

Complexity Theory

Part Two

Recap from Last Time

The Complexity Class **P**

- The complexity class **P** (***polynomial time***) is defined as

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

- Intuitively, **P** contains all decision problems that can be solved efficiently.
- This is like class **R**, except with “efficiently” tacked onto the end.

The Complexity Class **NP**

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

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

- Intuitively, **NP** is the set of problems where “yes” answers can be checked efficiently.
- This is like the class **RE**, but with “efficiently” tacked on to the definition.

The Biggest Unsolved Problem in
Theoretical Computer Science:

$$\mathbf{P} \stackrel{?}{=} \mathbf{NP}$$

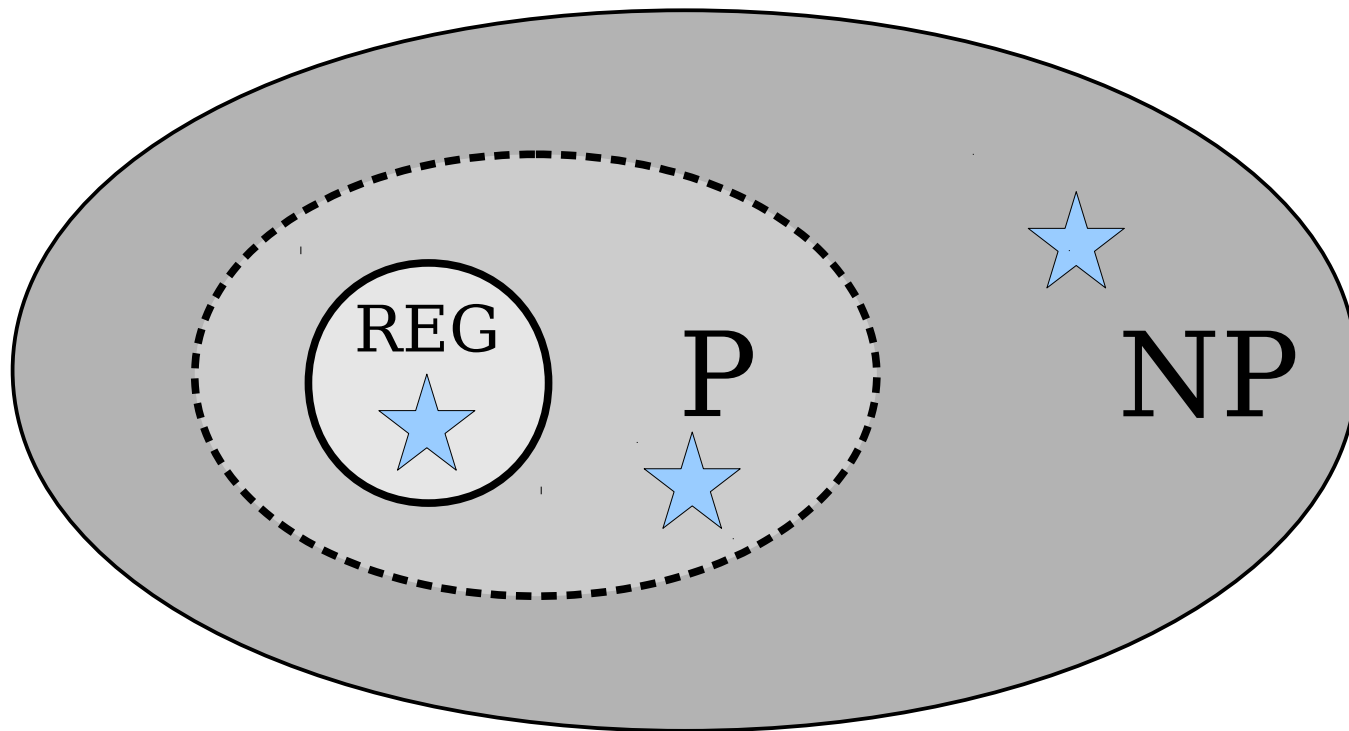
Theorem (Baker-Gill-Solovay): Any proof that purely relies on universality and self-reference cannot resolve $\mathbf{P} \stackrel{?}{=} \mathbf{NP}$.

Proof: Take CS154!

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?

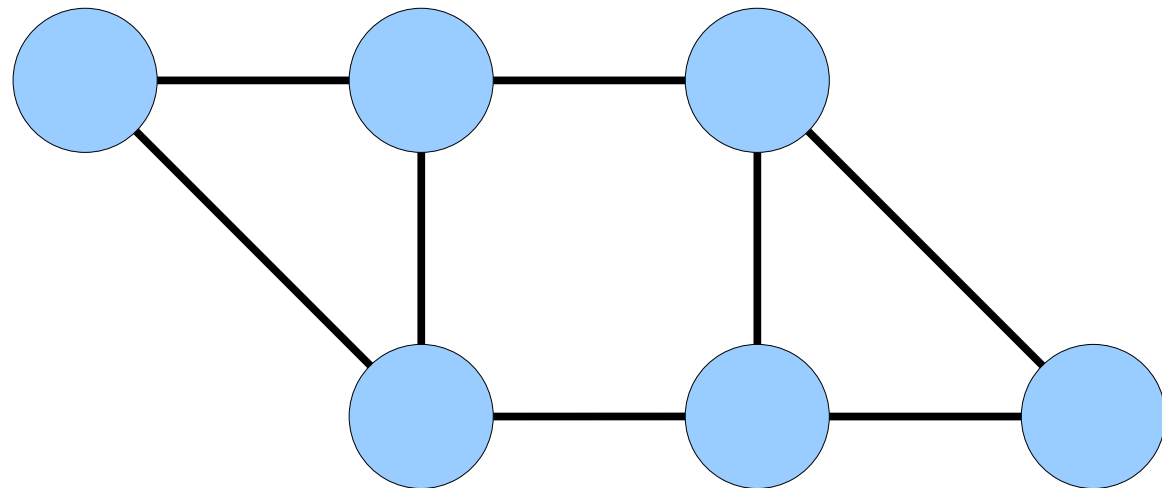
Reducibility

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.

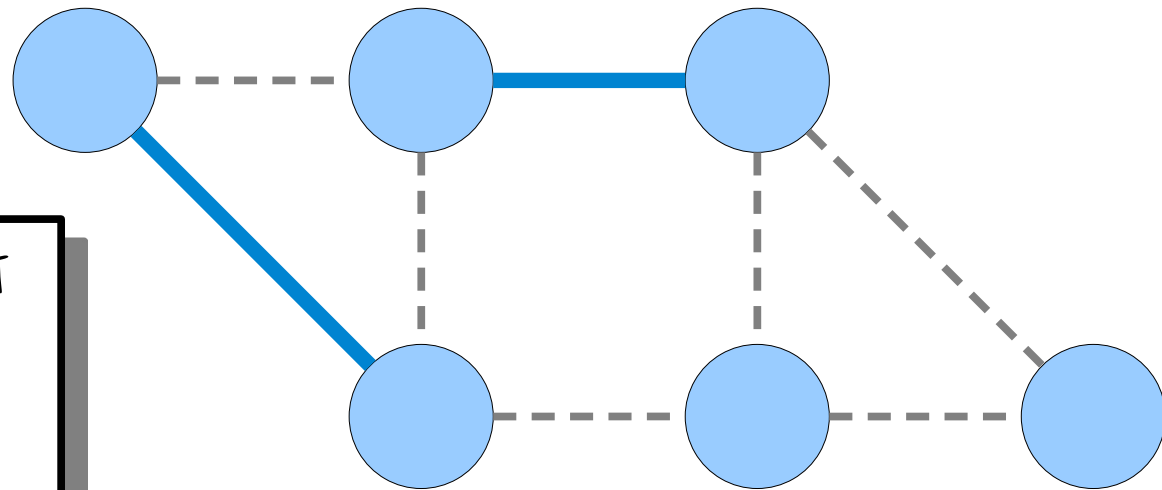
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Maximum Matching

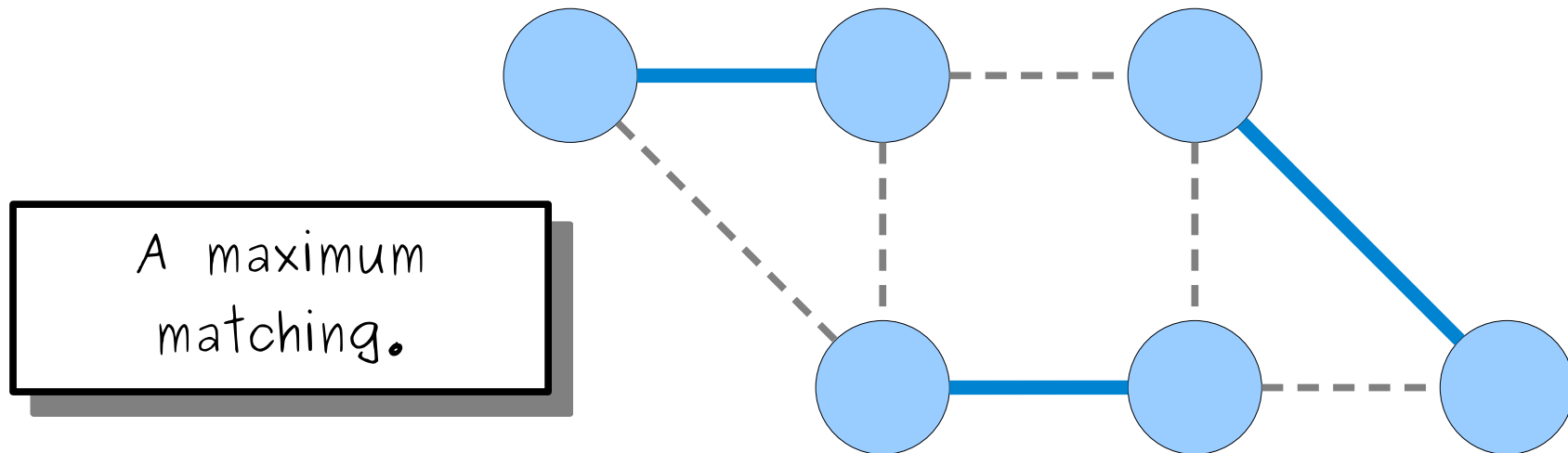
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A matching, but
not a maximum
matching.

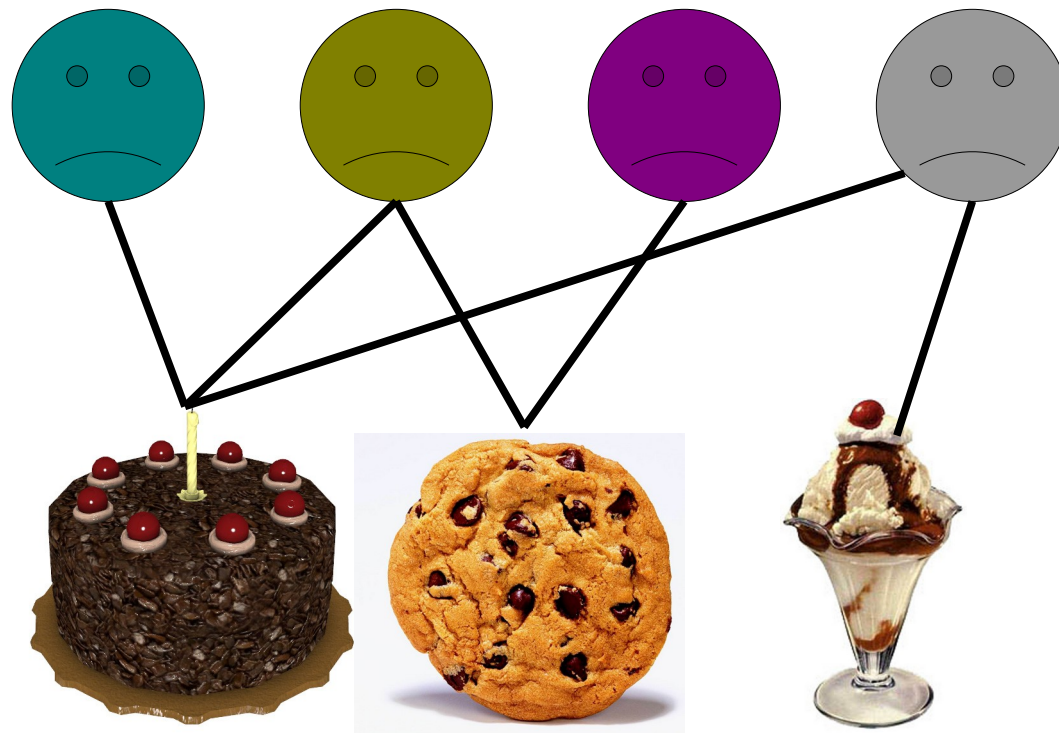
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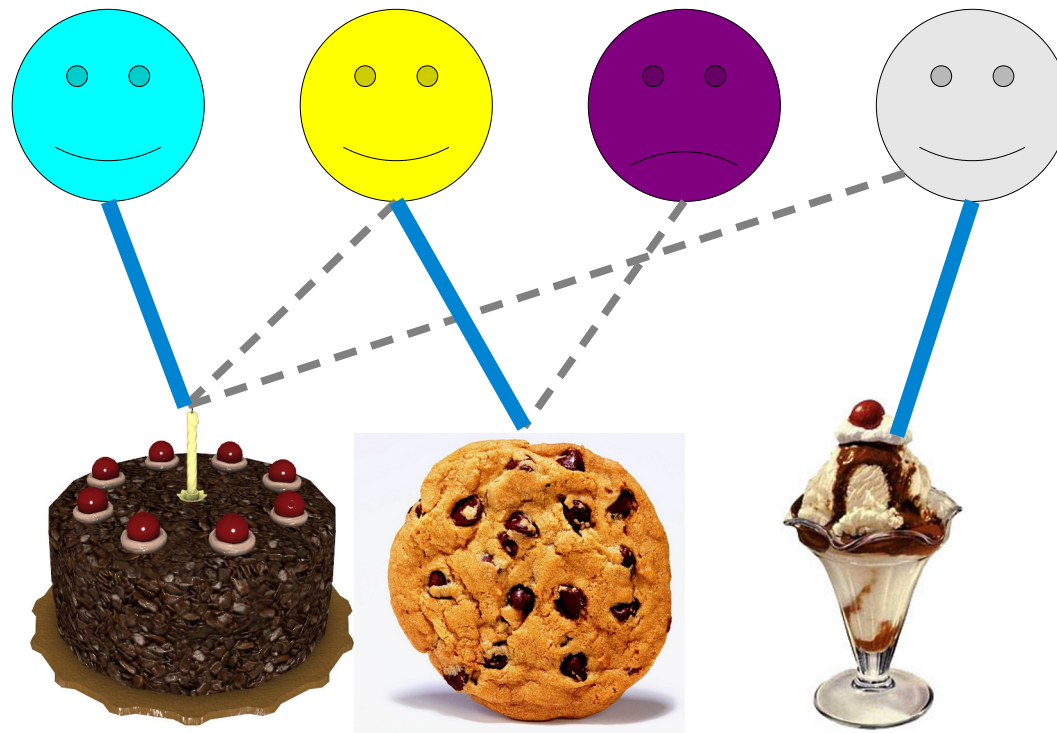
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Maximum Matching

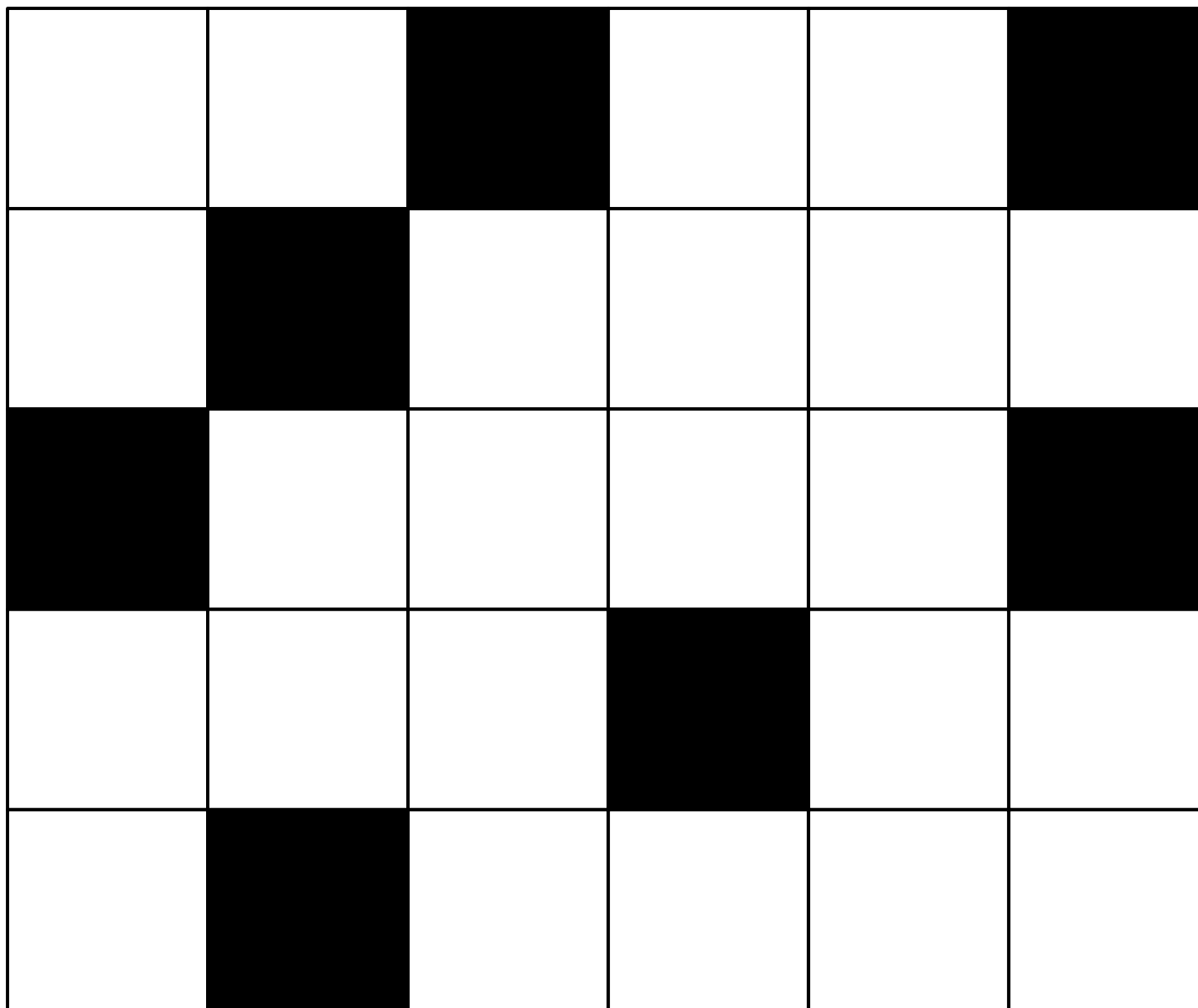
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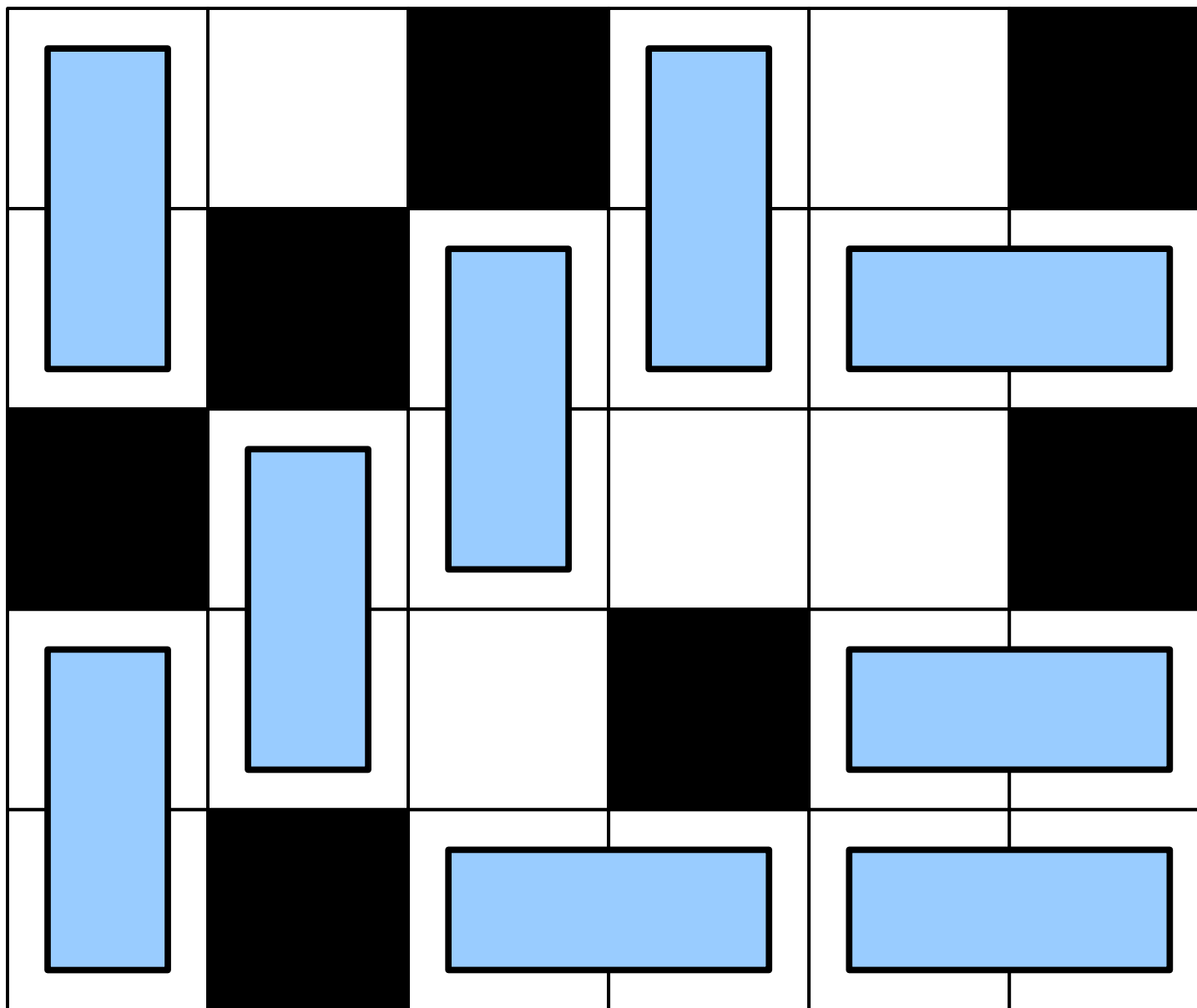
Maximum Matching

- Jack Edmonds' paper “Paths, Trees, and Flowers” gives a polynomial-time algorithm for finding maximum matchings.
 - He's the guy from last time with the quote about “better than decidable.”
- Using this fact, what other problems can we solve?

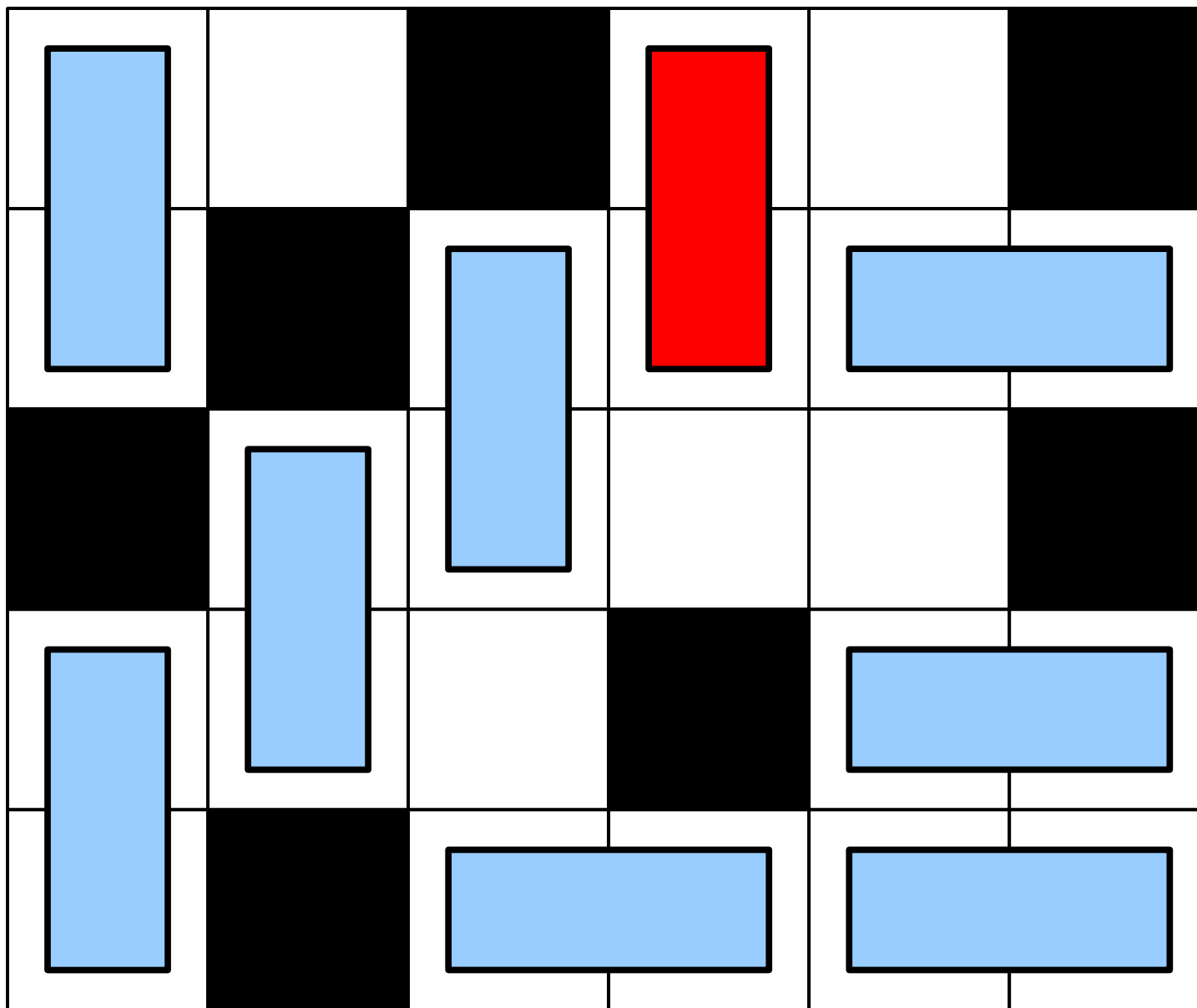
Domino Tiling



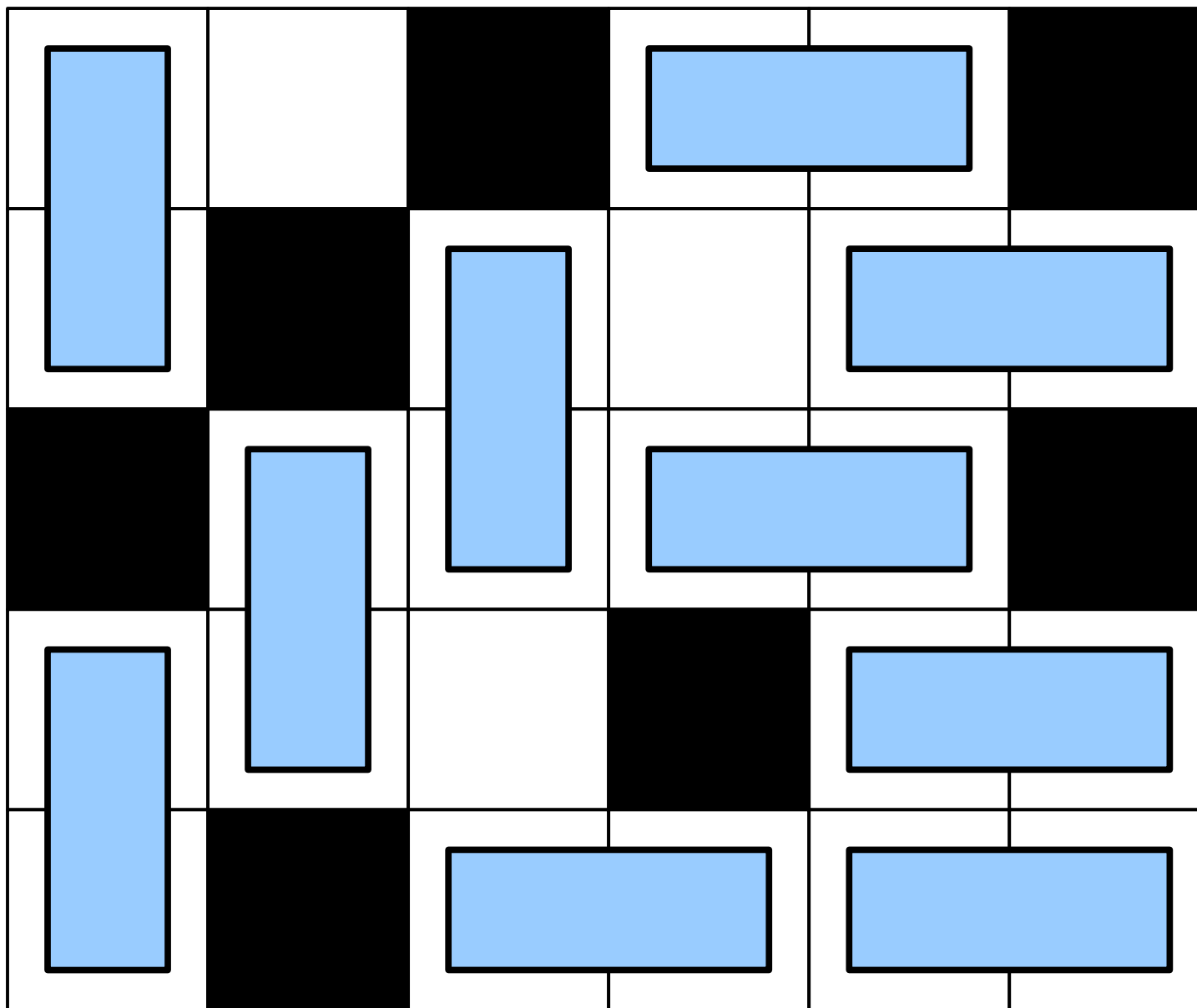
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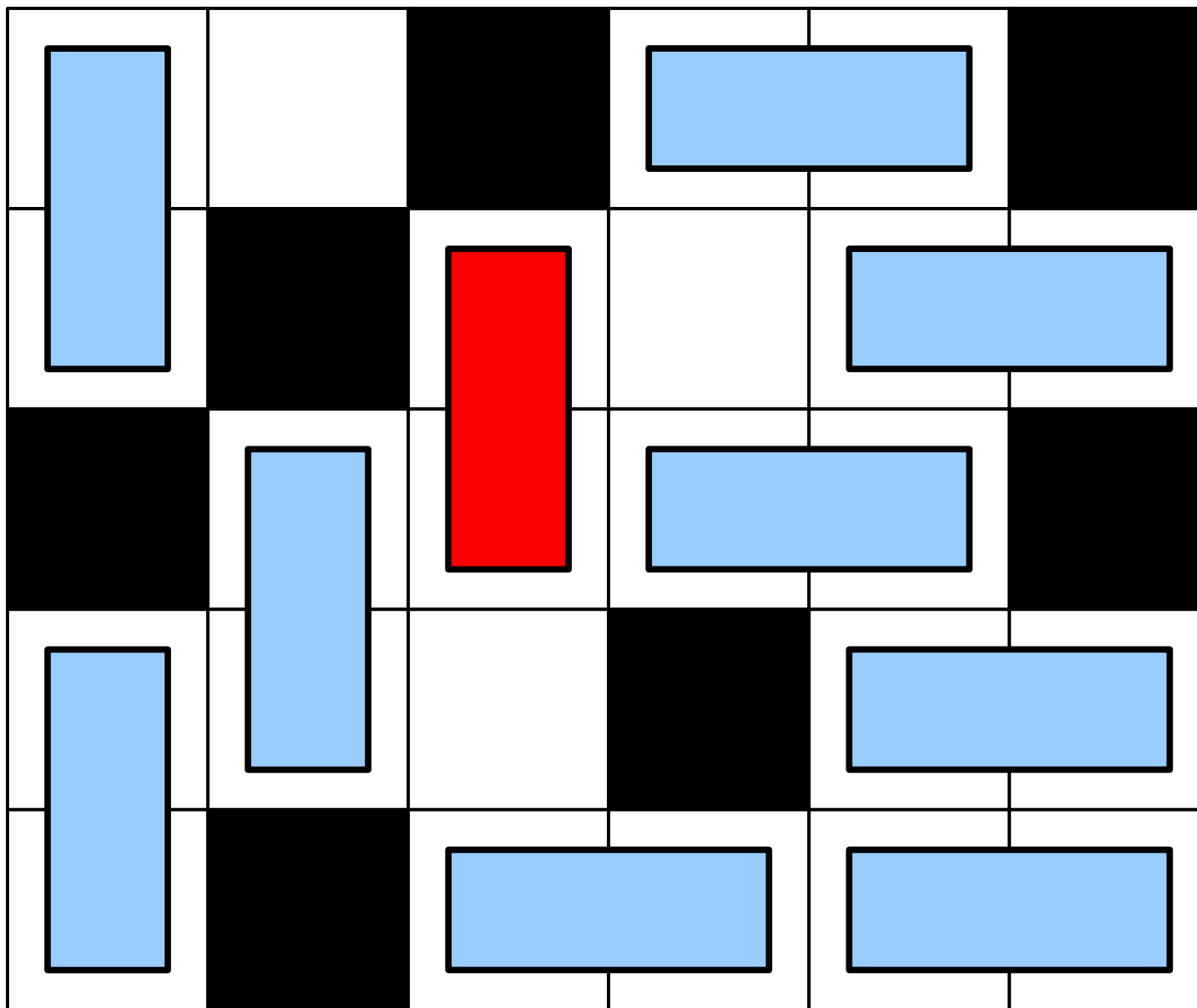
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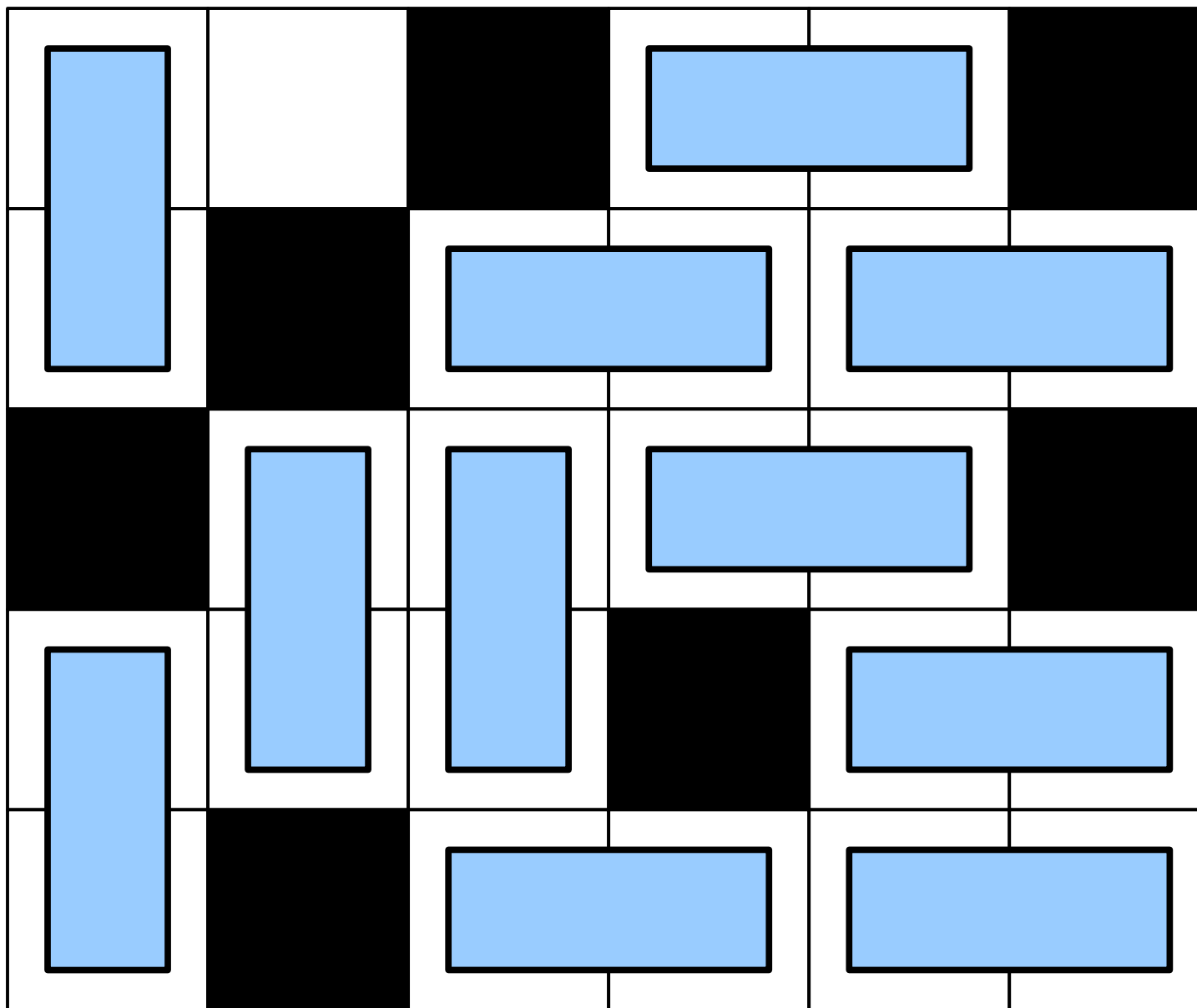
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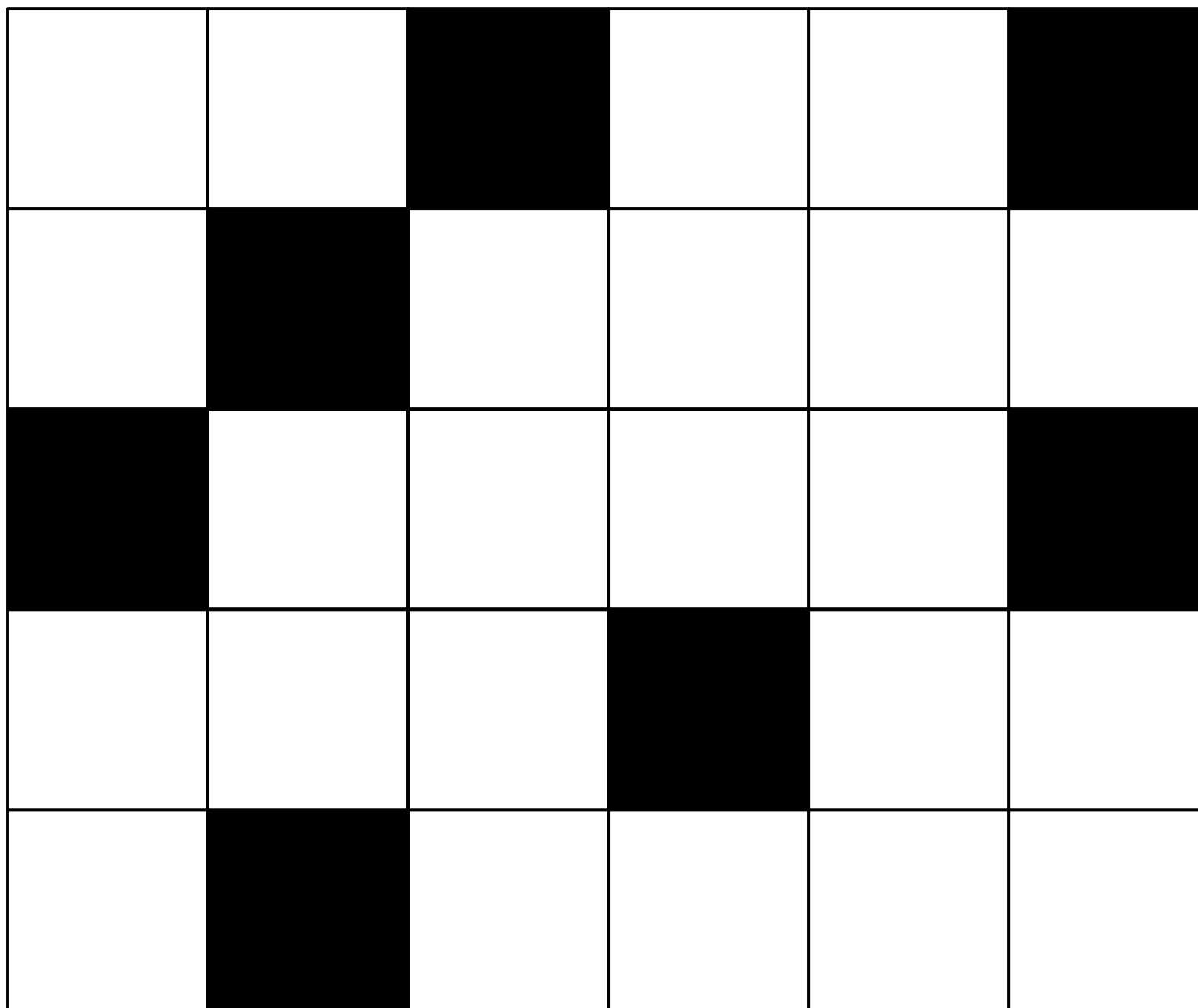
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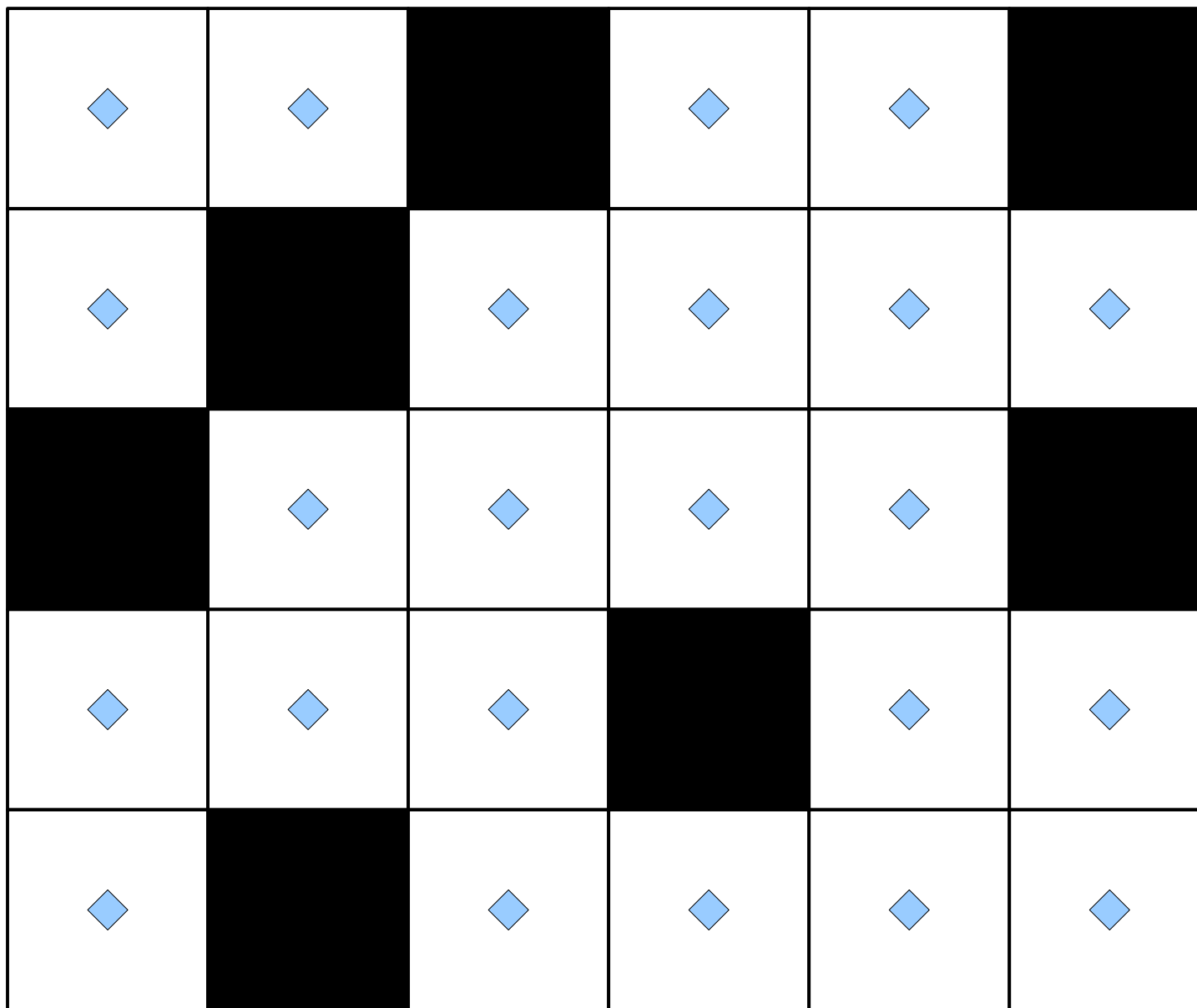
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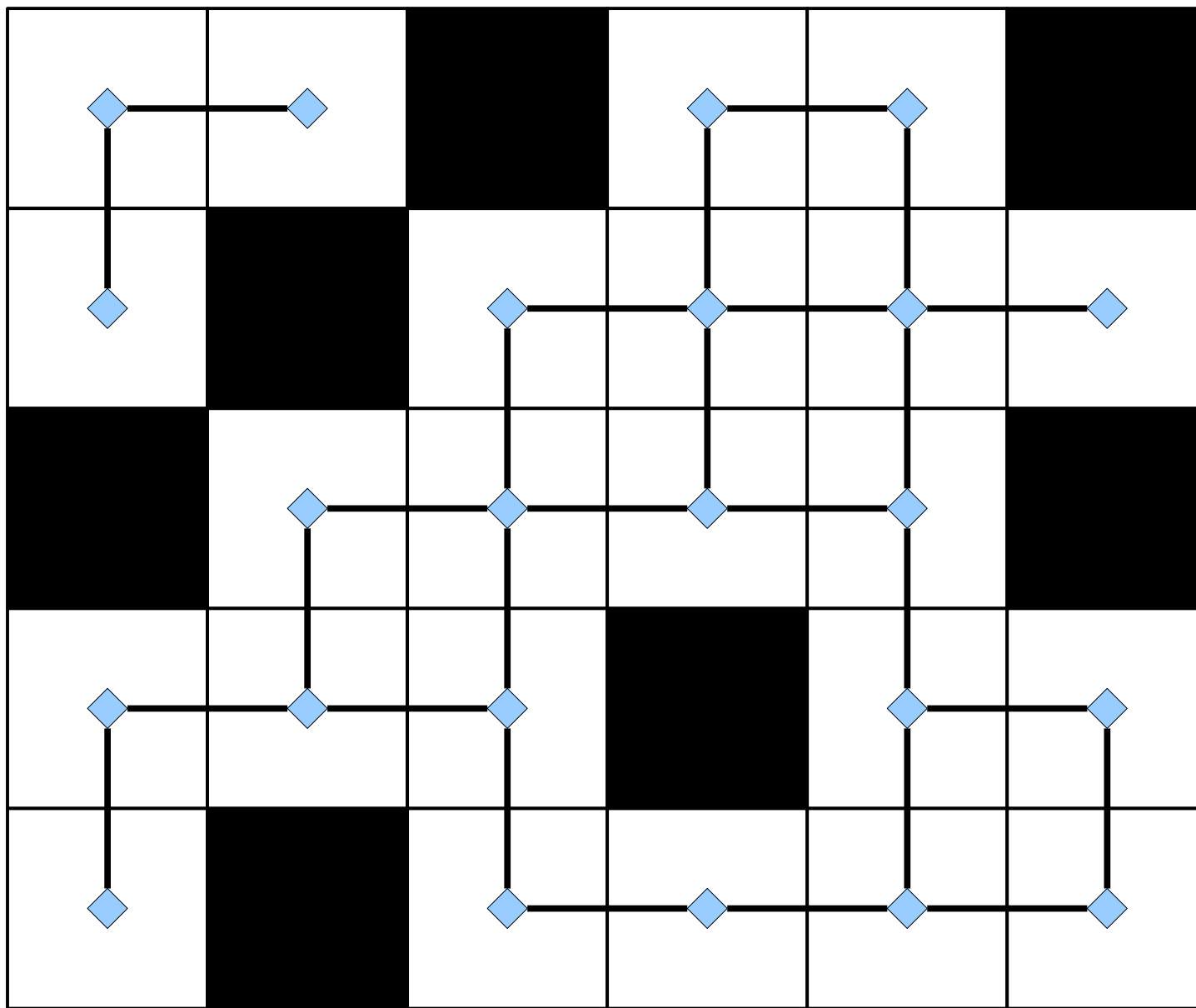
Solving Domino Tiling



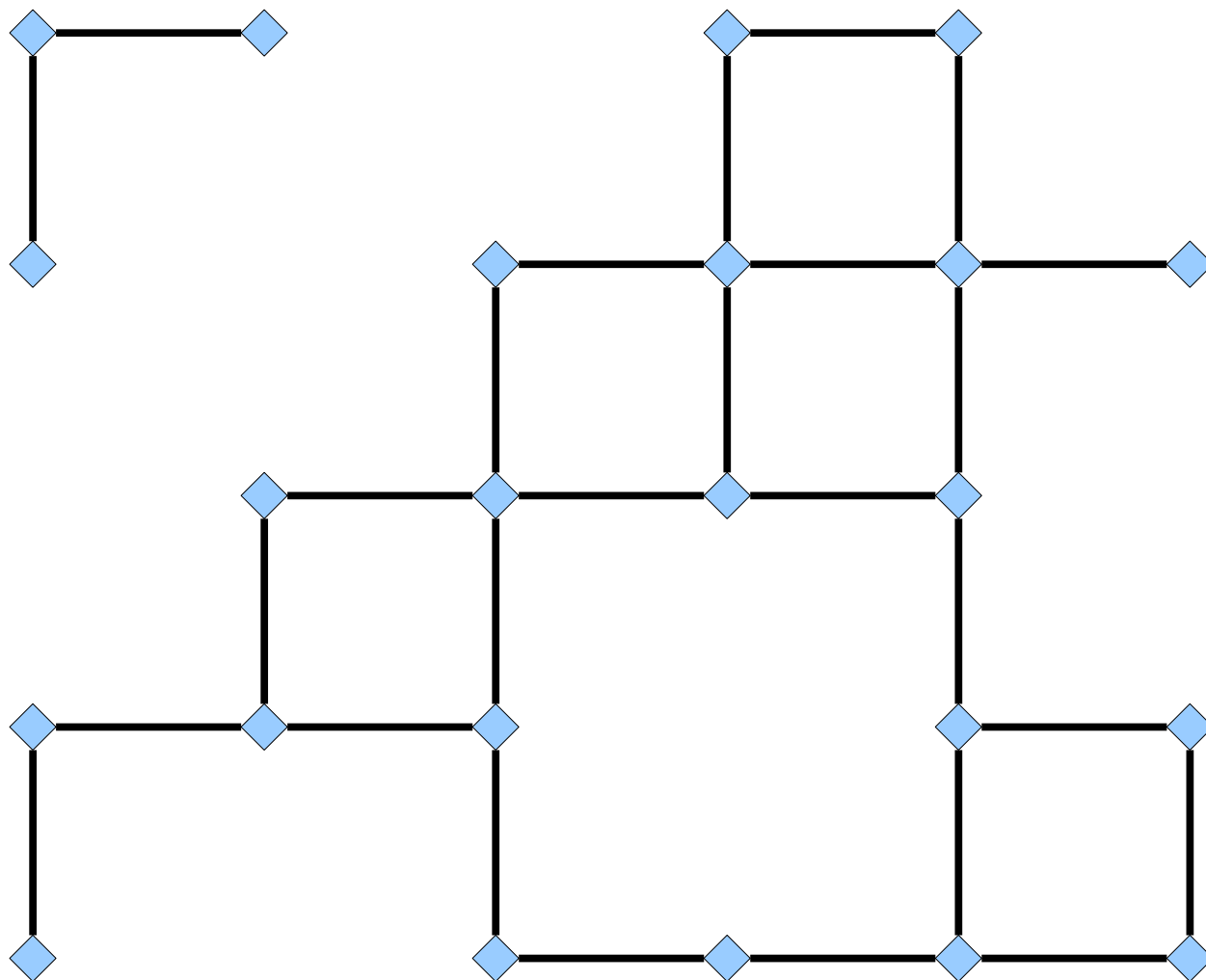
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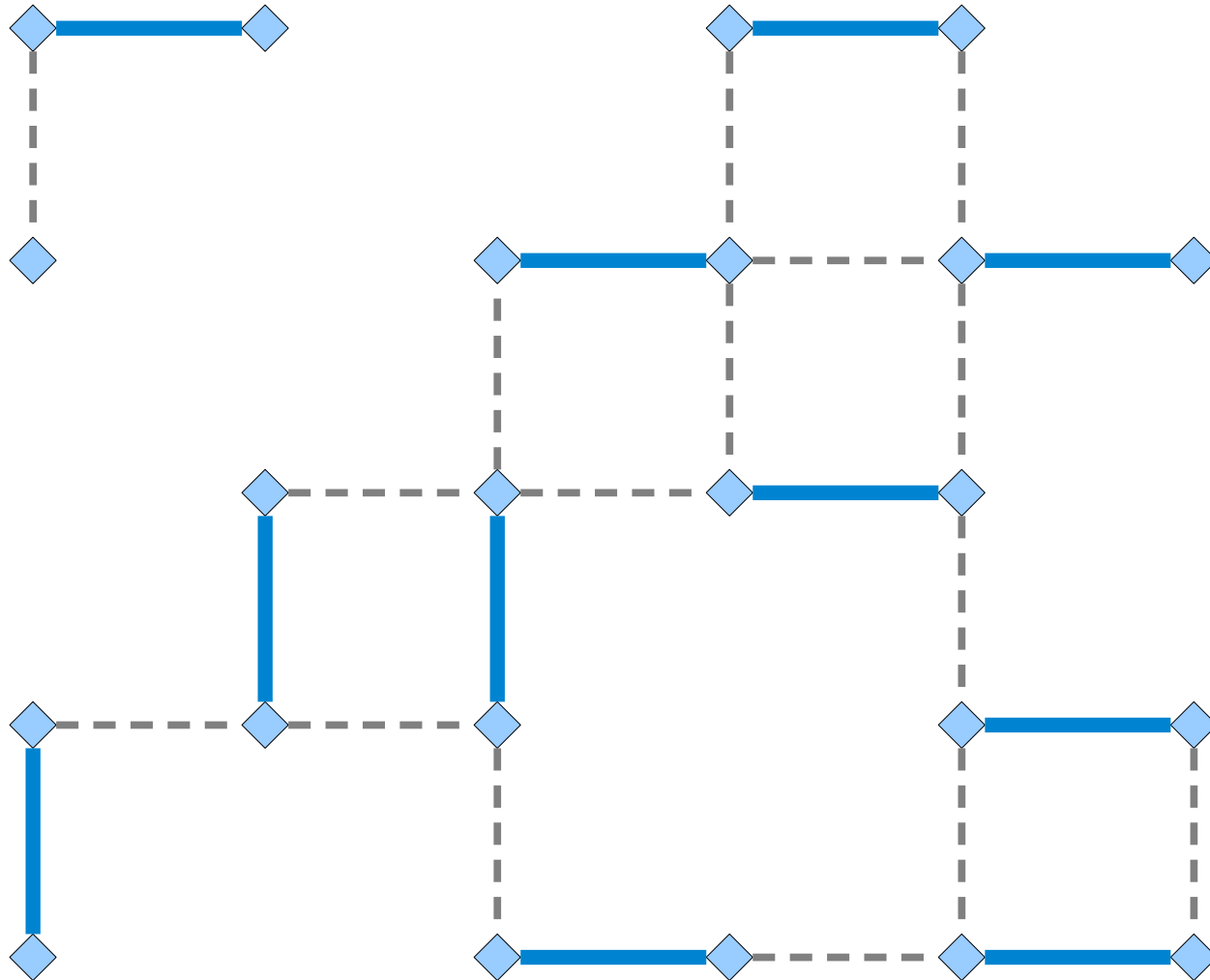
Solving Domino Tiling



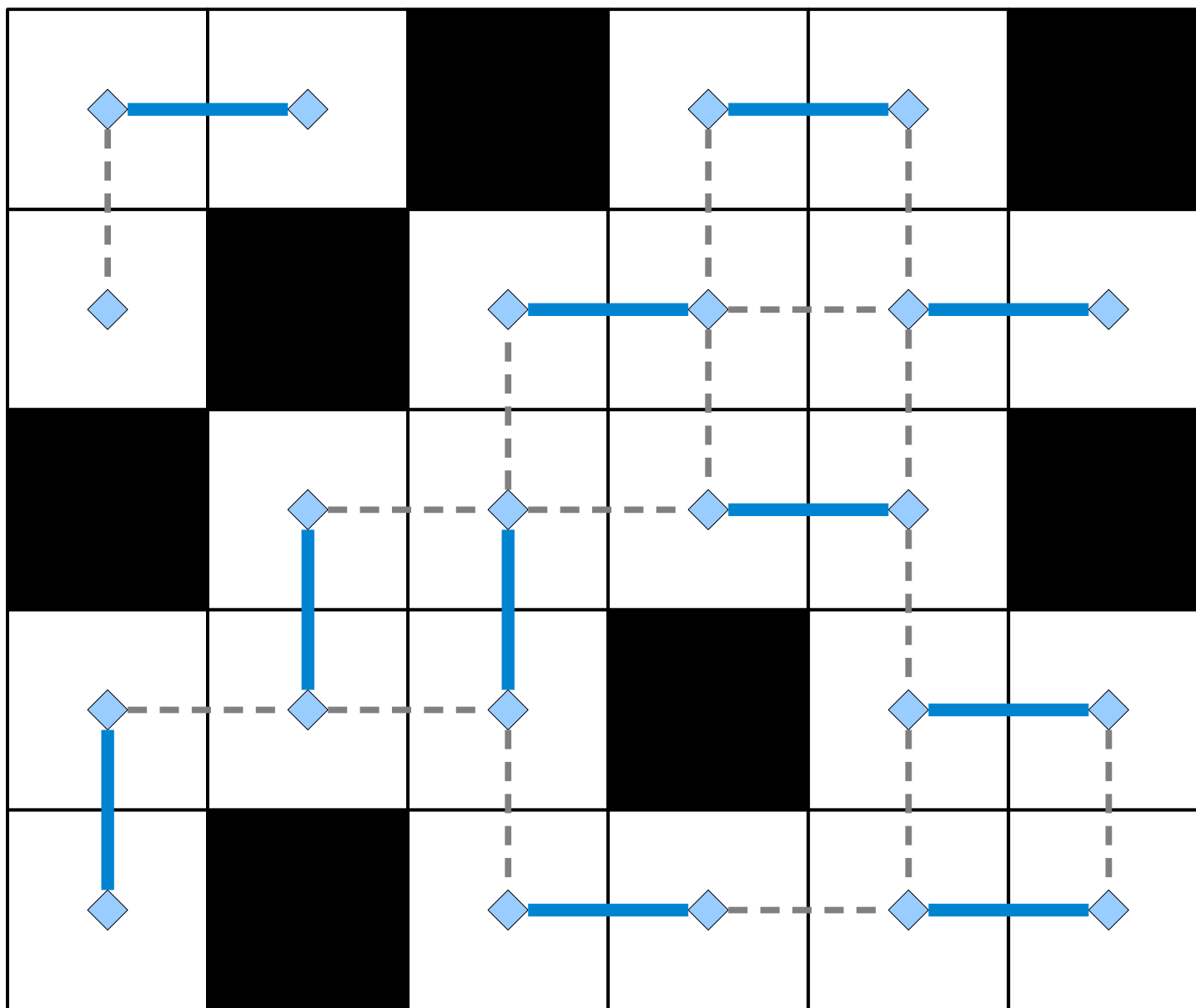
Solving Domino Tiling



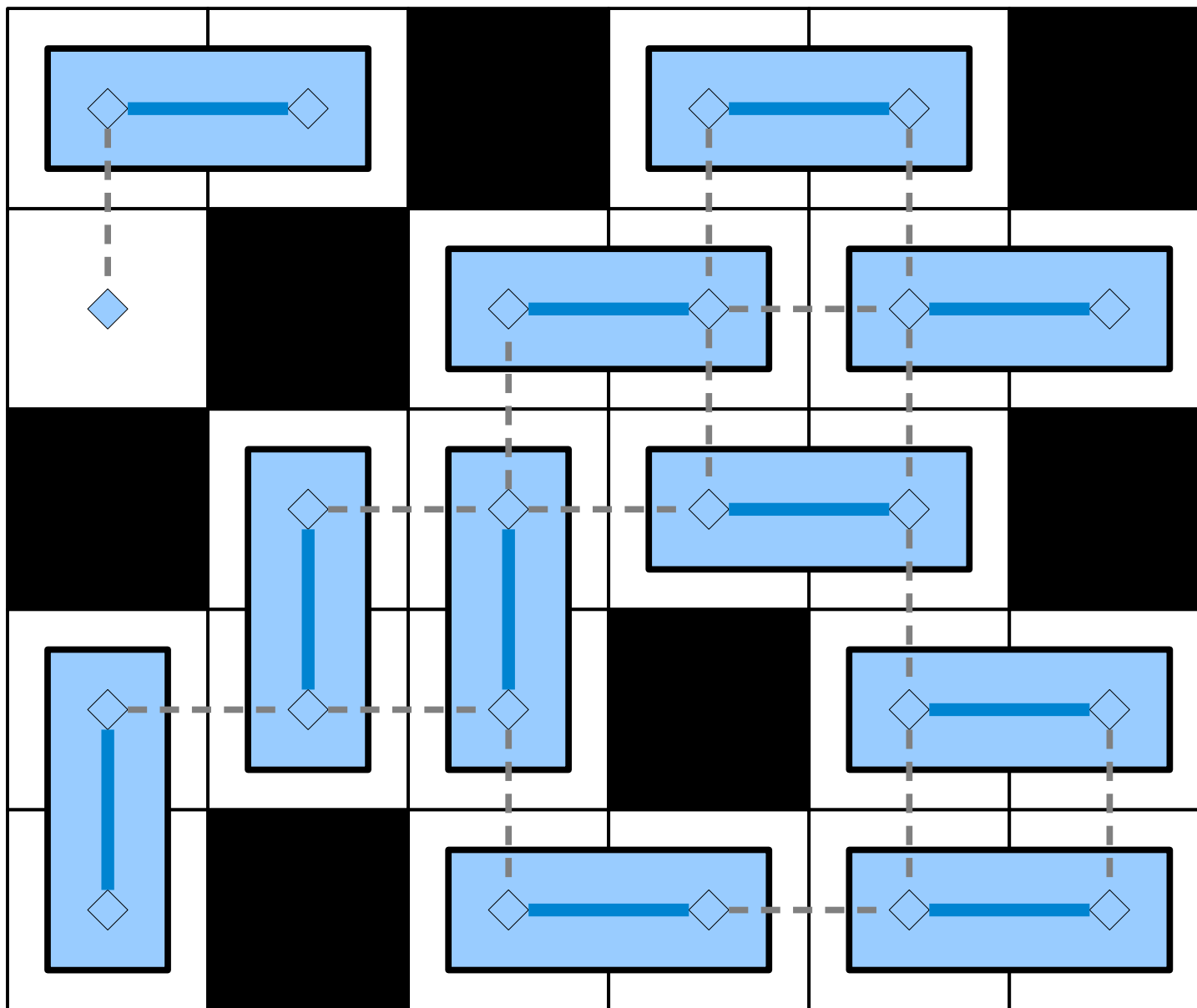
Solving Domino Tiling



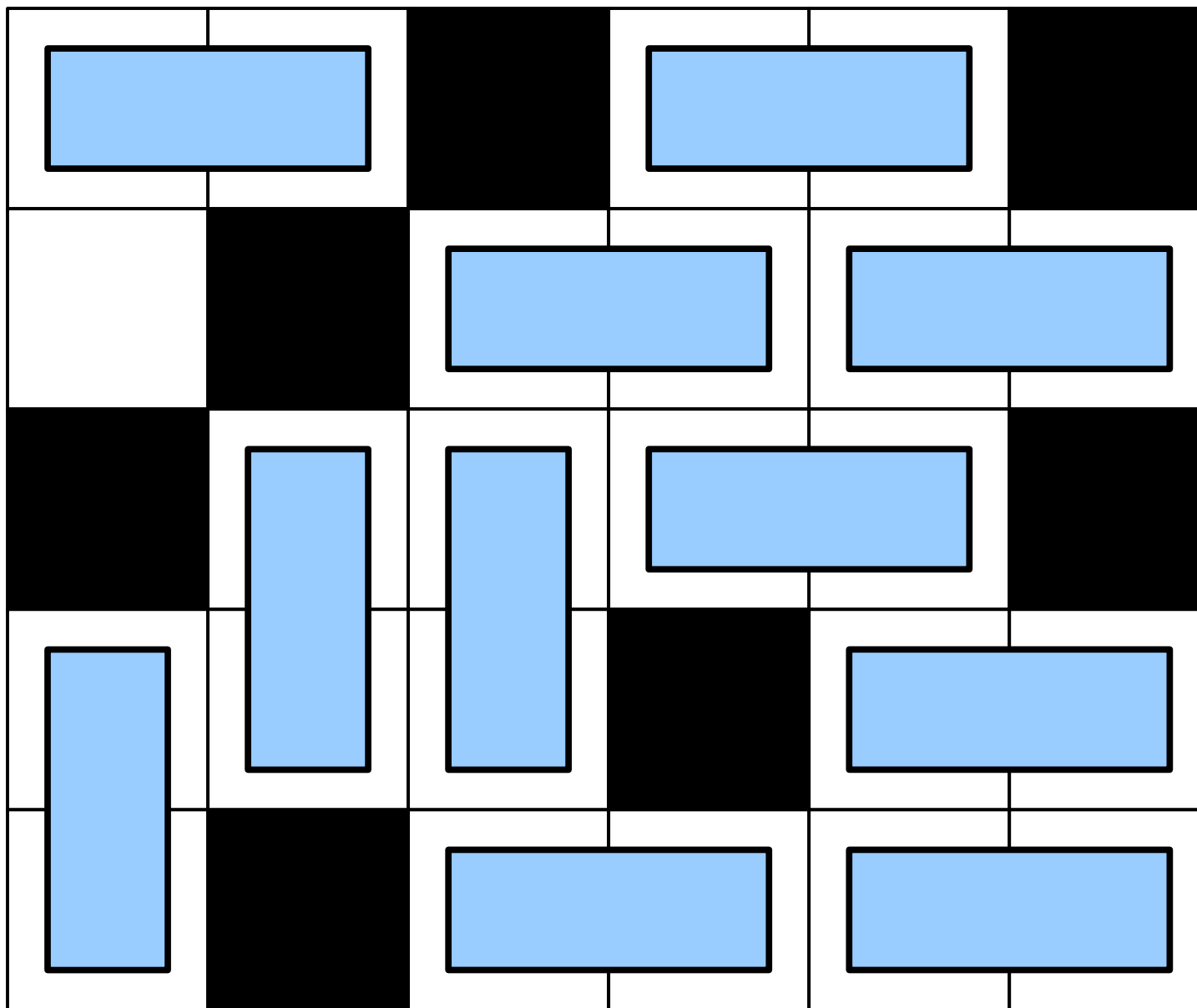
Solving Domino Tiling



Solving Domino Tiling



Solving Domino Tiling



In Pseudocode

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

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 ?
- This problem can be solved in polynomial time (use DFS or BFS).

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

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S-Video to DVI

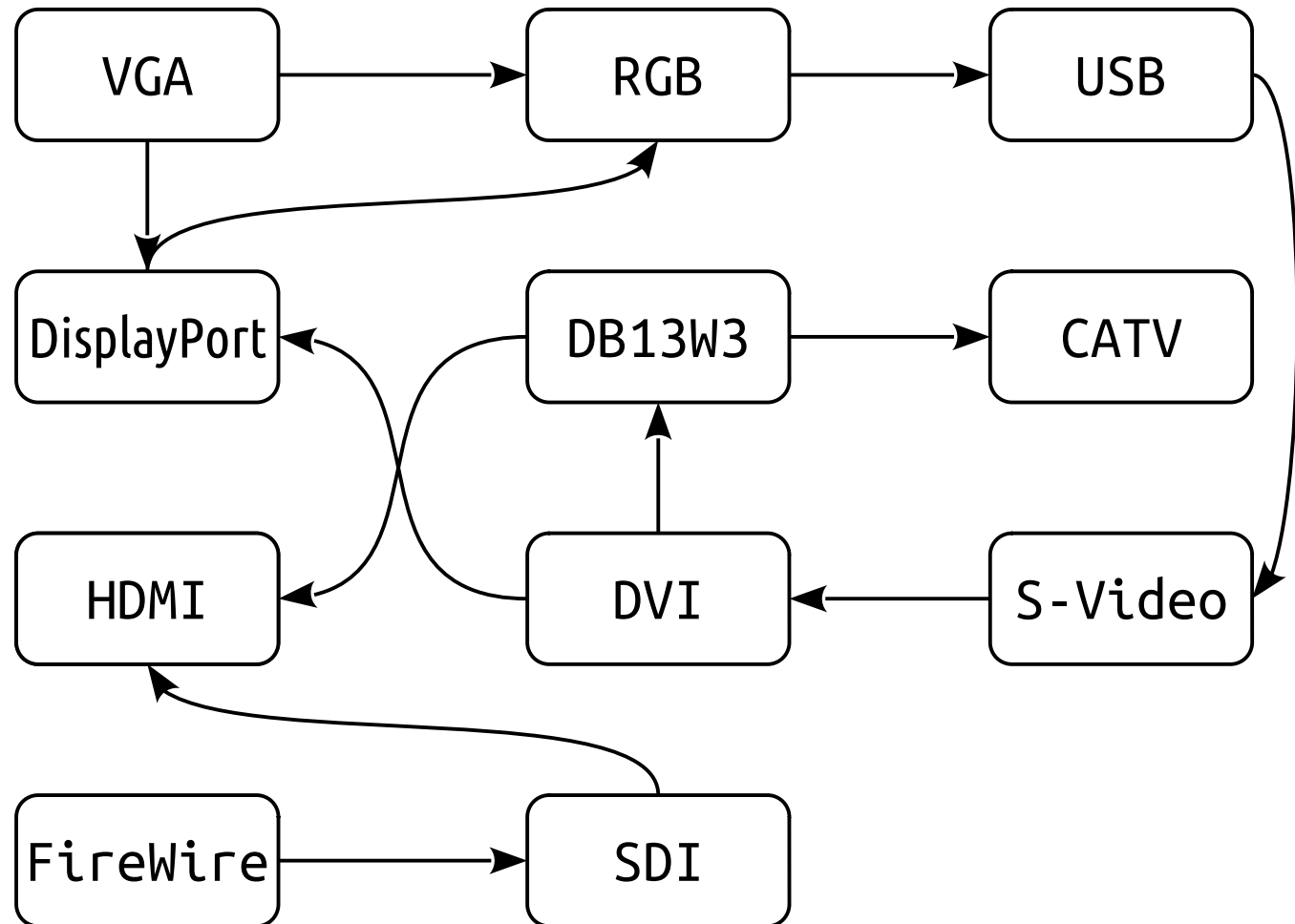
FireWire to SDI

VGA to RGB

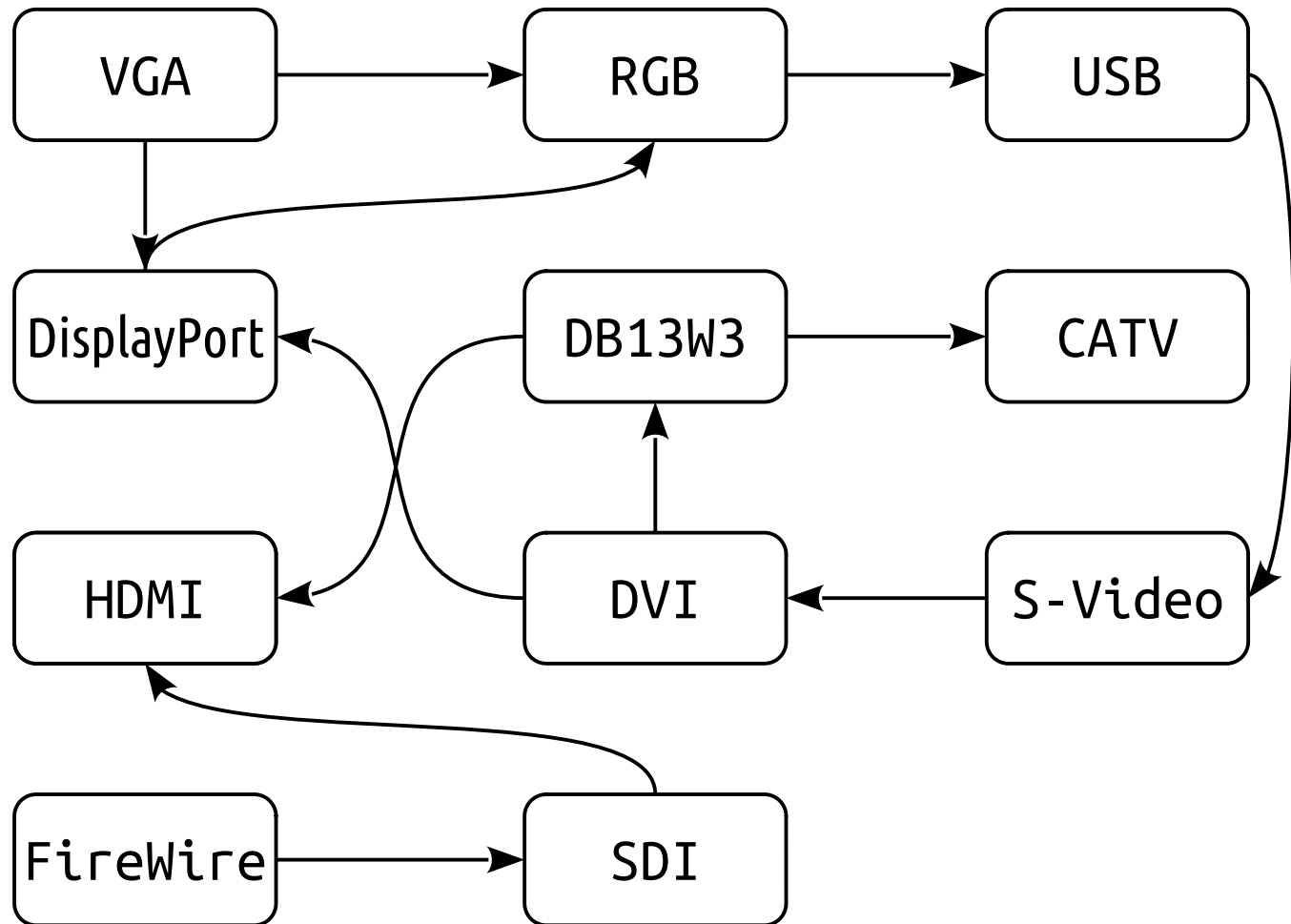
DVI to DisplayPort

USB to S-Video

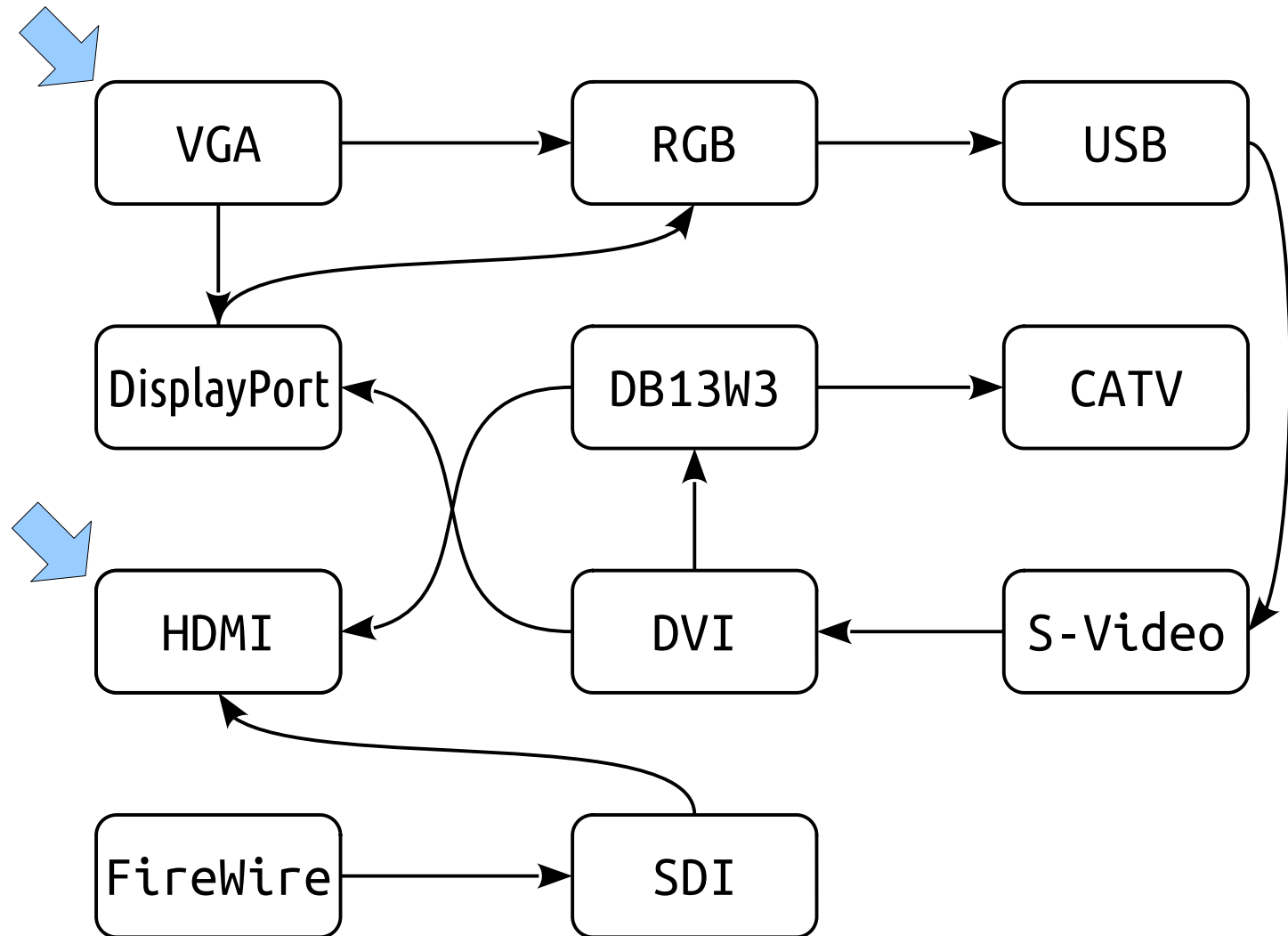
SDI to HDMI



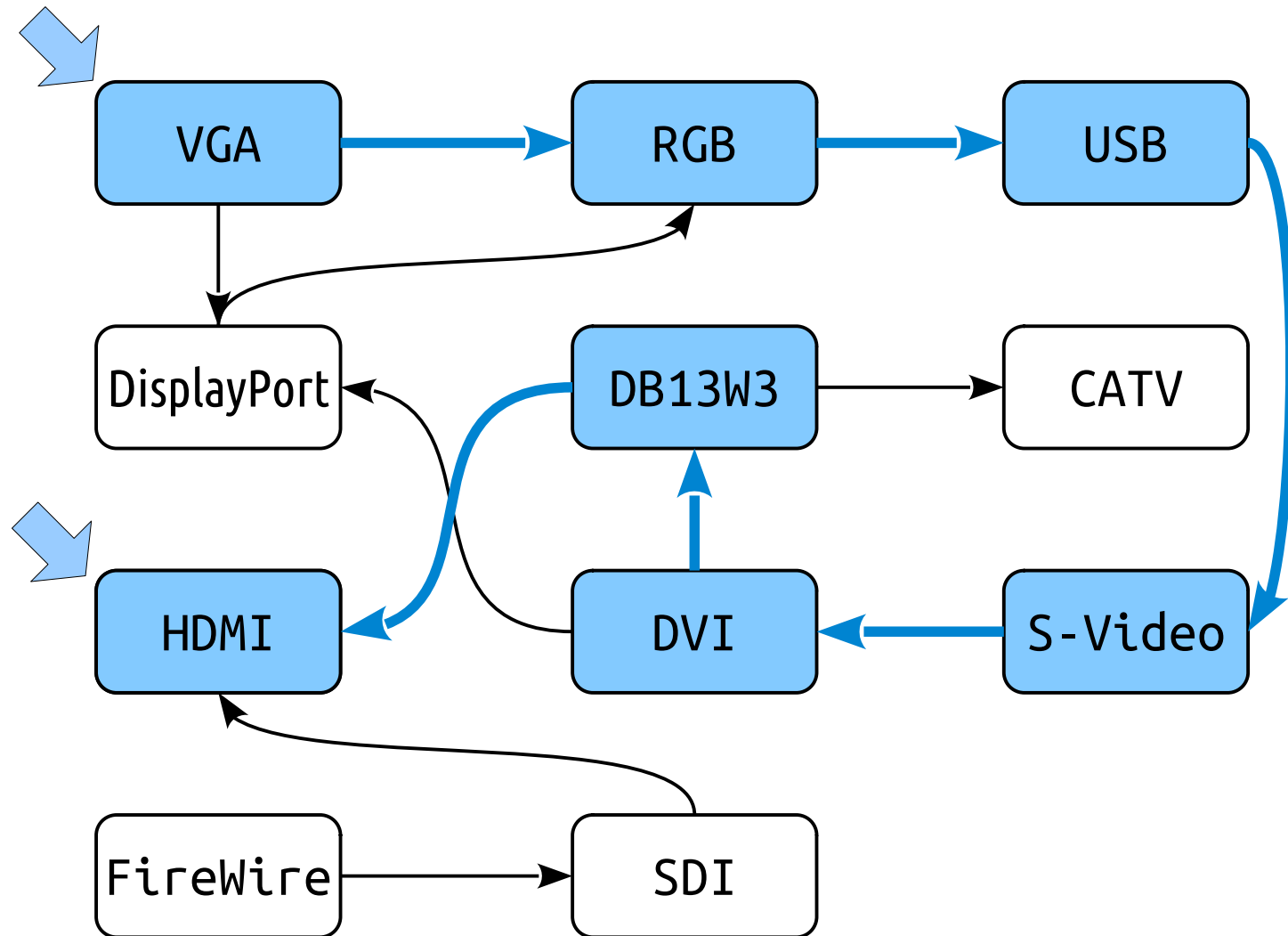
Converter Conundrums



Converter Conundrums



Converter Conundrums



Converter Conundrums

Connectors

RGB to USB

VGA to DisplayPort

DB13W3 to CATV

DisplayPort to RGB

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S-Video to DVI

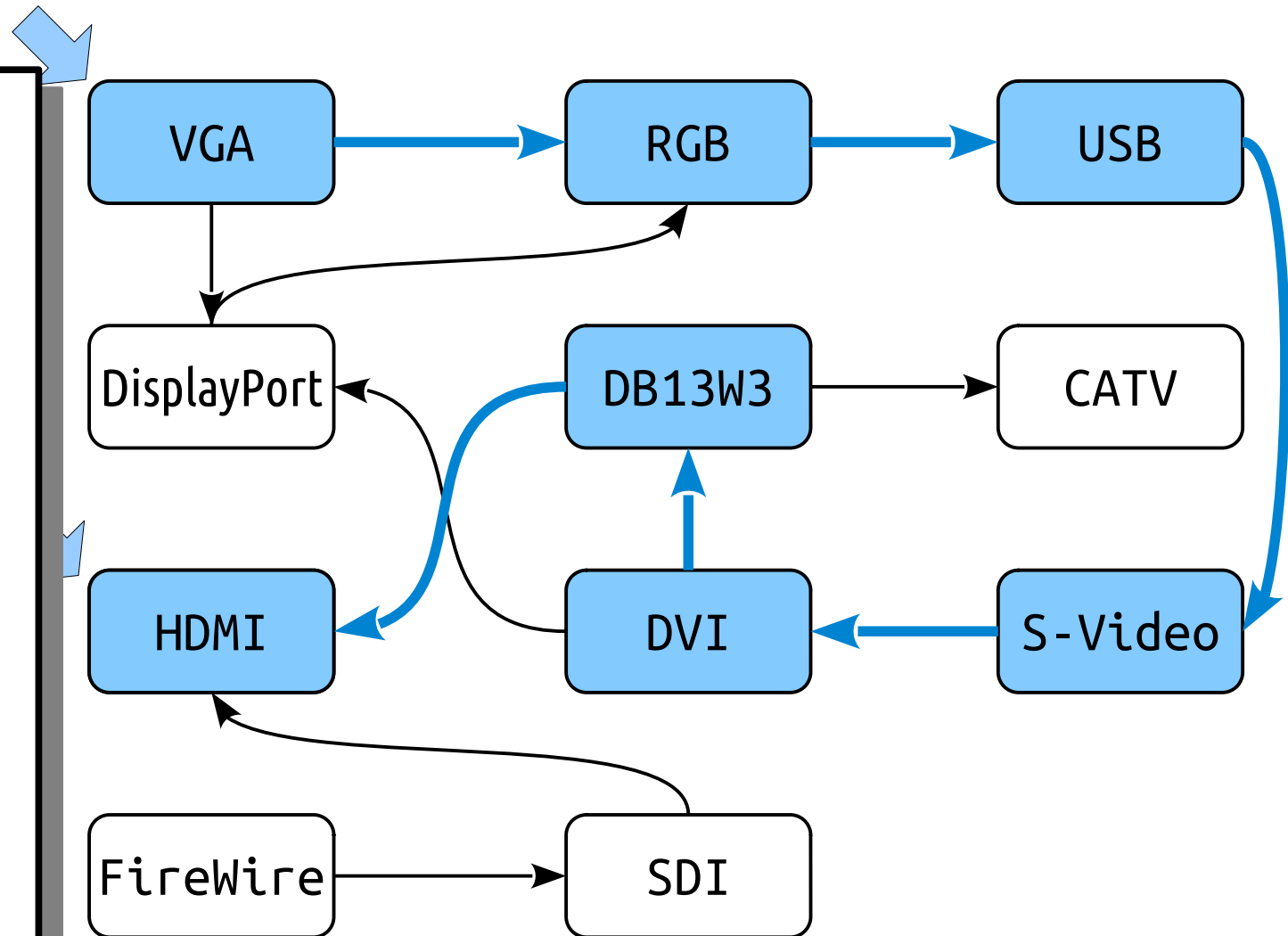
FireWire to SDI

VGA to RGB

DVI to DisplayPort

USB to S-Video

SDI to HDMI



In Pseudocode

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

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(translate(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(translate(input));  
}
```

- 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 .***

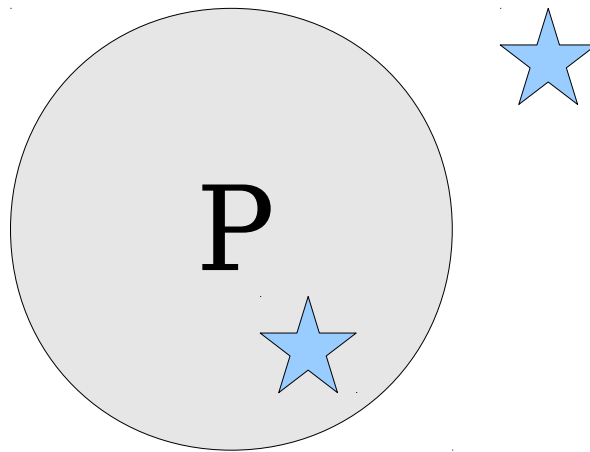
* Assuming that `translate` runs in polynomial time.

```
bool solveProblemA(string input) {  
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}
```

- This is a powerful general problem-solving technique. You'll see it a lot in CS161.

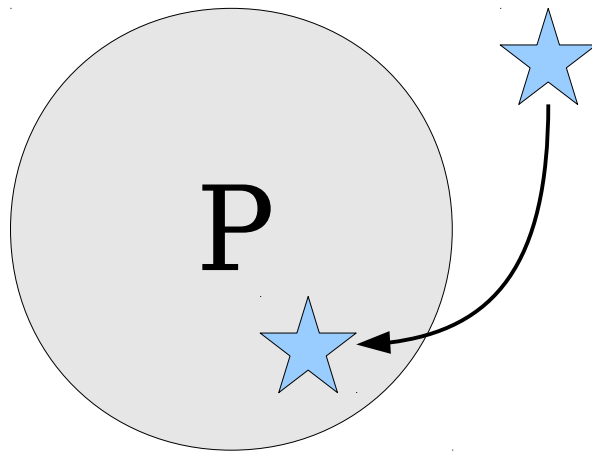
Polynomial-Time Reductions

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



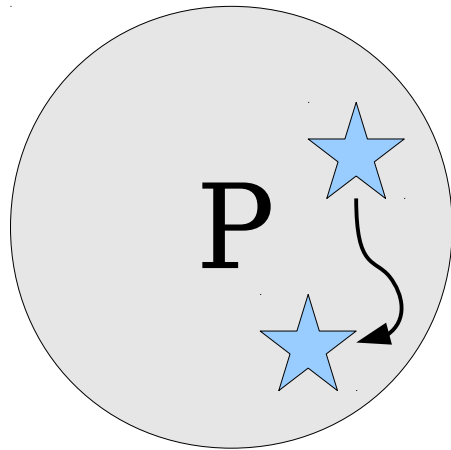
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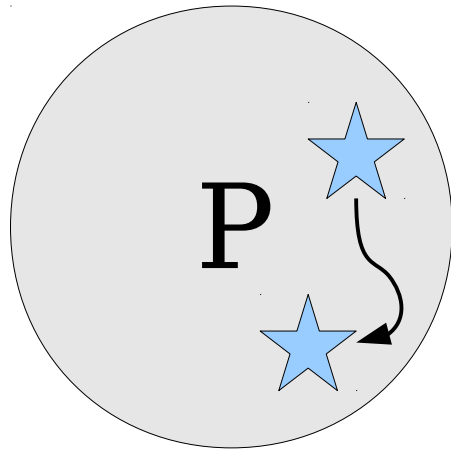
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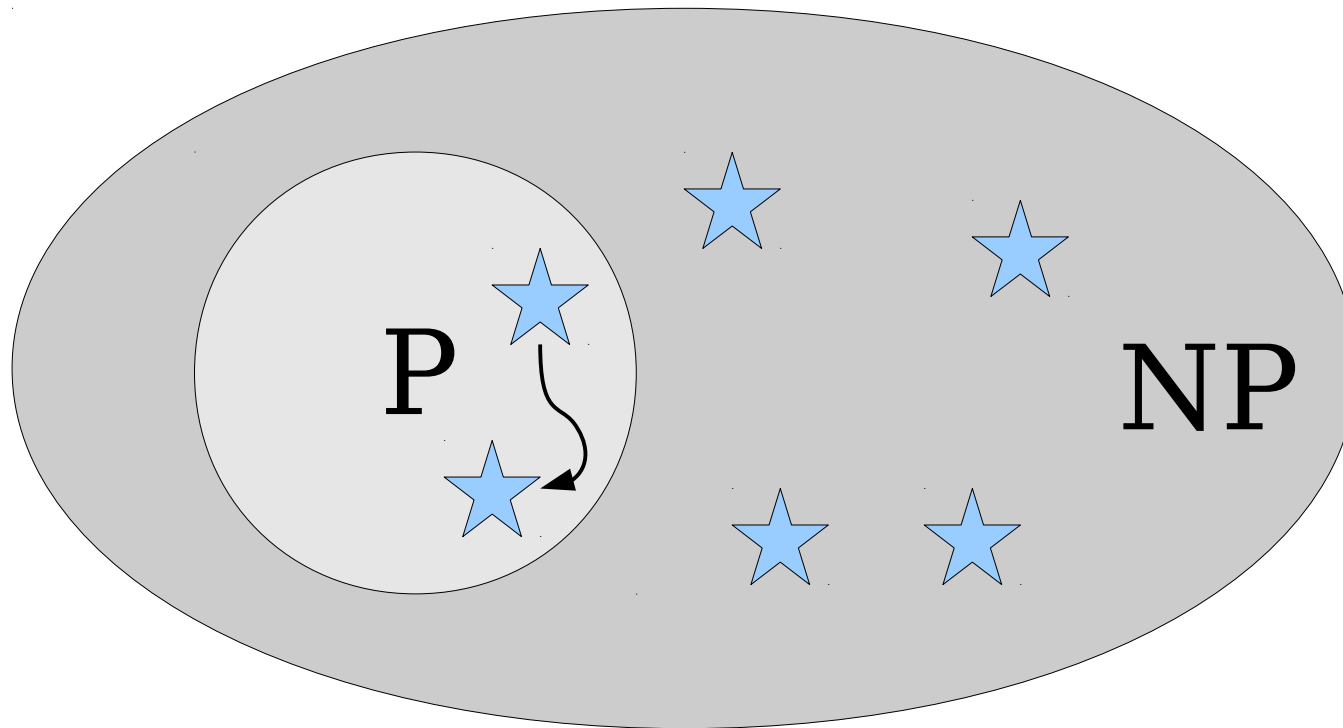
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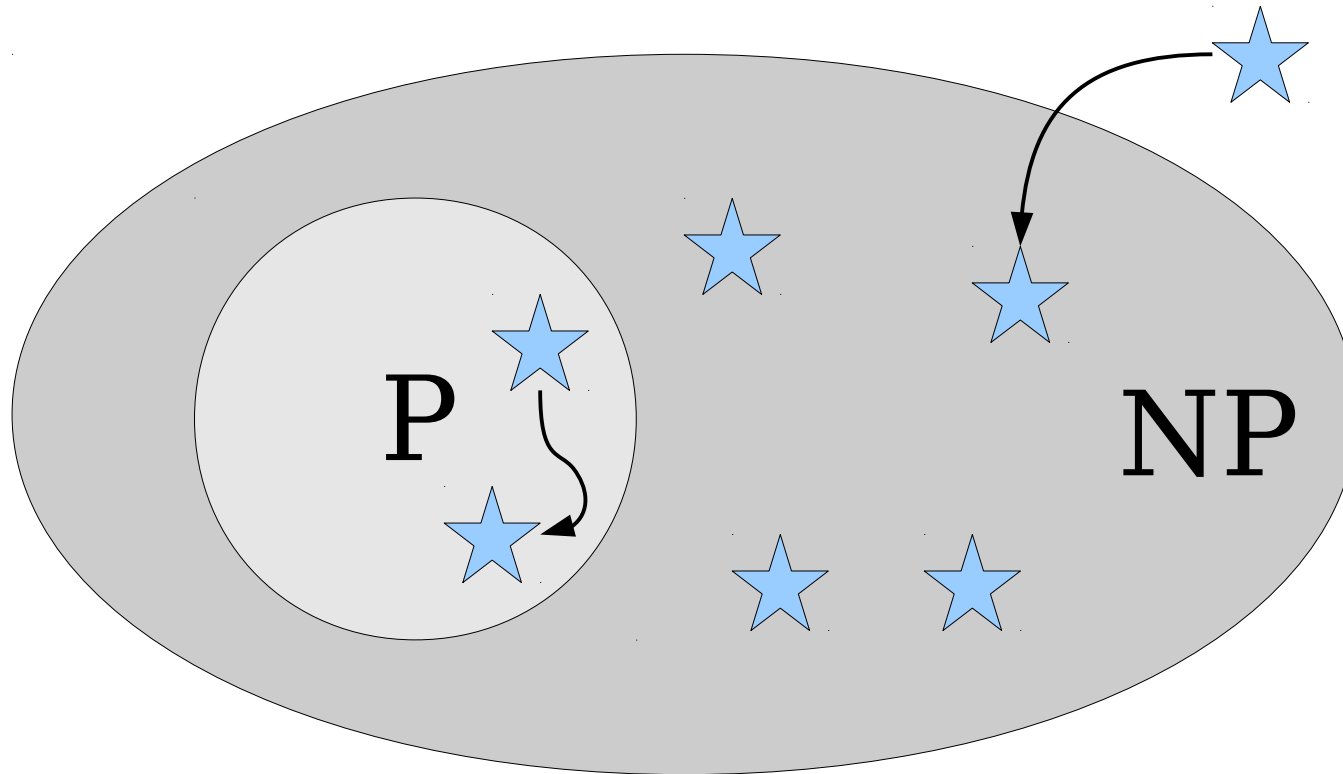
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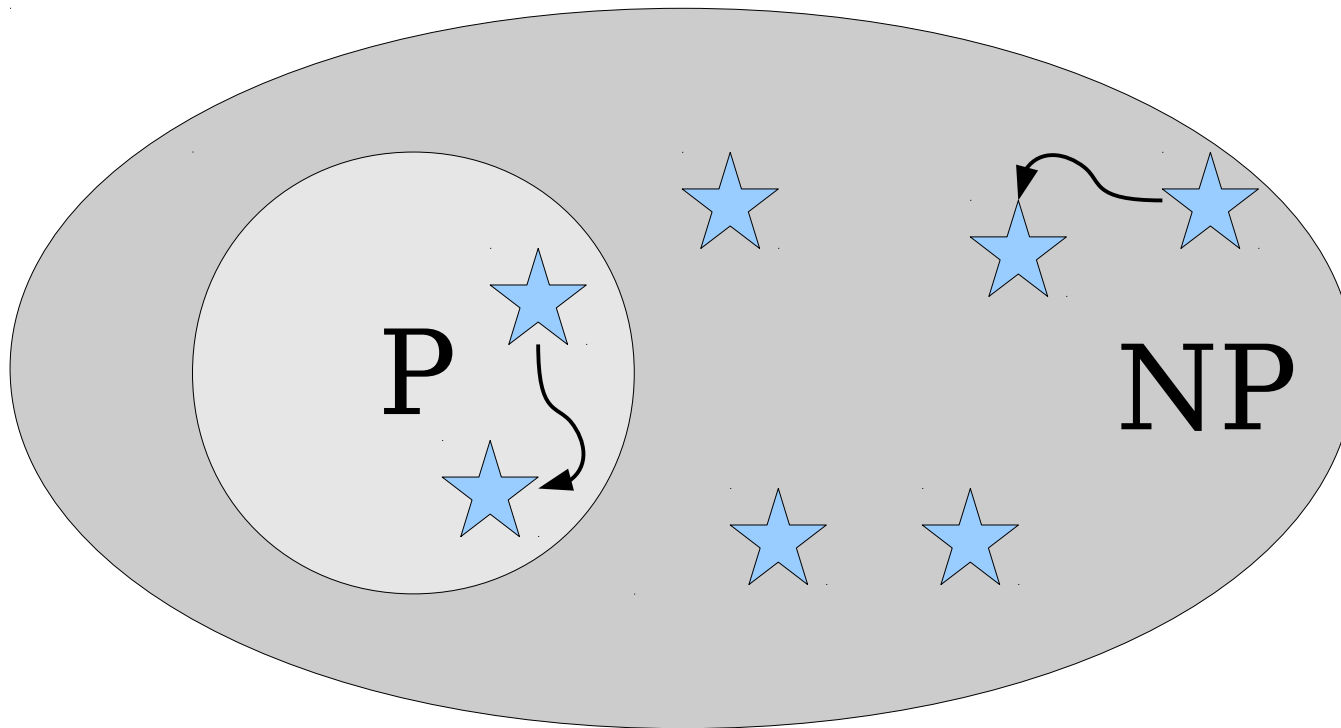
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Polynomial-Time Reductions

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



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

What else can we do with it?

Time-Out for Announcements!

Please evaluate this course on Axess.

Your feedback makes a difference.

Final Exam Logistics

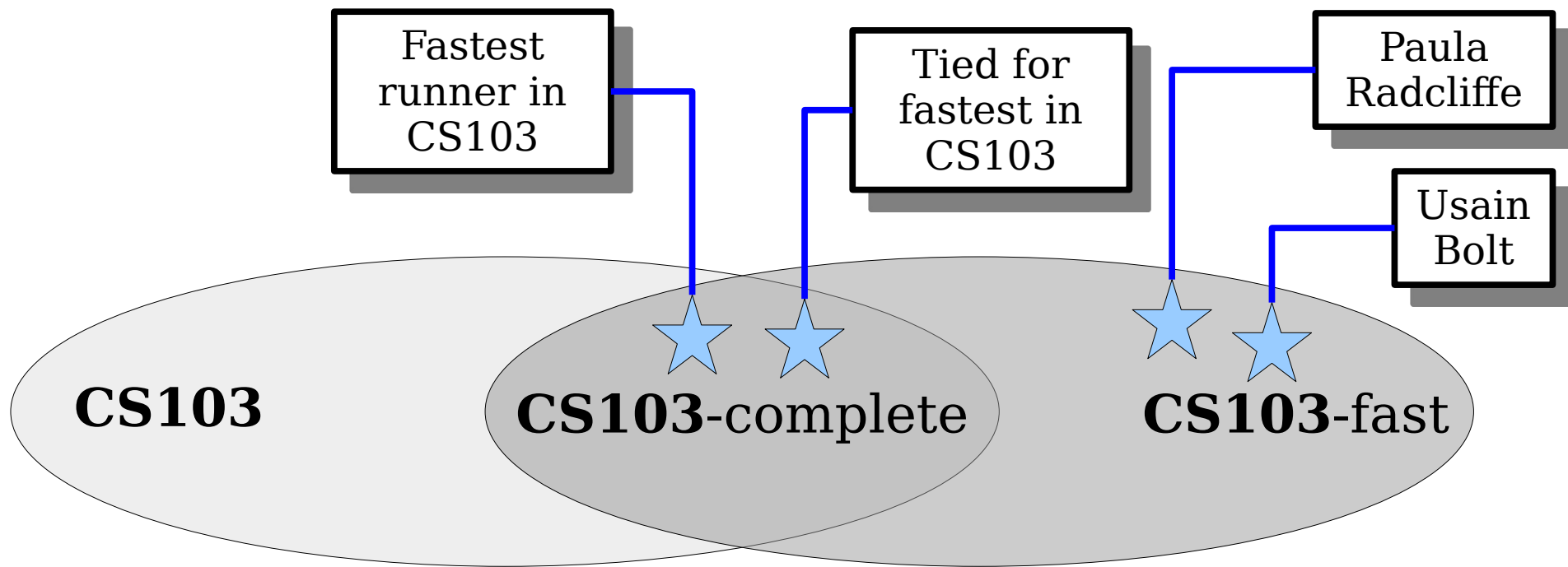
- Our final exam is a take-home exam that goes out this Friday at 2:30PM and comes due next Thursday (December 9th) at 3:30PM.
- Like the midterms, you can work on the exam for any amount of time in that period.
- Like the midterms, the exam is open-book and open-note, but you cannot communicate with other humans or solicit solutions.
- Unlike the midterms, this exam is designed to take about six hours to complete and covers all topics from the course (PS1 – PS9, plus L00 – L27).

Preparing for the Final

- We've posted a gigantic list of cumulative review problems on the course website that you can use to get more practice with whatever topics you're interested in.
- Our recommendation: Look back over the exams and problem sets and redo any problems that you didn't really get the first time around.
- Keep the TAs in the loop: stop by office hours to have them review your answers and offer feedback.

Back to CS103!

An Analogy: Running Really Fast



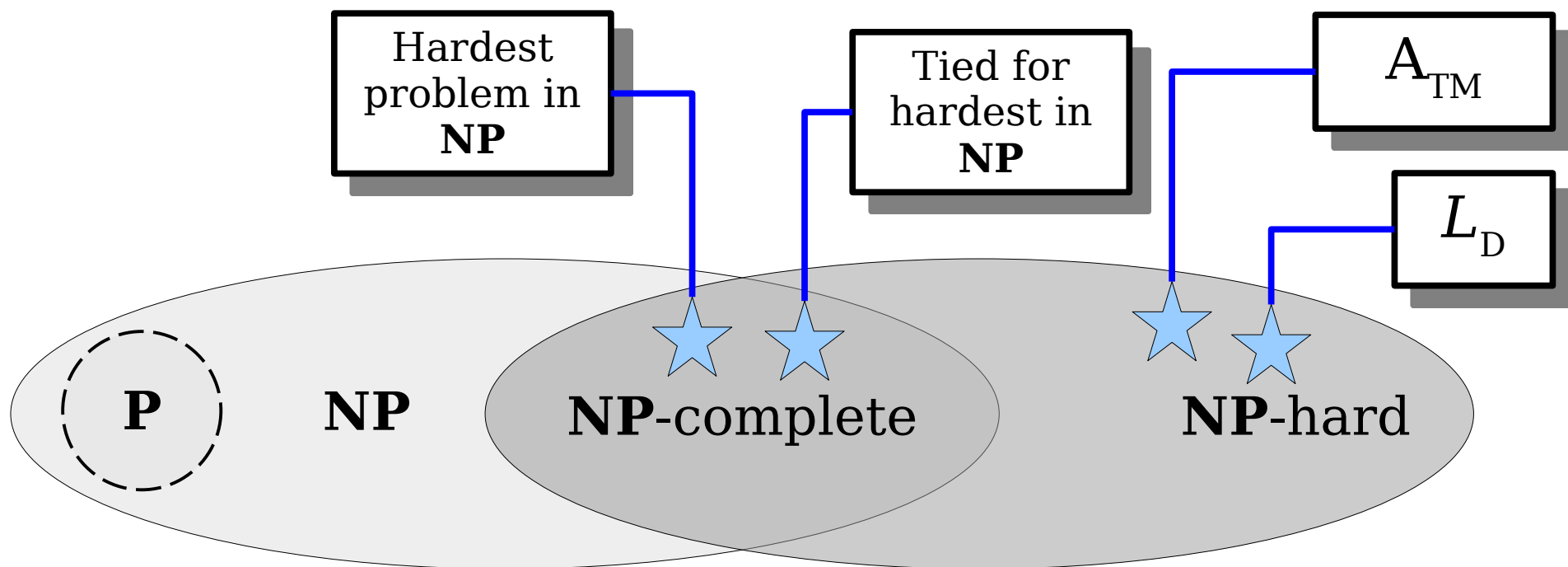
For people A and B , we say $A \leq_r B$ if A 's top running speed is at most B 's top speed.
(Intuitively: B can run at least as fast as A .)

We say that person P is **CS103-fast** if
 $\forall A \in \text{CS103}. A \leq_r P.$

(How fast are you if you're CS103-fast?)

We say that person P is **CS103-complete** if
 $P \in \text{CS103}$ and P is **CS103-fast**.

(How fast are you if you're CS103-complete?)



For languages A and B , we say $A \leq_p B$ if A reduces to B in polynomial time.

(Intuitively: B is at least as hard as A .)

We say that a language L is **NP-hard** if

$$\forall A \in \mathbf{NP}. A \leq_p L.$$

(How hard is a problem that's NP-hard?)

We say that a language L is **NP-complete** if

$$L \in \mathbf{NP} \text{ and } L \text{ is NP-hard.}$$

(How hard is a problem that's NP-complete?)

Intuition: The **NP**-complete problems are the hardest problems in **NP**.

If we can determine how hard those problems are, it would tell us a lot about the **P** $\stackrel{?}{=}$ **NP** question.

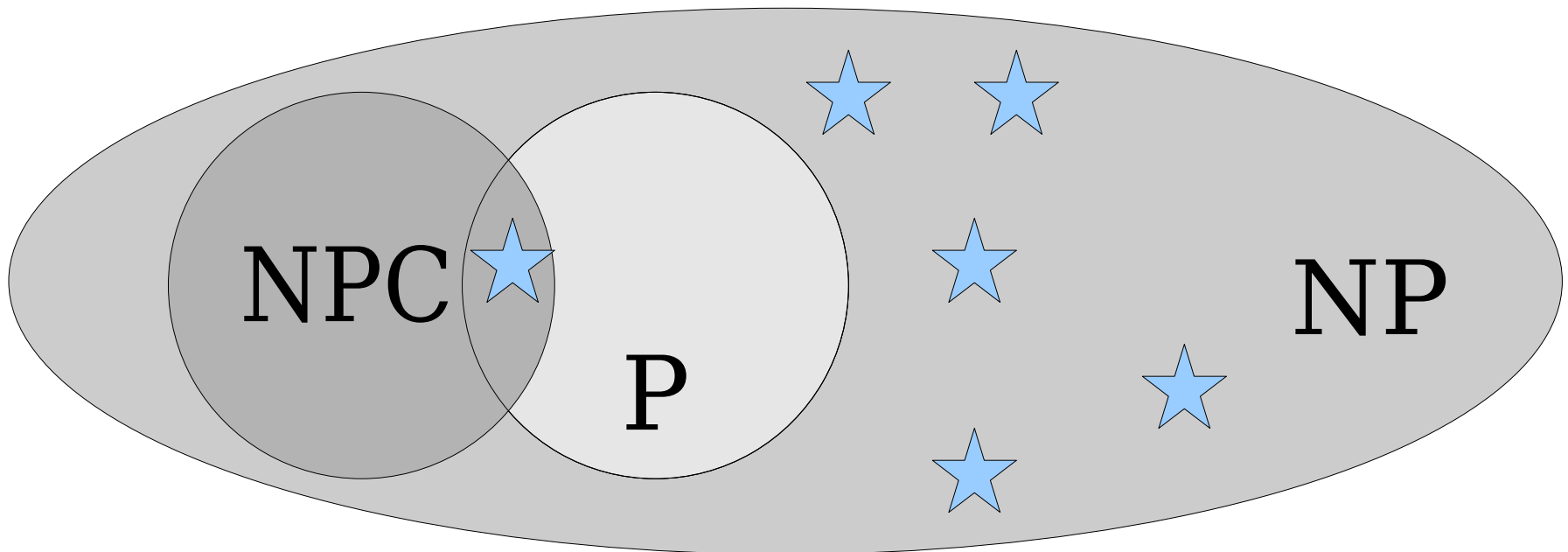
The Tantalizing Truth

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

Intuition: This means the hardest problems in **NP** aren't actually that hard. We can solve them in polynomial time. So that means we can solve all problems in **NP** in polynomial time.

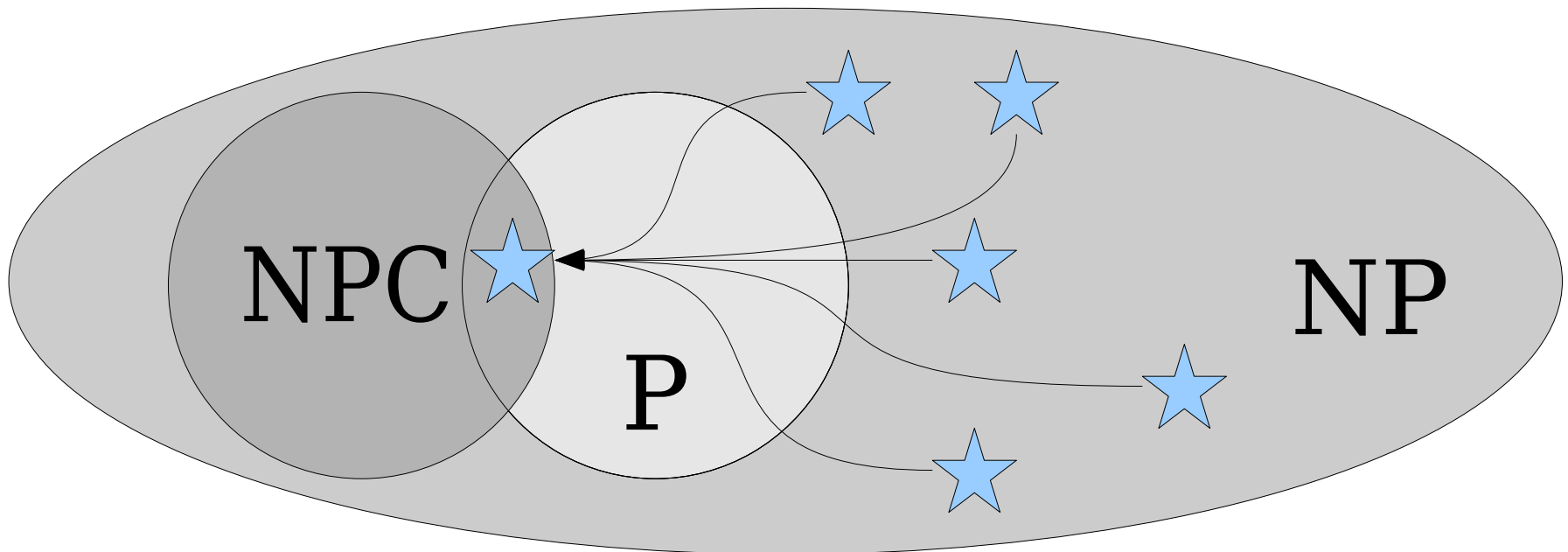
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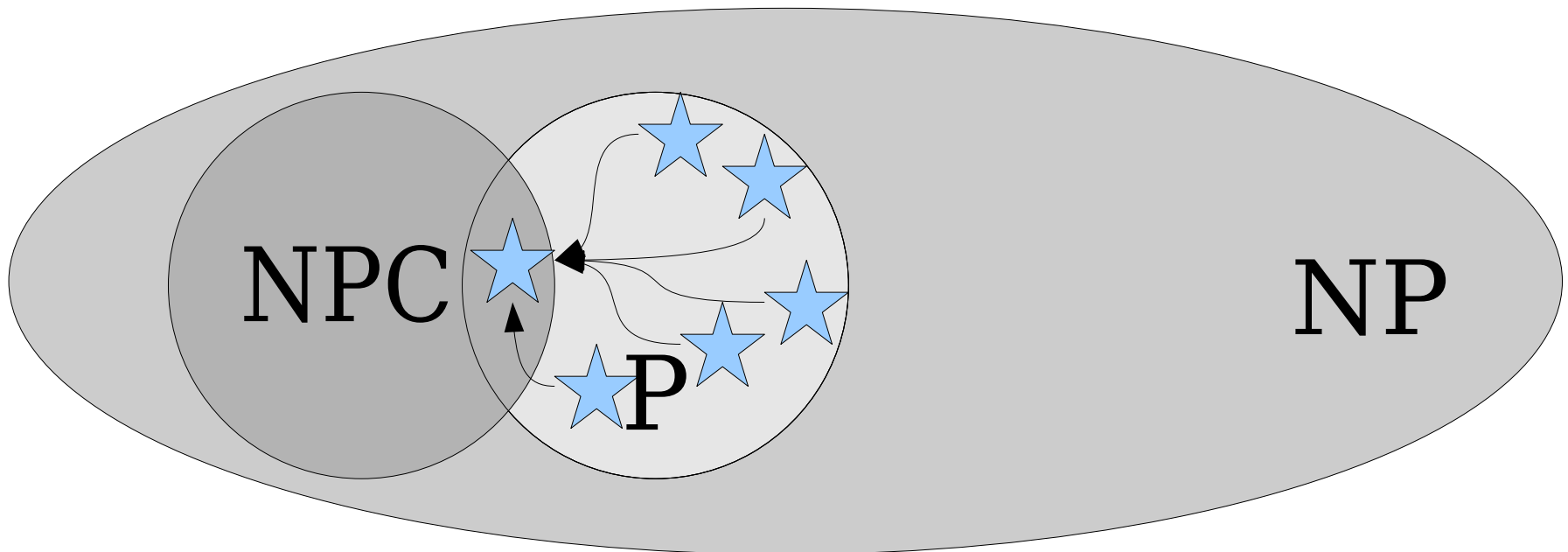
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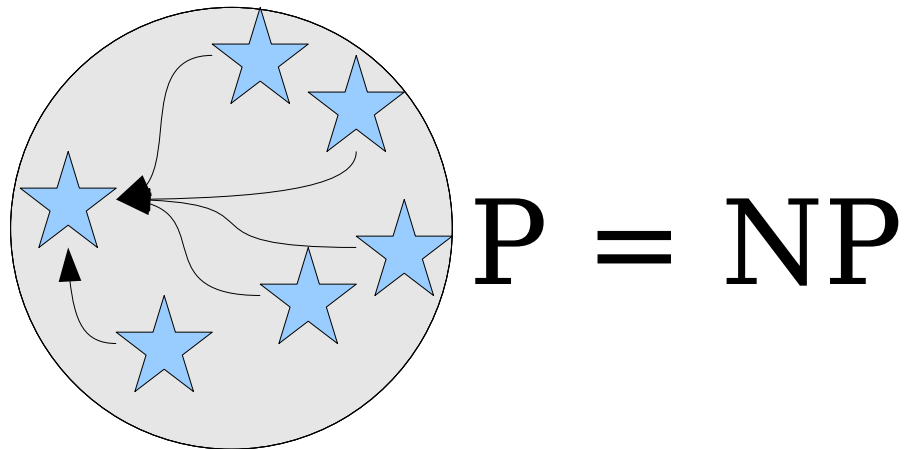
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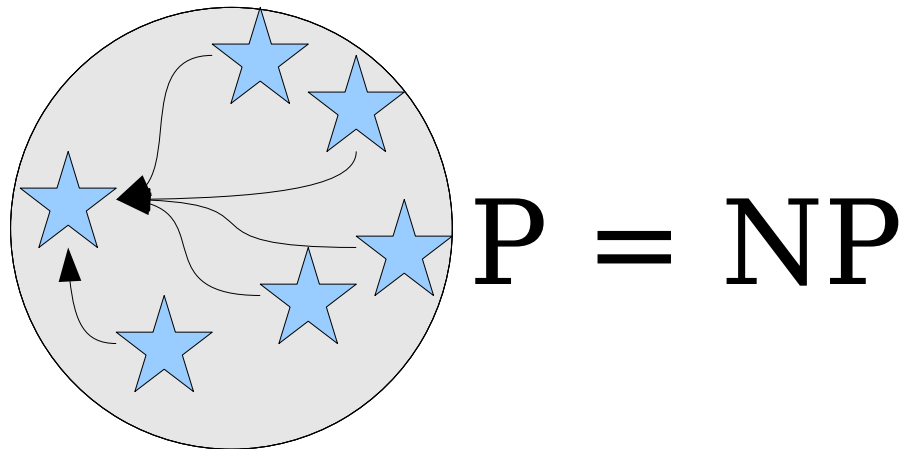
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The Tantalizing Truth

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

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

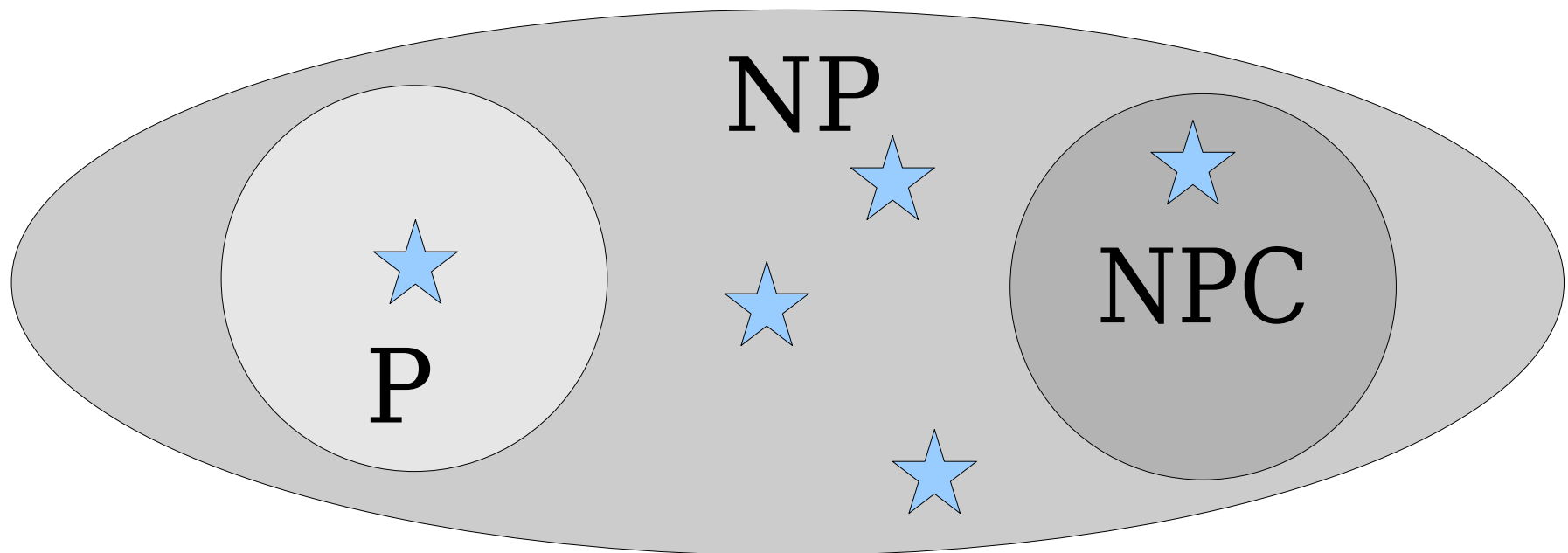
Theorem: If *any* **NP**-complete language is not in **P**, then $\mathbf{P} \neq \mathbf{NP}$.

Intuition: This means the hardest problems in **NP** are so hard that they can't be solved in polynomial time. So the hardest problems in **NP** aren't in **P**, meaning $\mathbf{P} \neq \mathbf{NP}$.

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 φ 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 φ 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 φ , is φ satisfiable?

- Formally:

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

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 . ■-ish

Proof: Take CS154!

Why All This Matters

- Resolving $\mathbf{P} \stackrel{?}{=} \mathbf{NP}$ is equivalent to just figuring out how hard SAT is.

$$\text{SAT} \in \mathbf{P} \quad \leftrightarrow \quad \mathbf{P} = \mathbf{NP}$$

- We've turned a huge, abstract, theoretical problem about solving problems versus checking solutions into the concrete task of seeing how hard one problem is.
- You can get a sense for how little we know about algorithms and computation given that we can't yet answer this question!

Why All This Matters

- You will almost certainly encounter **NP**-hard problems in practice – they're everywhere!
- If a problem is **NP**-hard, then there is no known algorithm for that problem that
 - is efficient on all inputs,
 - always gives back the right answer, and
 - runs deterministically.
- ***Useful intuition:*** If you need to solve an **NP**-hard problem, you will either need to settle for an approximate answer, an answer that's likely but not necessarily right, or have to work on really small inputs.

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 receive transplants. (*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?

Next Time

- ***Why All This Matters***
- ***Where to Go from Here***
- ***A Final “Your Questions”***
- ***Parting Words!***