

# Complexity Theory

## Part Two

Recap from Last Time

# The Complexity Class **P**

- The ***complexity class P*** (for ***p***olynomial time) contains all problems that can be solved in polynomial time.
- Formally:

$$\mathbf{P} = \{ L \mid \text{There is a polynomial-time decider for } L \}$$

# The Complexity Class **NP**

- The complexity class **NP** (***nondeterministic polynomial time***) contains all problems that can be verified in polynomial time.

- Formally:

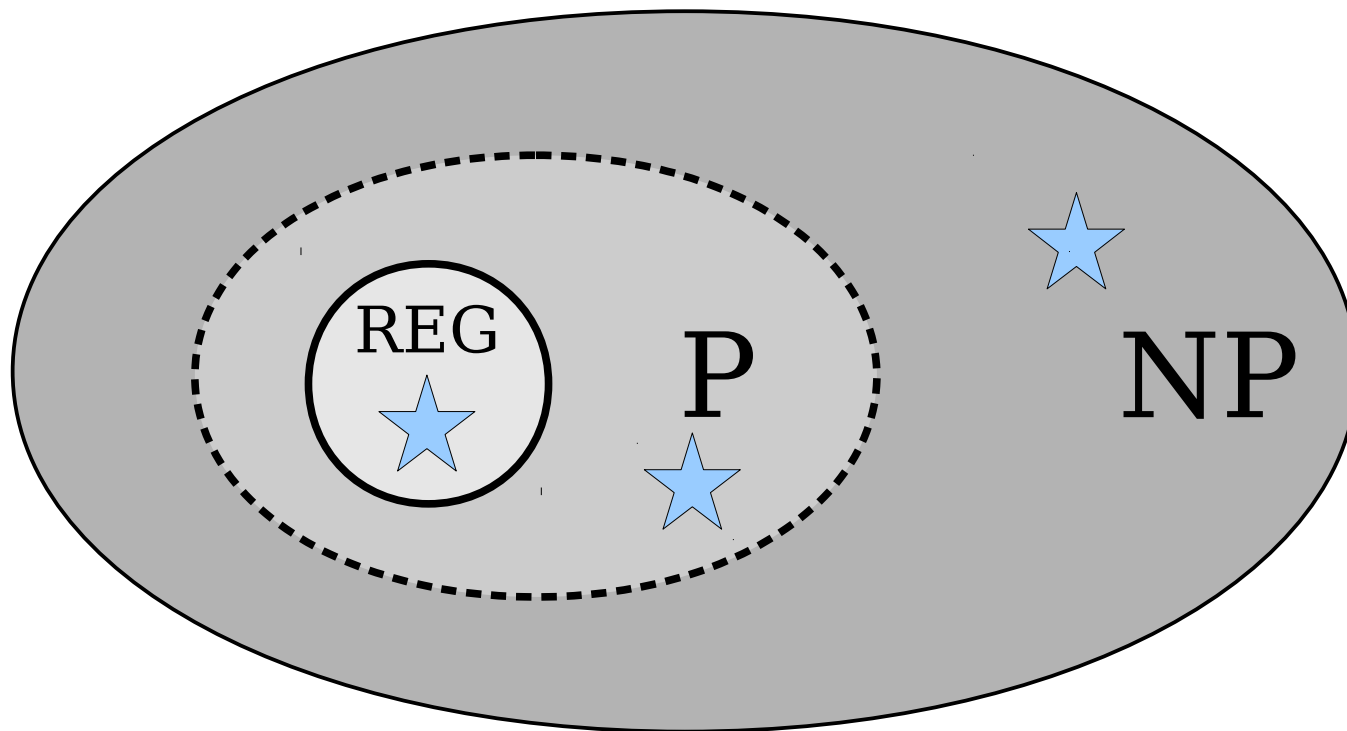
$$\mathbf{NP} = \{ L \mid \text{There is a polynomial-time verifier for } L \}$$

- This means a verifier  $V$ 's runtime is a polynomial in  $|w|$  (that is,  $V$ 's runtime is  $O(|w|^k)$  for some integer  $k$ ).

So how *are* we going to  
reason about **P** and **NP**?

New Stuff!

# A Challenge



Problems in **NP** vary widely in their difficulty, even if **P = NP**.

How can we rank the relative difficulties of problems?



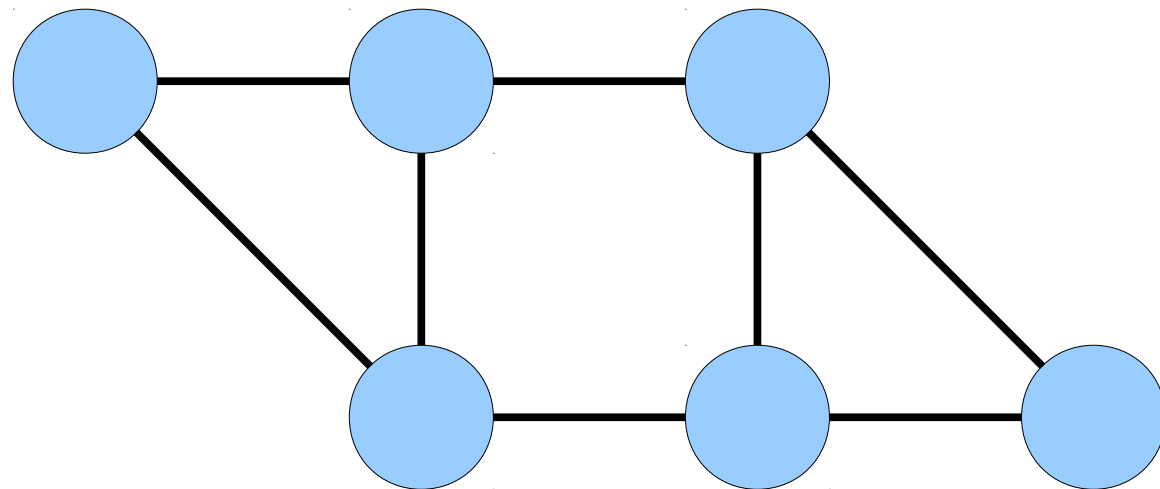
Reducibility

# Maximum Matching

- Given an undirected graph  $G$ , a **matching** in  $G$  is a set of edges such that no two edges share an endpoint.
- A **maximum matching** is a matching with the largest number of edges.

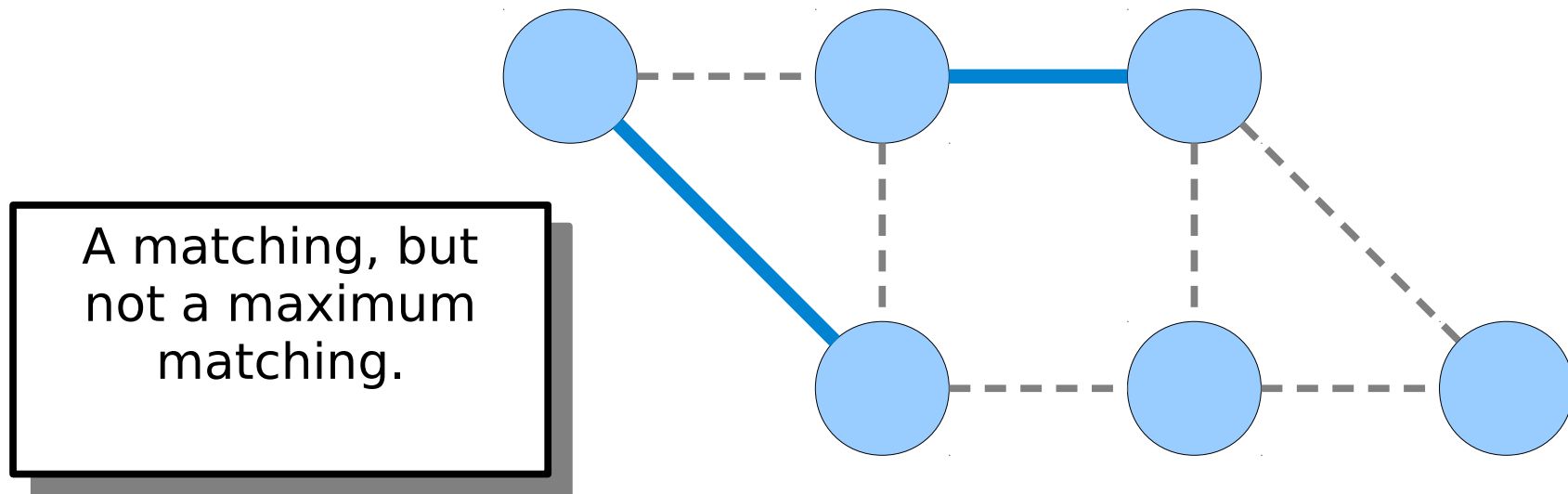
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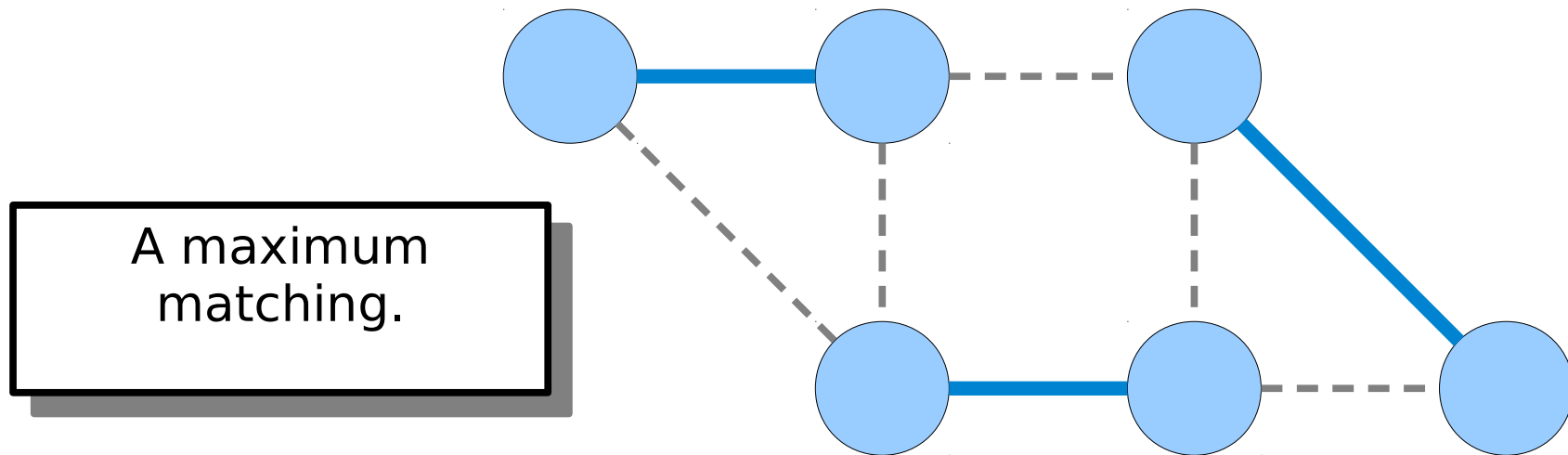
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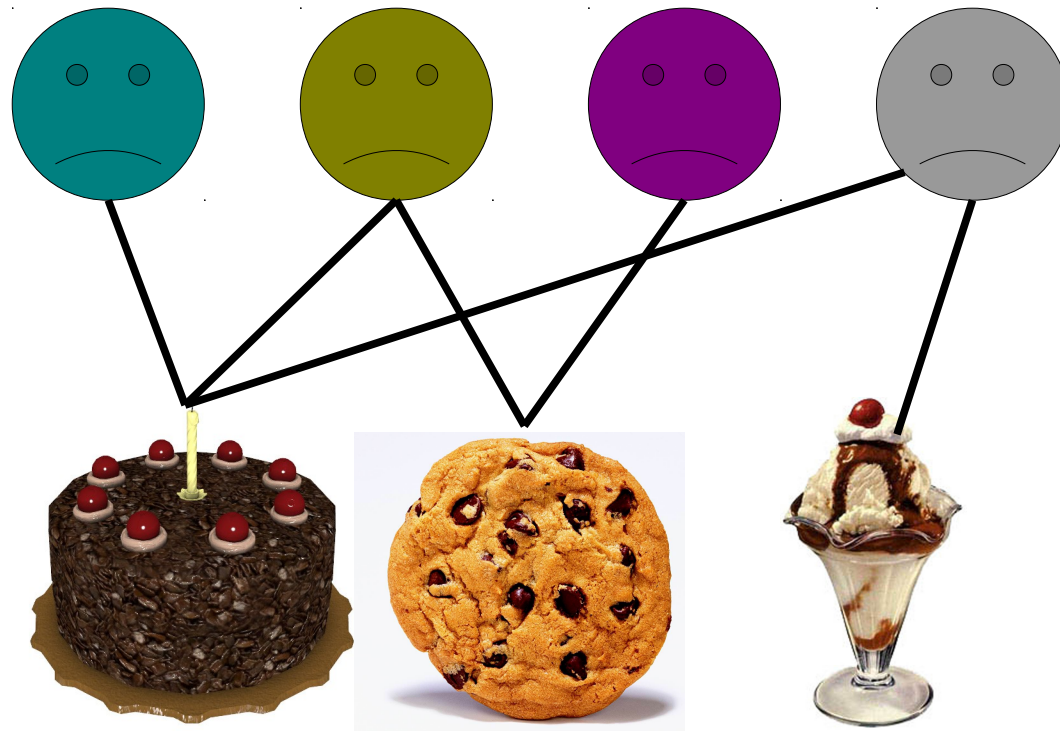
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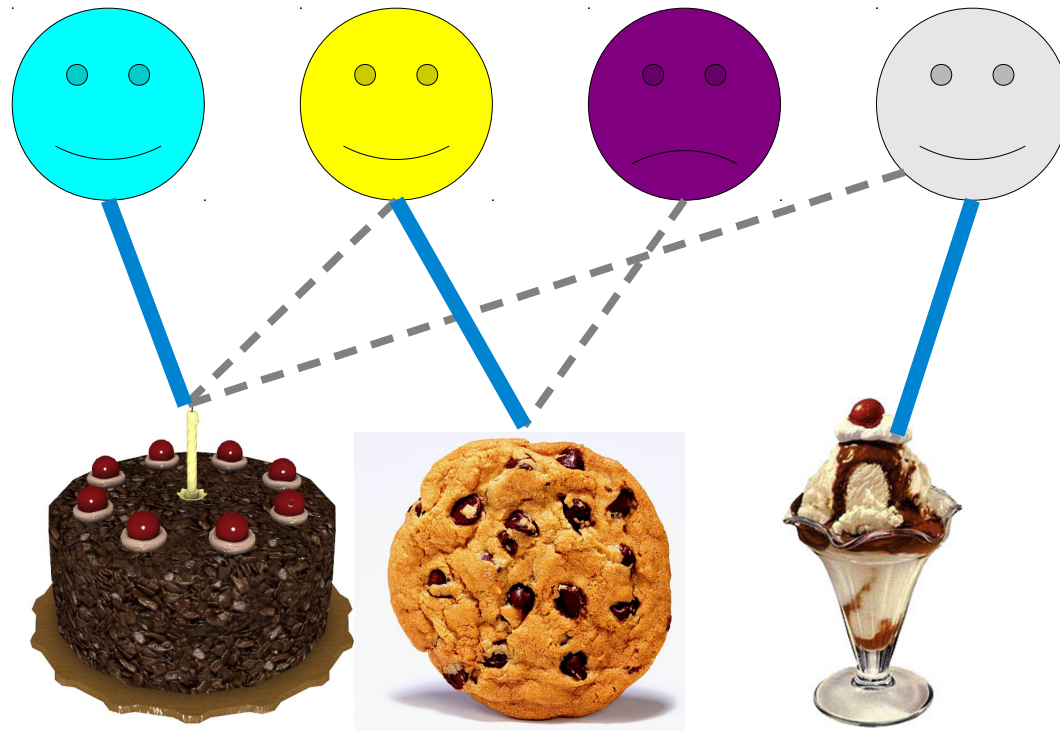
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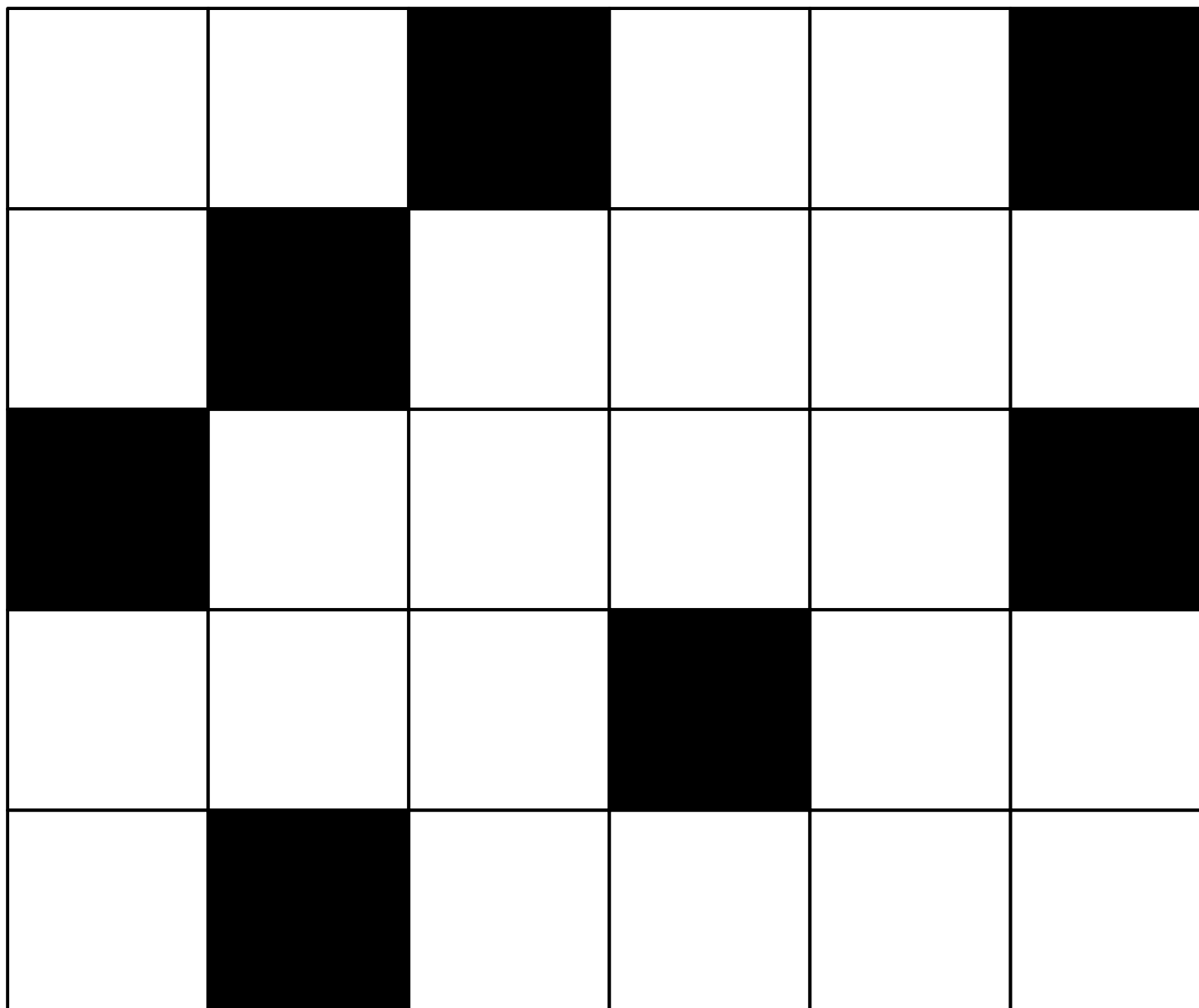


# Maximum Matching

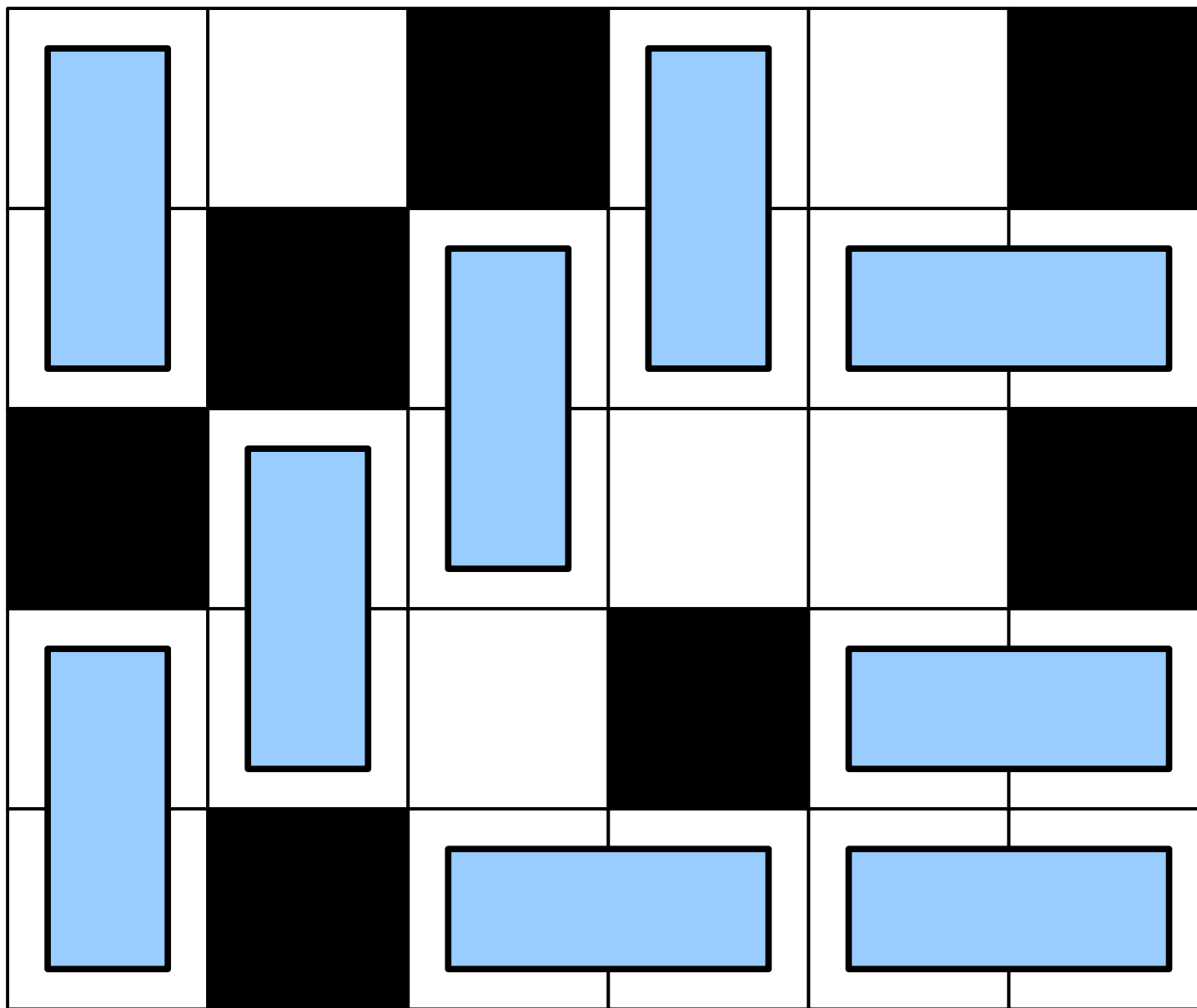
- Jack Edmonds' paper “Paths, Trees, and Flowers” gives a polynomial-time algorithm for finding maximum matchings.
  - (This is the same Edmonds as in “Cobham-Edmonds Thesis.”)
- Using this fact, what other problems can we solve?



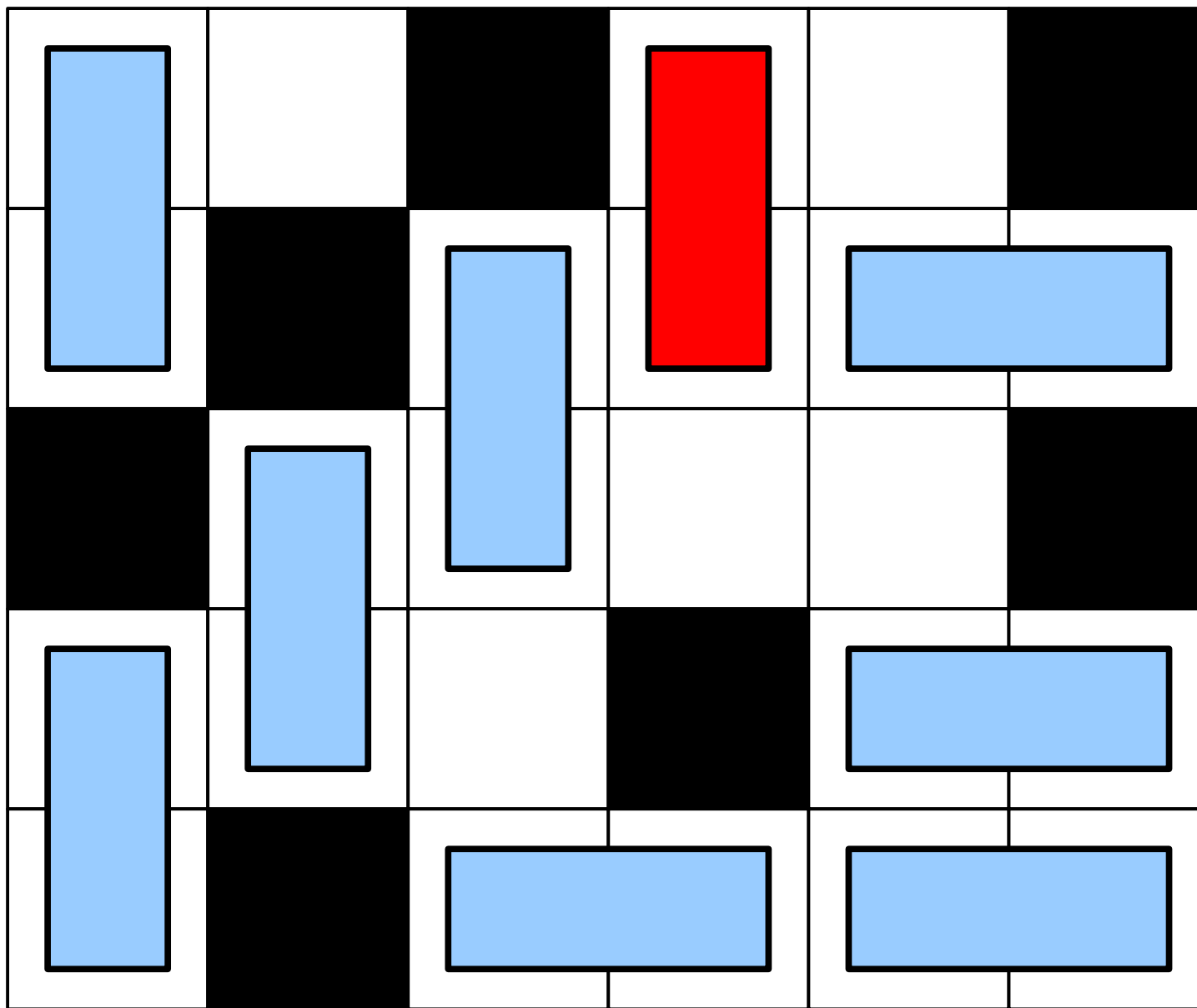
# Domino Tiling



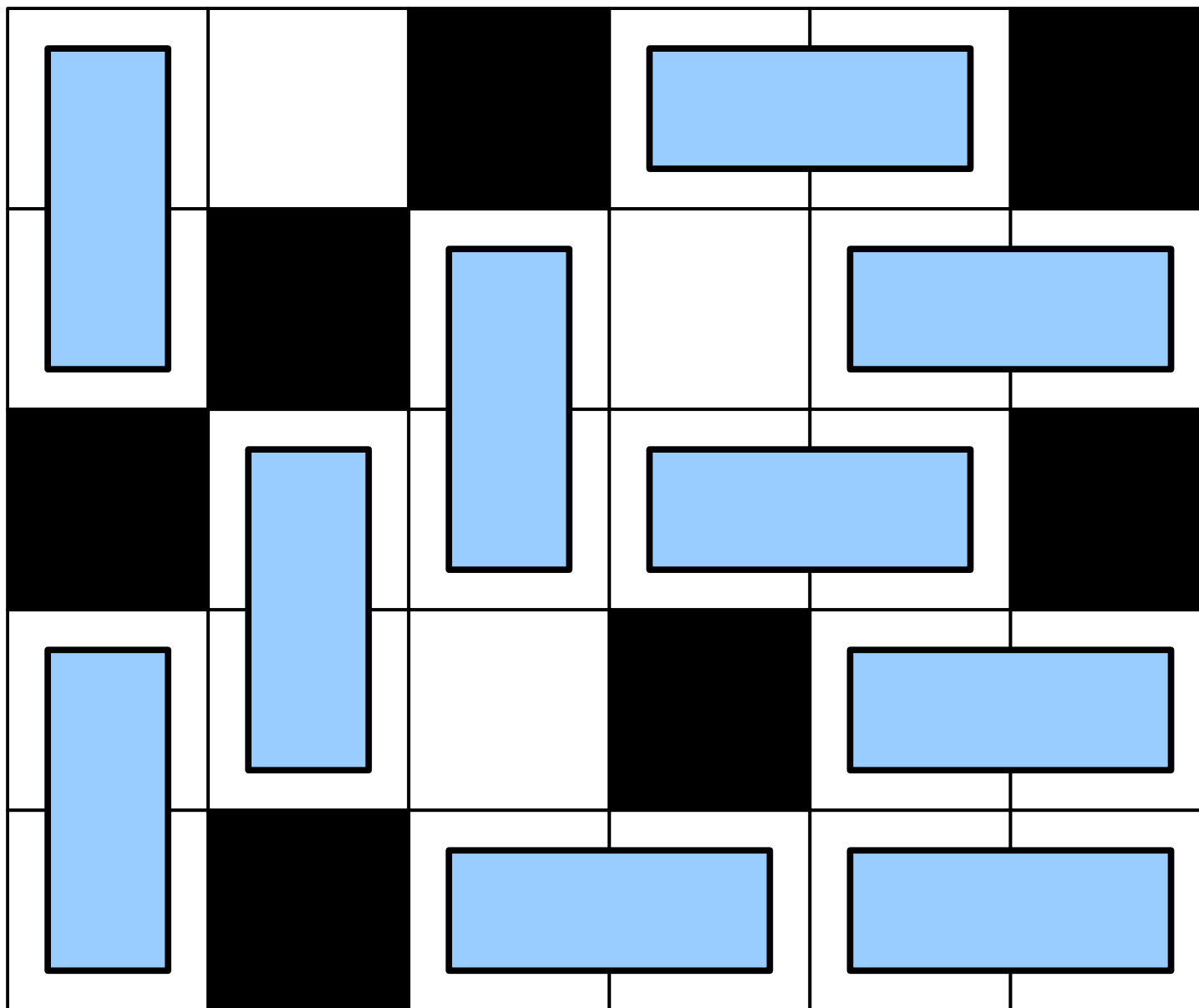
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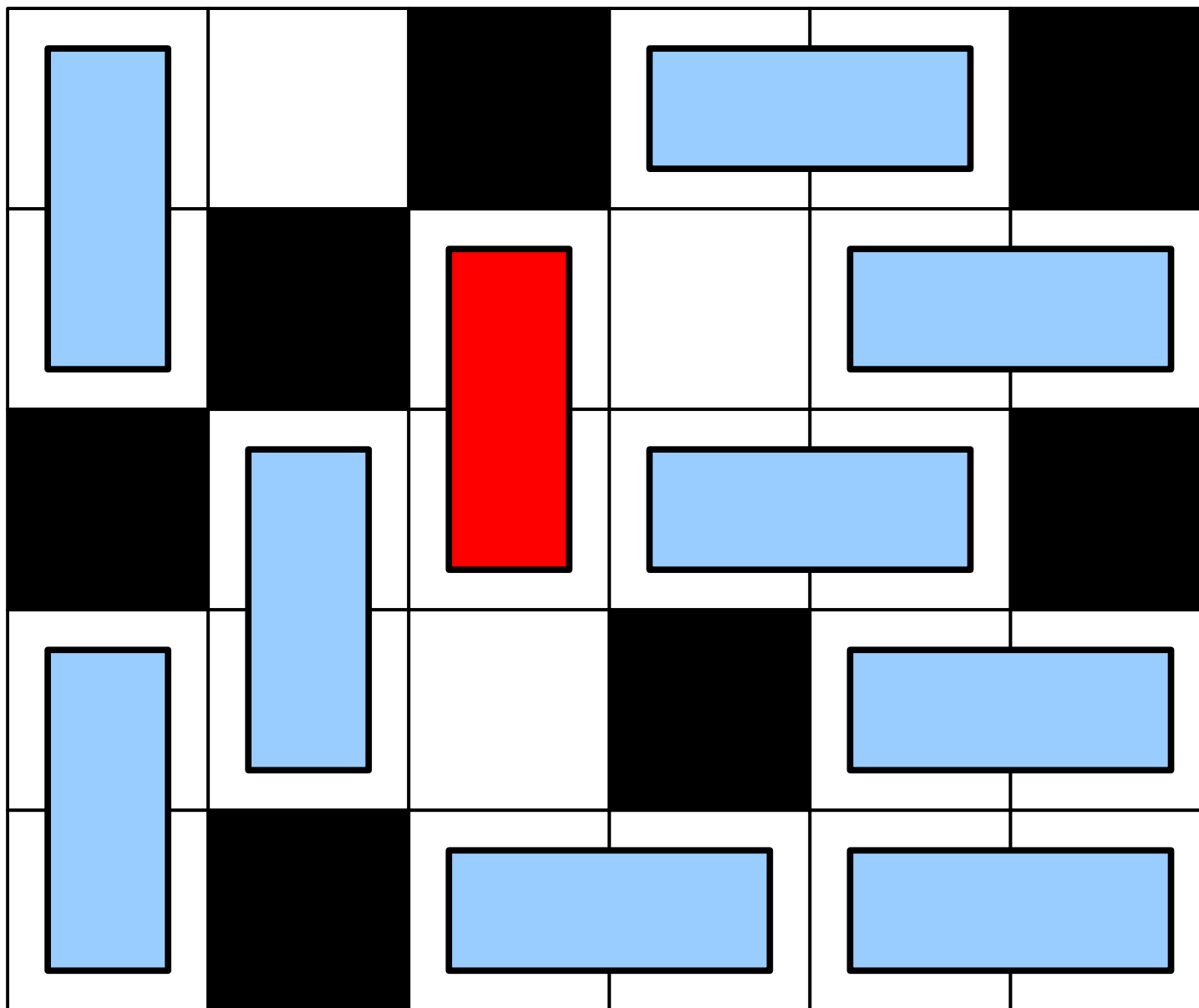
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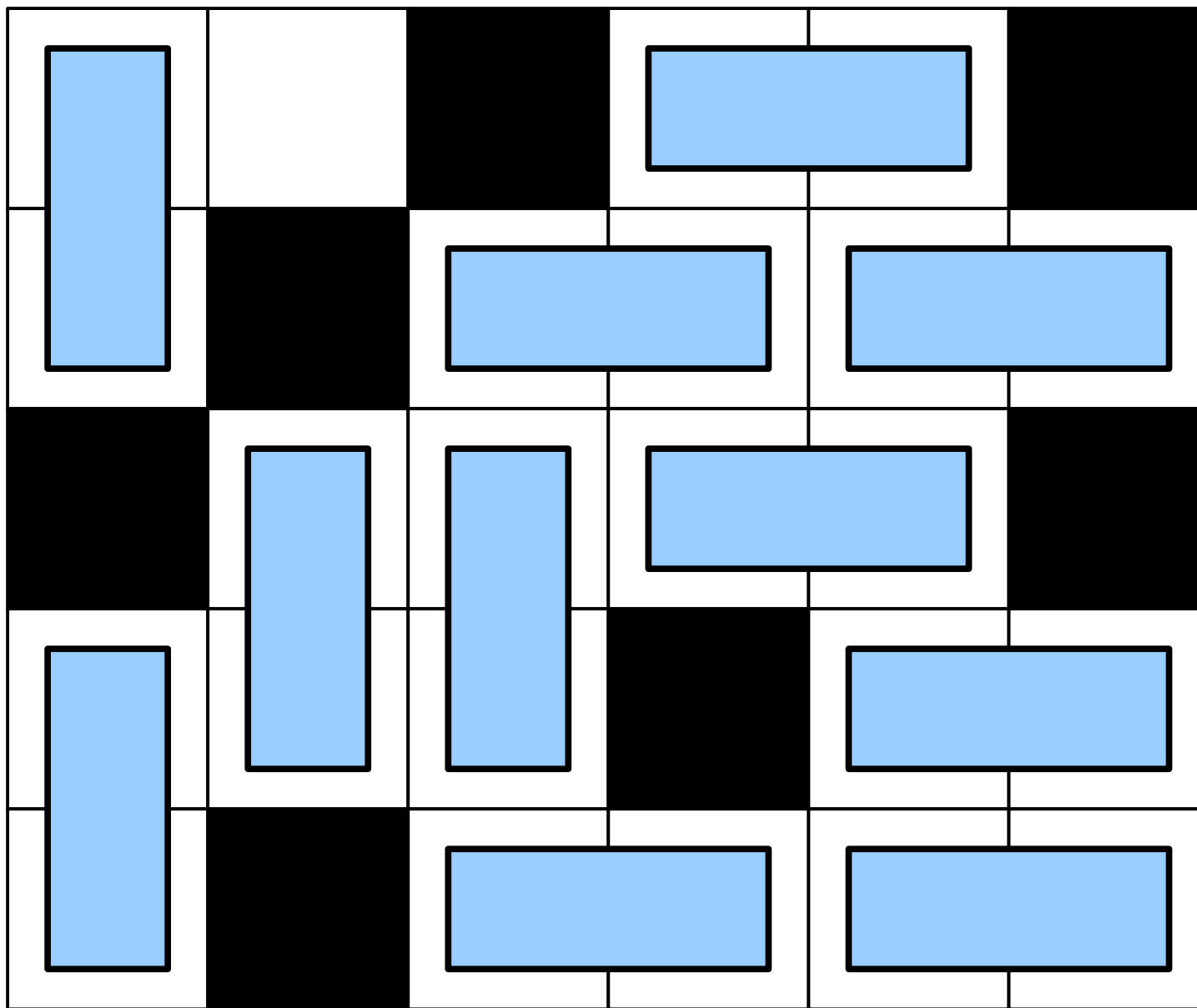
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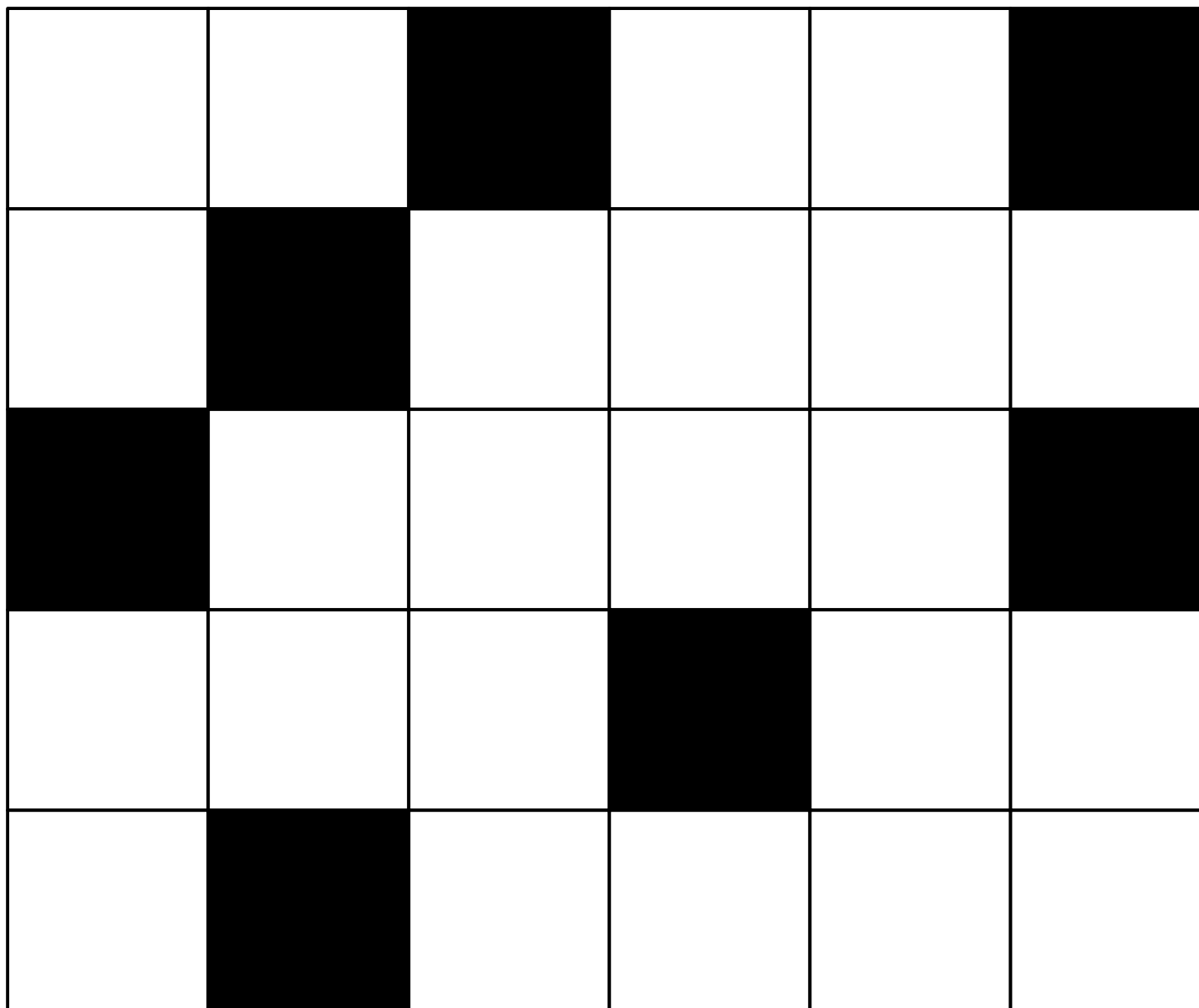
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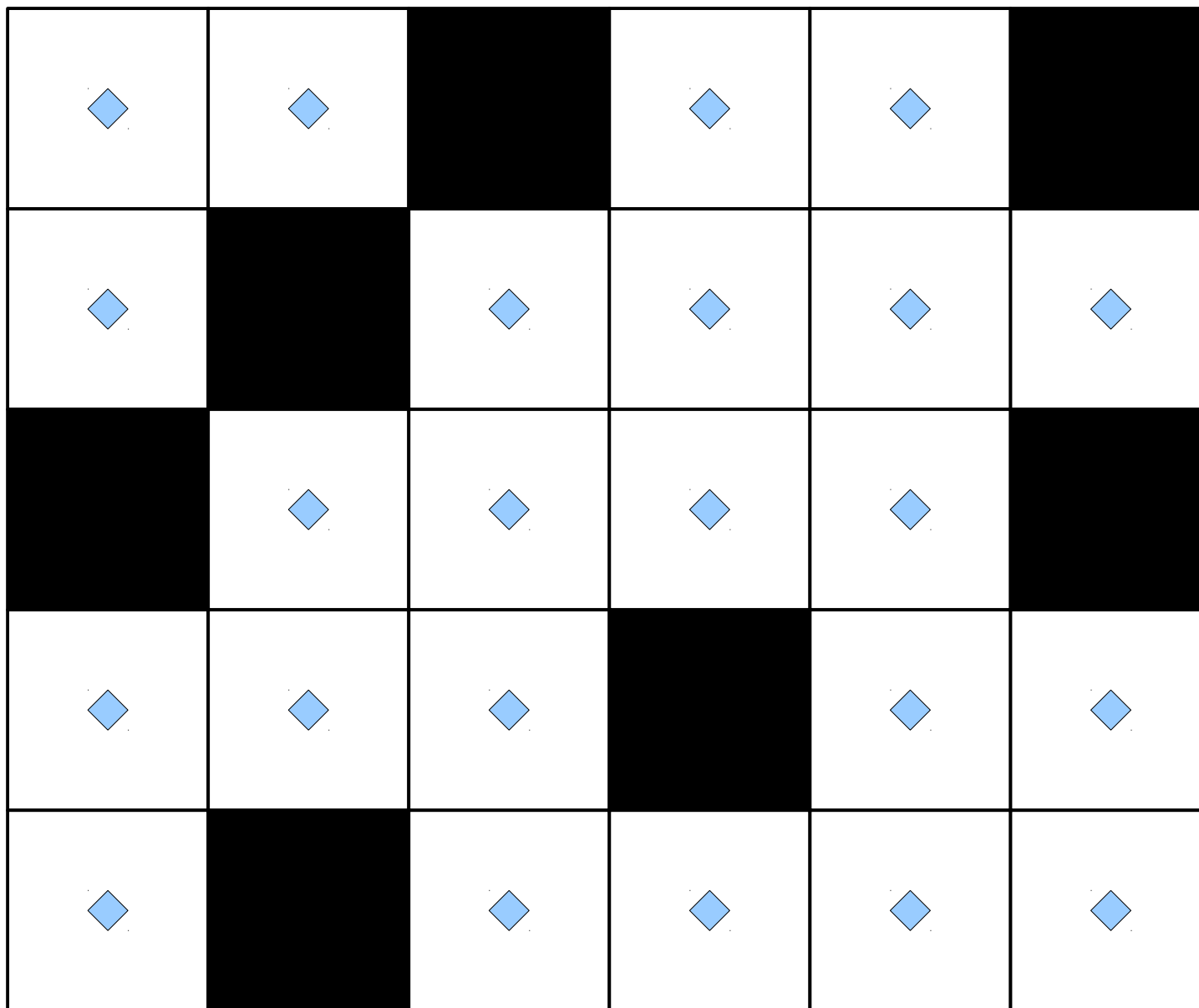
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# Solving Domino Tiling

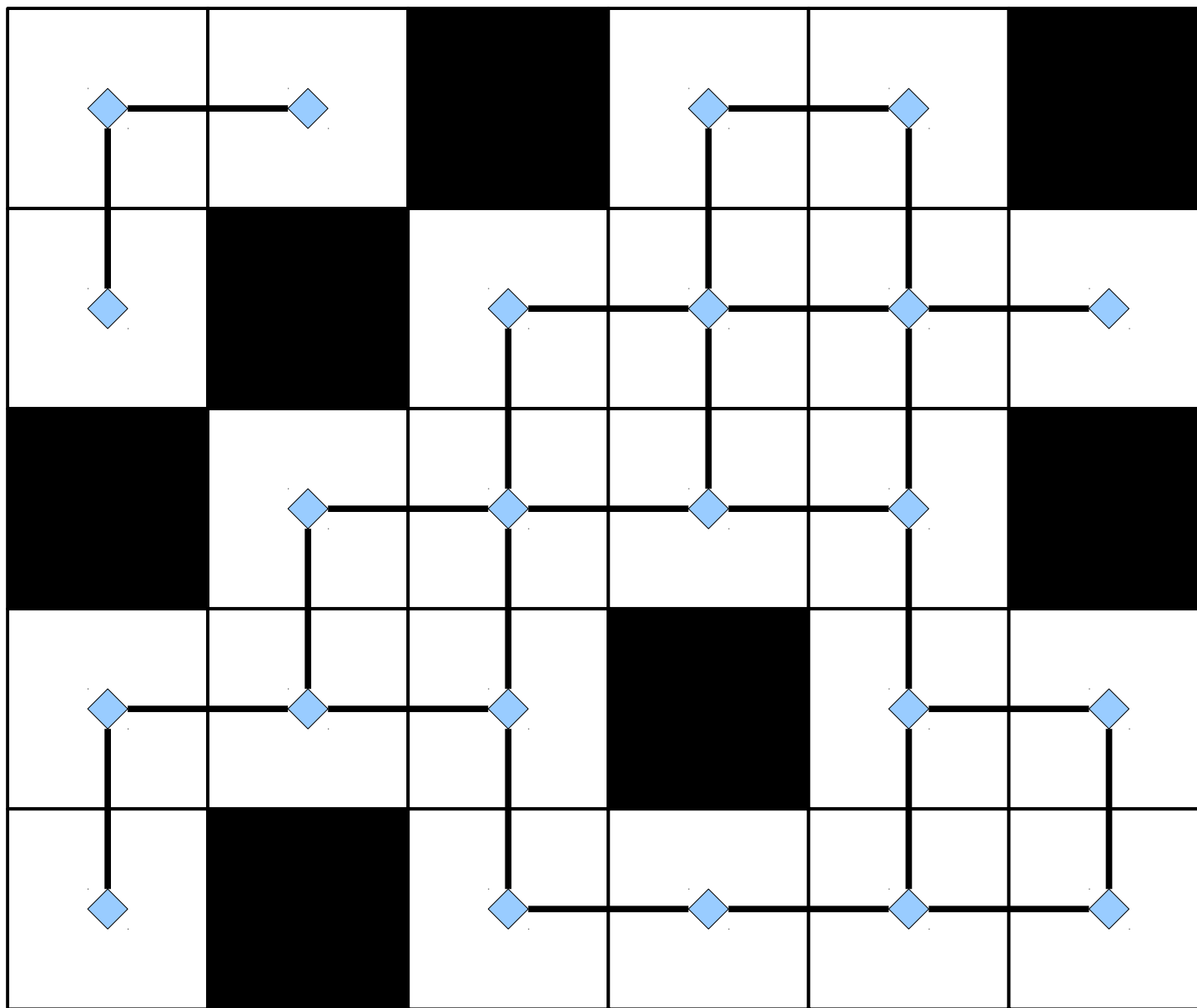


# Solving Domino Tiling

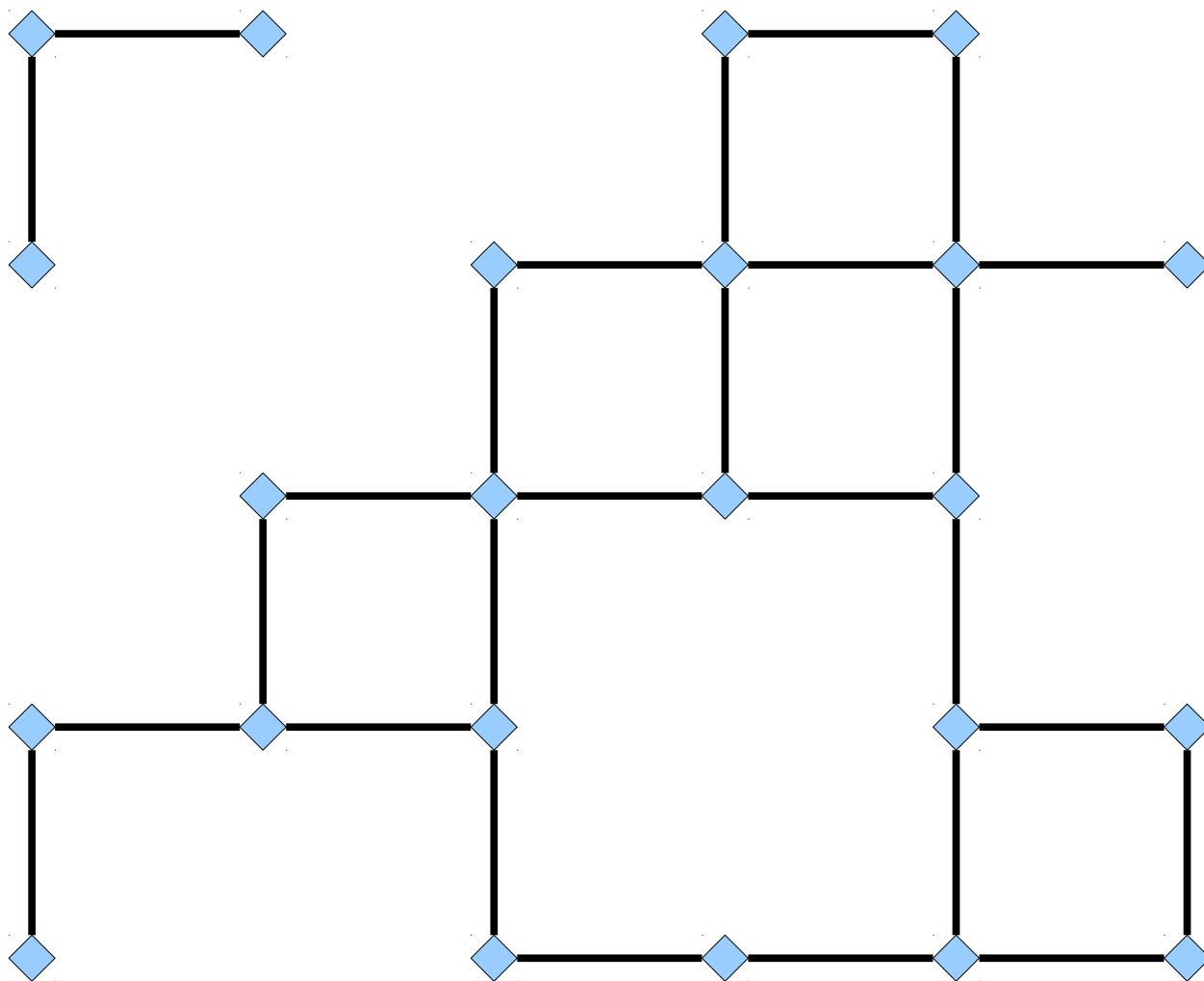




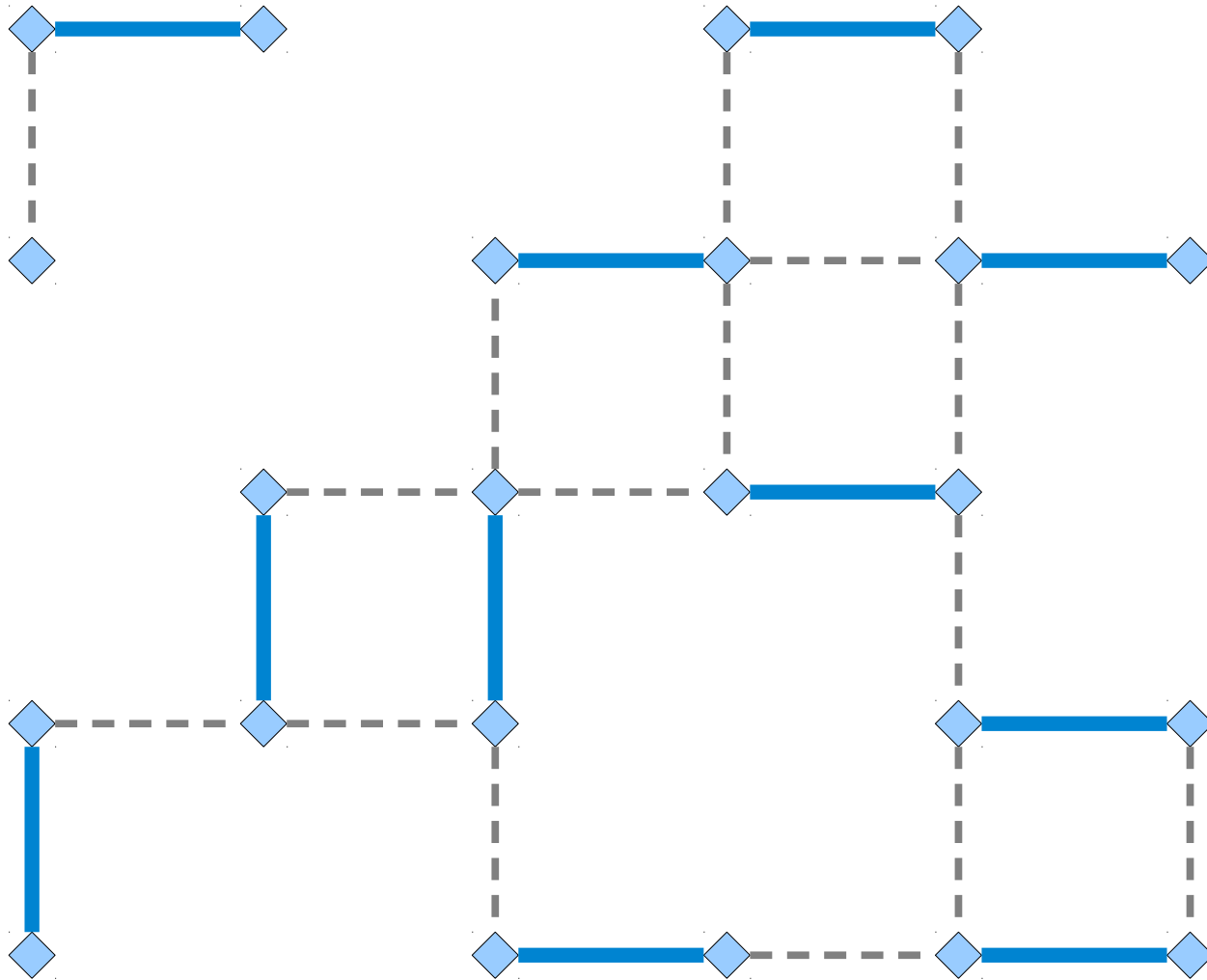
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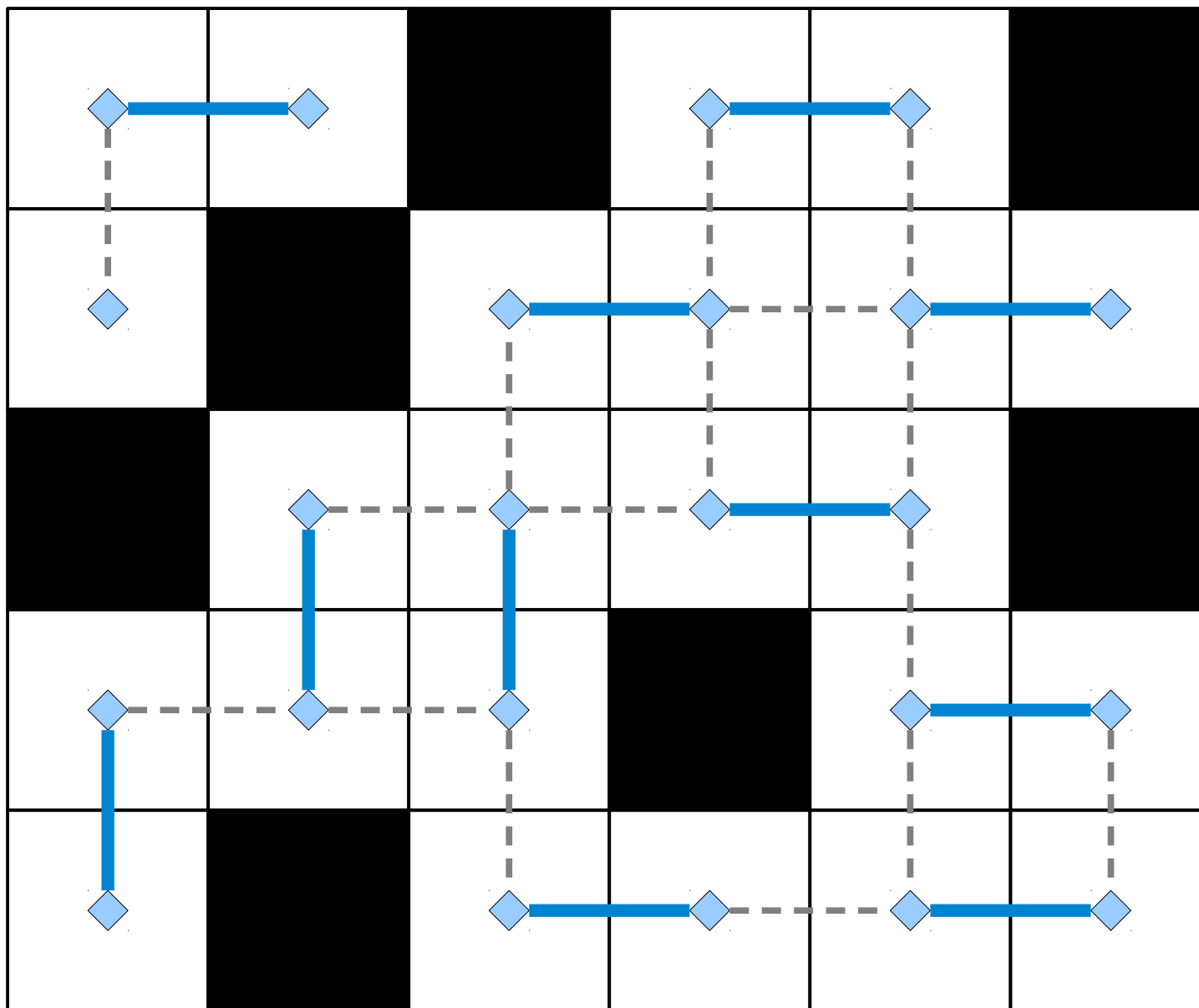
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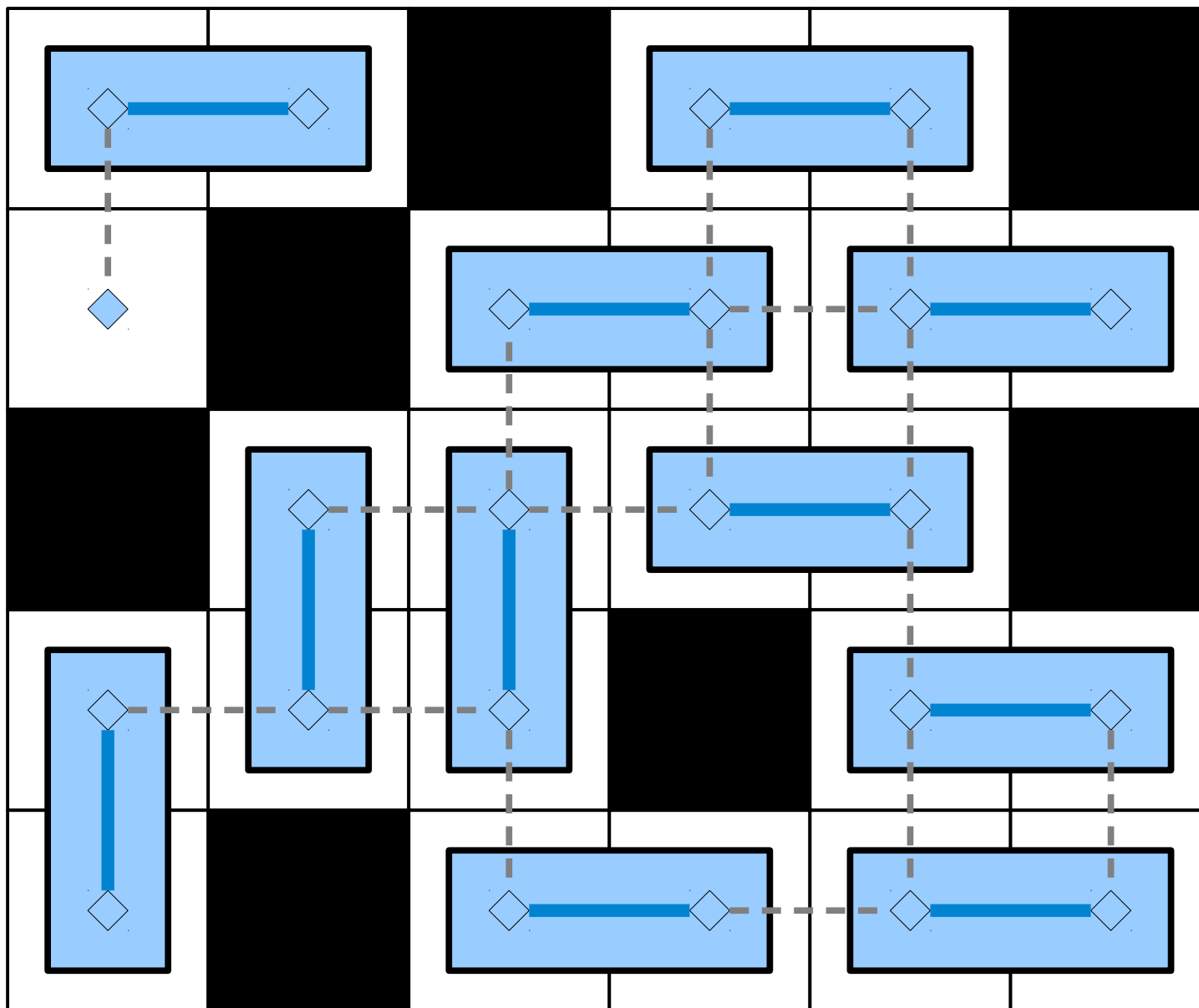
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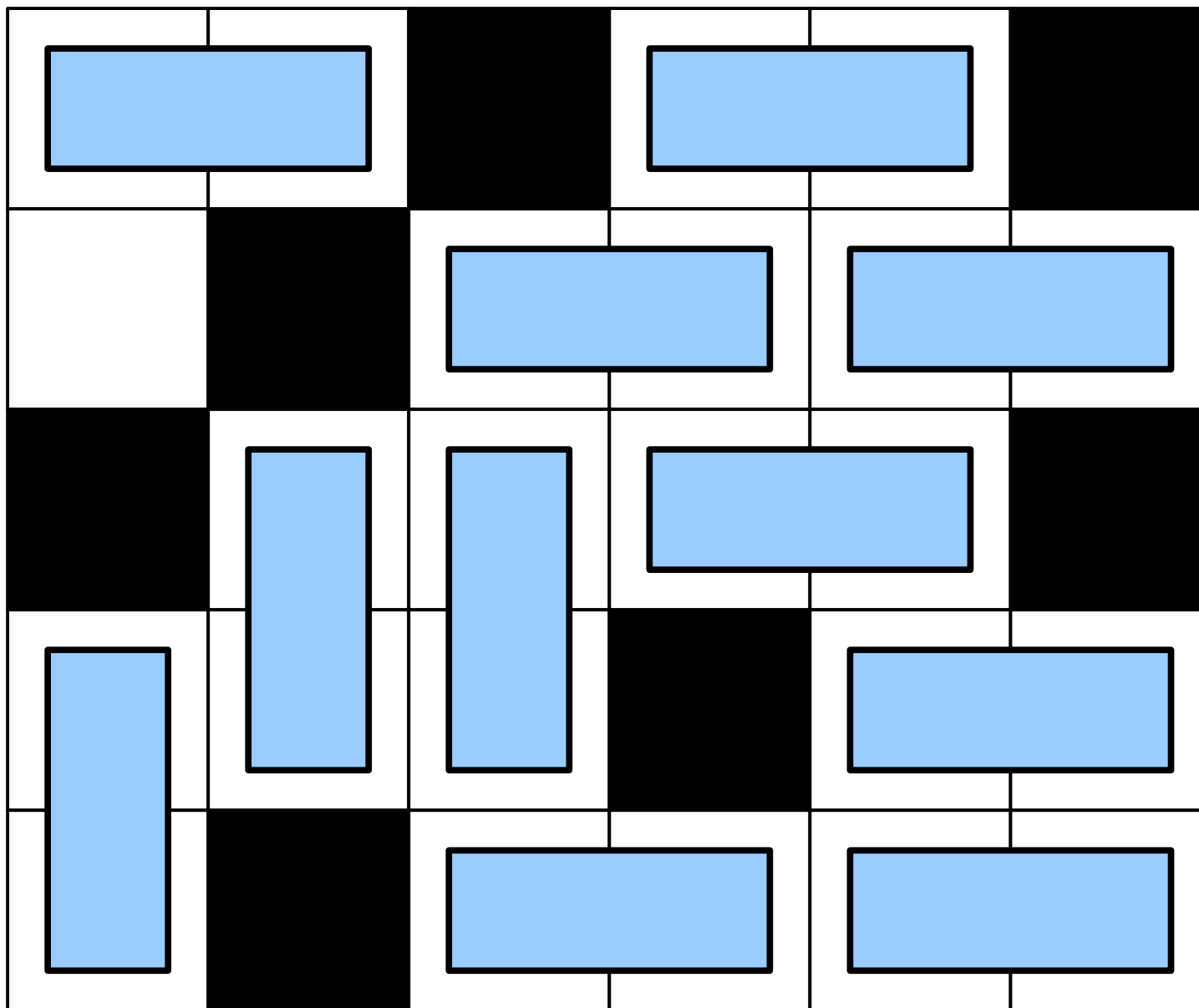
# Solving Domino Tiling



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# Solving Domino Tiling



# In Pseudocode

```
boolean canPlaceDominos(Grid  $G$ , int  $k$ ) {  
    return hasMatching(gridToGraph( $G$ ),  $k$ );  
}
```

Based on this connection between maximum matching and domino tiling, which of the following statements would be more proper to conclude?

- A. Finding a maximum matching isn't any more difficult (in BigO/P-NP terms) than tiling a grid with dominoes.
- B. Tiling a grid with dominoes isn't any more difficult (in BigO/P-NP terms) than finding a maximum matching.

Answer at **PollEv.com/cs103** or  
text **CS103** to **22333** once to join, then **A** or **B**.



## ***Intuition:***

Tiling a grid with dominoes can't be “harder” than solving maximum matching, because if we can solve maximum matching efficiently, we can solve domino tiling efficiently.

Another Example

# Reachability

- Consider the following problem:  
**Given an directed graph  $G$  and nodes  $s$  and  $t$  in  $G$ , is there a path from  $s$  to  $t$ ?**
- It's known that this problem can be solved in polynomial time (use DFS or BFS).
- Given that we can solve the reachability problem in polynomial time, what other problems can we solve in polynomial time?

# Converter Conundrums

- Suppose that you want to plug your laptop into a projector.
- Your laptop only has a VGA output, but the projector needs HDMI input.
- You have a box of connectors that convert various types of input into various types of output (for example, VGA to DVI, DVI to DisplayPort, etc.)
- **Question:** Can you plug your laptop into the projector?

# Converter Conundrums

## Connectors

RGB to USB

VGA to DisplayPort

DB13W3 to CATV

DisplayPort to RGB

DB13W3 to HDMI

DVI to DB13W3

S-Video to DVI

FireWire to SDI

VGA to RGB

DVI to DisplayPort

USB to S-Video

SDI to HDMI

# Converter Conundrums

# Connectors

# RGB to USB

# VGA to DisplayPort

# DB13W3 to CATV

# DisplayPort to RGB

# DB13W3 to HDMI

# DVI to DB13W3

## S-Video to DVI

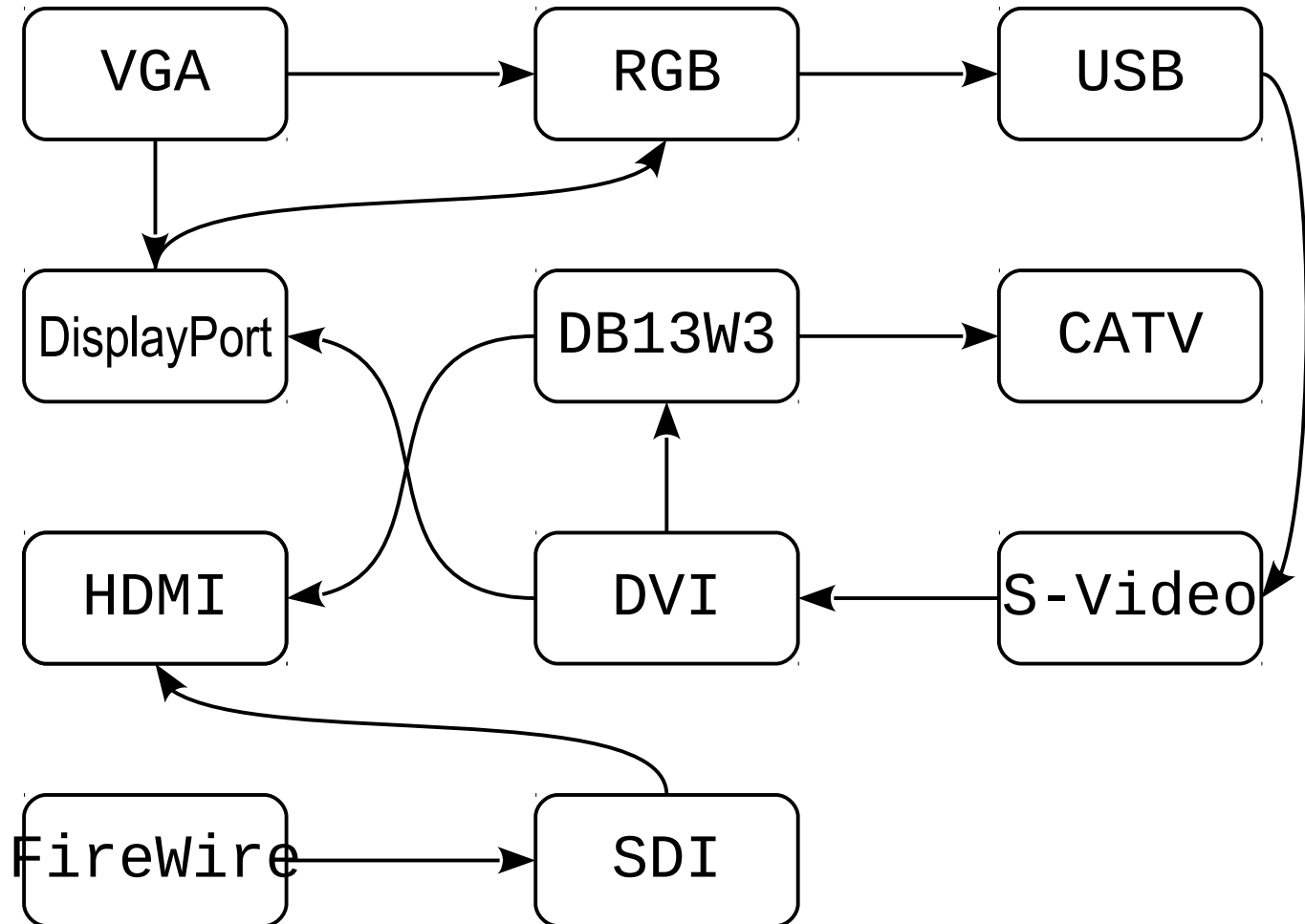
# FireWire to SDI

# VGA to RGB

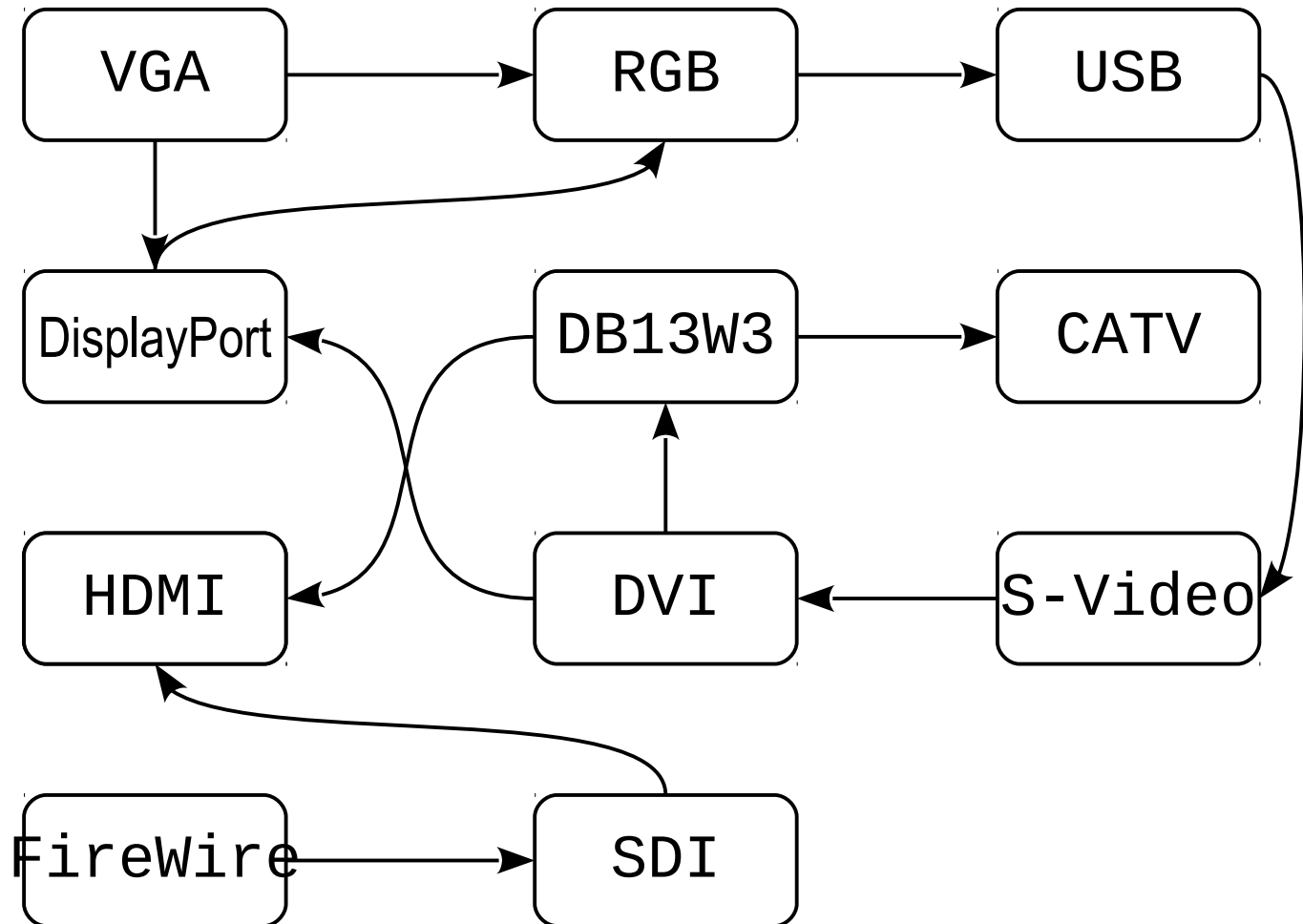
# DVI to DisplayPort

# USB to S-Video

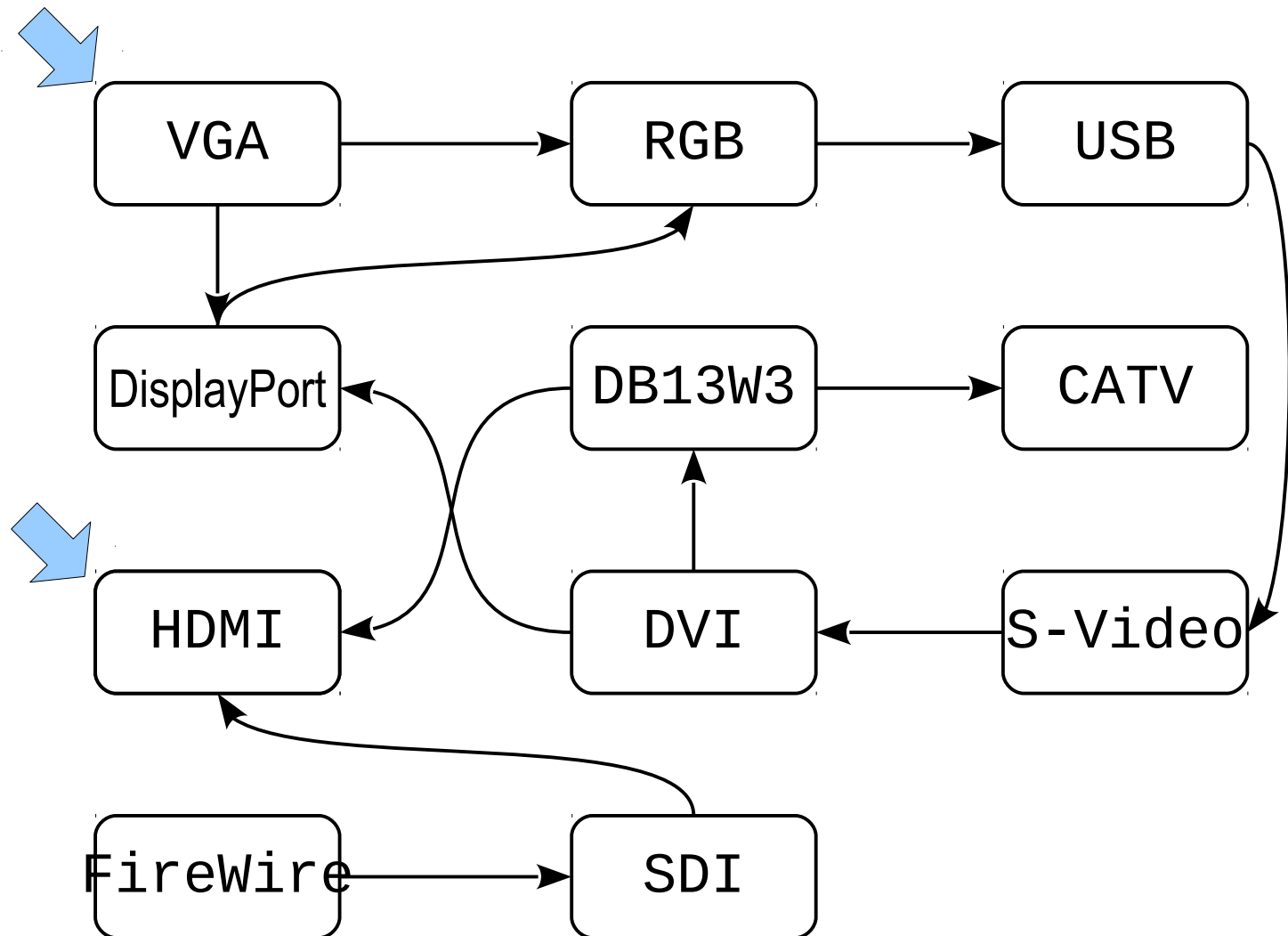
SDI to HDMI



# Converter Conundrums

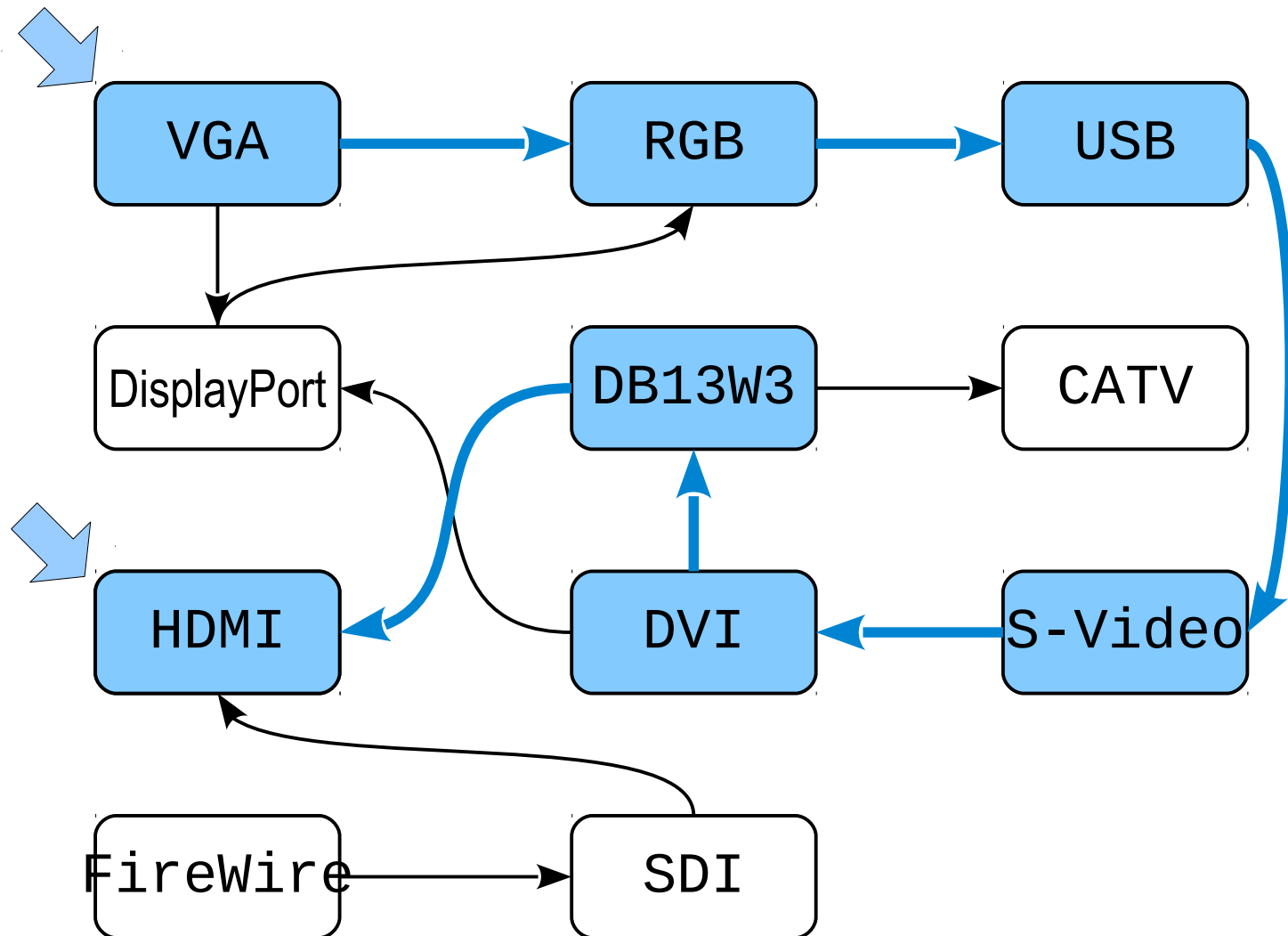


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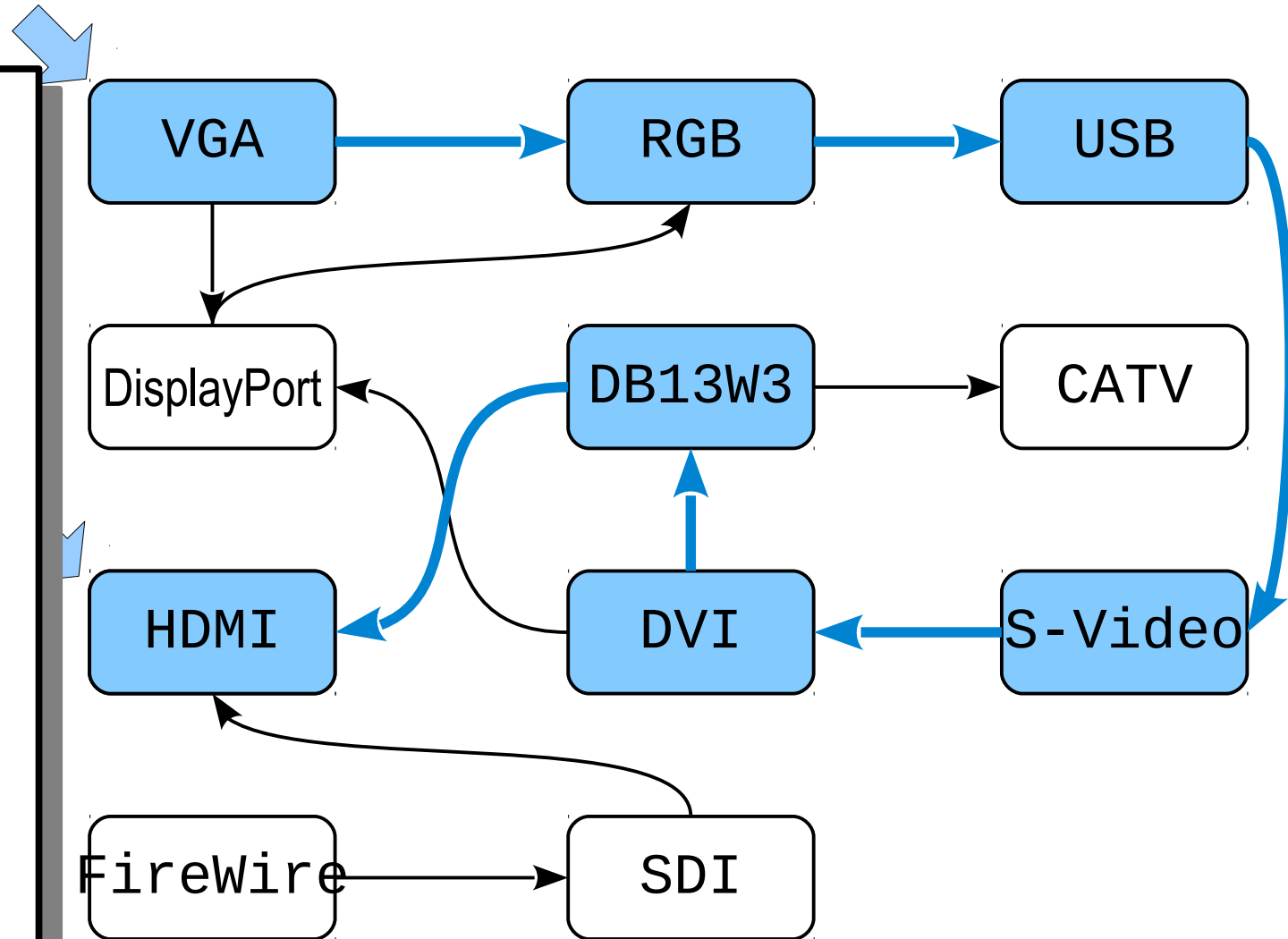
FireWire to SDI

VGA to RGB

DVI to DisplayPort

USB to S-Video

SDI to HDMI



# In Pseudocode

```
boolean canPlugIn(List<Plug> plugs) {  
    return isReachable(plugsToGraph(plugs),  
                        VGA, HDMI);  
}
```

Based on this connection between plugging a laptop into a projector and determining reachability, which of the following statements would be more proper to conclude?

- A. Plugging a laptop into a projector isn't any more difficult than computing reachability in a directed graph.
- B. Computing reachability in a directed graph isn't any more difficult than plugging a laptop into a projector.

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## ***Intuition:***

Finding a way to plug a computer into a projector can't be “harder” than determining reachability in a graph, since if we can determine reachability in a graph, we can find a way to plug a computer into a projector.

```
bool solveProblemA(string input) {  
    return solveProblemB(transform(input));  
}
```

### ***Intuition:***

Problem  $A$  can't be “harder” than problem  $B$ , because solving problem  $B$  lets us solve problem  $A$ .

```
bool solveProblemA(string input) {  
    return solveProblemB(transform(input));  
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```

- If  $A$  and  $B$  are problems where it's possible to solve problem  $A$  using the strategy shown above\*, we write

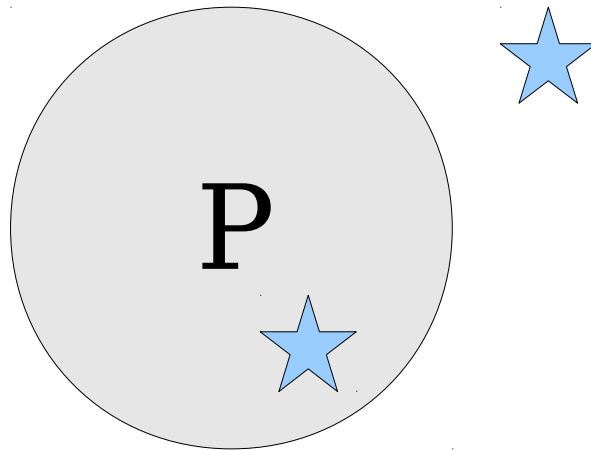
$$A \leq_p B.$$

- We say that ***A is polynomial-time reducible to B.***

\* Assuming that transform runs in polynomial time.

# Polynomial-Time Reductions

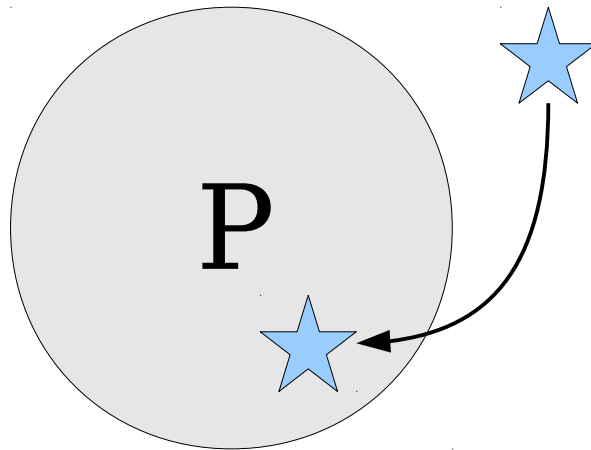
- If  $A \leq_p B$  and  $B \in \mathbf{P}$ , then  $A \in \mathbf{P}$ .





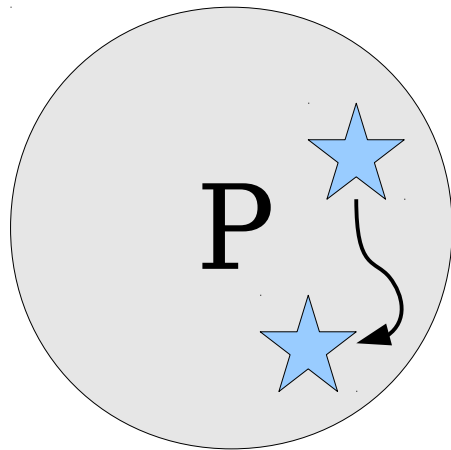
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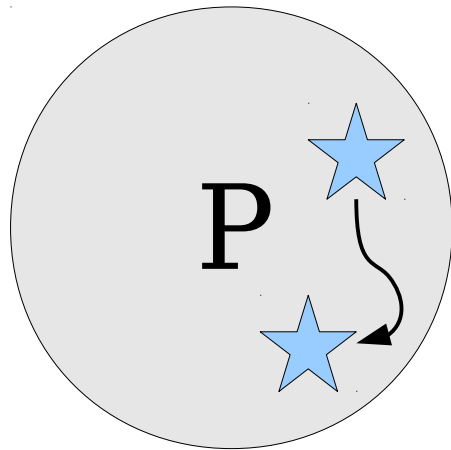
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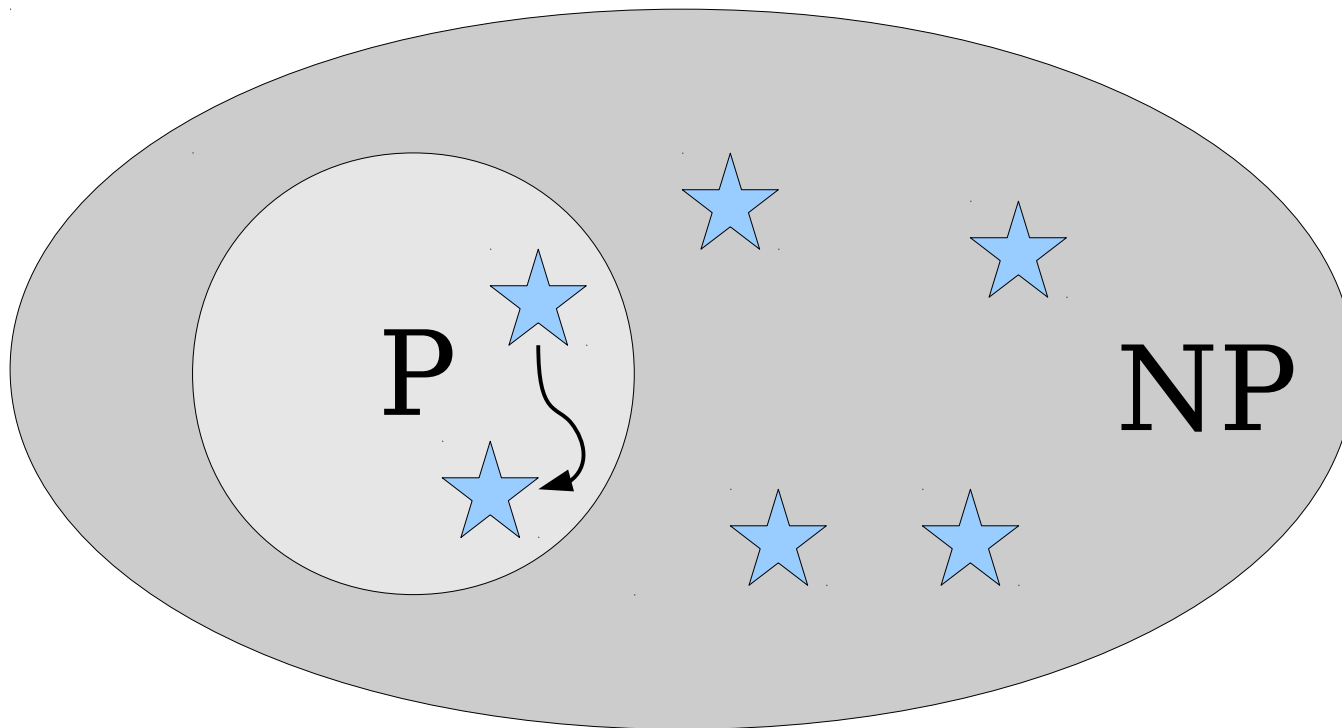
# Polynomial-Time Reductions

- If  $A \leq_p B$  and  $B \in \mathbf{P}$ , then  $A \in \mathbf{P}$ .
- If  $A \leq_p B$  and  $B \in \mathbf{NP}$ , then  $A \in \mathbf{NP}$ .



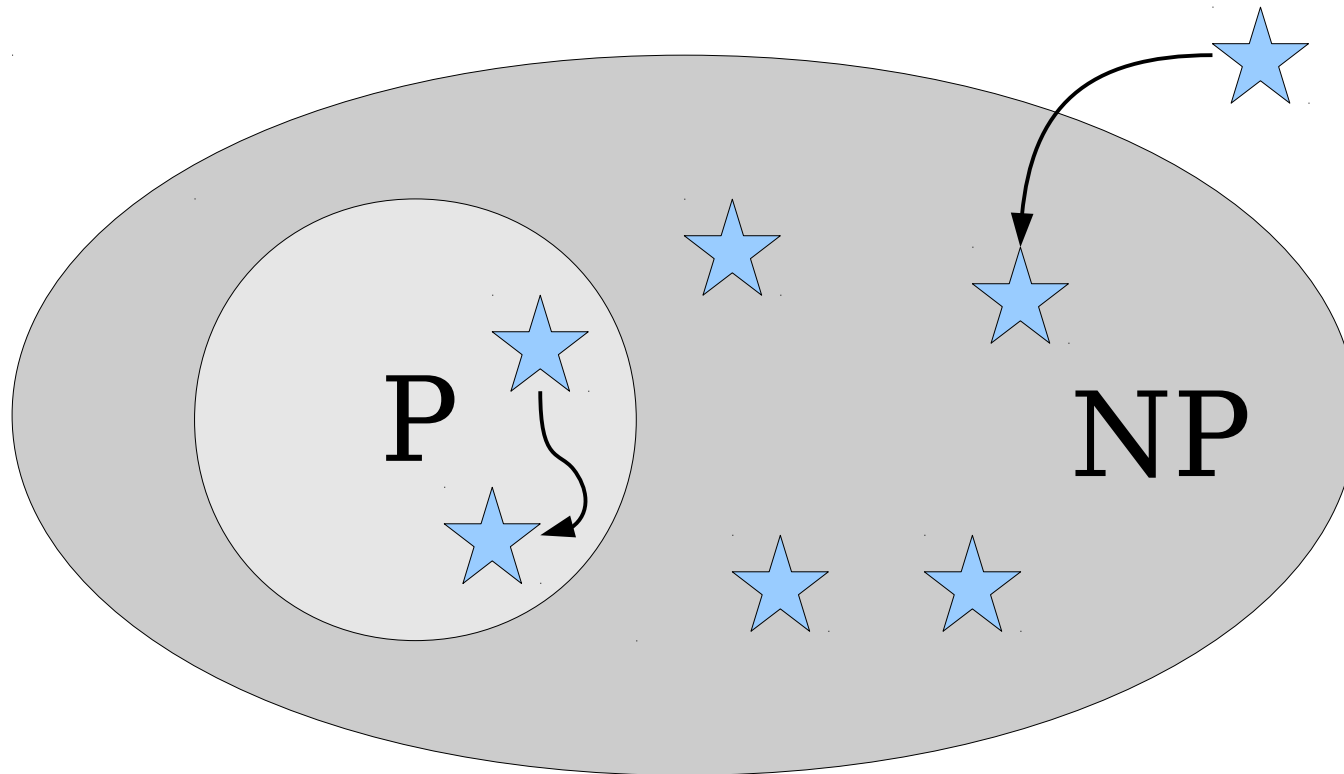
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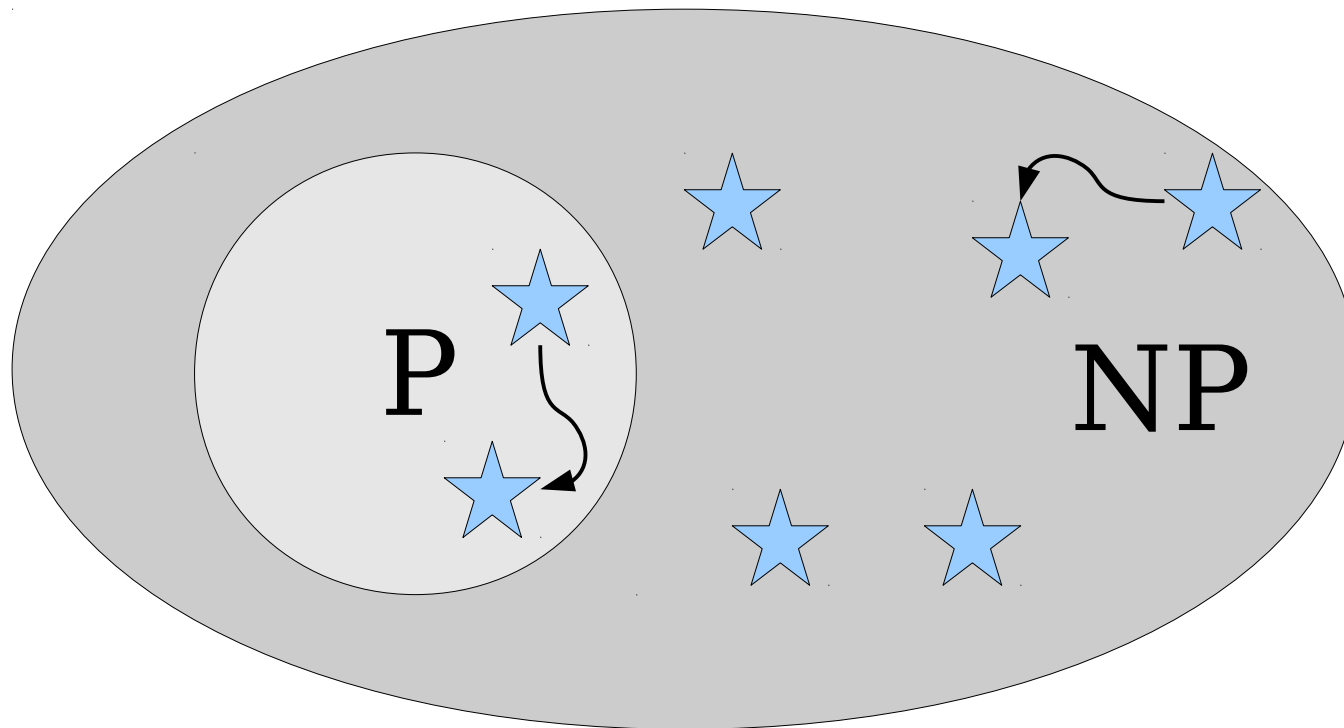
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This  $\leq_p$  relation lets us rank the relative difficulties of problems in **P** and **NP**.

What else can we do with it?

# **NP**-Hardness and **NP**-Completeness



***Question:*** What makes a problem  
hard to solve?

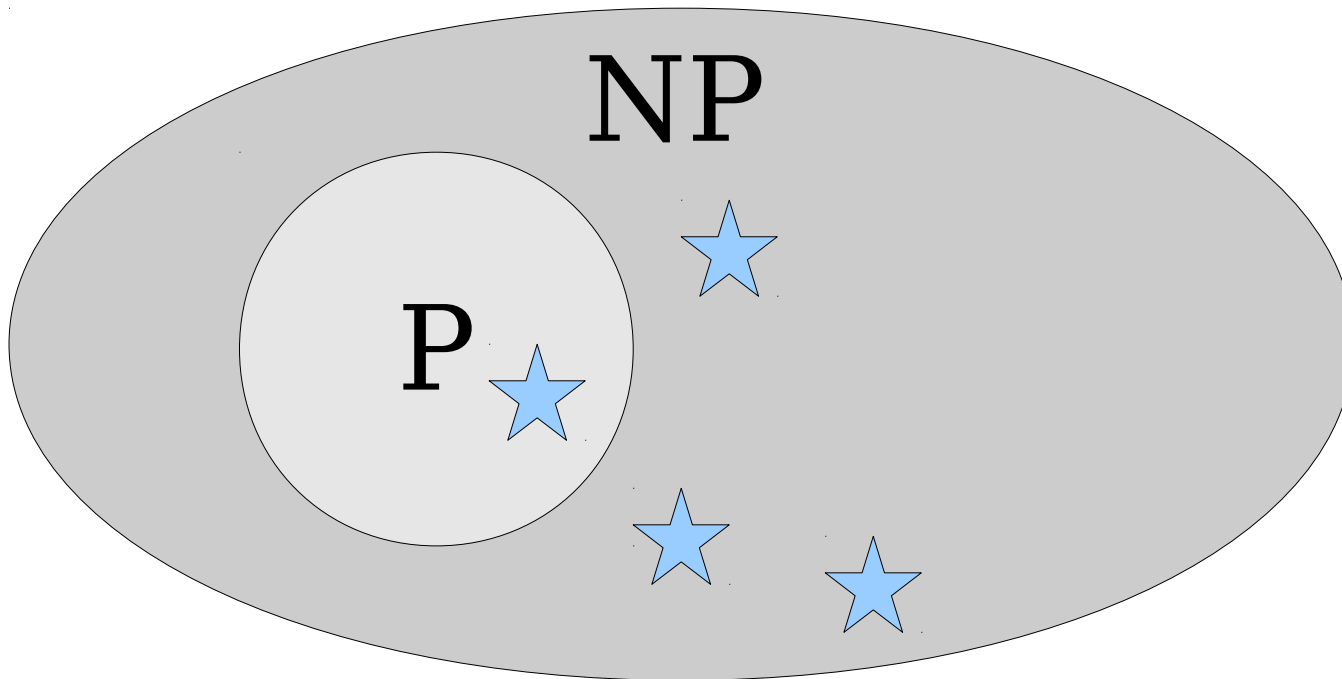
***Intuition:*** If  $A \leq_p B$ , then problem  $B$  is at least as hard\* as problem  $A$ .

\* for some definition of “at least as hard as.”

***Intuition:*** To show that some problem is hard, show that lots of other problems reduce to it.

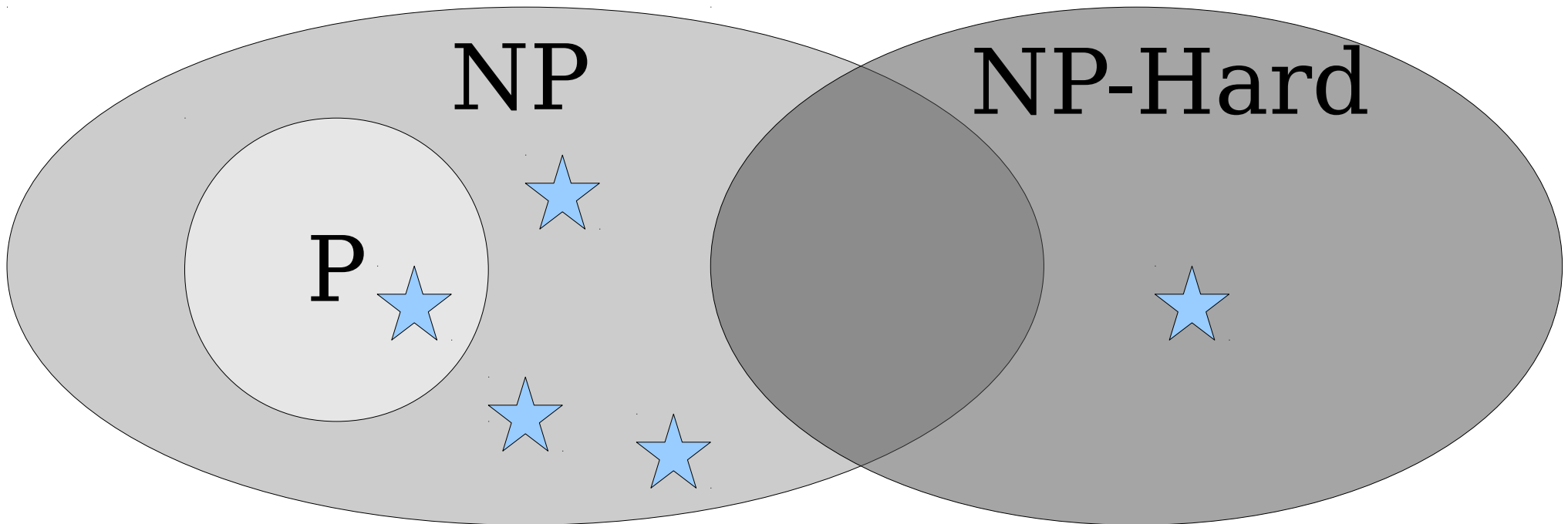
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- A language  $L$  is called **NP-hard** if for *every*  $A \in \mathbf{NP}$ , we have  $A \leq_p L$ .



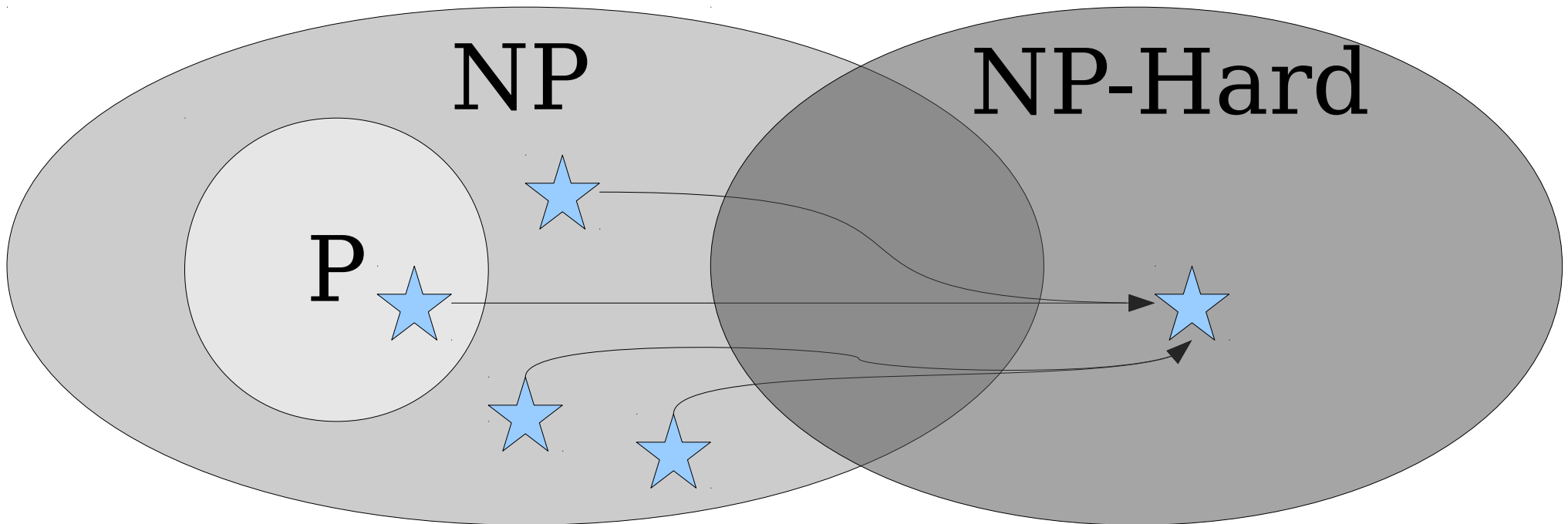
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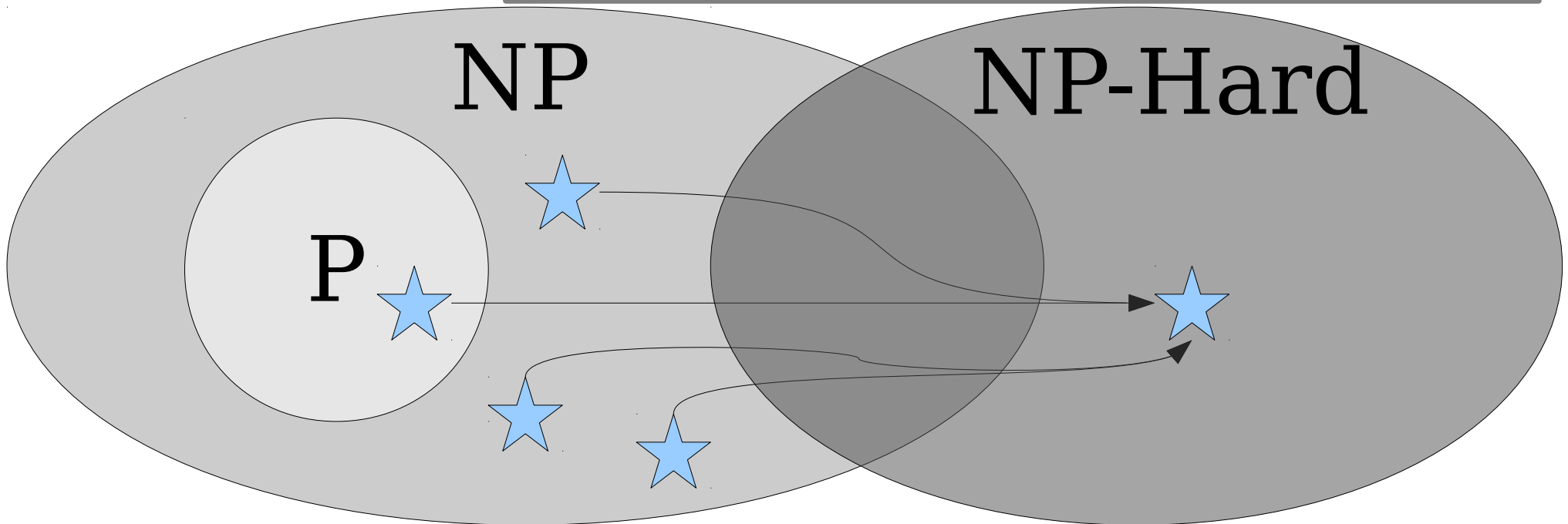
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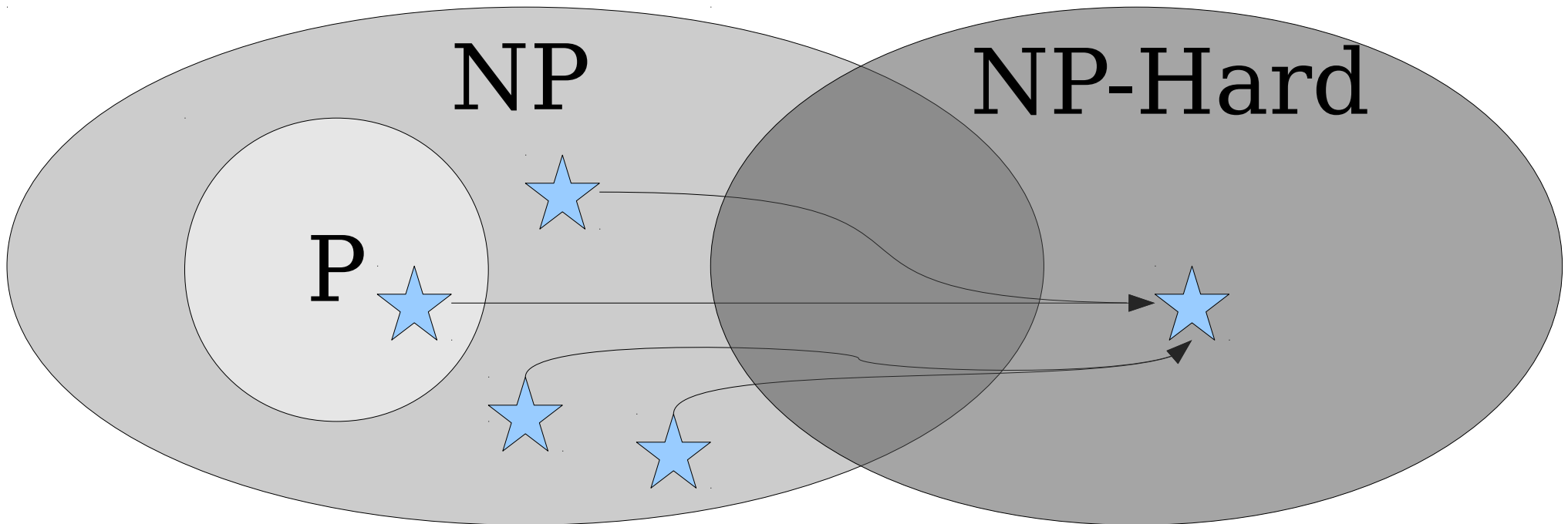
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Intuitively:  $L$  has to be at least as hard as every problem in  $\mathbf{NP}$ , since an algorithm for  $L$  can be used to decide all problems in  $\mathbf{NP}$ .



# NP-Hardness

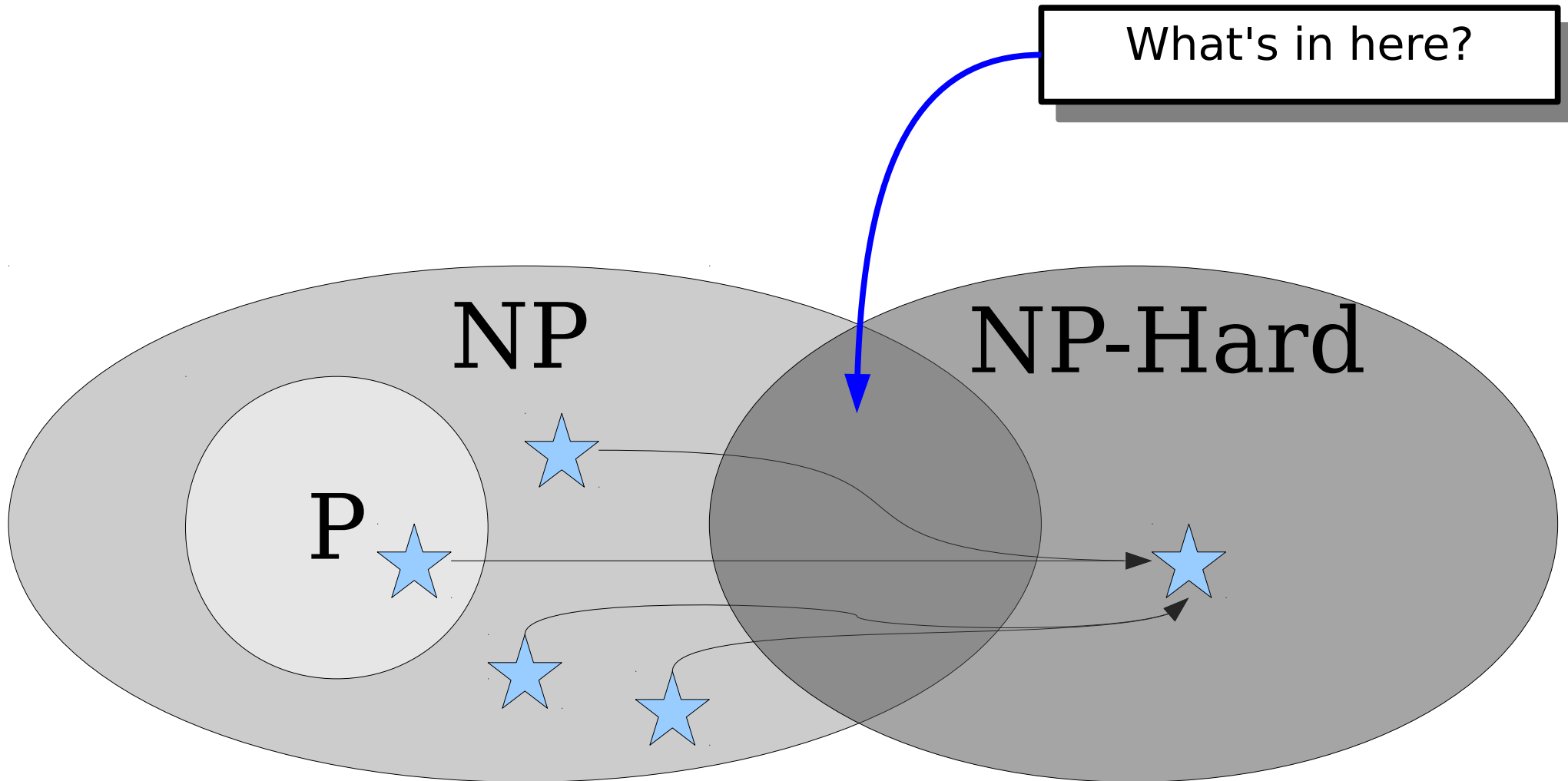
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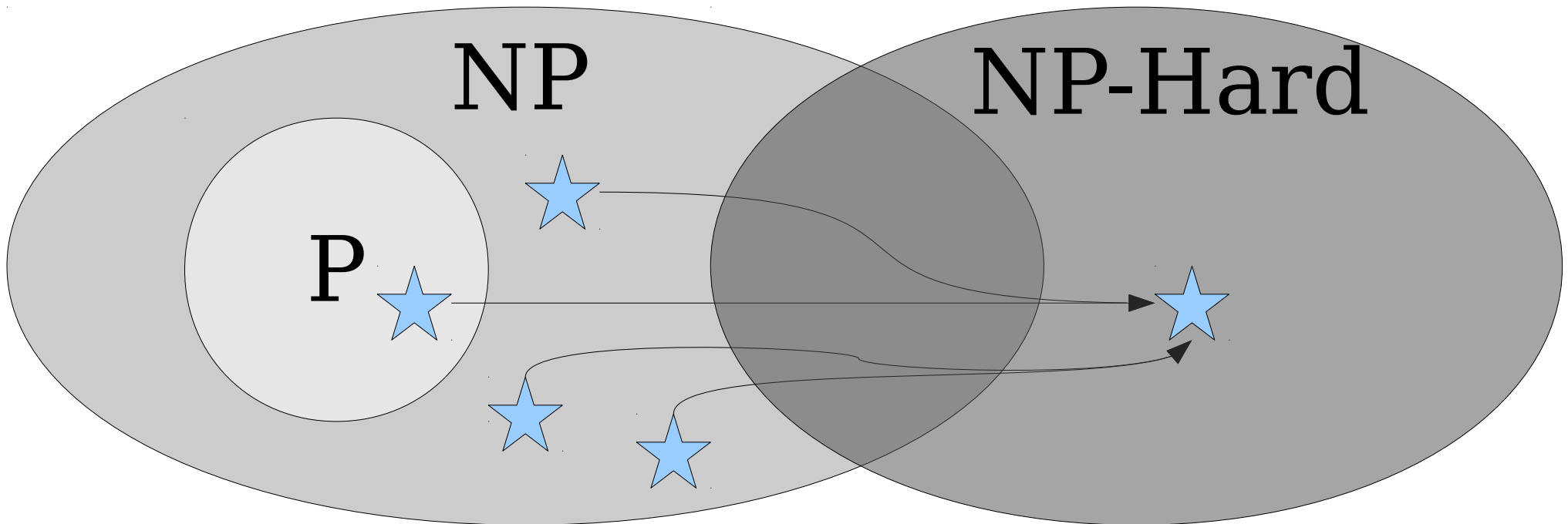
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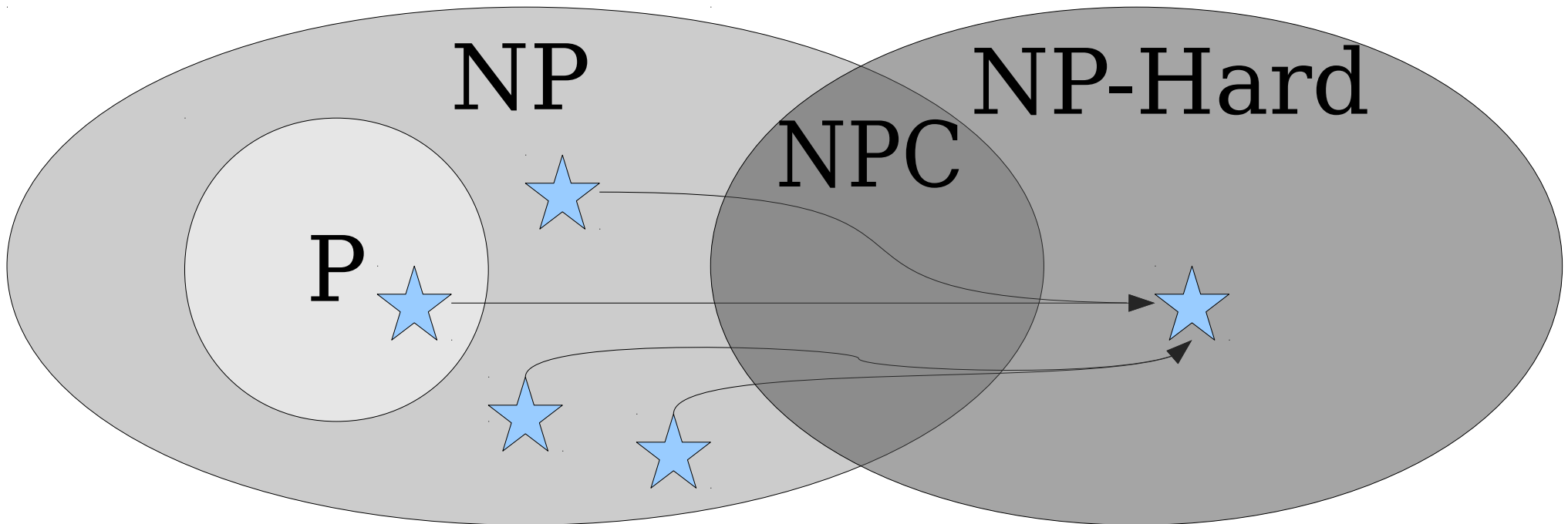
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- A language  $L$  is called **NP-hard** if for *every*  $A \in \mathbf{NP}$ , we have  $A \leq_p L$ .
- A language in  $L$  is called **NP-complete** if  $L$  is **NP-hard** and  $L \in \mathbf{NP}$ .
- The class **NPC** is the set of **NP-complete** problems.



# NP-Hardness

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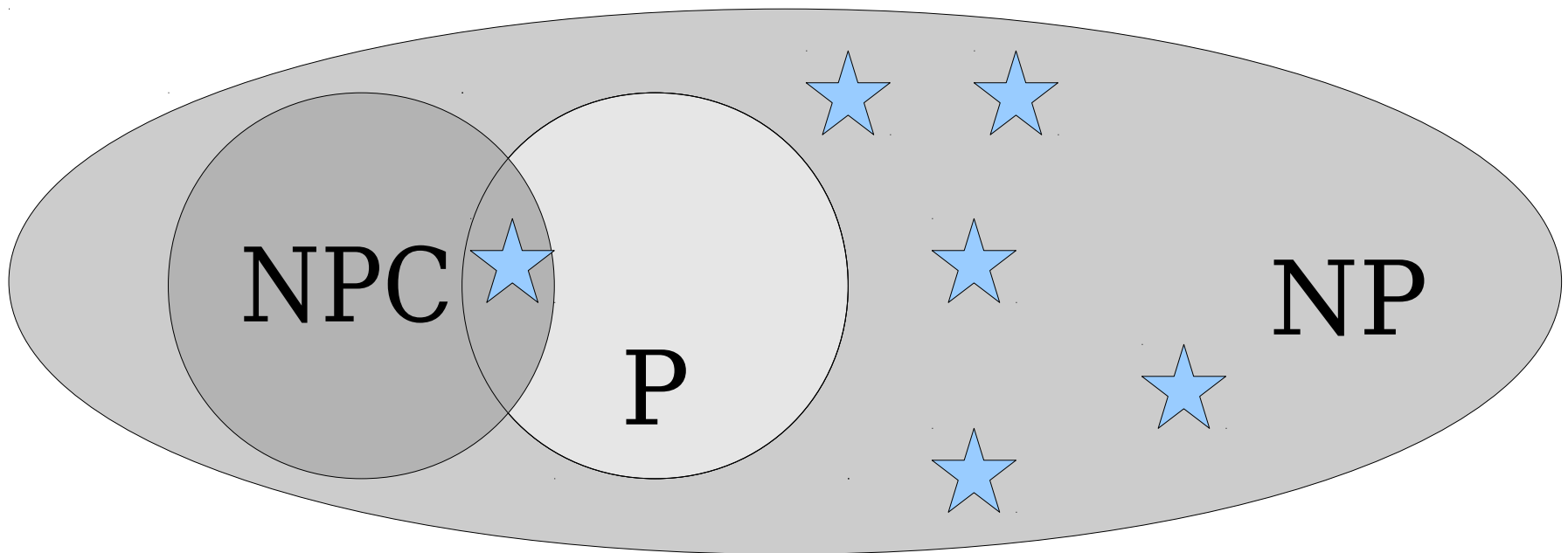


# The Tantalizing Truth

***Theorem:*** If *any* **NP**-complete language is in **P**, then **P** = **NP**.

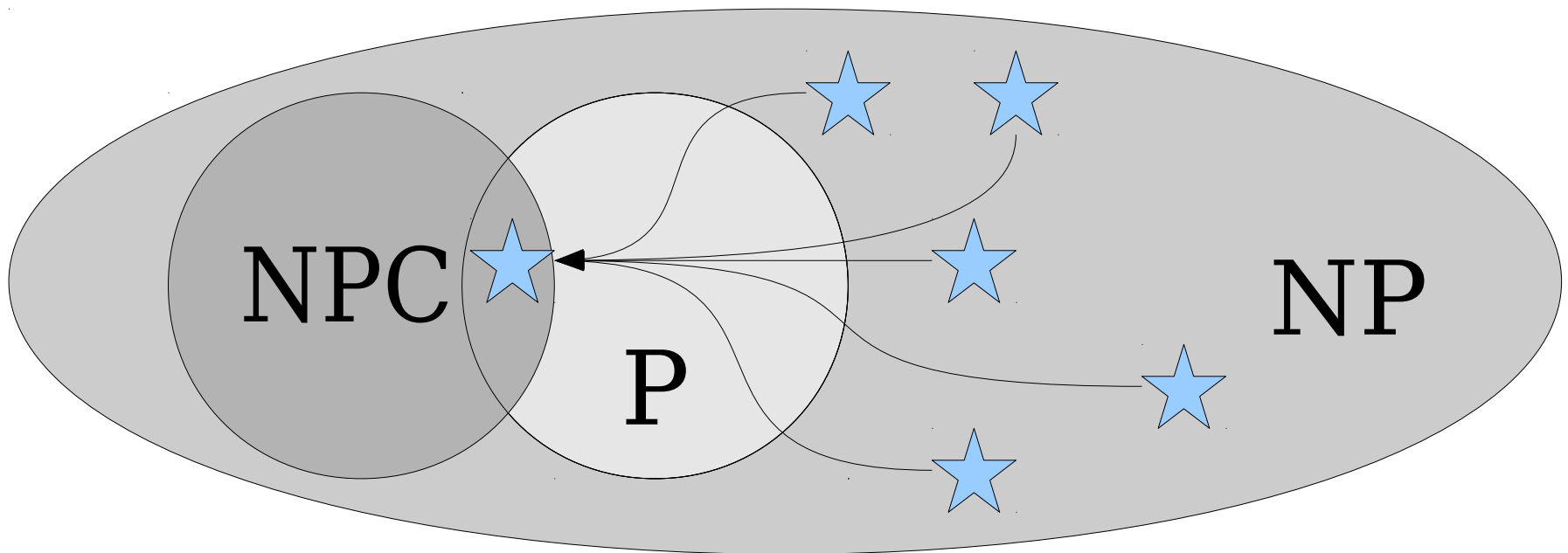
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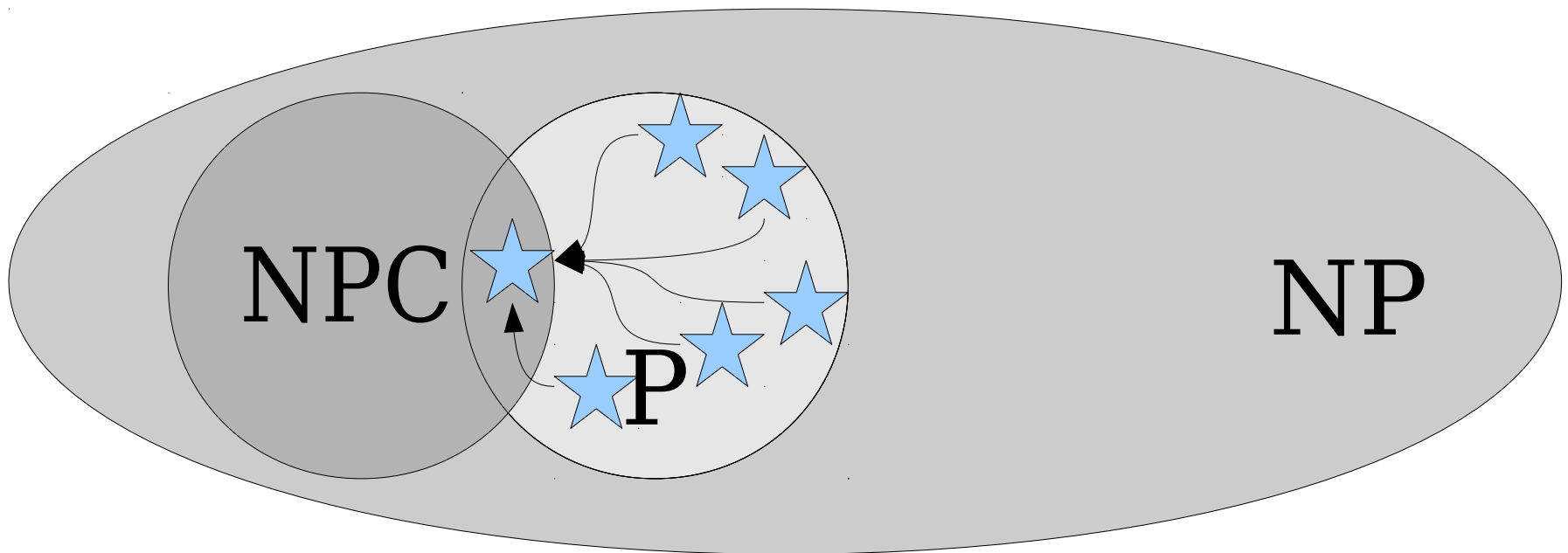
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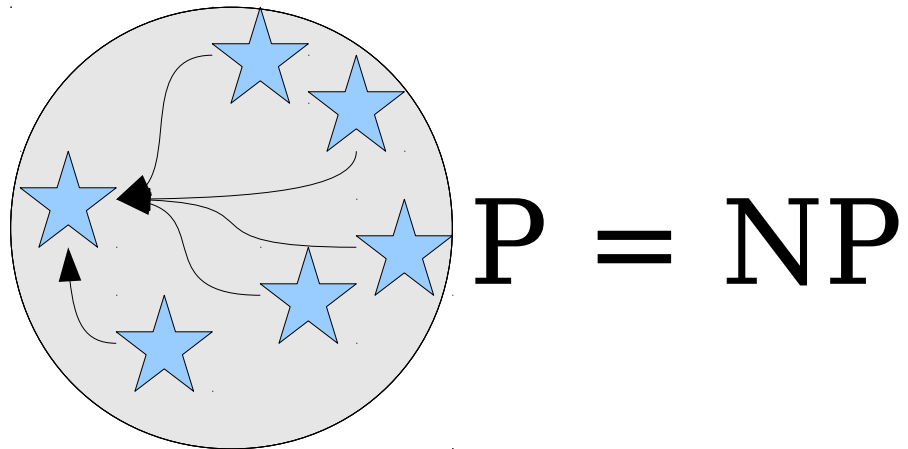
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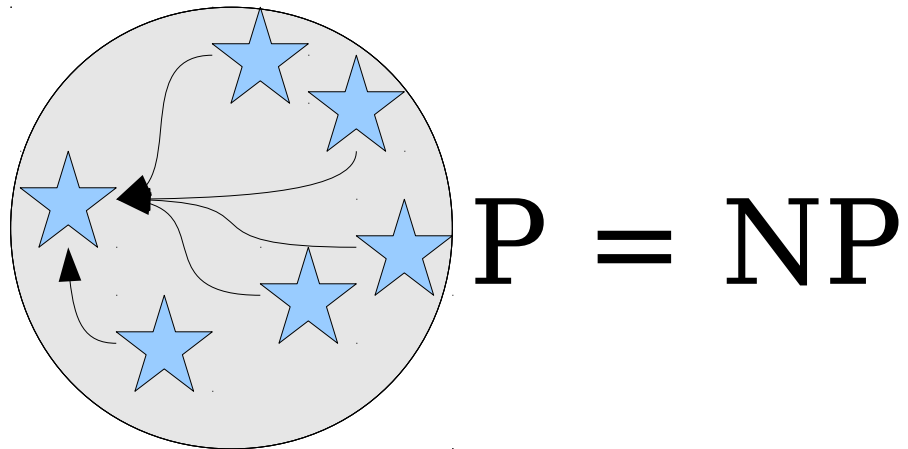




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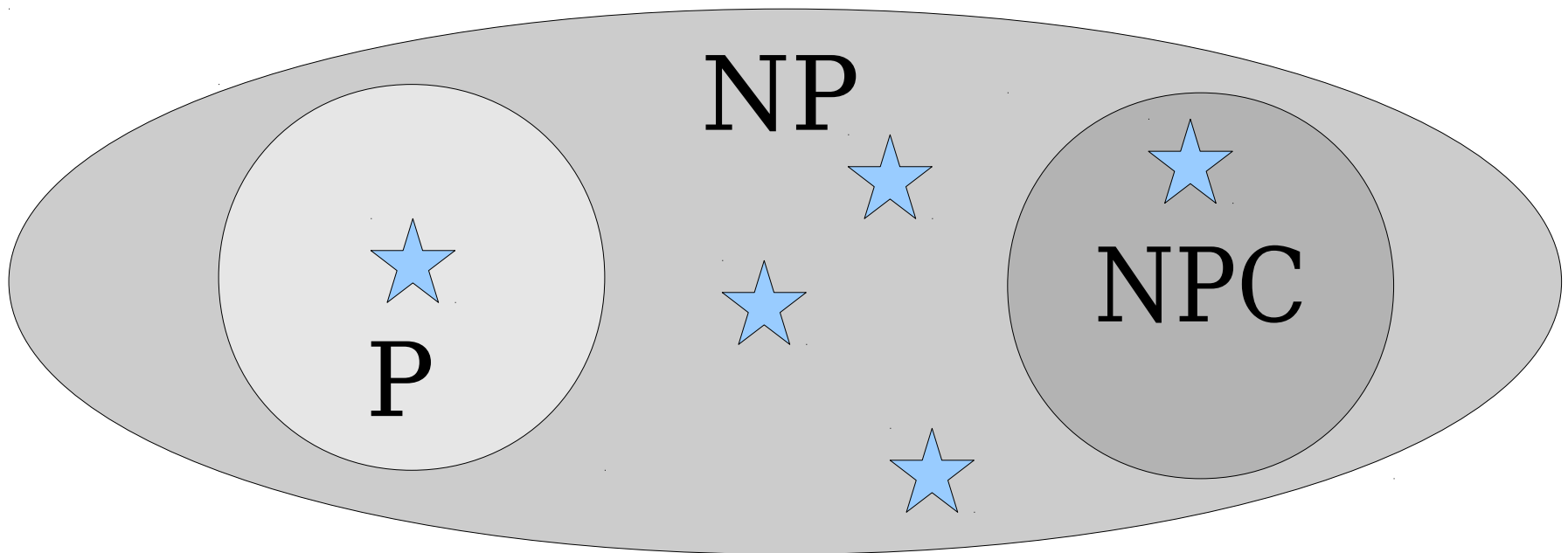
**Proof:** Suppose that  $L$  is **NP**-complete and  $L \in \mathbf{P}$ . Now consider any arbitrary **NP** problem  $A$ . Since  $L$  is **NP**-complete, we know that  $A \leq_p L$ . Since  $L \in \mathbf{P}$  and  $A \leq_p L$ , we see that  $A \in \mathbf{P}$ . Since our choice of  $A$  was arbitrary, this means that **NP**  $\subseteq$  **P**, so **P** = **NP**. ■



# The Tantalizing Truth

**Theorem:** If *any* **NP**-complete language is not in **P**, then  $\mathbf{P} \neq \mathbf{NP}$ .

**Proof:** Suppose that  $L$  is an **NP**-complete language not in **P**. Since  $L$  is **NP**-complete, we know that  $L \in \mathbf{NP}$ . Therefore, we know that  $L \in \mathbf{NP}$  and  $L \notin \mathbf{P}$ , so  $\mathbf{P} \neq \mathbf{NP}$ . ■



***How do we even know NP-complete problems exist in the first place?***

# Satisfiability

- A propositional logic formula  $\varphi$  is called **satisfiable** if there is some assignment to its variables that makes it evaluate to true.
  - $p \wedge q$  is satisfiable.
  - $p \wedge \neg p$  is unsatisfiable.
  - $p \rightarrow (q \wedge \neg q)$  is satisfiable.
- An assignment of true and false to the variables of  $\varphi$  that makes it evaluate to true is called a **satisfying assignment**.

# SAT

- The *boolean satisfiability problem* (***SAT***) is the following:

**Given a propositional logic formula  $\varphi$ , is  $\varphi$  satisfiable?**

- Formally:

**$SAT = \{ \langle \varphi \rangle \mid \varphi \text{ is a satisfiable PL formula} \}$**

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The language SAT happens to be in **NP**. Think about how a polynomial-time verifier for SAT might work. Which of the following would work as certificates for such a verifier, given that the input is a propositional formula  $\varphi$ ?

- A. The truth table of  $\varphi$ .
- B. One possible variable assignment to  $\varphi$ .
- C. A list of all possible variable assignments for  $\varphi$ .
- D. None of the above, or two or more of the above.

Answer at **PollEv.com/cs103** or  
text **CS103** to **22333** once to join, then **A**, **B**, **C**, or **D**.

***Theorem (Cook-Levin):*** SAT is **NP**-complete.

***Proof Idea:*** To see that **SAT**  $\in$  **NP**, show how to make a polynomial-time verifier for it. Key idea: have the certificate be a satisfying assignment.

To show that **SAT** is **NP**-hard, given a polynomial-time verifier  $V$  for an arbitrary **NP** language  $L$ , for any string  $w$  you can construct a polynomially-sized formula  $\varphi(w)$  that says “there is a certificate  $c$  where  $V$  accepts  $\langle w, c \rangle$ .” This formula is satisfiable if and only if  $w \in L$ , so deciding whether the formula is satisfiable decides whether  $w$  is in  $L$ .

***Proof:*** Take CS154!

# Why All This Matters

- Resolving  $\mathbf{P} \stackrel{?}{=} \mathbf{NP}$  is equivalent to just figuring out how hard SAT is.
  - If  $\text{SAT} \in \mathbf{P}$ , then  $\mathbf{P} = \mathbf{NP}$ .  
If  $\text{SAT} \notin \mathbf{P}$ , then  $\mathbf{P} \neq \mathbf{NP}$ .



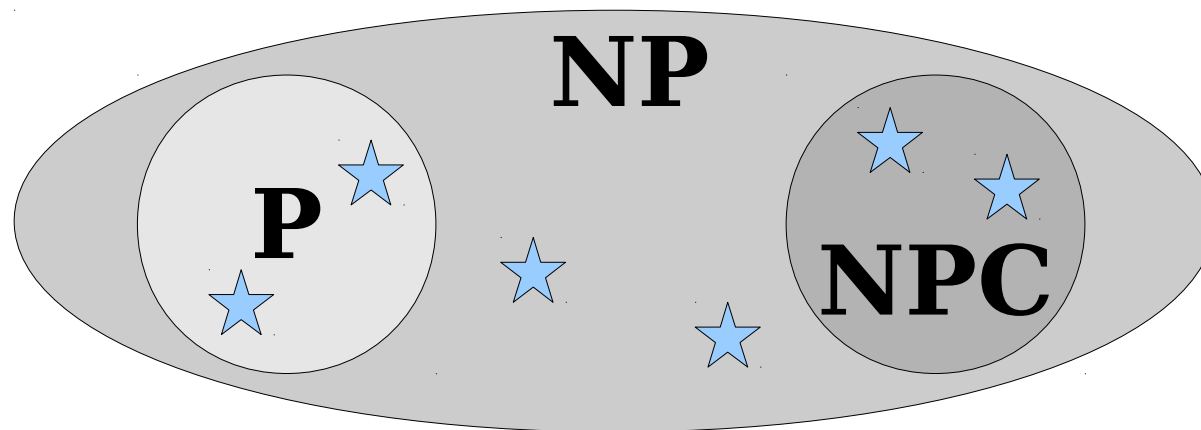
# Sample NP-Hard Problems

- **Computational biology:** Given a set of genomes, what is the most probable evolutionary tree that would give rise to those genomes? (*Maximum parsimony problem*)
- **Game theory:** Given an arbitrary perfect-information, finite, two player game, who wins? (*Generalized geography problem*)
- **Operations research:** Given a set of jobs and workers who can perform those tasks in parallel, can you complete all the jobs within some time bound? (*Job scheduling problem*)
- **Machine learning:** Given a set of data, find the simplest way of modeling the statistical patterns in that data (*Bayesian network inference problem*)
- **Medicine:** Given a group of people who need kidneys and a group of kidney donors, find the maximum number of people who can end up with kidneys (*Cycle cover problem*)
- **Systems:** Given a set of processes and a number of processors, find the optimal way to assign those tasks so that they complete as soon as possible (*Processor scheduling problem*)

***Coda:*** What if  $\mathbf{P} \stackrel{?}{=} \mathbf{NP}$  is resolved?

# Intermediate Problems

- With few exceptions, every problem we've discovered in **NP** has either
  - definitely been proven to be in **P**, or
  - definitely been proven to be **NP**-complete.
- A problem that's **NP**, not in **P**, but not **NP**-complete is called ***NP-intermediate***.
- ***Theorem (Ladner)***: There are **NP**-intermediate problems if and only if **P**  $\neq$  **NP**.



What if **P**  $\neq$  **NP**?

## ***A Good Read:***

“A Personal View of Average-Case Complexity” by Russell Impagliazzo

What if **P** = **NP**?

And a Dismal Third Option