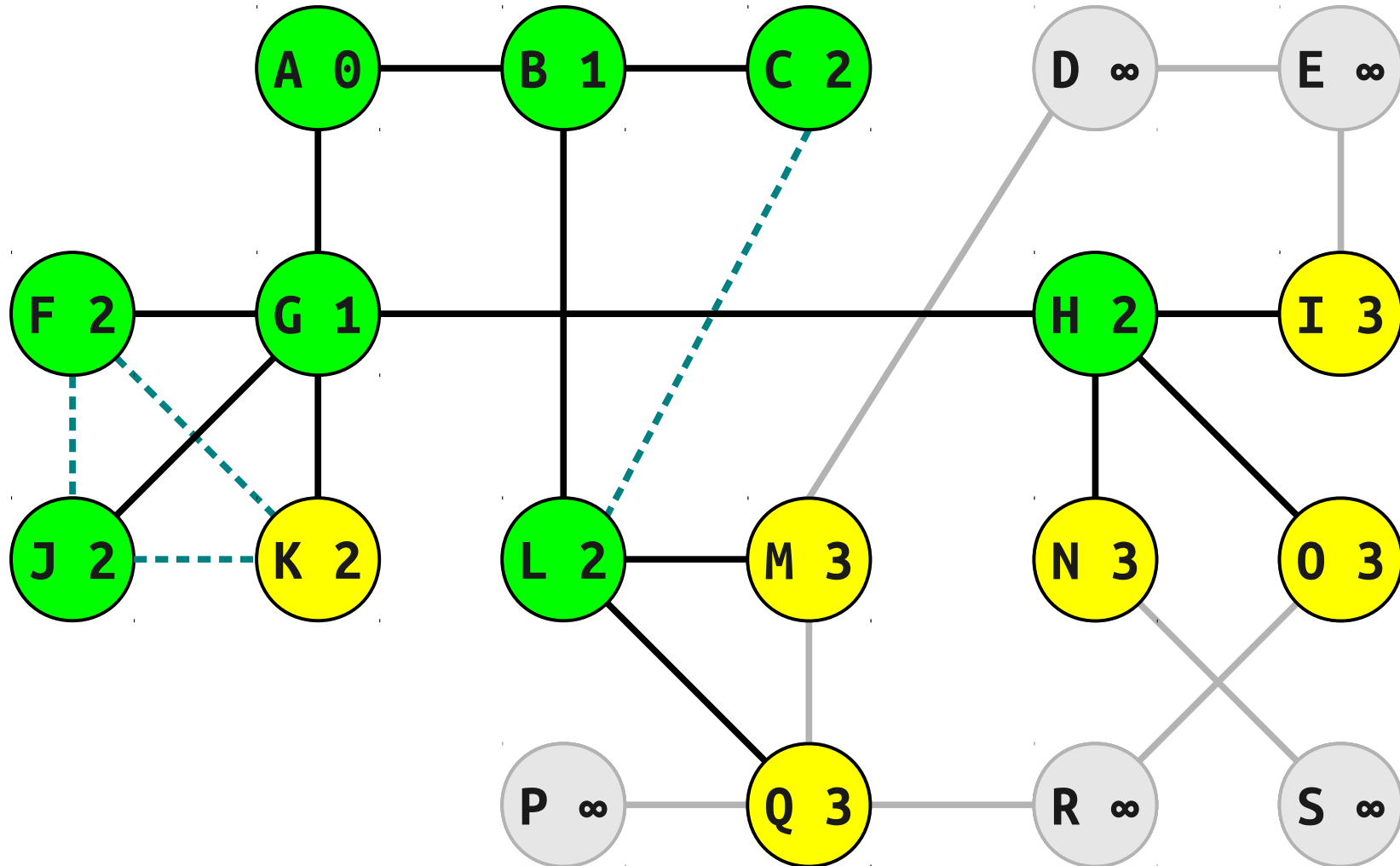
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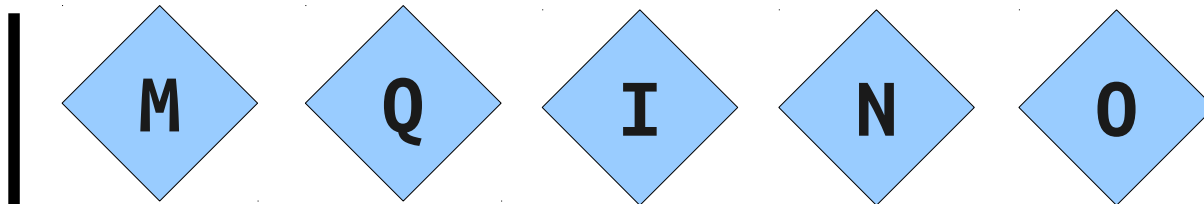
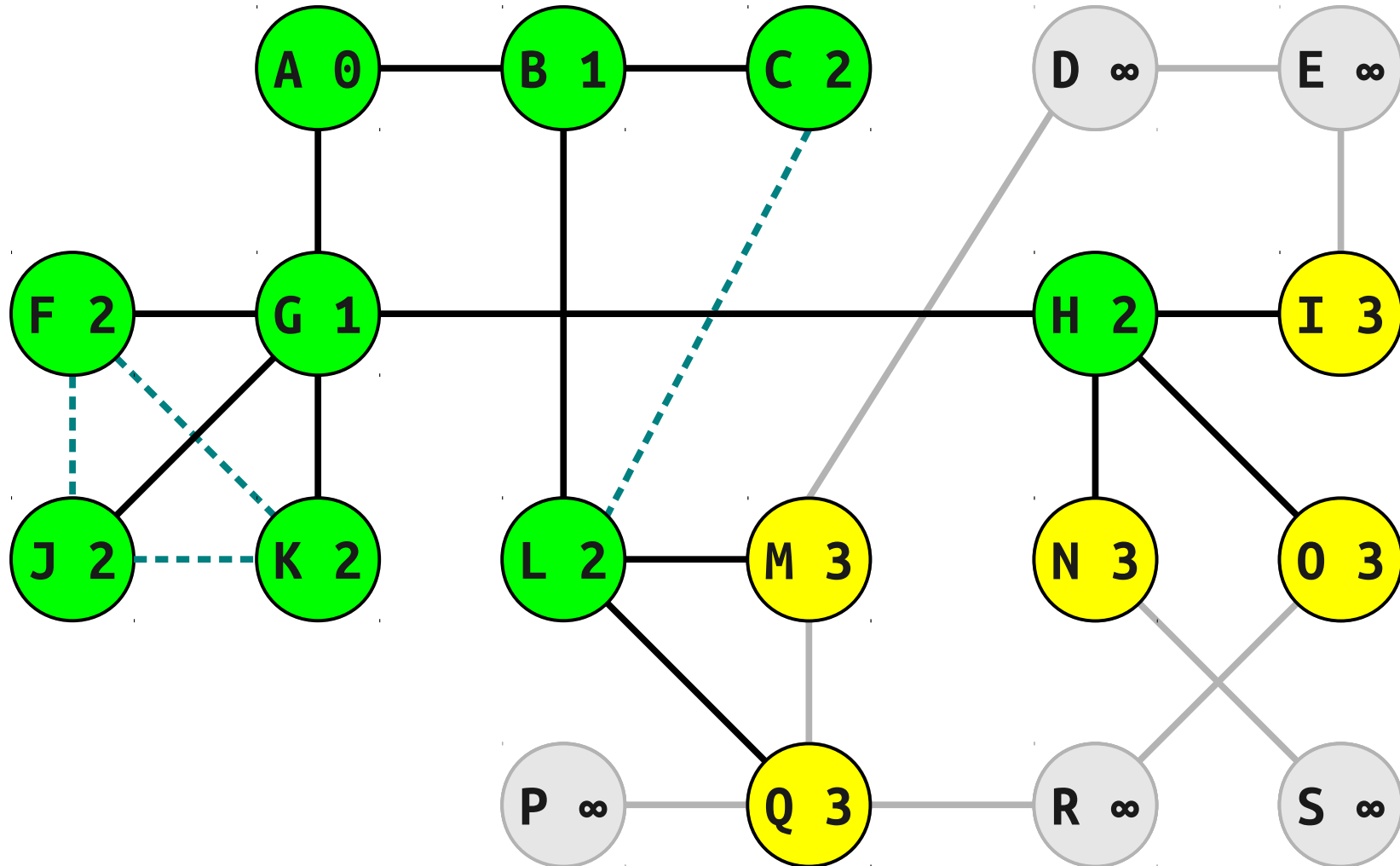
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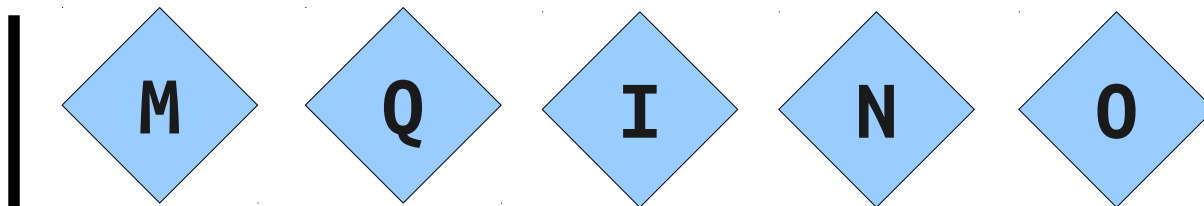
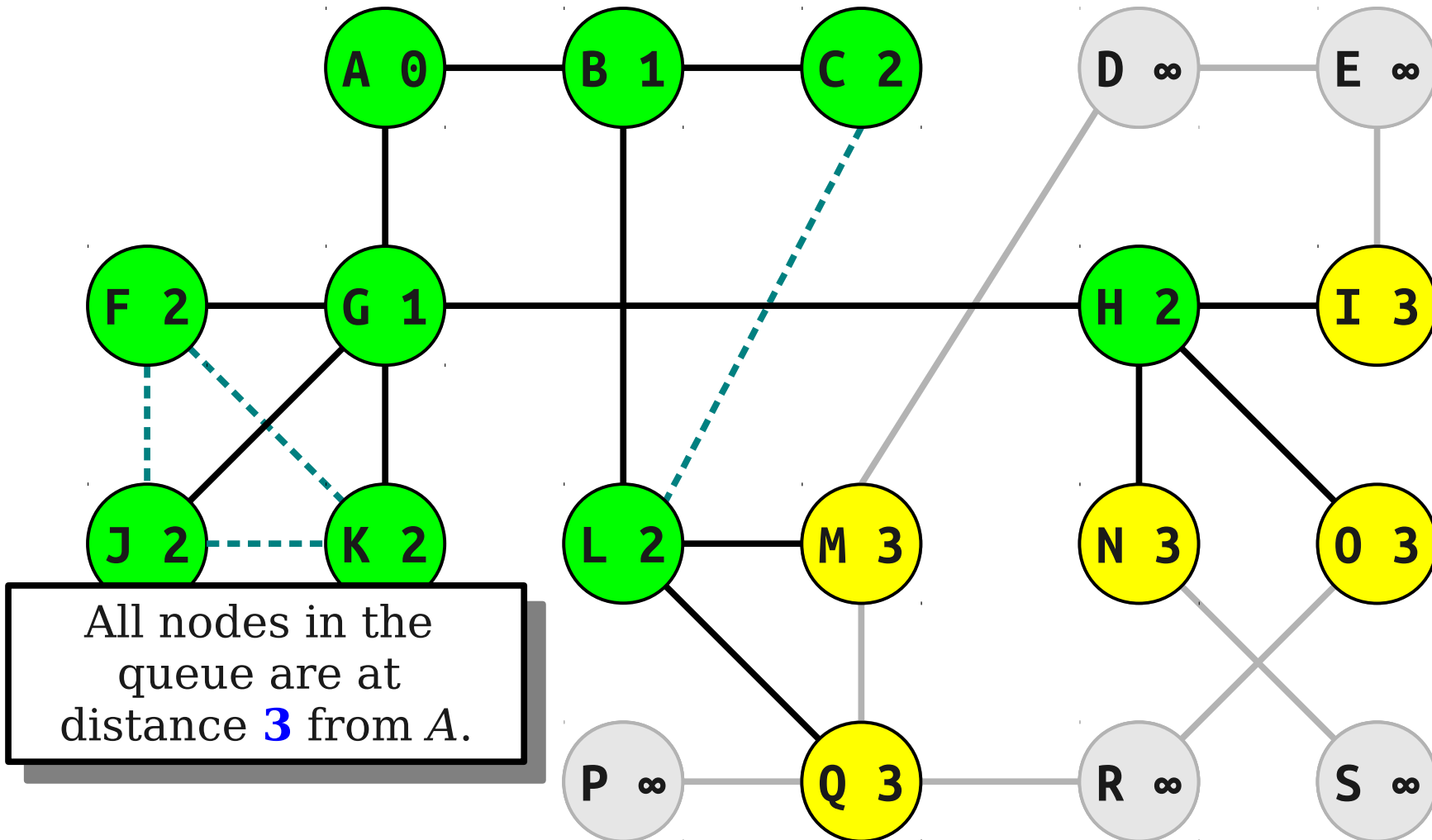
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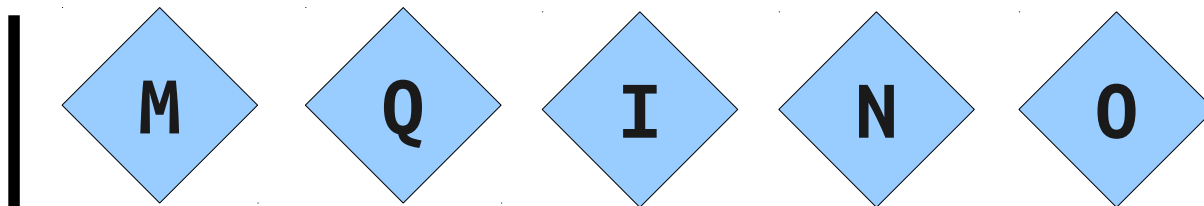
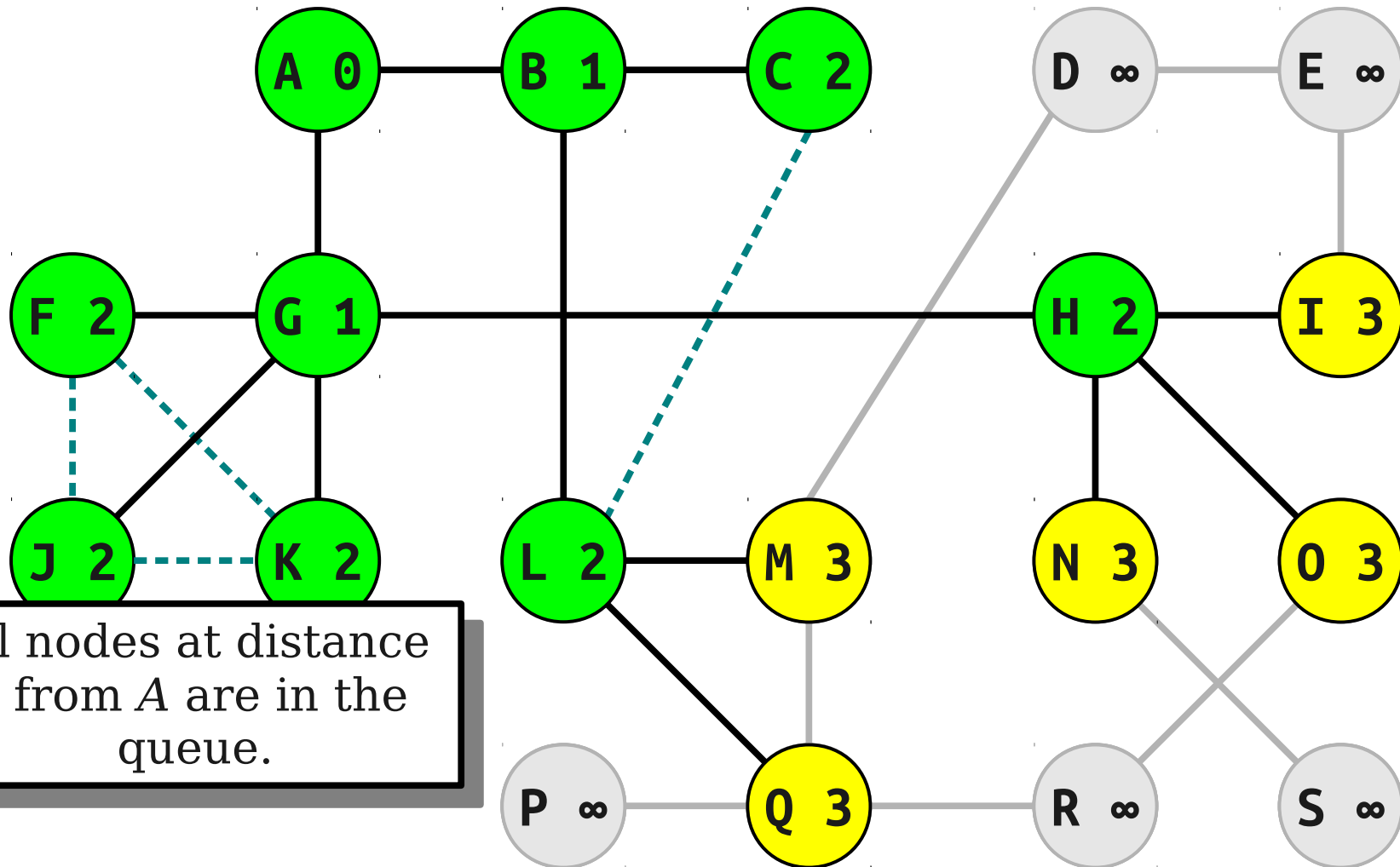


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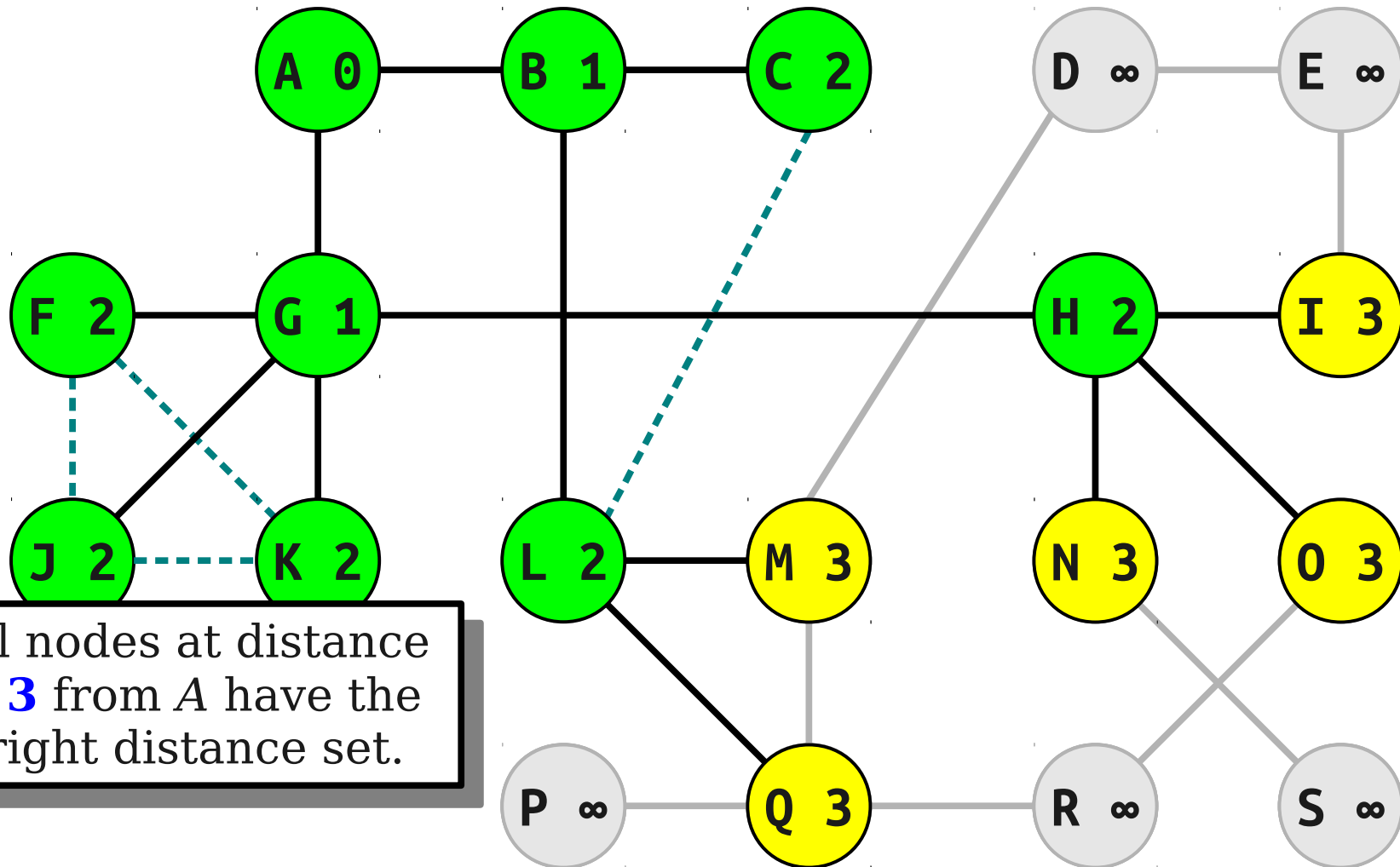




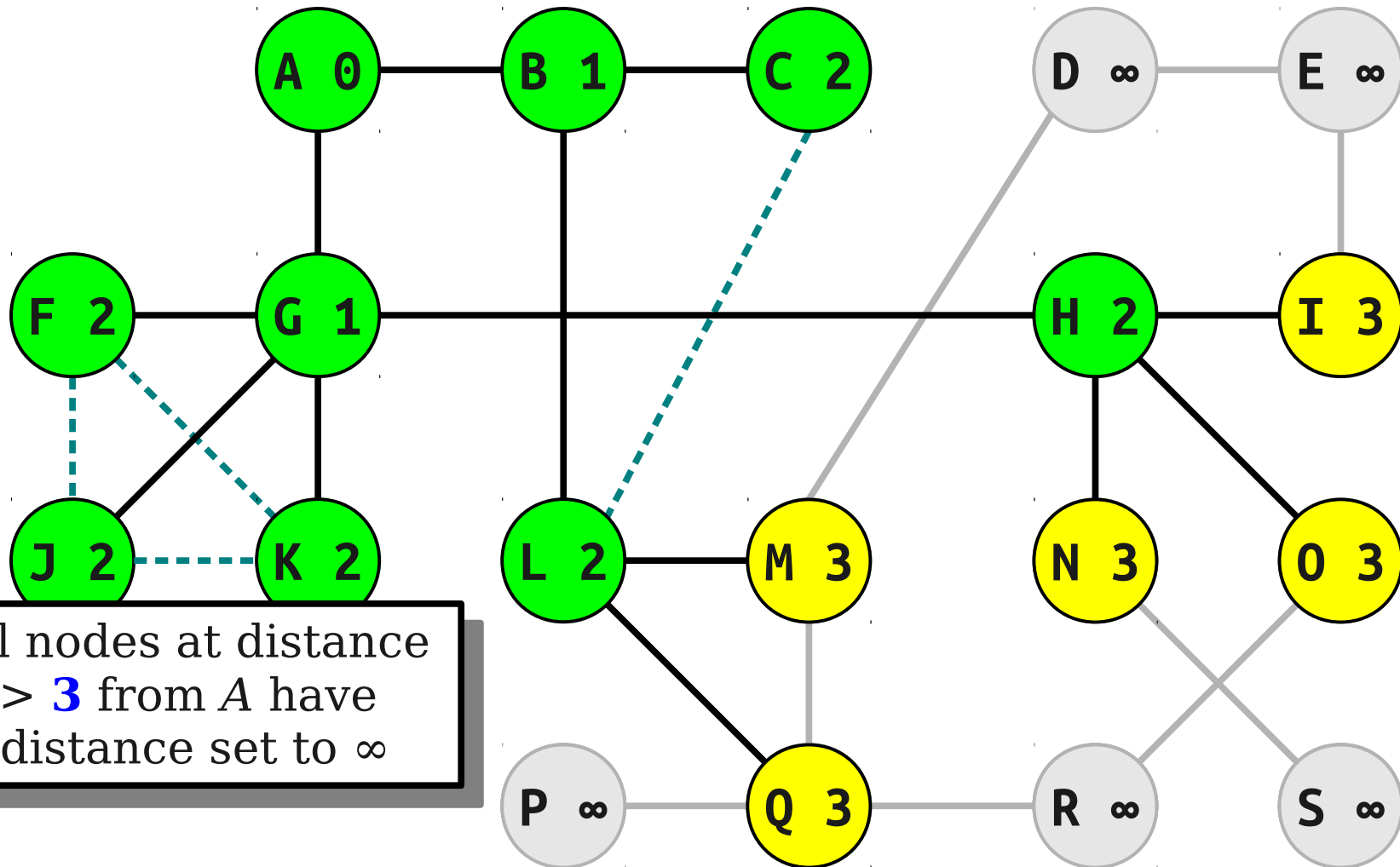
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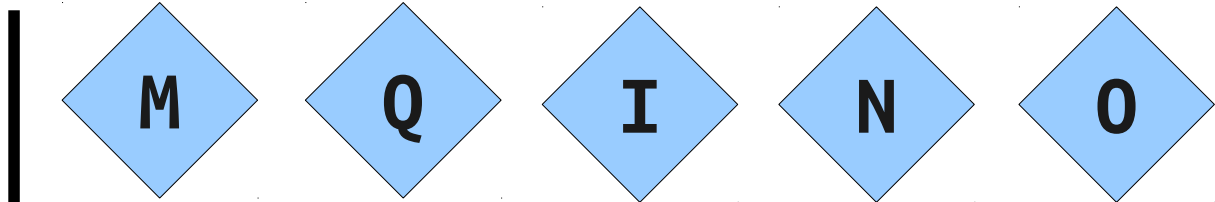
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All nodes at distance  $> 3$  from A have distance set to  $\infty$



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- (2) All nodes  $v$  where  $d(s, v) \leq n$  have  $\text{dist}[v] = d(s, v)$ .
- (3) All nodes  $v$  where  $d(s, v) > n$  have  $\text{dist}[v] = \infty$

Let  $k$  be the maximum finite distance of any node from node  $s$ . Note the following:

- Any node  $v$  where  $d(s, v)$  is finite satisfies  $d(s, v) \leq k$ , and any node  $v$  where  $d(s, v) > k$  satisfies  $d(s, v) = \infty$ . This follows from the fact that we picked the maximum possible finite  $k$ .
- There must be nodes at distances  $0, 1, 2, \dots, k$  from  $s$ . A simple inductive argument using property (1) shows that there will be exactly  $k + 1$  rounds, corresponding to distances  $0, 1, \dots, k$ .

So consider  $\text{dist}[v]$  for any node  $v$  after the algorithm terminates (that is, after  $k+1$  rounds). If  $d(s, v)$  is finite, then  $d(s, v) \leq k \leq k+1$ , and so by (1) we have  $\text{dist}[v] = d(s, v)$ . If  $d(s, v) = \infty$ , then  $d(s, v) > k + 1$ , so by (2) we have  $\text{dist}[v] = \infty$ . Thus  $d(s, v) = \text{dist}[v]$  for all  $v \in V$ , as required.

*Theorem:* Breadth-first search always terminates with  $\text{dist}[v] = d(s, v)$  for all  $v \in V$ .

*Proof:* Define “round  $n$ ” of BFS to be an instance where at the start of the loop, all nodes  $v$  in the queue satisfy  $\text{dist}[v] = n$ . We will prove in an lemma the following are always true after the first  $n$  rounds:

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*Lemma:* After  $n$  rounds, the following hold:

- (1) For any node  $v$ ,  $d(s, v) = n$  iff  $v$  is in the queue.
- (2) All nodes  $v$  where  $d(s, v) \leq n$  have  $\text{dist}[v] = d(s, v)$ .
- (3) All nodes  $v$  where  $d(s, v) > n$  have  $\text{dist}[v] = \infty$

*Proof:* By induction  $n$ . After 0 rounds,  $\text{dist}[s] = 0$ ,  $\text{dist}[v] = \infty$  for any  $v \neq s$ , and the queue holds only  $s$ . Since  $s$  is the only node at distance 0, (1) - (3) hold.

For the inductive step, assume for some  $n$  that (1) - (3) hold after  $n$  rounds. We will prove (1) - (3) hold after  $n + 1$  rounds. We need to show the following:

- (a) For any node  $v$ ,  $d(s, v) = n + 1$  iff  $v$  is in the queue.
- (b) All nodes  $v$  where  $d(s, v) \leq n + 1$  have  $\text{dist}[v] = d(s, v)$ .
- (c) All nodes  $v$  where  $d(s, v) > n + 1$  have  $\text{dist}[v] = \infty$

To prove (a), note that at the end of round  $n$ , all nodes of distance  $n$  will have been dequeued, so we need to show all nodes  $v$  where  $d(s, v) = n + 1$  are enqueued and nothing else is. Note that if a node  $u$  is enqueued in round  $n + 1$ , then at the start of round  $n + 1$   $\text{dist}[u] = \infty$  (so by (2) and (3), its distance is at least  $n + 1$ ) and  $u$  must have been adjacent to a node  $v$  in the queue (by (1),  $d(s, v) = n$ ). Thus there is a path of length  $n + 1$  to  $u$  (take the path of length  $n$  to  $v$ , then follow the edge to  $u$ ), and there is no shorter path, so this is the shortest path to  $u$ . Thus,  $d(s, u) = n + 1$ . Also note that if a node  $u$  satisfies  $d(s, u) = n + 1$ , then by (3) at the start of round  $n + 1$  it must have  $\text{dist}[u] = \infty$ . Also, it must be adjacent to some node at distance  $n$ , which by (1) must be in the queue at the start of the round. Thus at the end of round  $n + 1$ ,  $u$  will be enqueued and  $\text{dist}[u]$  set to  $n + 1$ .

By our above argument, we know that (a) must hold. Since we didn't change any  $\text{dist}$  values for nodes at distance  $n$  or less, and we set  $\text{dist}$  values for all enqueued nodes to  $n + 1$ , (b) holds. Finally, since we only changed labels for nodes at distance  $n + 1$ , (c) holds as well. This completes the induction. ■



Question 1: How do we prove this always finds the right distances?

Question 2: How *efficiently* does this find the right distances?

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Question 2: How *efficiently* does this find the right distances?

# Graph Terminology

- When analyzing algorithms on a graph, there are (usually) two parameters we care about:
  - The number of nodes, denoted  **$n$** . ( $n = |V|$ )
  - The number of edges, denoted  **$m$** . ( $m = |E|$ )
- Note that  $m = O(n^2)$ . (*Why?*)
- A graph is called **dense** if  $m = \Theta(n^2)$ . A graph is called **sparse** if it is not dense.

```
procedure breadthFirstSearch( $s$ ,  $G$ ):  
  let  $q$  be a new queue.  
  for each node  $v$  in  $G$ :  
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   $\text{dist}[s] = 0$   
  enqueue( $s$ ,  $q$ )  
  
  while  $q$  is not empty:  
    let  $v = \text{dequeue}(q)$   
    for each neighbor  $u$  of  $v$ :  
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**O(1)**

**O(n)**

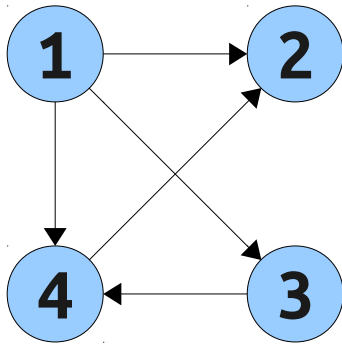
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How are our graphs represented?

# Adjacency Matrices

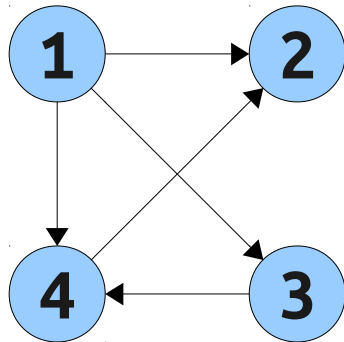
- An **adjacency matrix** is a representation of a graph as an  $n \times n$  matrix  $M$  of 0s and 1s, where
  - $M_{uv} = 1$  if  $(u, v) \in E$ .
  - $M_{uv} = 0$  otherwise.



$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

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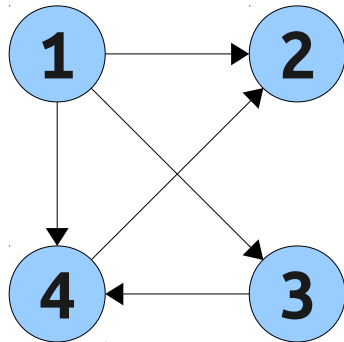
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- Memory usage:



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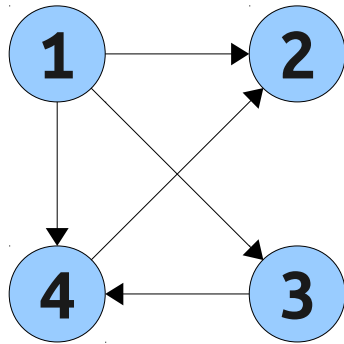


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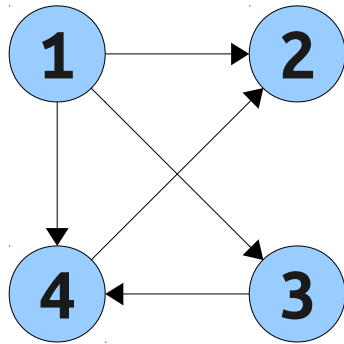


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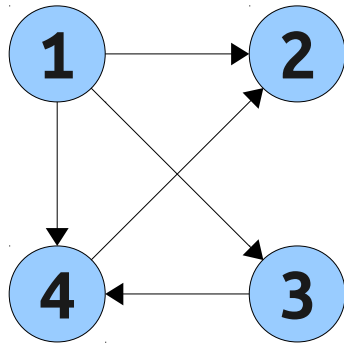


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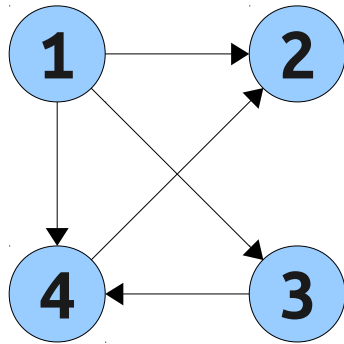


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- Time to check if an edge exists:  $O(1)$
- Time to find all outgoing edges for a node:  $\Theta(n)$

**O(1)**

**O(n)**

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**procedure** breadthFirstSearch( $s$ ,  $G$ ):

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**let**  $q$  be a new queue.

**$O(n)$**

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**$O(1)$**

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  enqueue( $s$ ,  $q$ )

**$O(n^2)$**

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$+ O(n^2)$

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**enqueue**( $u, q$ )

$\Theta(n)$

$O(n^2)$

Why isn't the runtime  $\Theta(n^2)$ ?

# Linear Time on Graphs

- With an adjacency matrix, BFS runs in time  $O(n^2)$ . Is that efficient?
- In a graph with  $n$  nodes and  $m$  edges, we say that an algorithm runs in **linear time** iff the algorithm runs in time  $O(m + n)$ .
  - This is linear in the number of “pieces” of the graph, which is the number of nodes plus the number of edges.
- On a dense graph, this implementation of BFS runs in linear time:

$$O(n^2) = O(n^2 + n) = O(m + n)$$

- On sparser graphs (say,  $m = O(n)$ ), though, this is not linear time:

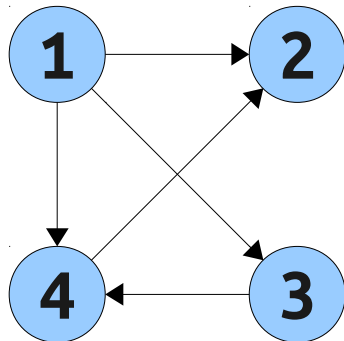
$$O(n^2) \neq O(n) = O(m + n)$$

# The Issue

- Our algorithm is slow because this step always takes  $\Theta(n)$  time:

**for** each neighbor  $u$  of  $v$ :

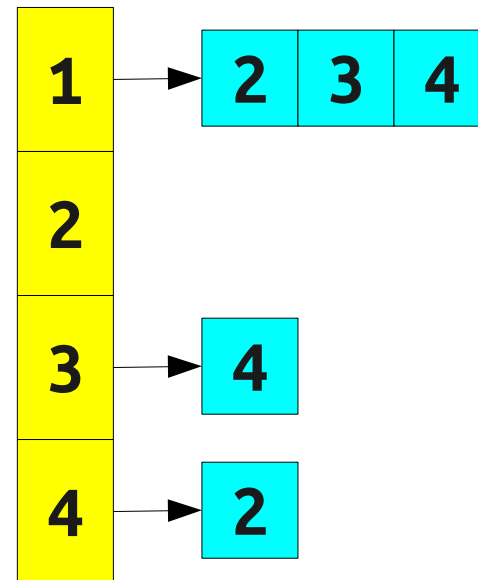
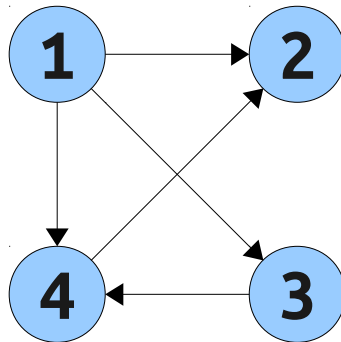
- Can we refine our data structure for storing the graph so that we can easily find all edges incident to a node?



$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

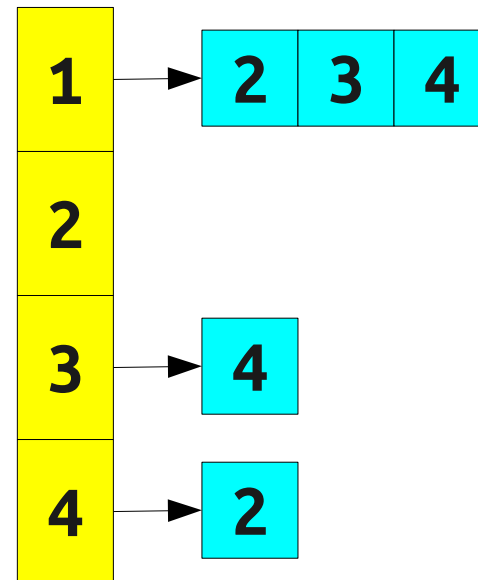
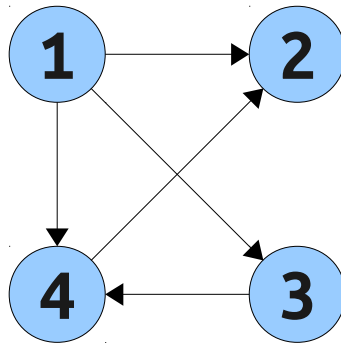
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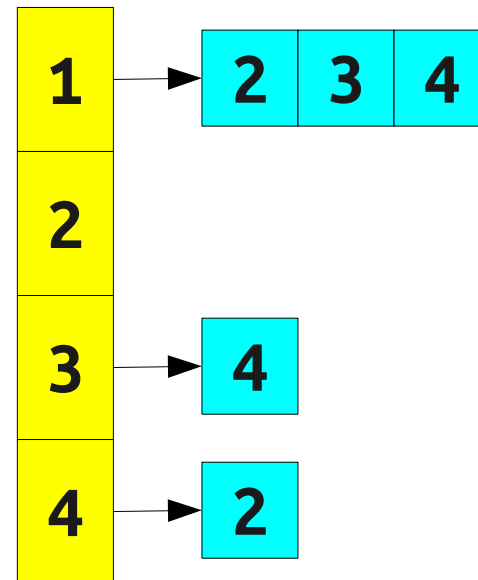
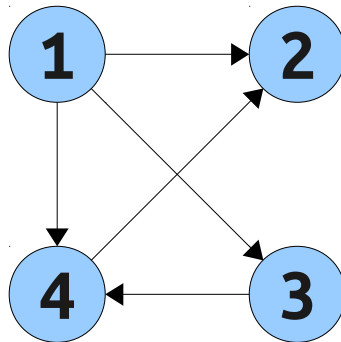


- Memory usage:



# Adjacency Lists

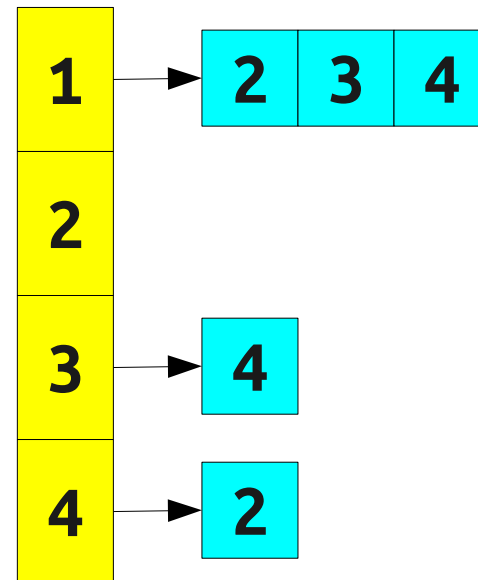
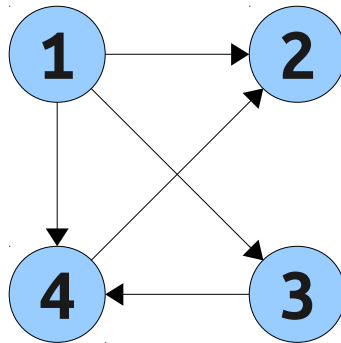
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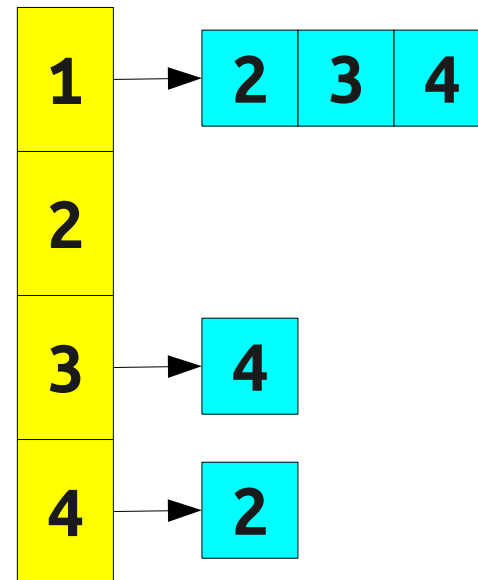
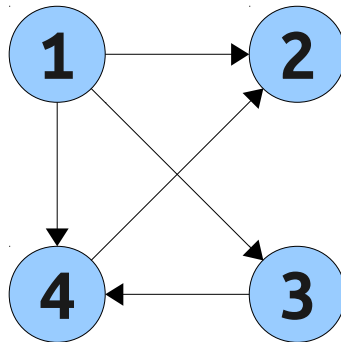
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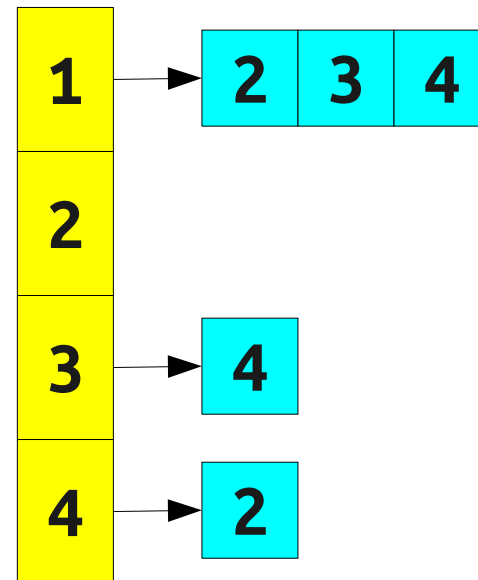
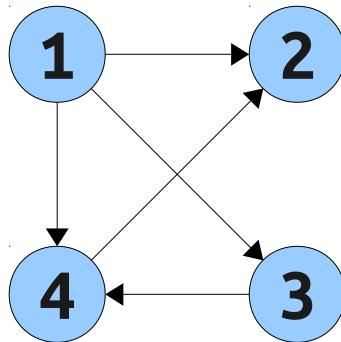
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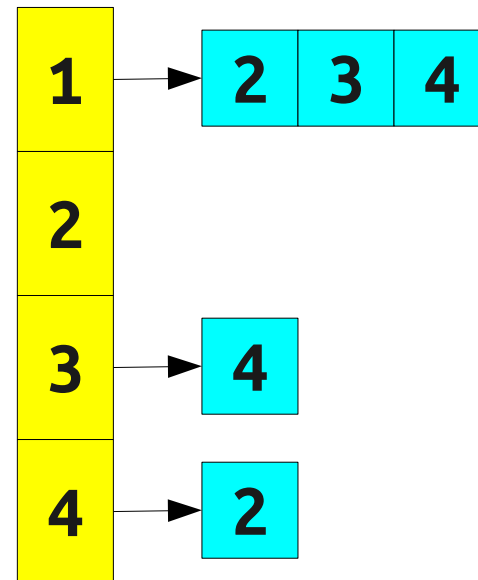
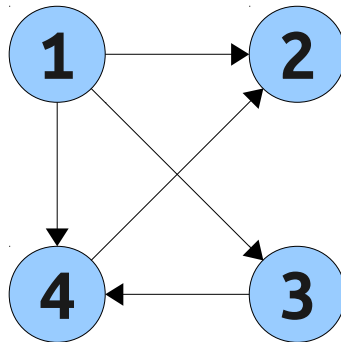
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**O(n)**

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procedure breadthFirstSearch(s, G):  
  let q be a new queue.  
  for each node v in G:  
    dist[v] = ∞  
  
  dist[s] = 0  
  enqueue(s, q)  
  
  while q is not empty:  
    let v = dequeue(q)  
    for each neighbor u of v:  
      if dist[u] = ∞:  
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**$O(n^2)$**

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# A Better Analysis

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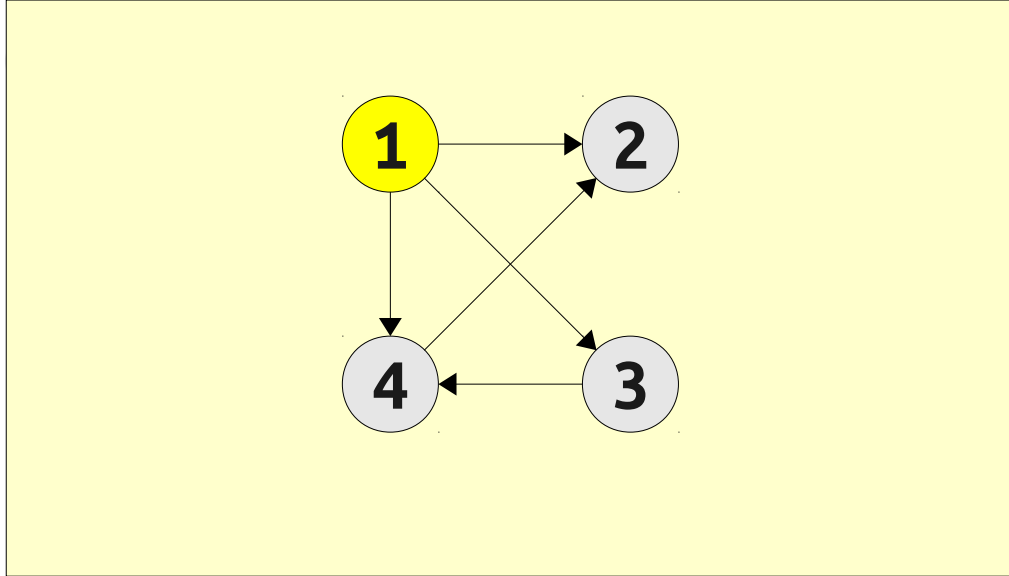
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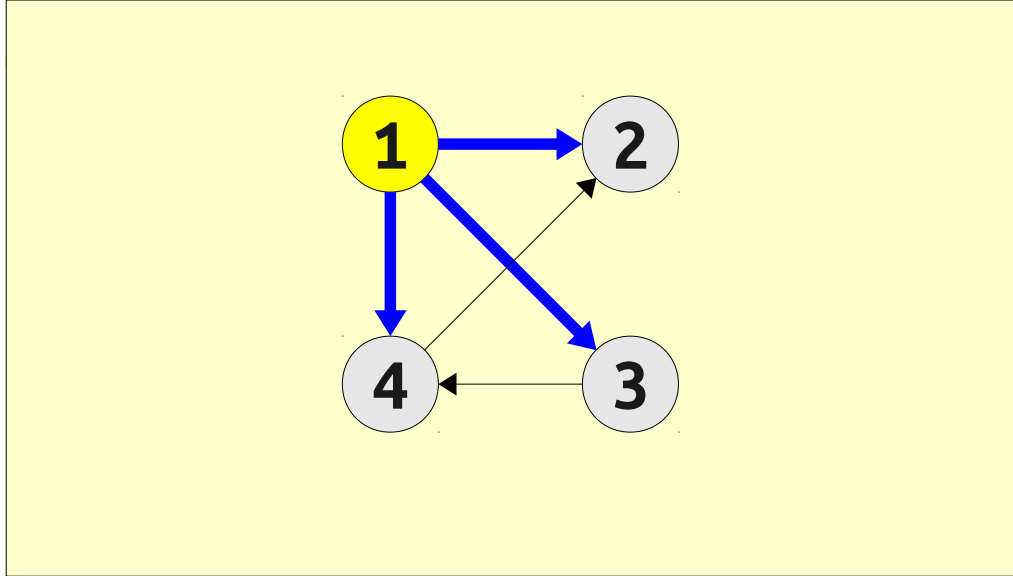
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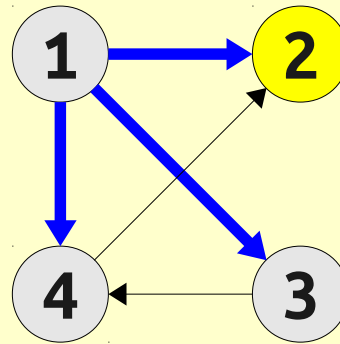
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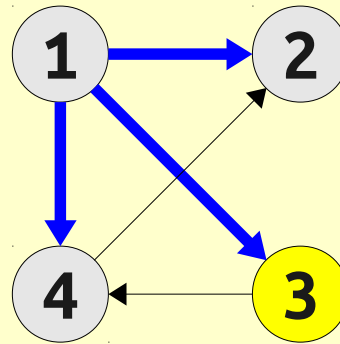
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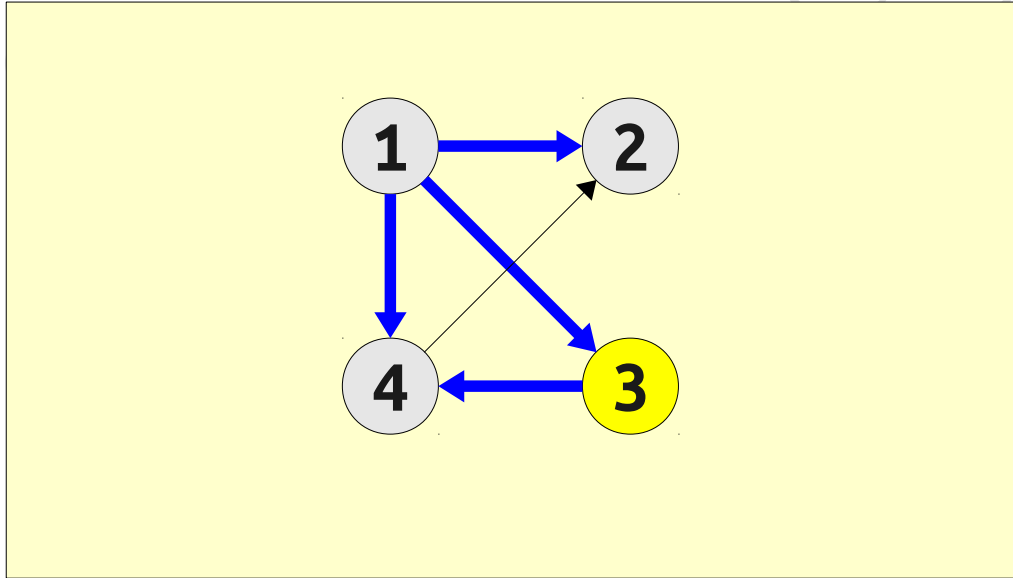
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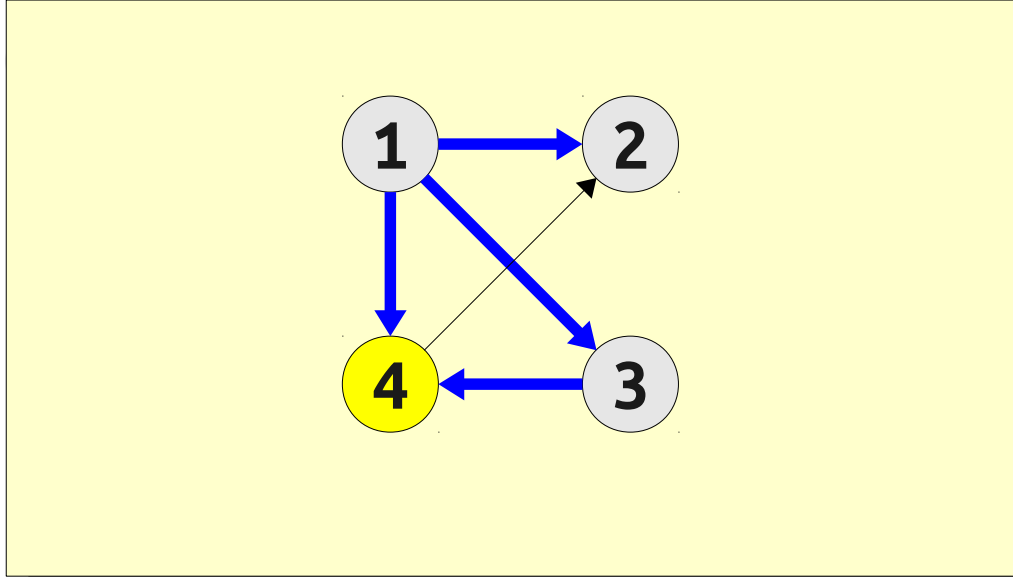
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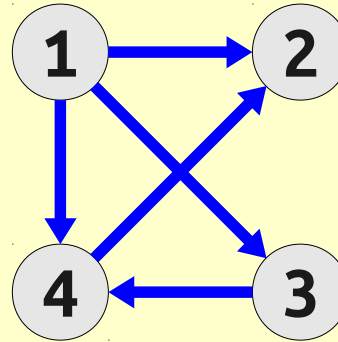
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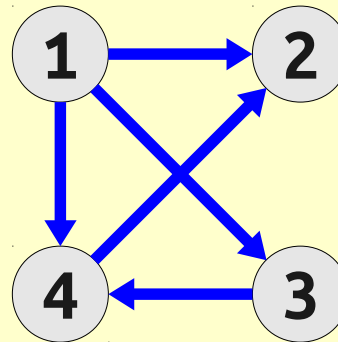
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**$O(m + n)$**



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enqueue( $u$ ,  $q$ )

$O(m + n)$

# A Better Analysis

- Using adjacency lists, BFS runs in time  $O(m + n)$ .
  - This is linear time!
- **Key Idea:** Do a more precise accounting of the work done by an algorithm.
  - Determine how much work is done *across all iterations* to determine total work.
  - Don't just find worst-case runtime and multiply by number of iterations.
- Going forward, we will use adjacency lists rather than adjacency matrices as our graph representation unless stated otherwise.

# Next Time

- Dijkstra's Algorithm
- Depth-First Search
- Directed Acyclic Graphs