

# Complexity Theory

## Part II

# Time Complexity

- The **time complexity** of a TM  $M$  is a function denoting the *worst-case* number of steps  $M$  takes on any input of length  $n$ .
  - By convention,  $n$  denotes the length of the input.
  - Assume we're only dealing with deciders, so there's no need to handle looping TMs.
- We often use **big-O notation** to describe growth rates of functions (and time complexity in particular).
  - Found by discarding leading coefficients and low-order terms.

# Polynomials and Exponentials

- A TM runs in **polynomial time** iff its runtime is some polynomial in  $n$ .
  - That is, time  $O(n^k)$  for some constant  $k$ .
- Polynomial functions “scale well.”
  - Small changes to the size of the input do not typically induce enormous changes to the overall runtime.
- Exponential functions scale terribly.
  - Small changes to the size of the input induce huge changes in the overall runtime.

# The Cobham-Edmonds Thesis

A language  $L$  can be **decided efficiently** iff there is a TM that decides it in polynomial time.

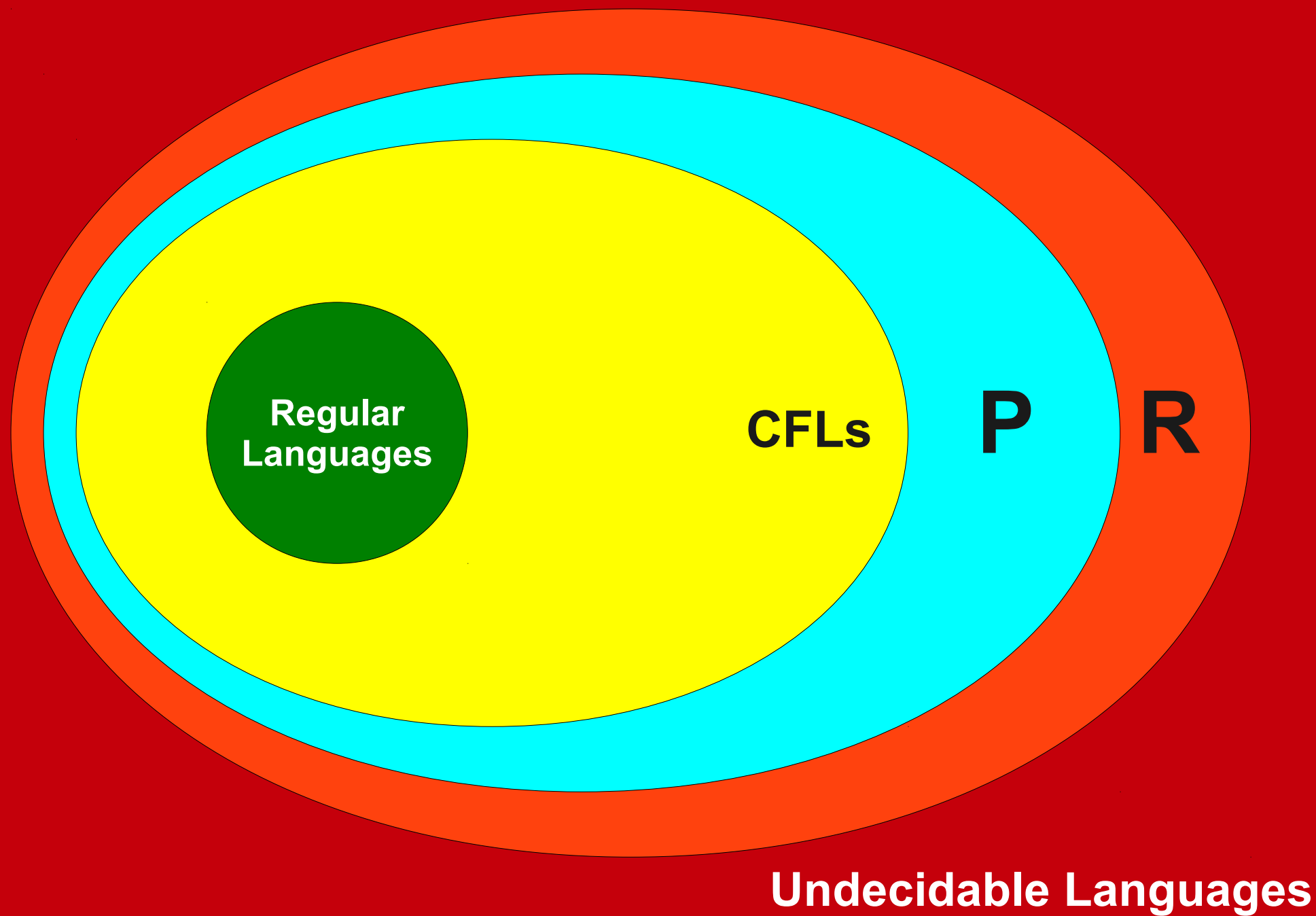
Equivalently,  $L$  can be decided efficiently iff it can be decided in time  $O(n^k)$  for some  $k \in \mathbb{N}$ .

Like the Church-Turing thesis, this is **not** a theorem!

It's an assumption about the nature of efficient computation, and it is somewhat controversial.

# The Complexity Class **P**

- The **complexity class P** (for **p**olynomial time) contains all problems that can be solved in polynomial time.
- Formally:
$$\mathbf{P} = \{ L \mid \text{There is a polynomial-time decider for } L \}$$
- Assuming the Cobham-Edmonds thesis, a language is in **P** iff it can be decided efficiently.



# Problems in **P**

- **Graph connectivity:**

Given a graph  $G$  and nodes  $s$  and  $t$ ,  
is there a path from  $s$  to  $t$ ?

- **Primality testing:**

Given a number  $p$ , is  $p$  prime? (Best known  
TM for this takes time  $O(n^{72})$ .)

- **Maximum matching:**

Given a set of tasks and workers who can  
perform those tasks, can all of the tasks be  
completed in under  $n$  hours?

# Problems in **P**

- **Remoteness testing:**

Given a graph  $G$ , are all of the nodes in  $G$  within distance at most  $k$  of one another?

- **Linear programming:**

Given a linear set of constraints and linear objective function, is the optimal solution at least  $n$ ?

- **Edit distance:**

Given two strings, can the strings be transformed into one another in at most  $n$  single-character edits?



# Other Models of Computation

- **Theorem:**  $L \in \mathbf{P}$  iff there is a polynomial-time TM or computer program that decides it.
- Essentially – a problem is in  $\mathbf{P}$  iff you could solve it on a normal computer in polynomial time.
- Proof involves simulating a computer with a TM; come talk to me after lecture for details on how to do this.

# Proving Languages are in **P**

- **Directly prove the language is in P.**
  - Build a decider for the language  $L$ .
  - Prove that the decider runs in time  $O(n^k)$ .
- **Use closure properties.**
  - Prove that the language can be formed by appropriate transformations of languages in **P**.
- **Reduce the language to a language in P.**
  - Show how a polynomial-time decider for some language  $L'$  can be used to decide  $L$ .

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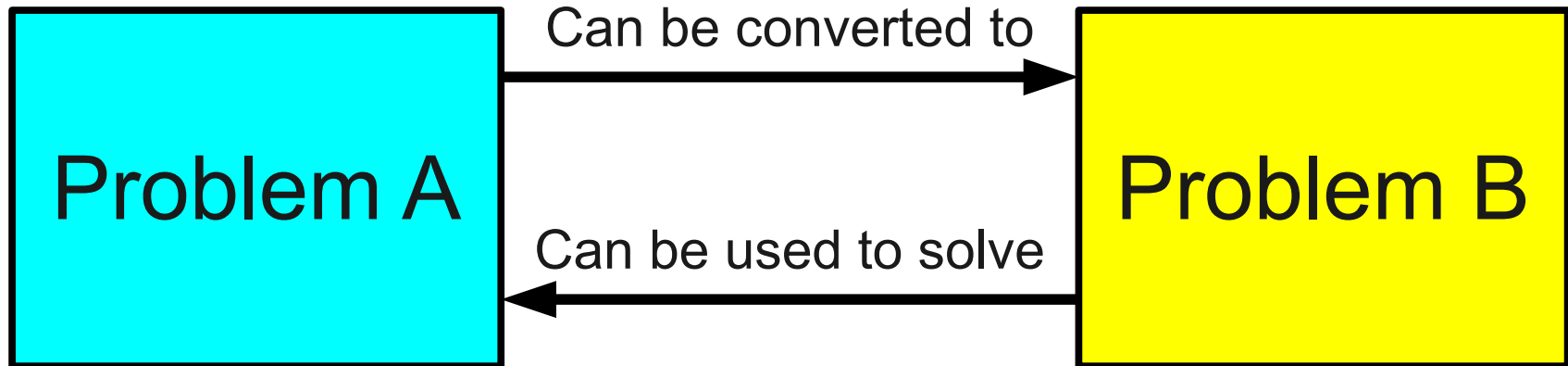
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- **Reduce the language to a language in P.**
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# Reductions



If any instance of  $A$  can be converted into an instance of  $B$ , we say that  $A$  **reduces** to  $B$ .

# Mapping Reductions and **P**

- When studying whether problems were in **R**, **RE**, or co-**RE**, we used mapping reductions.
- The construction we built using mapping reductions
  - computes the function  $f$  on some input string  $w$ , then
  - runs another TM on  $f(w)$ .
- When talking about class **P**, we need to make sure that this entire process doesn't take too much time.

# Polynomial-Time Reductions

- Let  $A \subseteq \Sigma_1^*$  and  $B \subseteq \Sigma_2^*$  be languages.
- A **polynomial-time mapping reduction** is a function  $f: \Sigma_1^* \rightarrow \Sigma_2^*$  with the following properties:
  - $f(w)$  can be computed **in polynomial time**.
  - $w \in A$  iff  $f(w) \in B$ .
- Informally:
  - A way of turning inputs to  $A$  into inputs to  $B$
  - that can be computed **in polynomial time**
  - that preserves the correct answer.
- Notation:  **$A \leq_p B$**  iff there is a polynomial-time mapping reduction from  $A$  to  $B$ .

# Polynomial-Time Reductions

- Suppose that we know that  $B \in \mathbf{P}$ .
- Suppose that  $A \leq_p B$  and that the reduction  $f$  can be computed in time  $O(n^k)$ .

$A$

Solvable?

$B$

Solvable  
in  $O(n^r)$

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Input size:  $n$

$A$

Solvable?

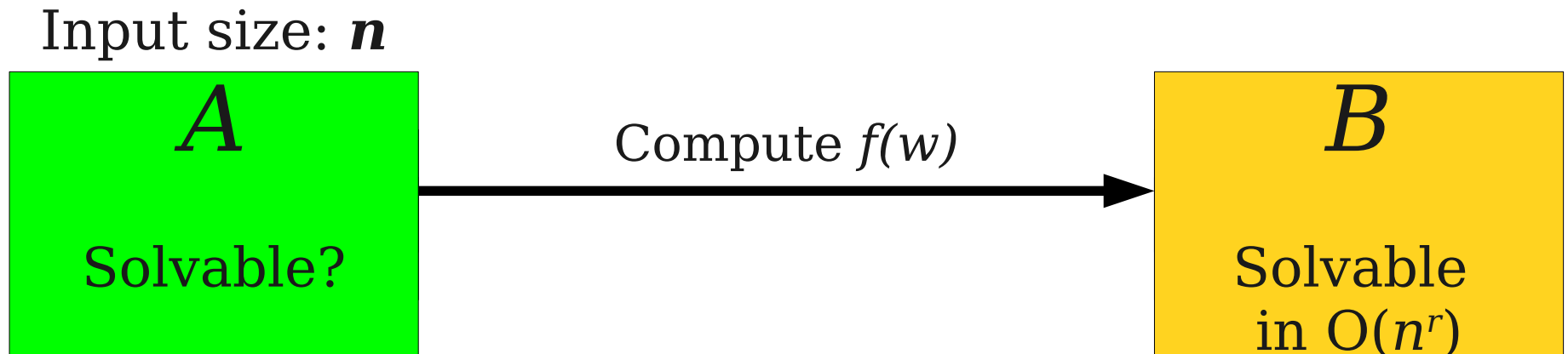
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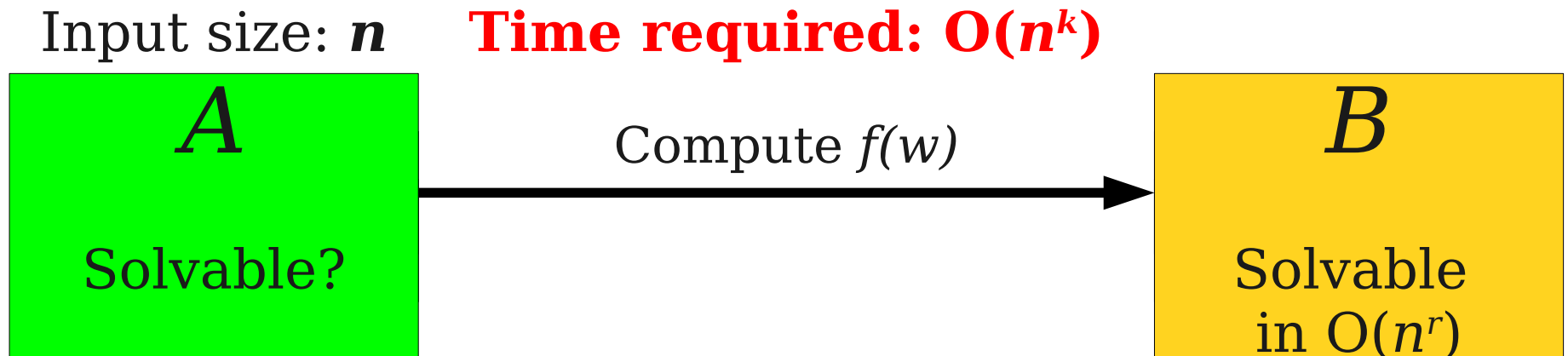
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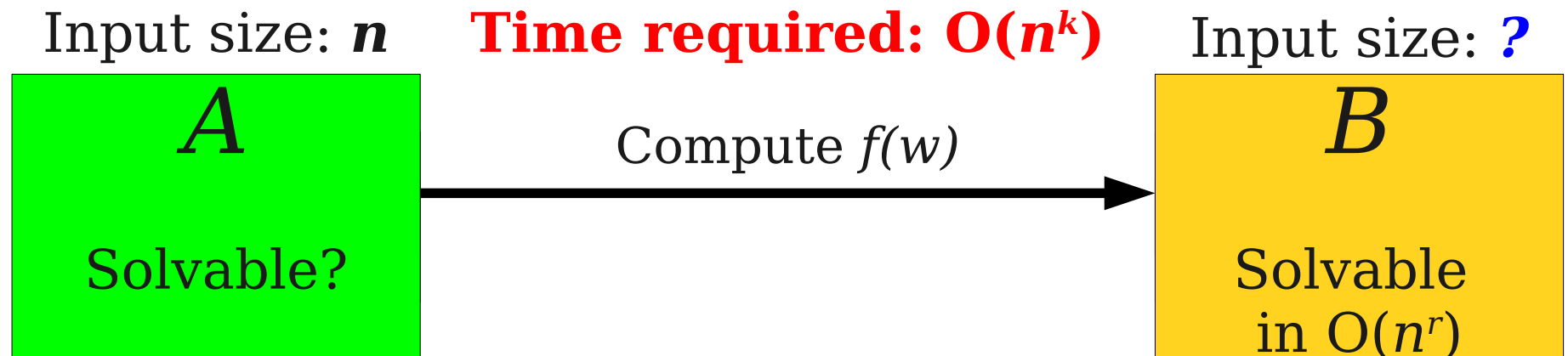
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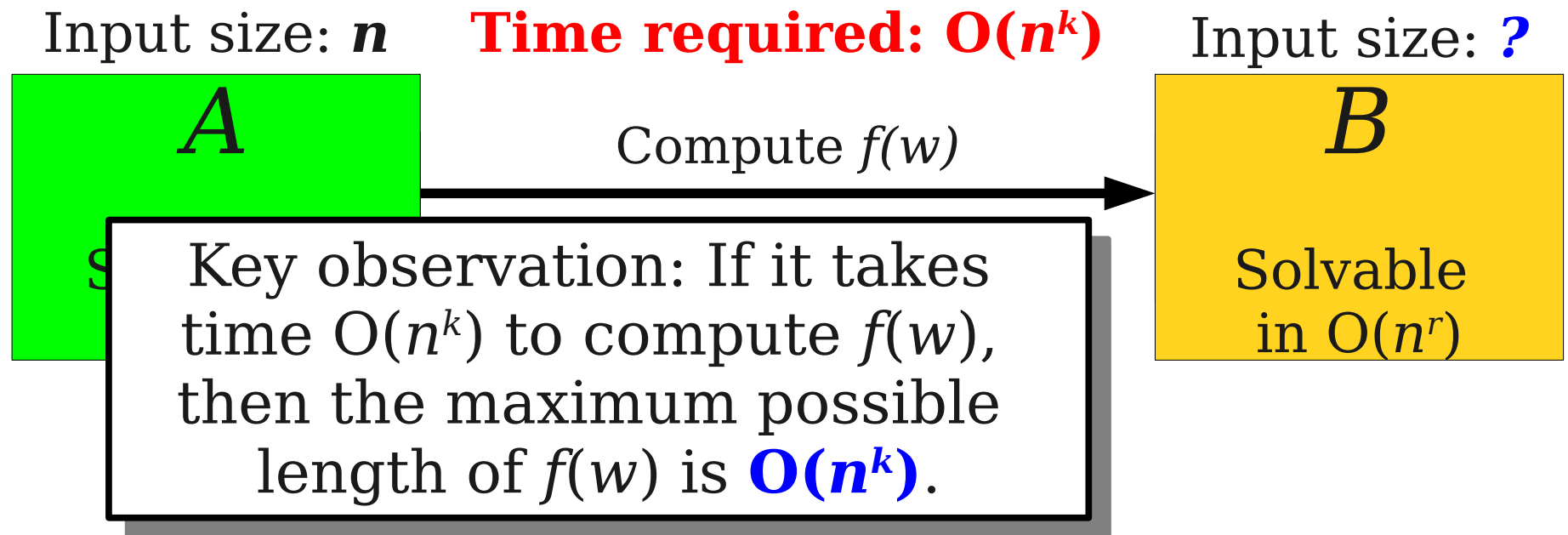
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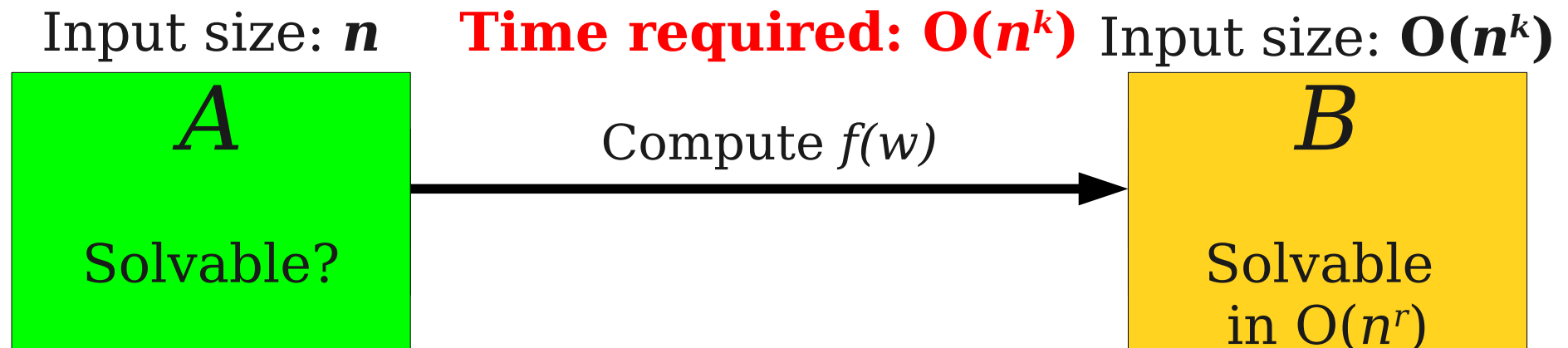
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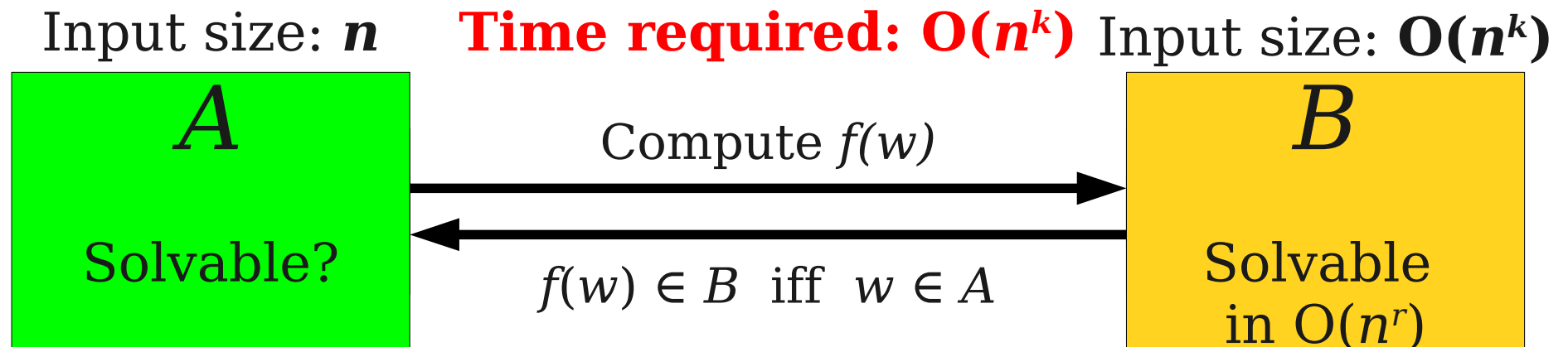
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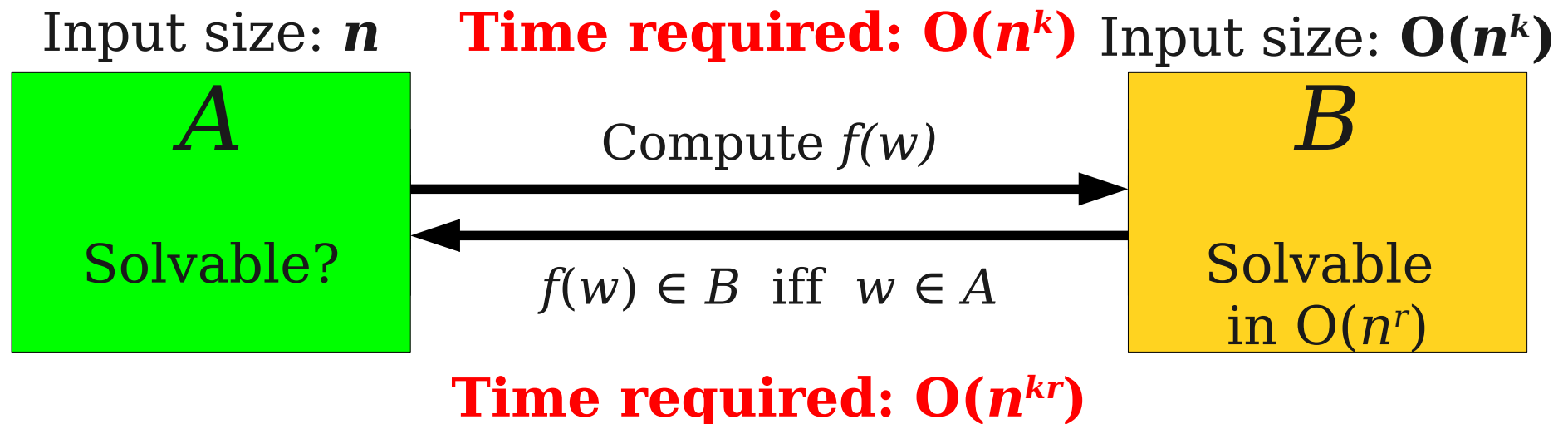
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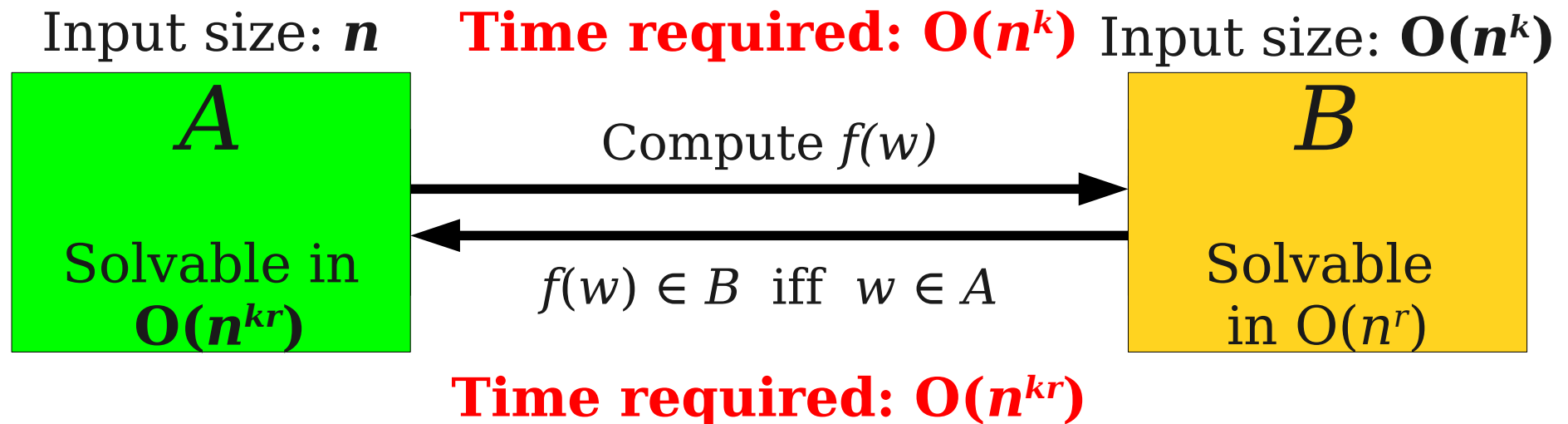
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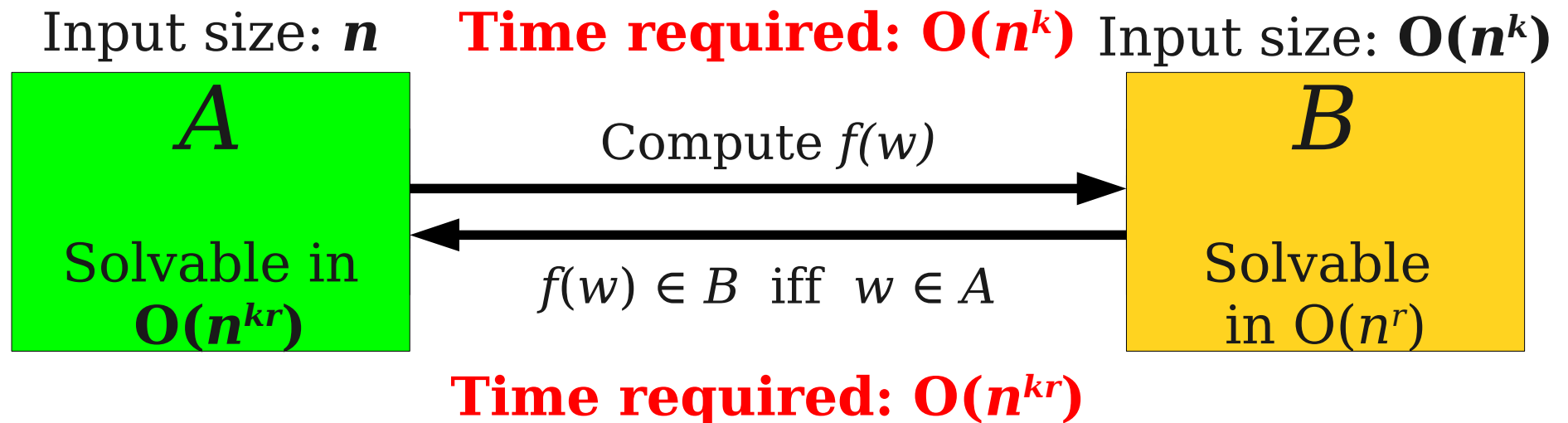
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# Polynomial-Time Reductions

- Suppose that we know that  $B \in \mathbf{P}$ .
- Suppose that  $A \leq_p B$  and that the reduction  $f$  can be computed in time  $O(n^k)$ .
- Then  $A \in \mathbf{P}$  as well.



*Theorem:* If  $B \in \mathbf{P}$  and  $A \leq_p B$ , then  $A \in \mathbf{P}$ .

*Proof:* Let  $H$  be a polynomial-time decider for  $B$ . Consider the following TM:

$M =$  “On input  $w$ :  
    Compute  $f(w)$ .  
    Run  $H$  on  $f(w)$ .  
    If  $H$  accepts, accept; if  $H$  rejects, reject.”

We claim that  $M$  is a polynomial-time decider for  $A$ . To see this, we prove that  $M$  is a polynomial-time decider, then that  $\mathcal{L}(M) = A$ . To see that  $M$  is a polynomial-time decider, note that because  $f$  is a polynomial-time reduction, computing  $f(w)$  takes time  $O(n^k)$  for some  $k$ . Moreover, because computing  $f(w)$  takes time  $O(n^k)$ , we know that  $|f(w)| = O(n^k)$ .  $M$  then runs  $H$  on  $f(w)$ . Since  $H$  is a polynomial-time decider,  $H$  halts in  $O(m^r)$  on an input of size  $m$  for some  $r$ . Since  $|f(w)| = O(n^k)$ ,  $H$  halts after  $O(|f(w)|^r) = O(n^{kr})$  steps. Thus  $M$  halts after  $O(n^k + n^{kr})$  steps, so  $M$  is a polynomial-time decider.

To see that  $\mathcal{L}(M) = A$ , note that  $M$  accepts  $w$  iff  $H$  accepts  $f(w)$  iff  $f(w) \in B$ . Since  $f$  is a polynomial-time reduction,  $f(w) \in B$  iff  $w \in A$ . Thus  $M$  accepts  $w$  iff  $w \in A$ , so  $\mathcal{L}(M) = A$ . ■

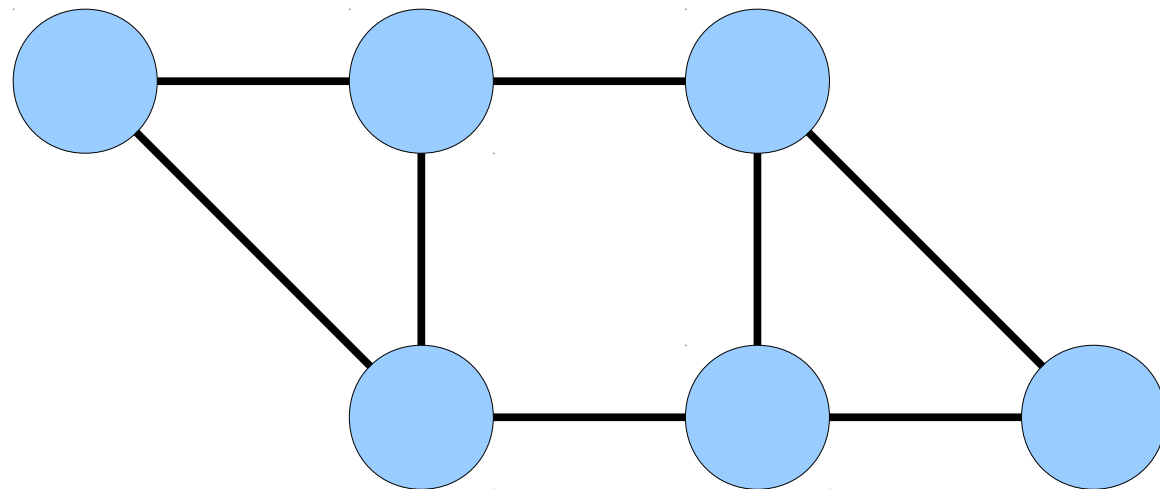
# A Sample Reduction

# 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.

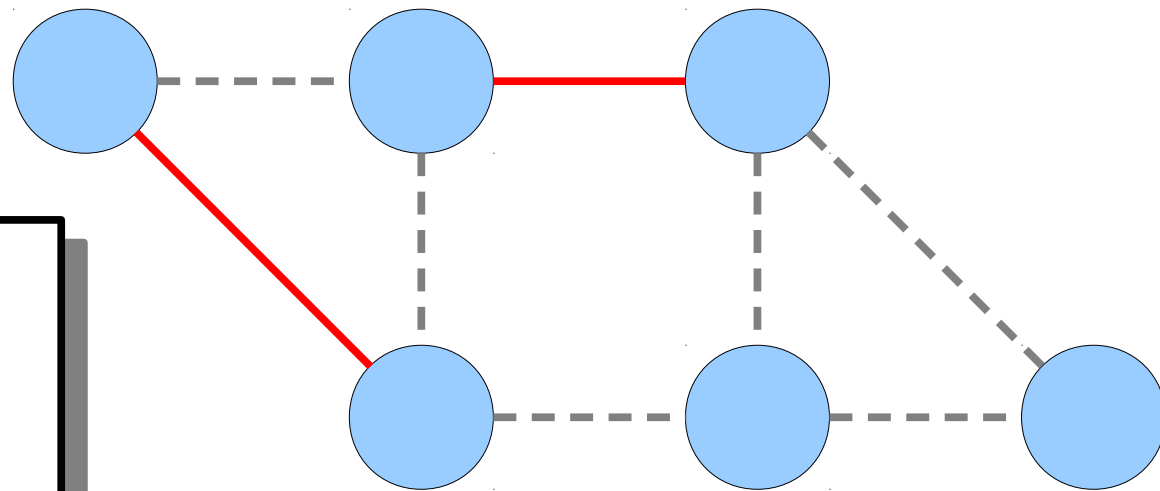
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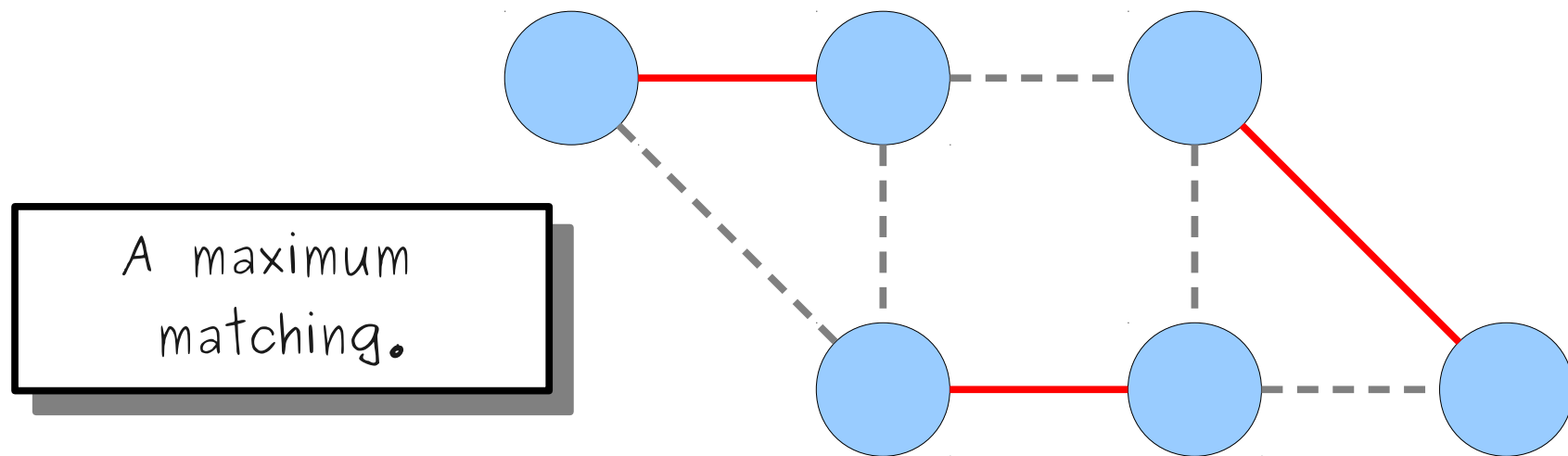
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A matching, but  
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matching.

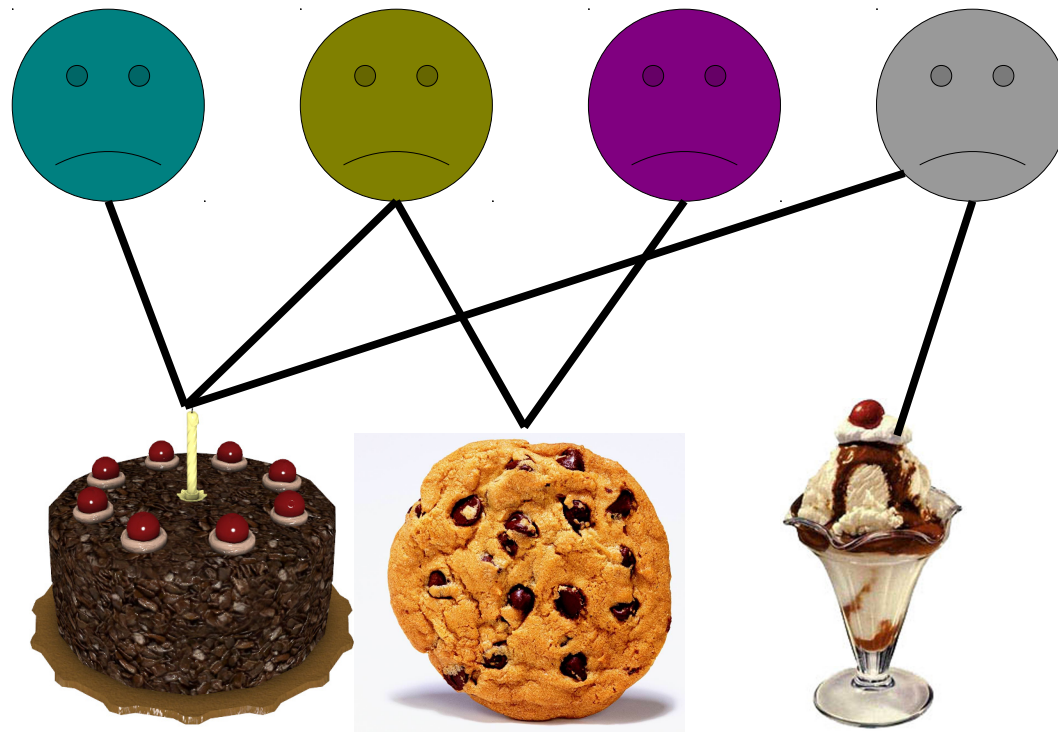
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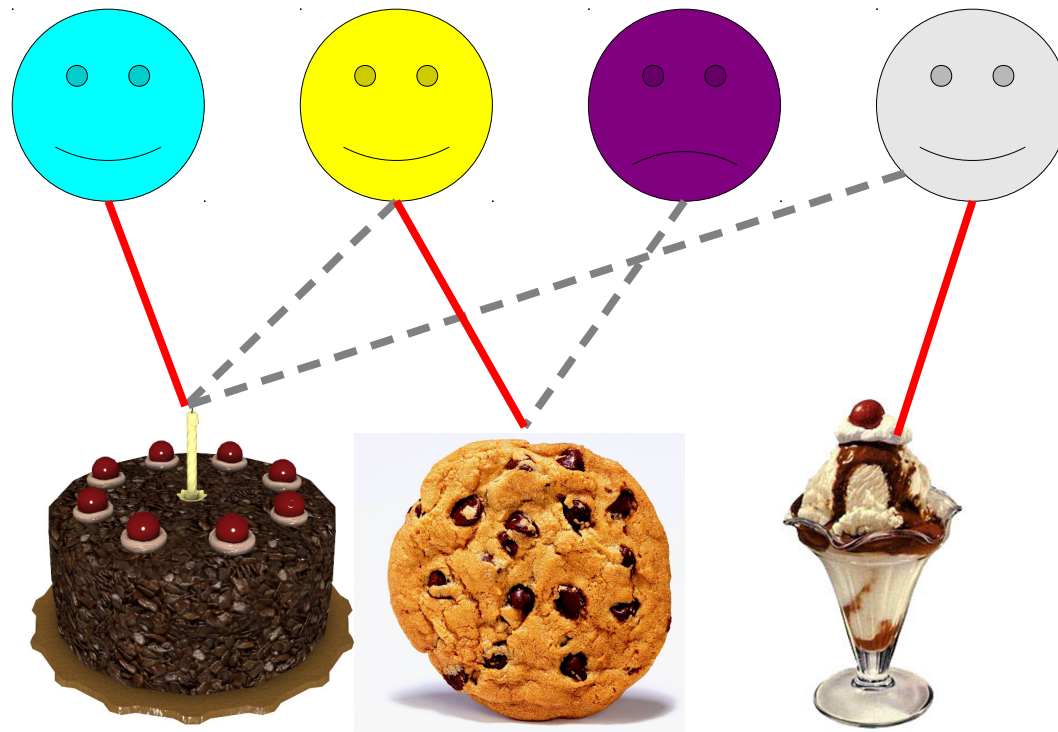
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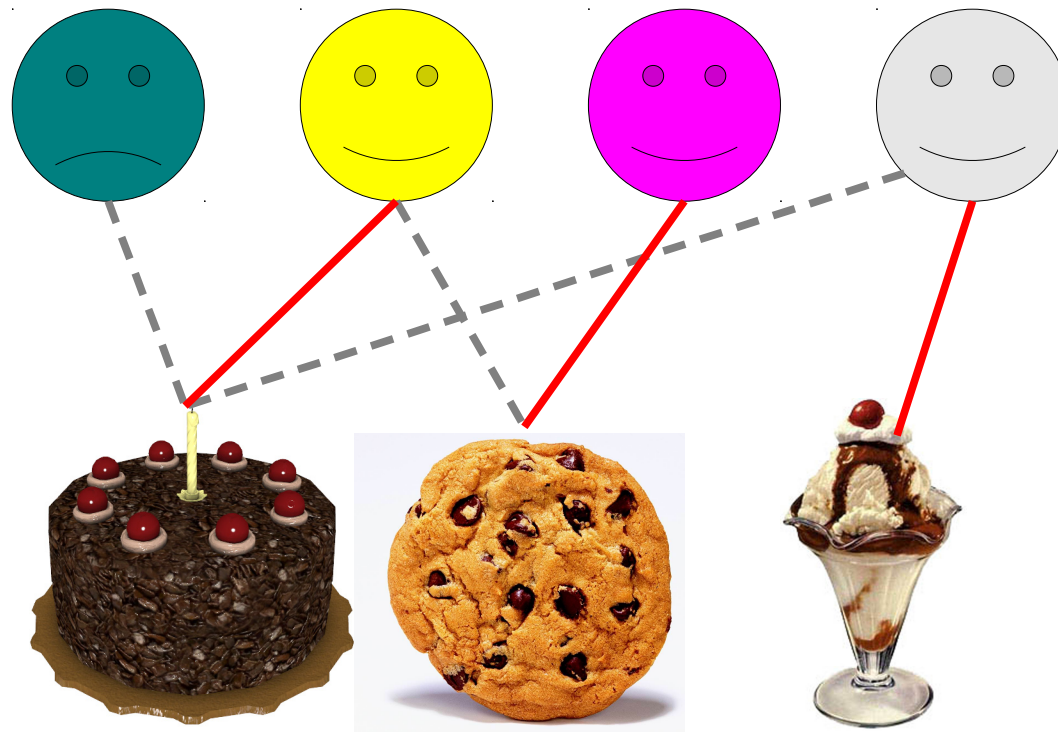
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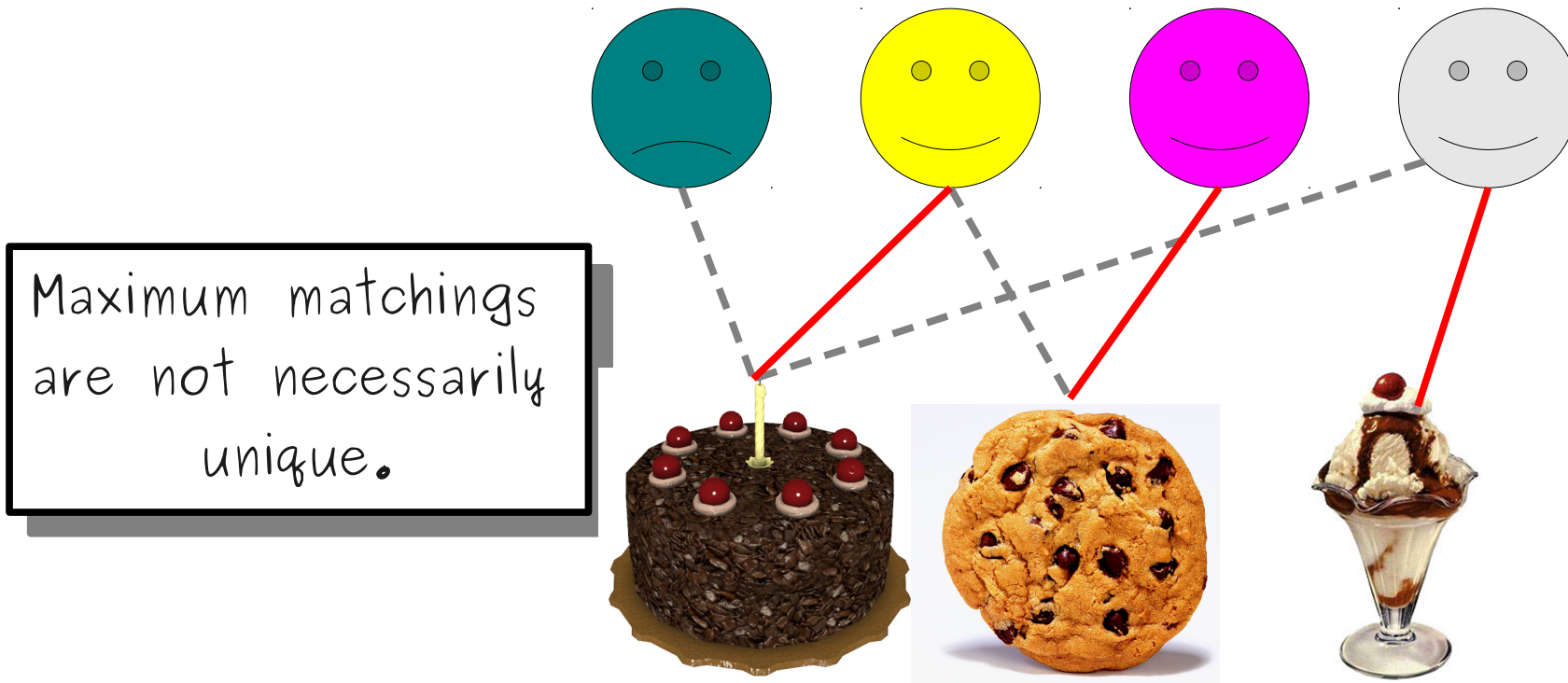
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# Maximum Matching

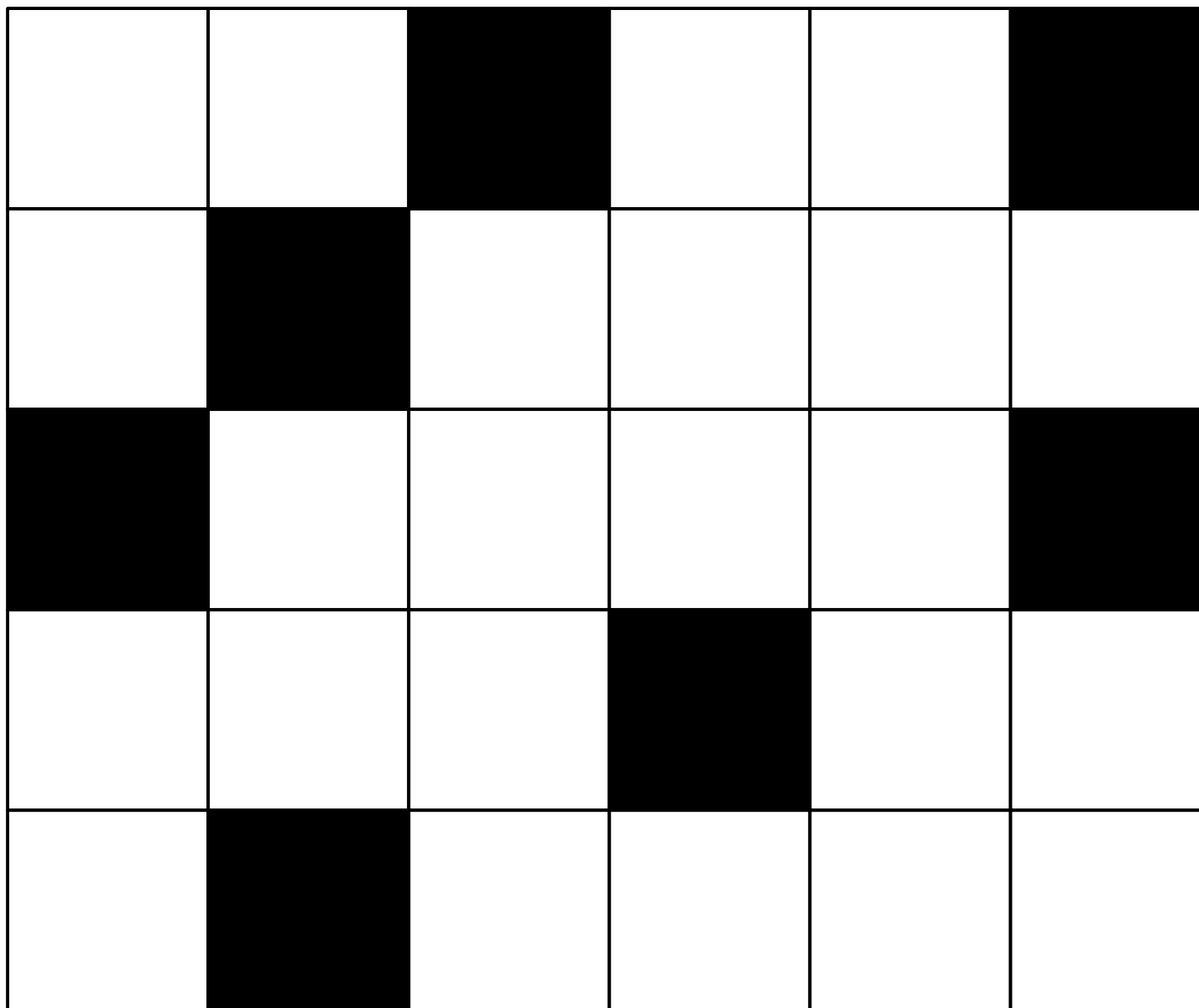
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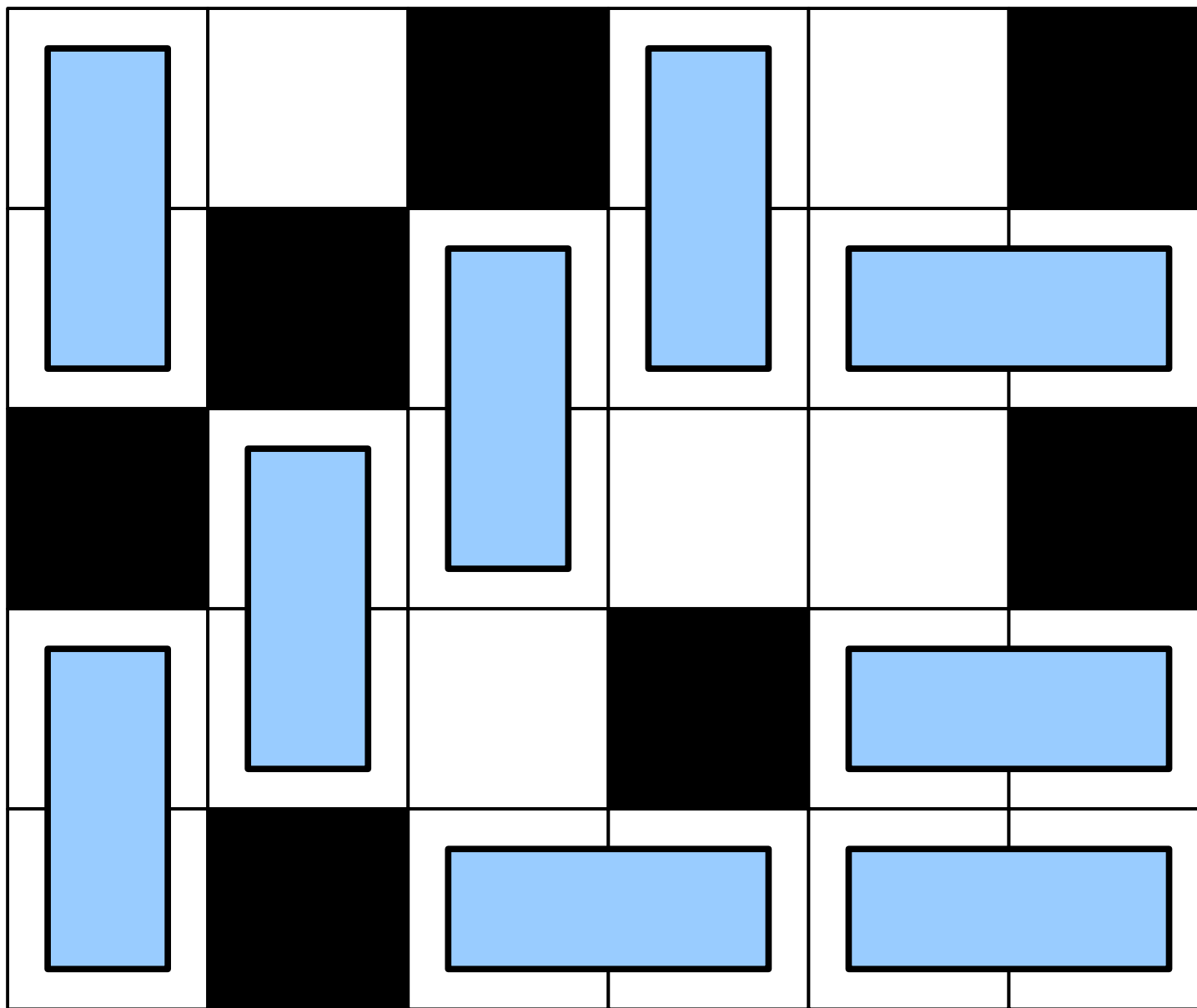
# Maximum Matching

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

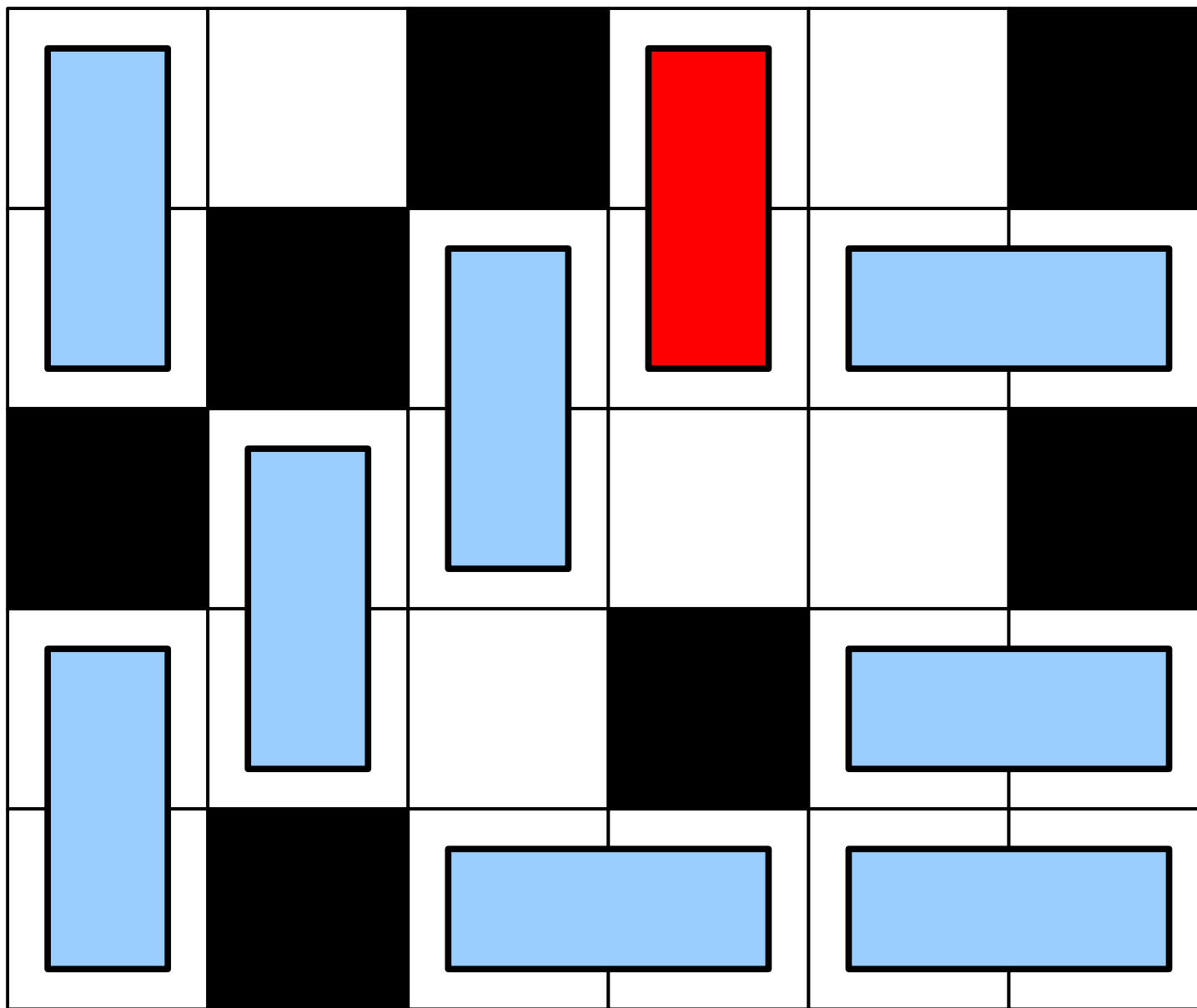
# Domino Tiling



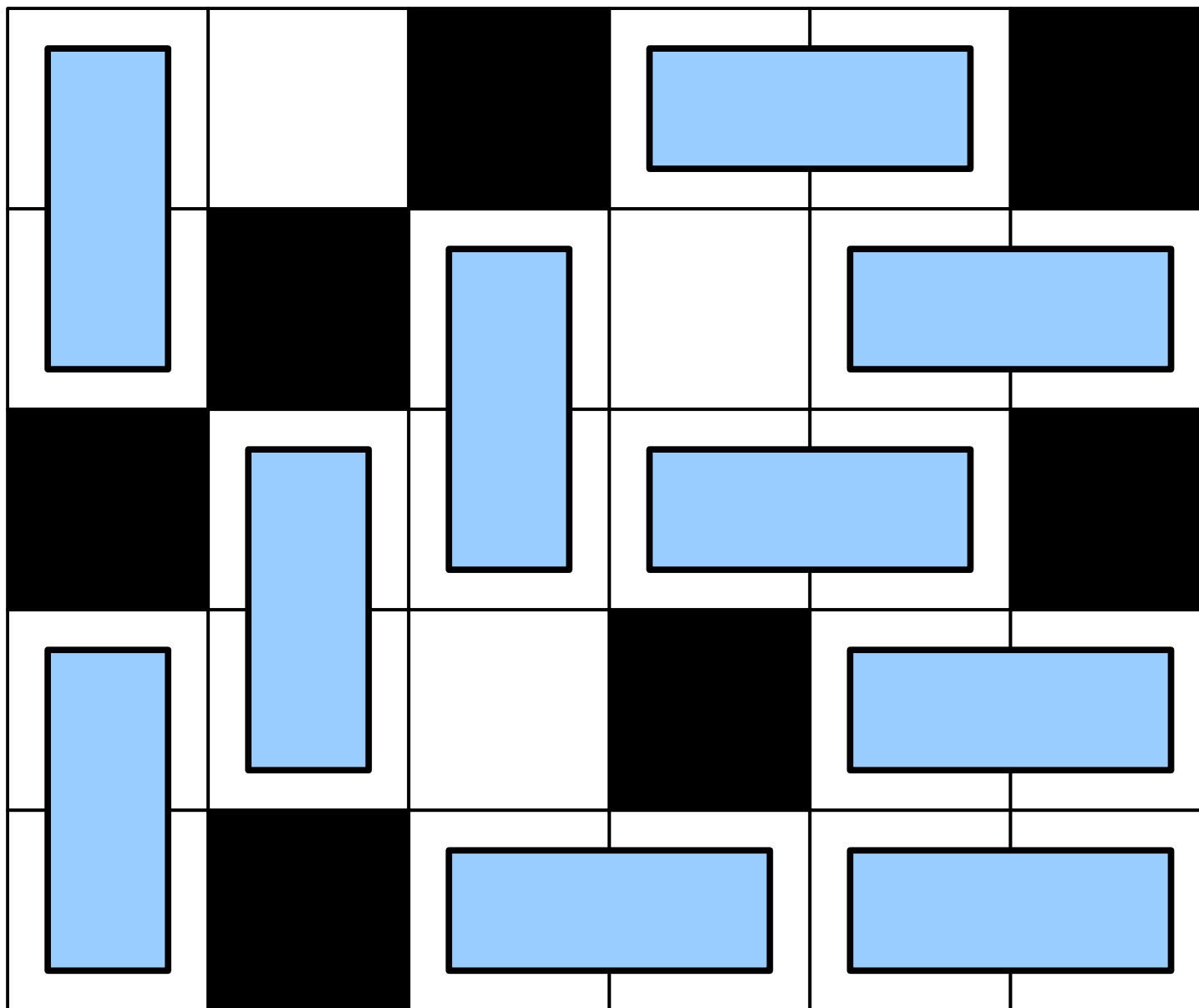
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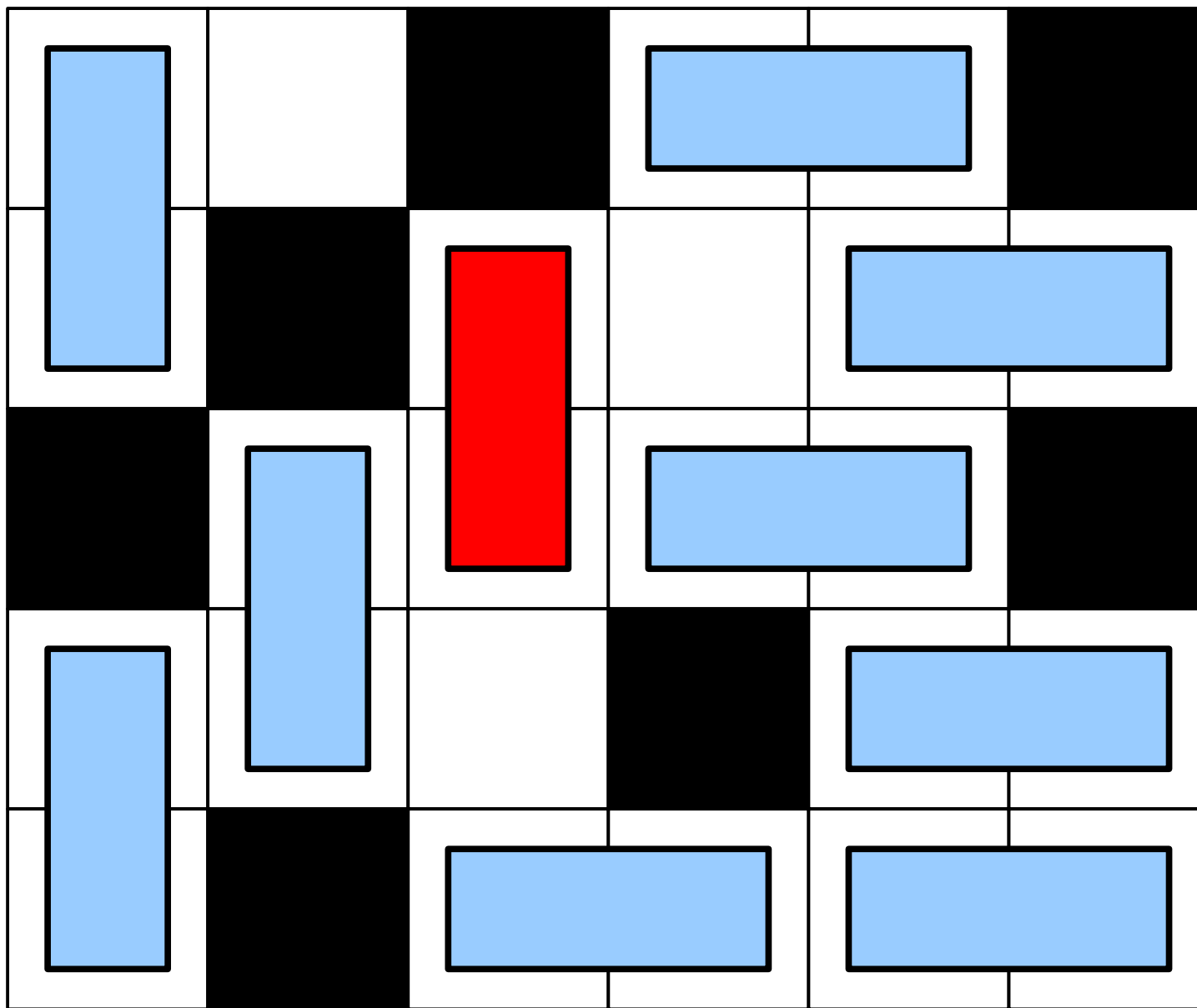


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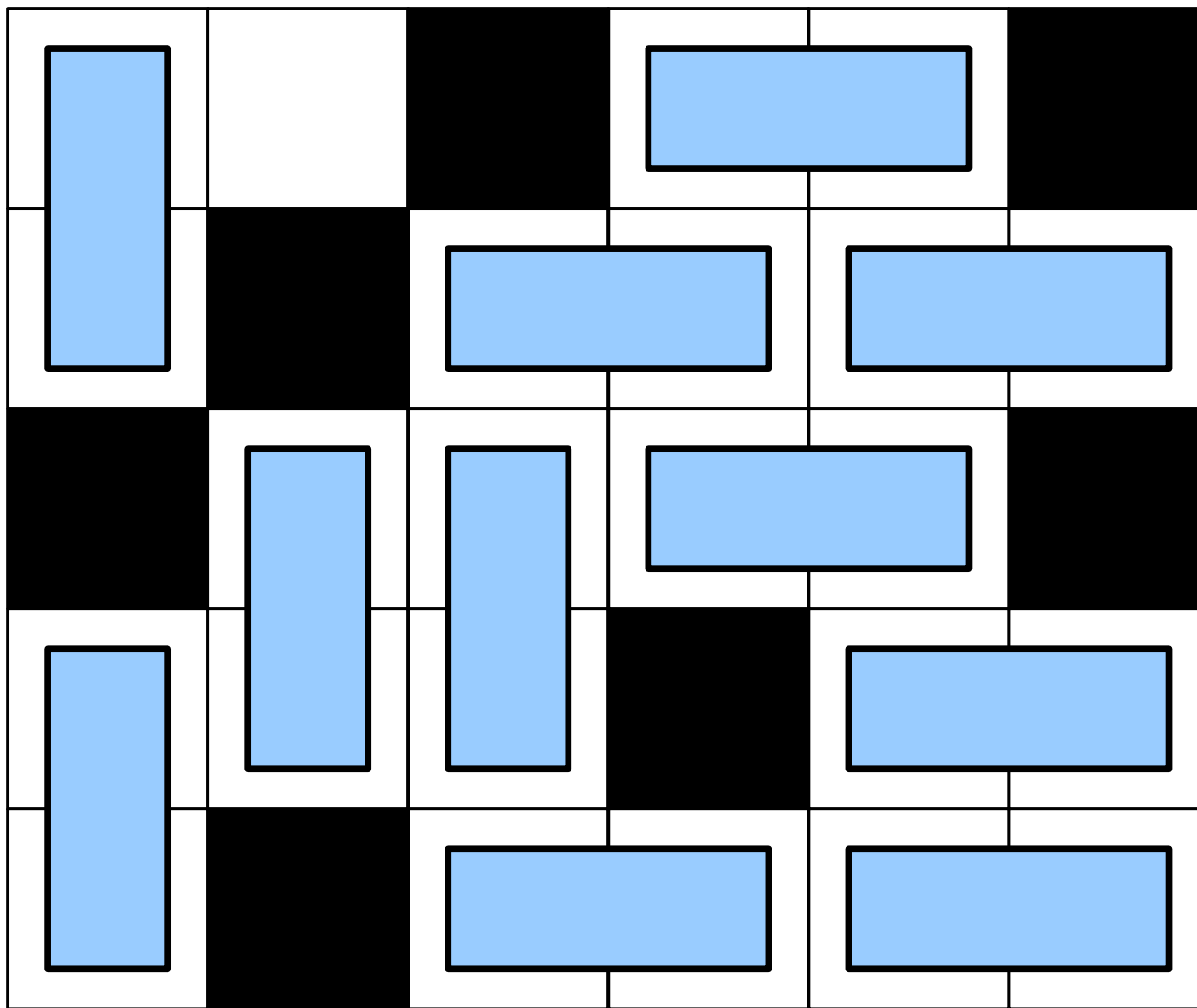




# Domino Tiling



# Domino Tiling



# A Domino Tiling Reduction

- Let *MATCHING* be the language defined as follows:

$MATCHING = \{ \langle G, k \rangle \mid G \text{ is an undirected graph with a matching of size at least } k \}$

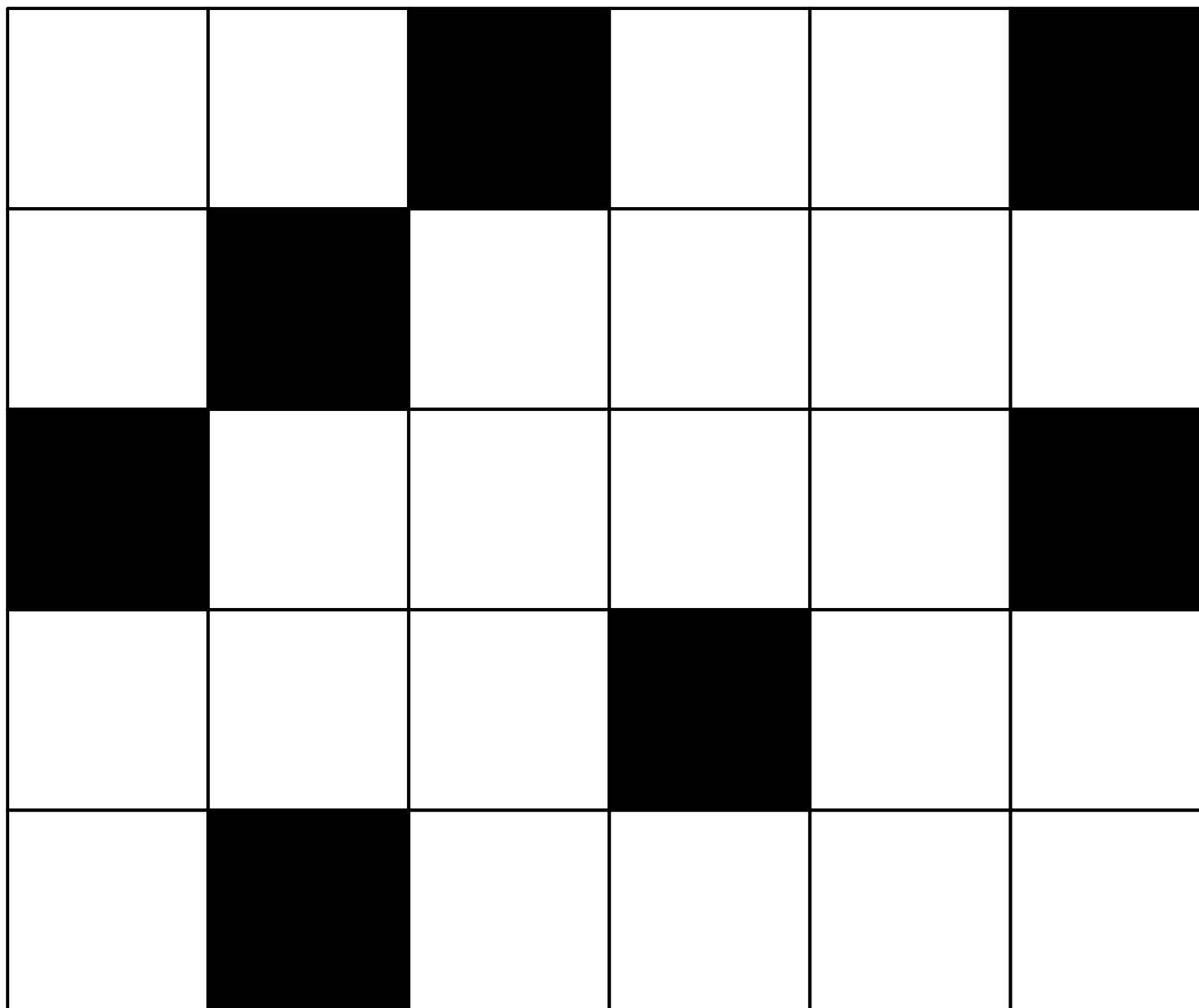
- **Theorem** (Edmonds):  $MATCHING \in \mathbf{P}$ .

- Let *DOMINO* be this language:

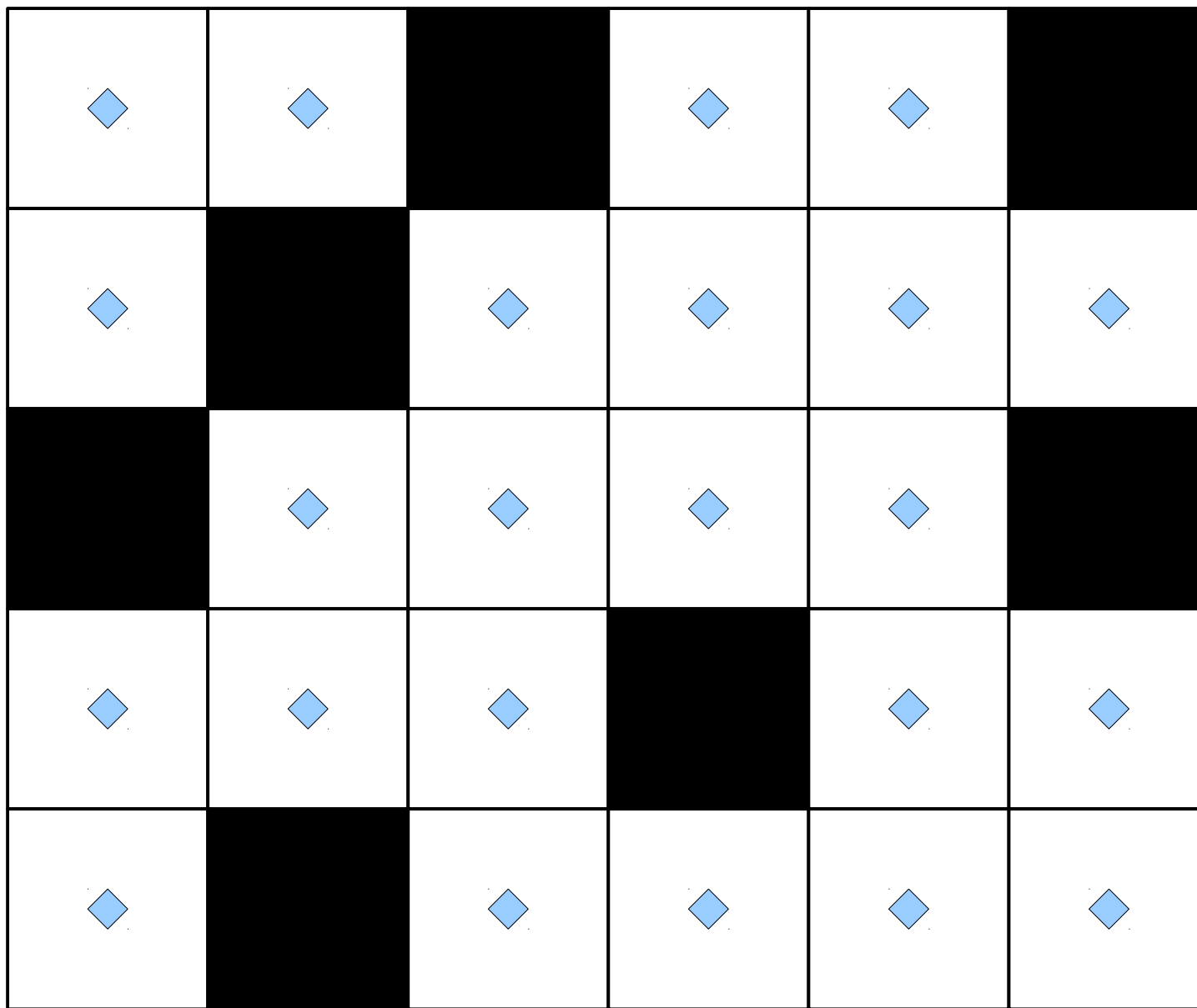
$DOMINO = \{ \langle D, k \rangle \mid D \text{ is a grid and } k \text{ nonoverlapping dominoes can be placed on } D. \}$

- We'll prove  $DOMINO \leq_P MATCHING$  to show that  $DOMINO \in \mathbf{P}$ .

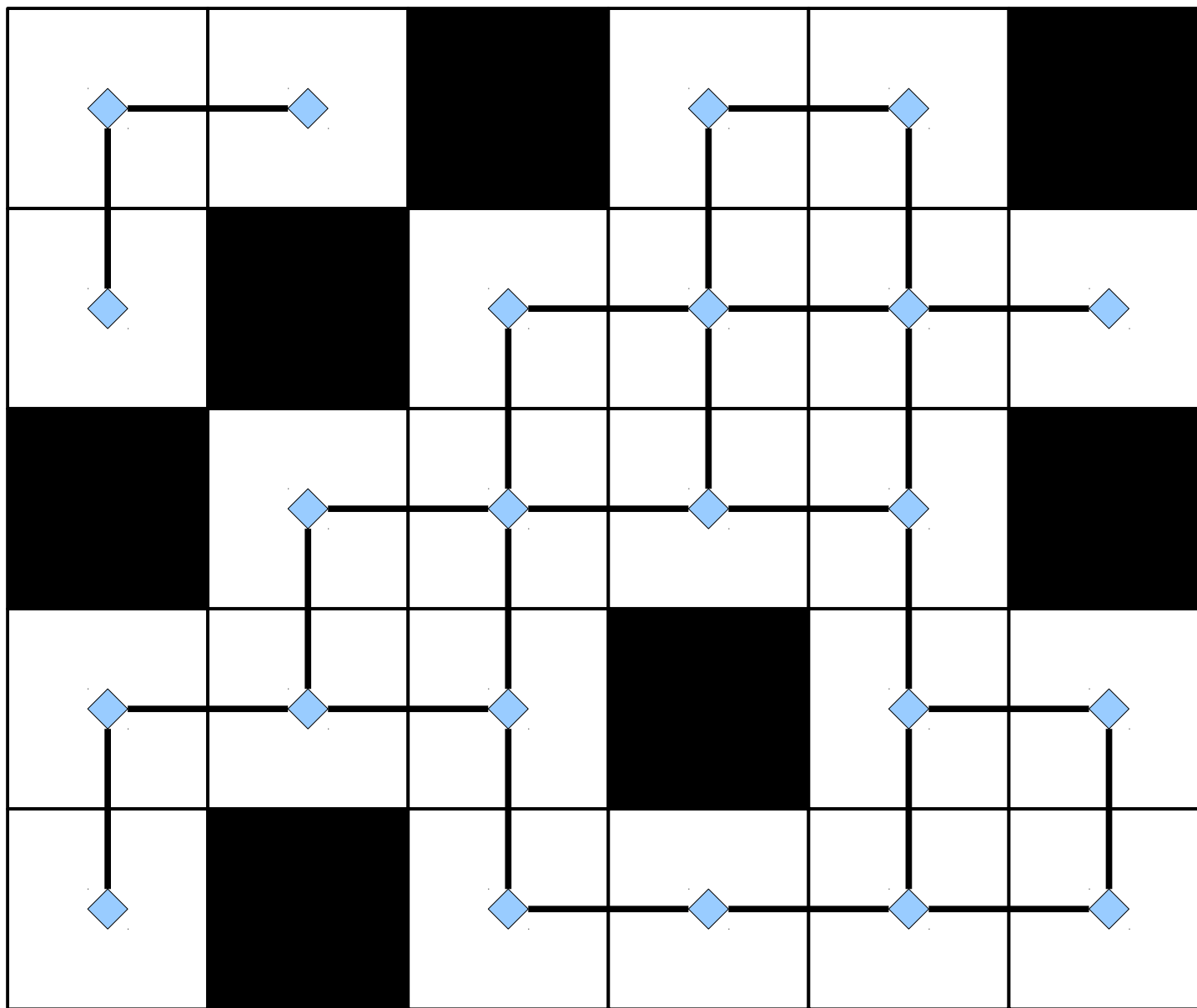
# Solving Domino Tiling



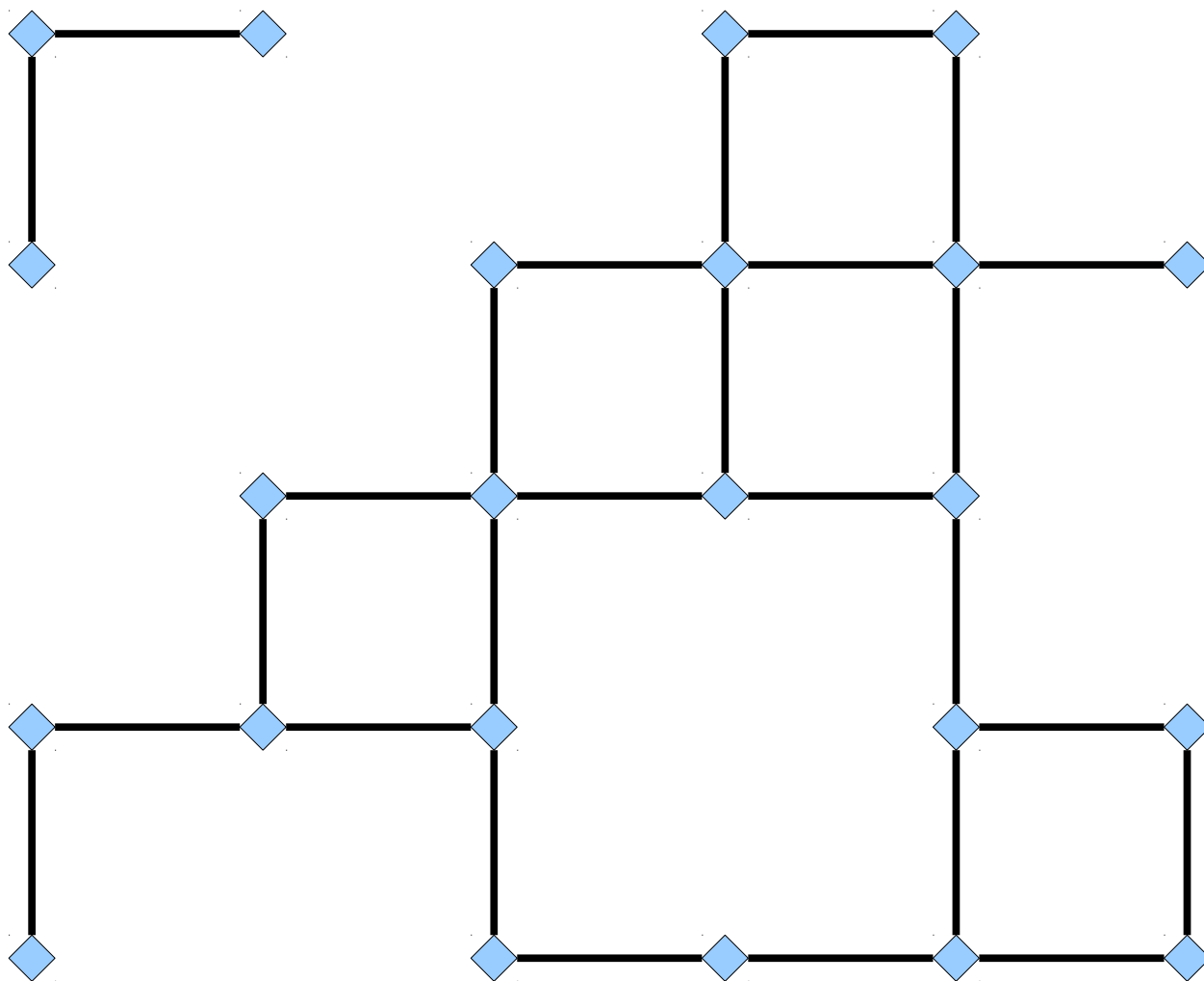
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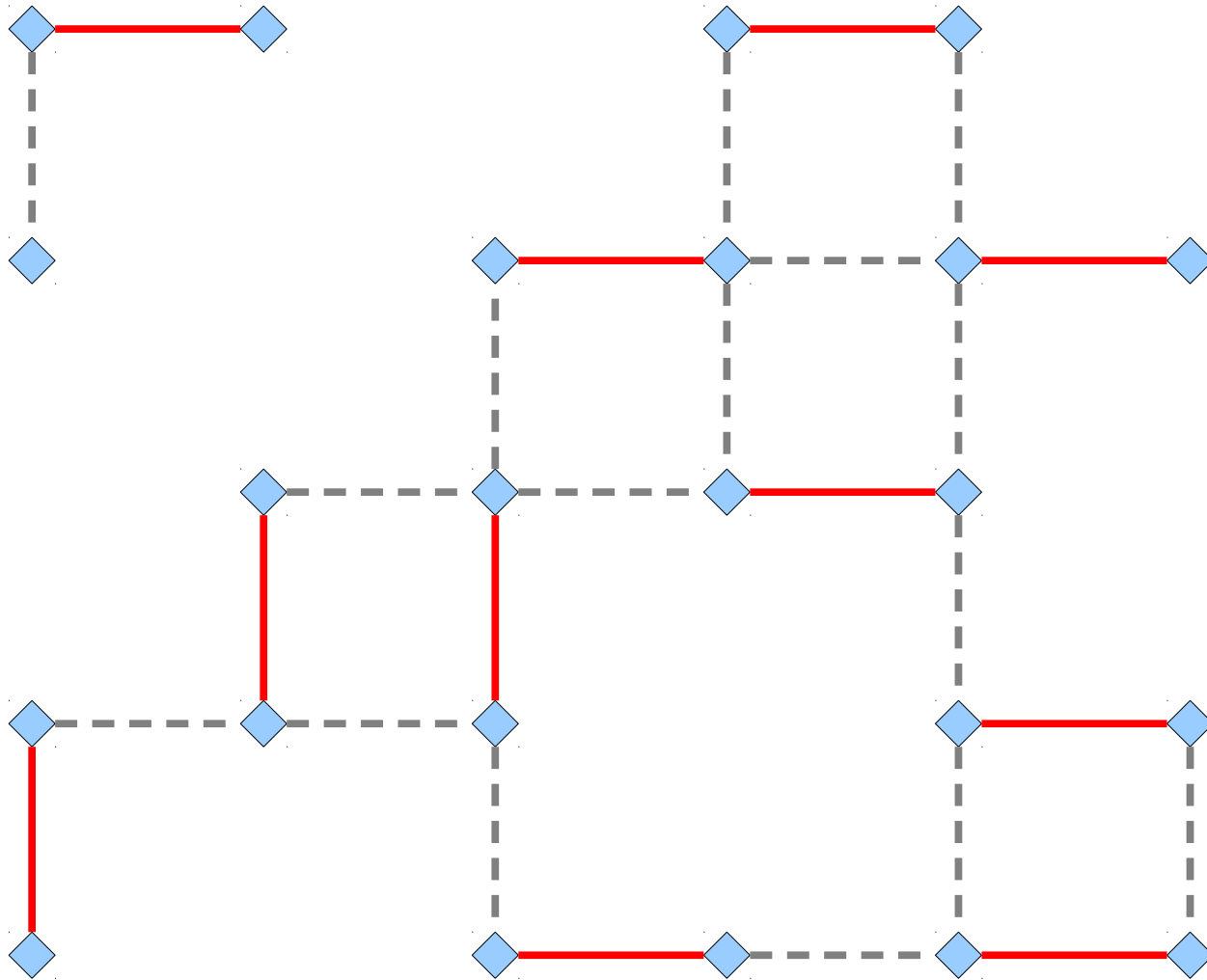
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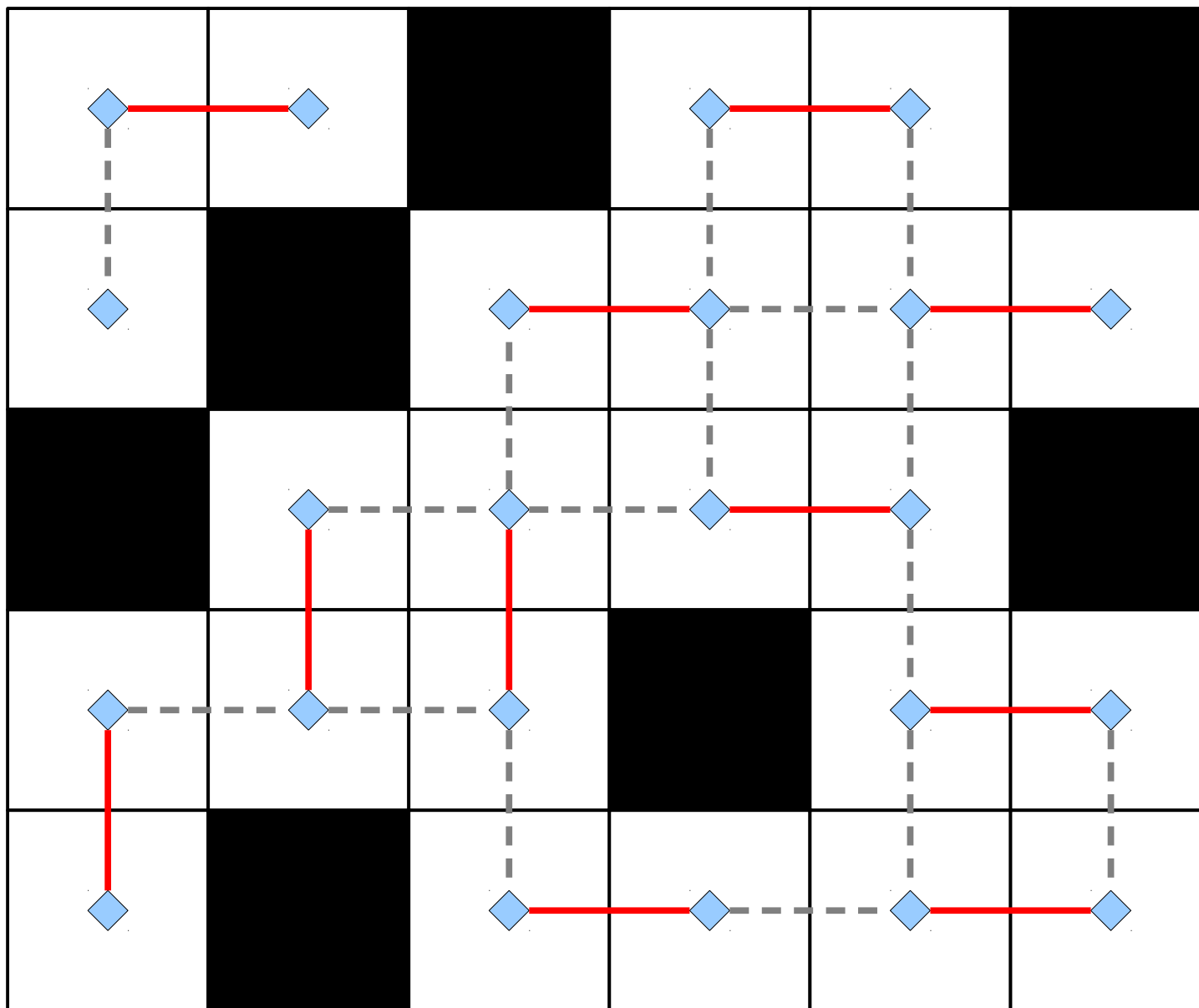


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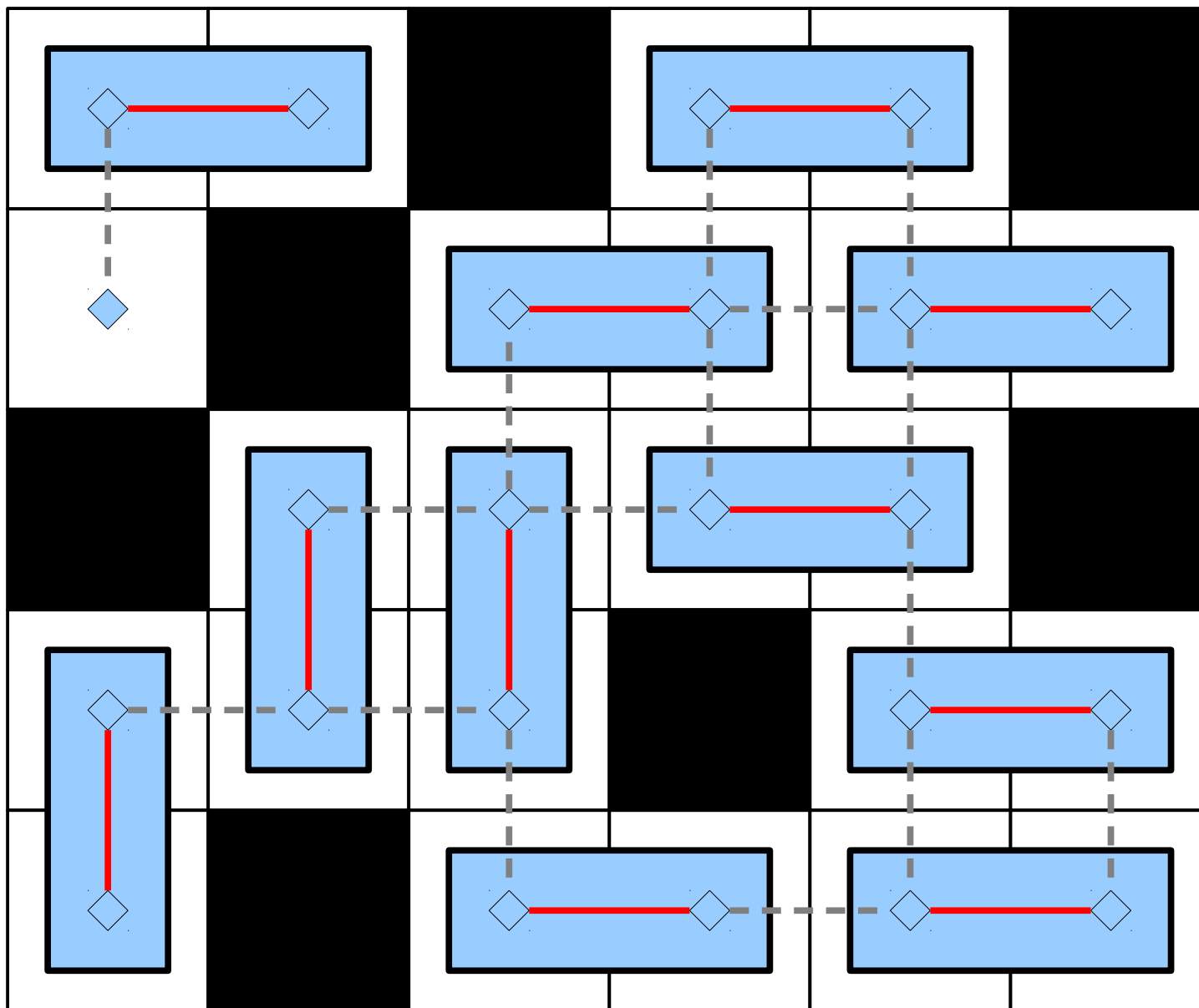




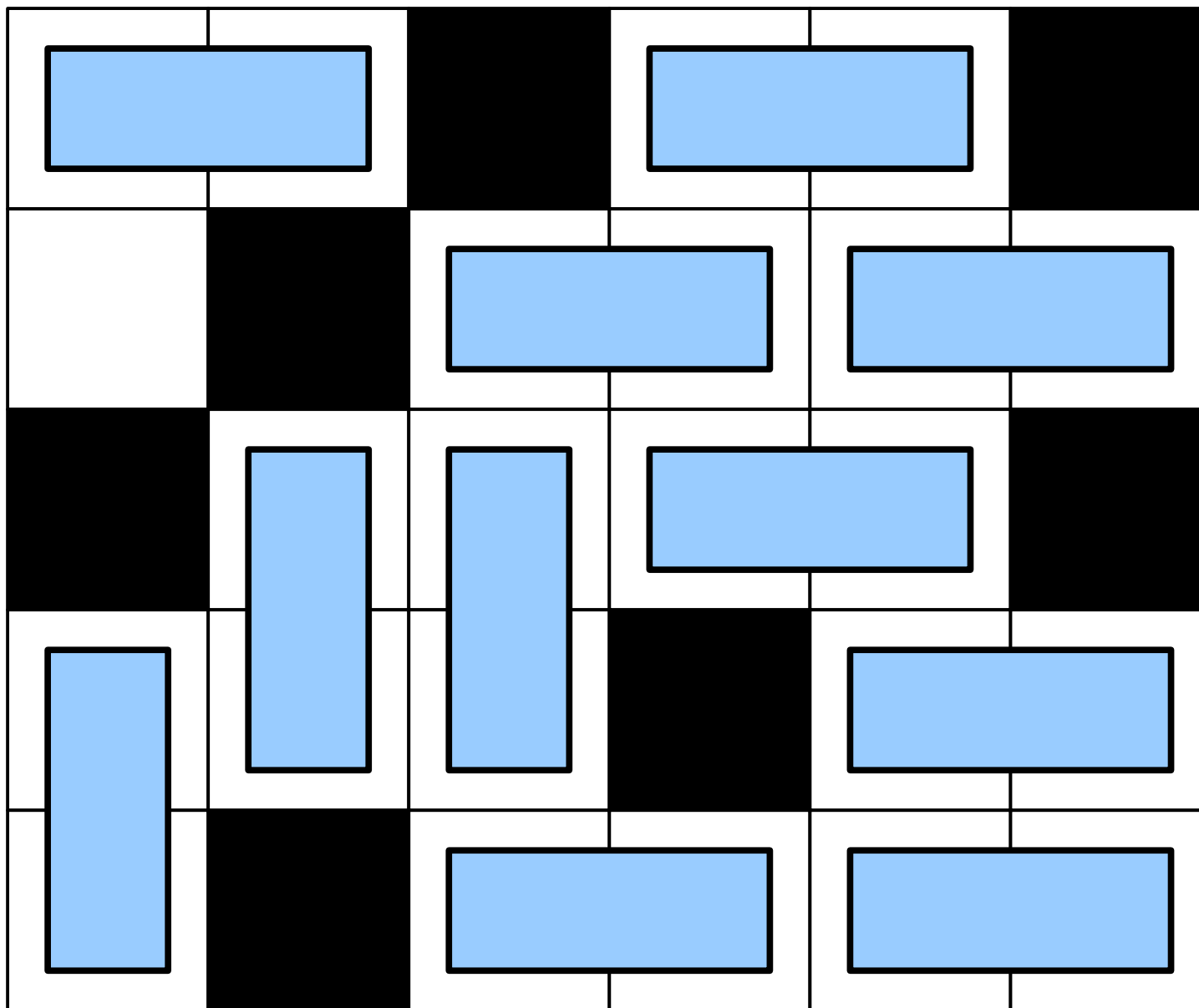
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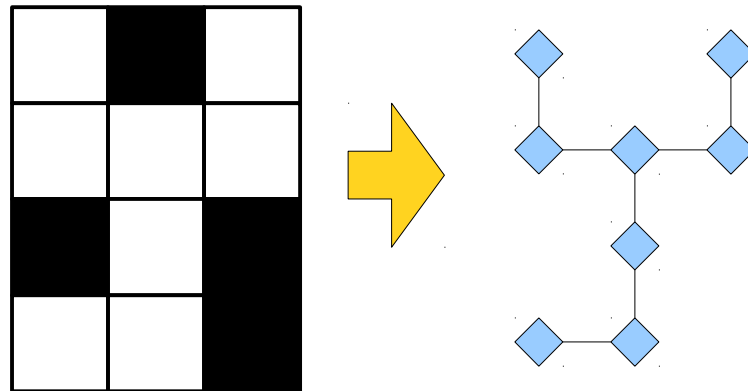


# Solving Domino Tiling



# Our Reduction

- Given as input  $\langle D, k \rangle$ , construct the graph  $G$  as follows:
  - For each empty cell, construct a node.
  - For each pair of adjacent empty cells, construct an edge between them.



- Let  $f(\langle D, k \rangle) = \langle G, k \rangle$ .

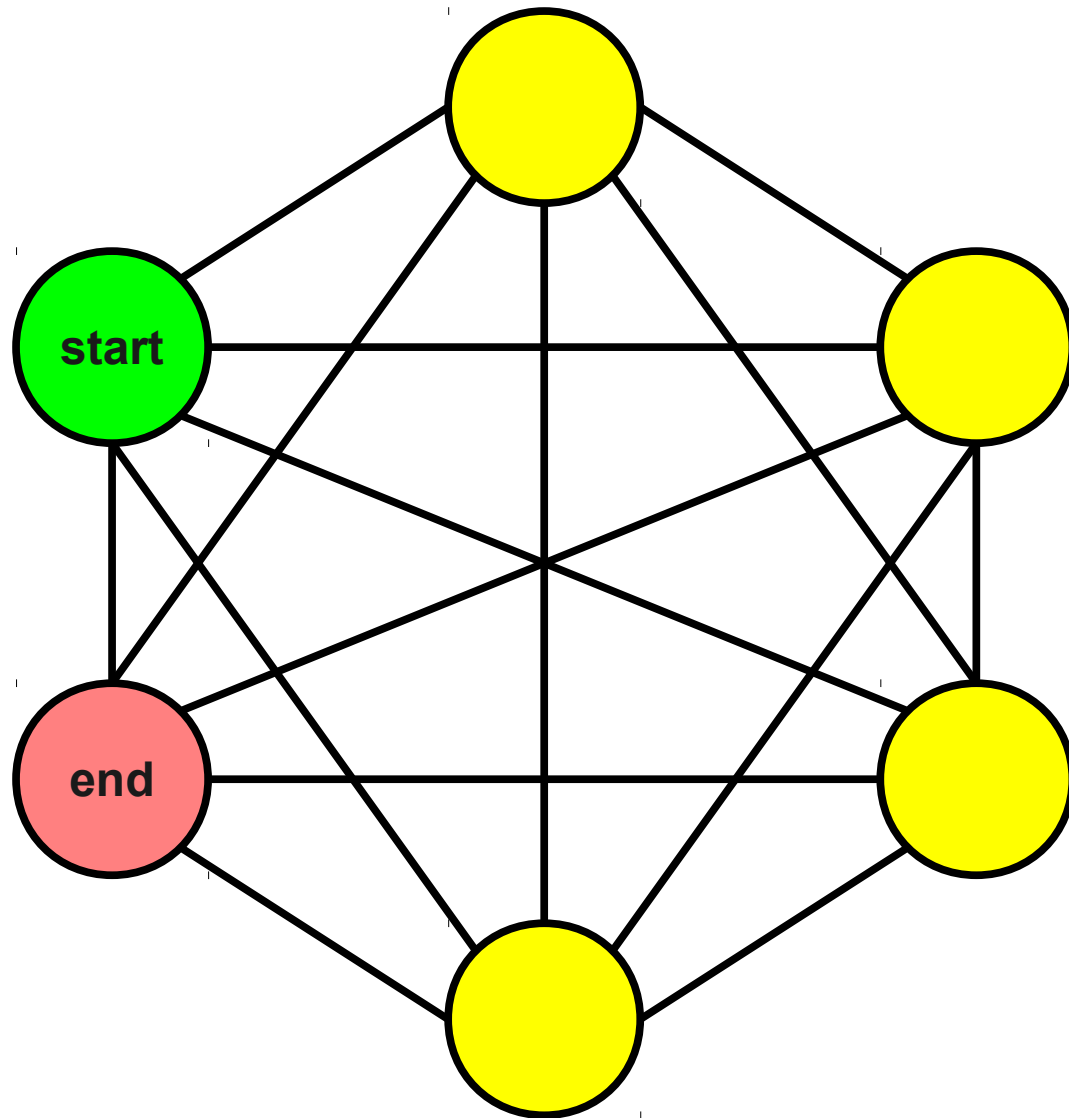
*Lemma:*  $f$  is computable in polynomial time.

*Proof:* We show that  $f(\langle D, k \rangle) = \langle G, k \rangle$  has size that is a polynomial in the size of  $\langle D, k \rangle$ .

For each empty cell  $x_i$  in  $D$ , we construct a single node  $v_i$  in  $G$ . Since there are  $O(|D|)$  cells, there are  $O(|D|)$  nodes in the graph. For each pair of adjacent, empty cells  $x_i$  and  $x_j$  in  $D$ , we add the edge  $(x_i, x_j)$ . Since each cell in  $D$  has four neighbors, the maximum number of edges we could add this way is  $O(|D|)$  as well. Thus the total size of the graph  $G$  is  $O(|D|)$ . Consequently, the total size of  $\langle G, k \rangle$  is  $O(|D| + |k|)$ , which is a polynomial in the size of the input.

Since each part of the graph could be constructed in polynomial time, the overall graph can be constructed in polynomial time. ■

What *can't* you do in polynomial time?



How many simple paths are there from the start node to the end node?



How many  
subsets of this  
set are there?



# An Interesting Observation

- There are (at least) exponentially many objects of each of the preceding types.
- However, each of those objects is not very large.
  - Each simple path has length no longer than the number of nodes in the graph.
  - Each subset of a set has no more elements than the original set.
- This brings us to our next topic...

**NIP**

The image features the letters 'NIP' in a bold, black, serif font. Behind the letter 'N' is a light gray graphic consisting of several concentric, overlapping curved lines that resemble a stylized eye or a series of ripples. Behind the letter 'P' is a light gray graphic of a stylized evergreen tree. Below the tree and the letter 'P' is a light gray landscape element consisting of a solid gray base with a white, wavy line representing a path or a body of water.

What if you could magically  
guess which element of the  
search space was the one  
you wanted?

# A Sample Problem

4	3	11	9	7	13	5	6	1	12	2	8	0	10
---	---	----	---	---	----	---	---	---	----	---	---	---	----

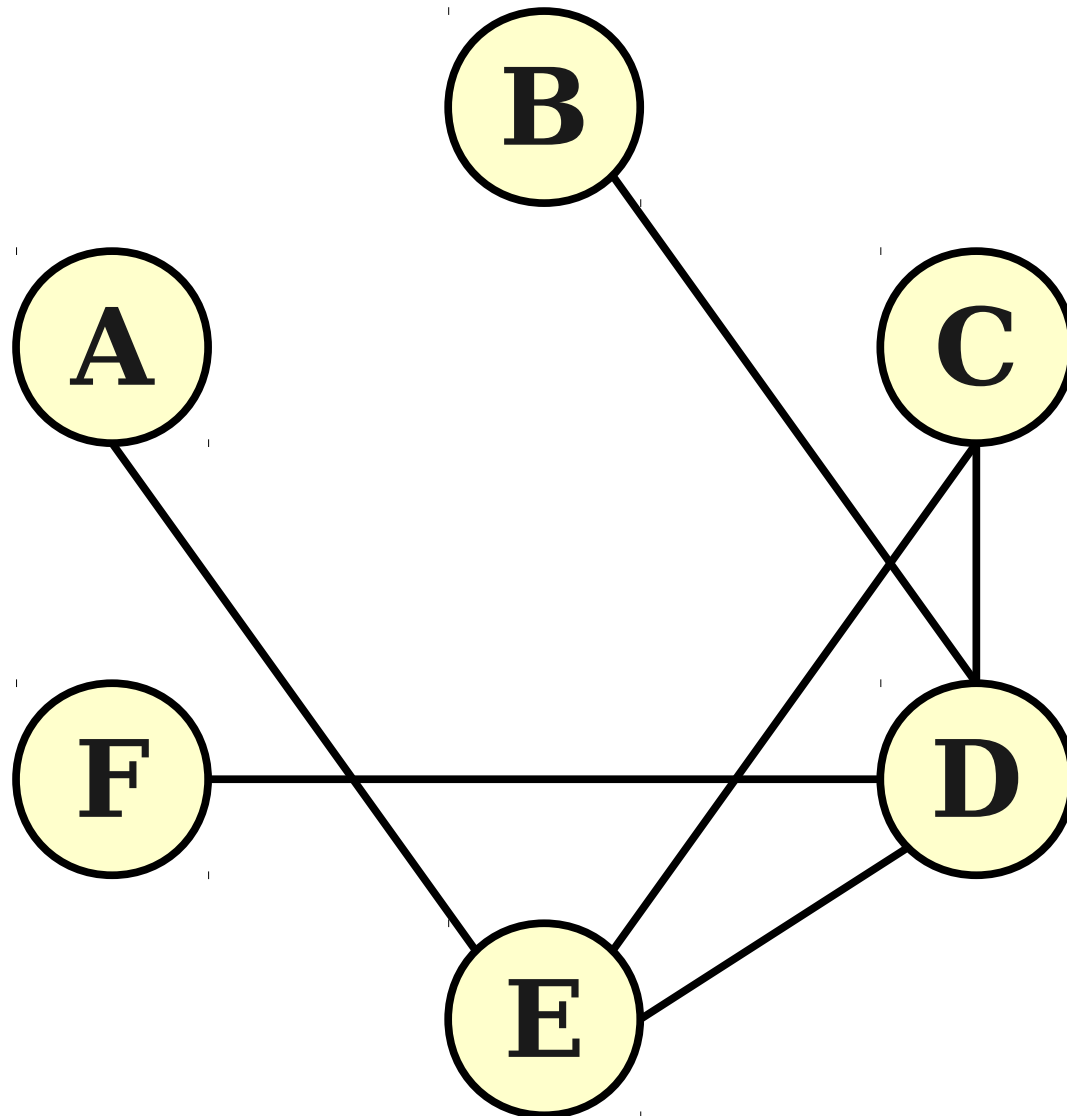
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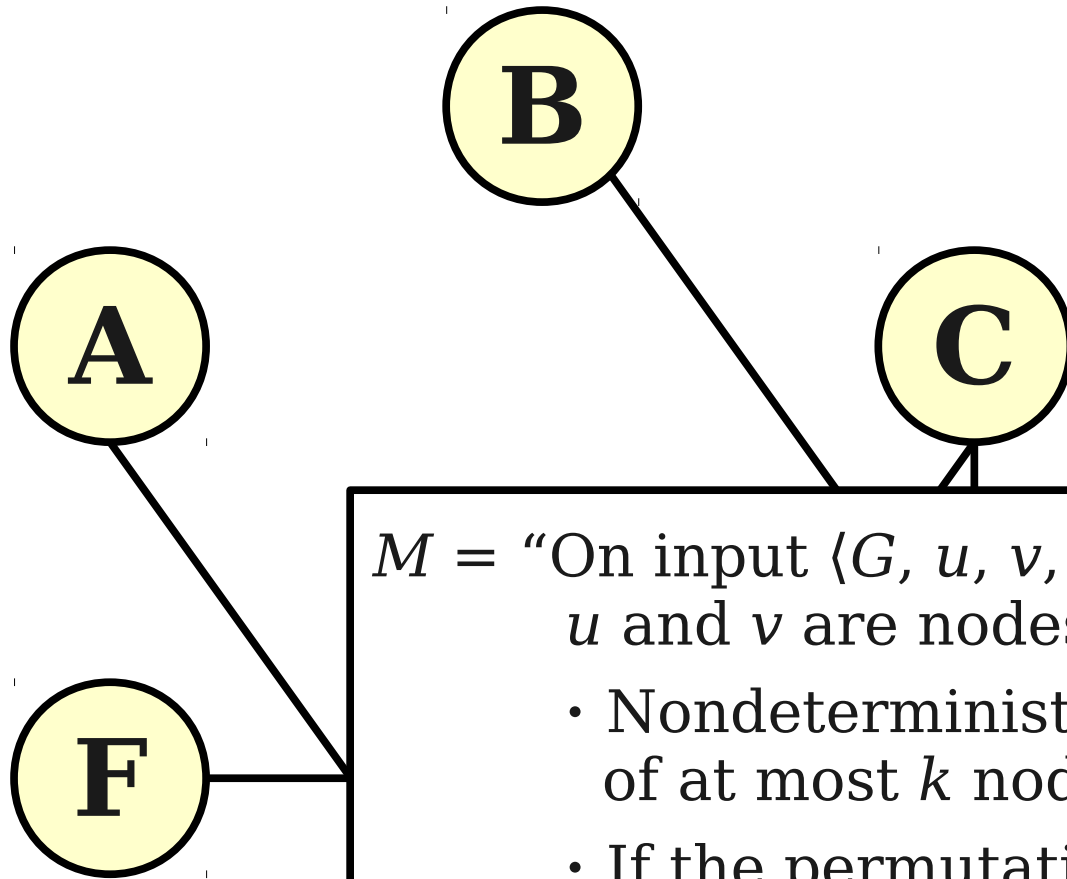
$M =$  “On input  $\langle S, k \rangle$ , where  $S$  is a sequence of numbers and  $k$  is a natural number:

- Nondeterministically guess a subsequence of  $S$ .
- If it is an ascending subsequence of length at least  $k$ , accept.
- Otherwise, reject.”

# Another Problem



# Another Problem



$M =$  “On input  $\langle G, u, v, k \rangle$ , where  $G$  is a graph,  $u$  and  $v$  are nodes in  $G$ , and  $k \in \mathbb{N}$ :

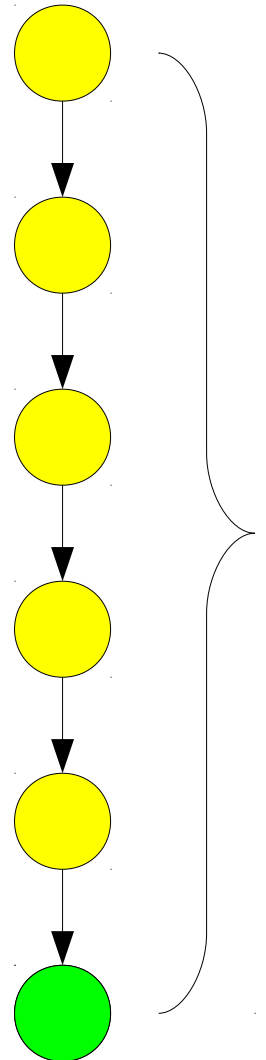
- Nondeterministically guess a permutation of at most  $k$  nodes from  $G$ .
- If the permutation is a path from  $u$  to  $v$ , accept.
- Otherwise, reject.

How do we measure NTM efficiency?



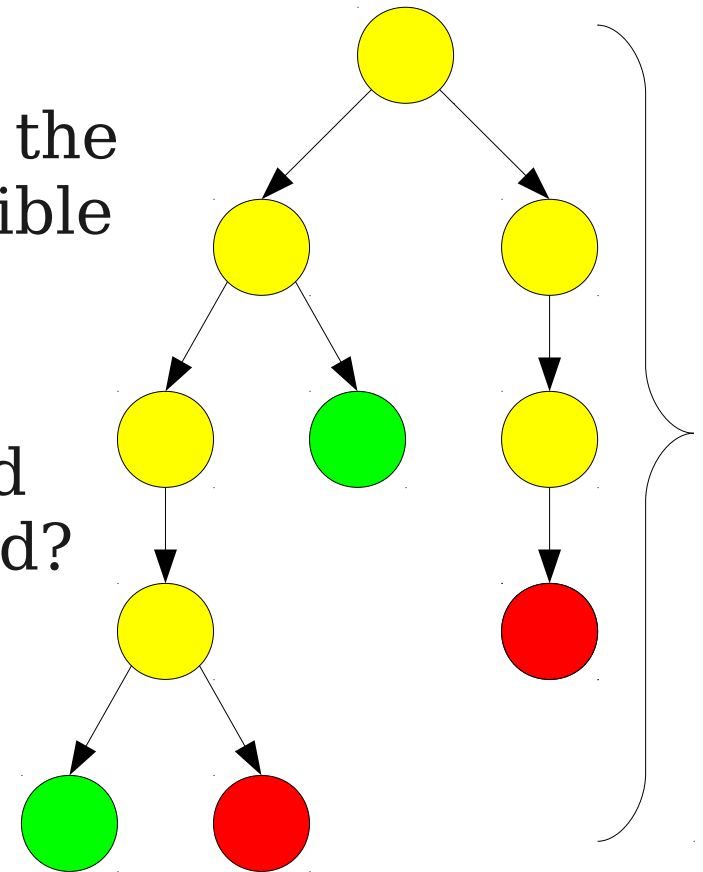
# Analyzing NTMs

- When discussing deterministic TMs, the notion of time complexity is (reasonably) straightforward.
- **Recall:** One way of thinking about nondeterminism is as a tree.
- In a **deterministic** computation, the tree is a straight line.
- The time complexity is the height of that straight line.

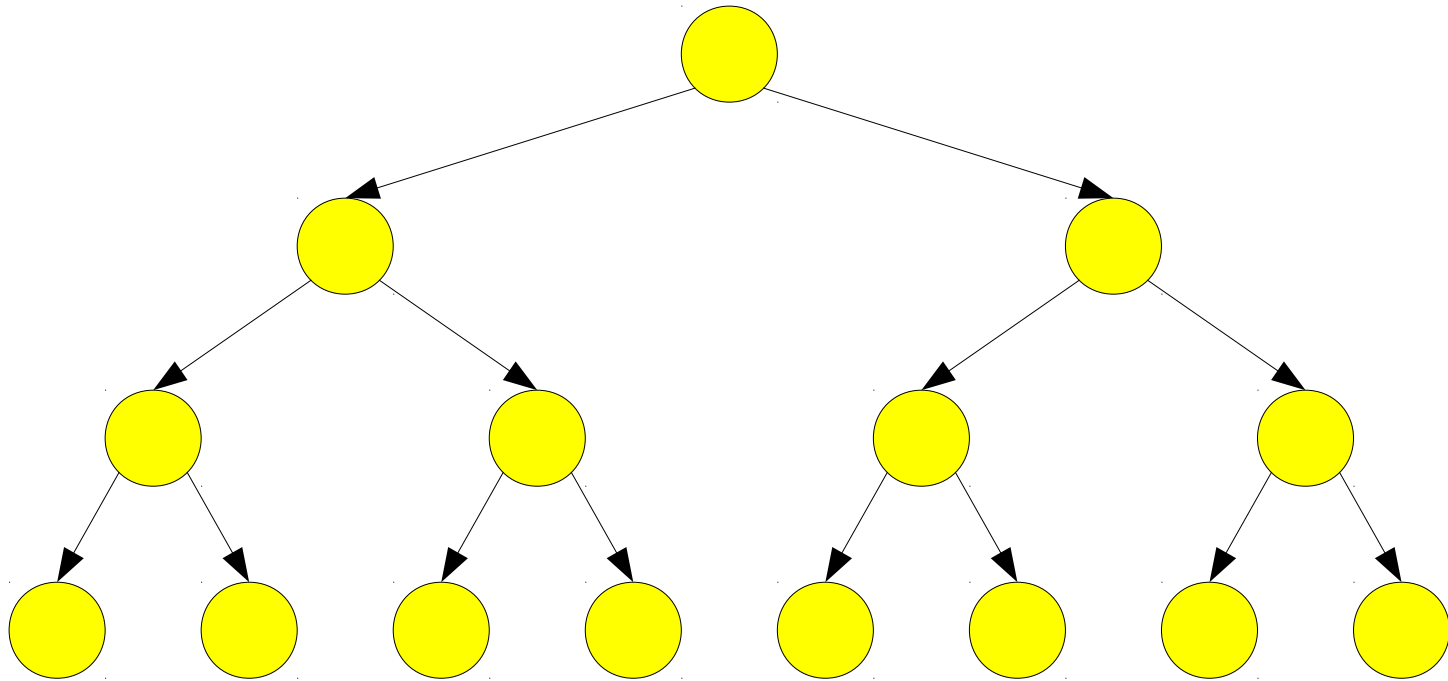


# Analyzing NTMs

- When discussing deterministic TMs, the notion of time complexity is (reasonably) straightforward.
- **Recall:** One way of thinking about nondeterminism is as a tree.
- The time complexity is the height of the tree (the length of the **longest** possible choice we could make).
- Intuition: If you ran all possible branches in parallel, how long would it take before all branches completed?



# The Size of the Tree



# From NTMs to TMs

- **Theorem:** For any NTM with time complexity  $f(n)$ , there is a TM with time complexity  $2^{O(f(n))}$ .
- **It is unknown whether it is possible to do any better than this in the general case.**
- NTMs are capable of exploring multiple options in parallel; this “seems” inherently faster than deterministic computation.

# The Complexity Class **NP**

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

$$\mathbf{NP} = \{ L \mid \text{There is a nondeterministic TM that decides } L \text{ in polynomial time.} \}$$

What types of problems are in **NP**?

# A Problem in NP

- Does a Sudoku grid have a solution?
  - $M$  = “On input  $\langle S \rangle$ , an encoding of a Sudoku puzzle:
    - **Nondeterministically** guess how to fill in all the squares.
    - **Deterministically** check whether the guess is correct.
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		7		6		1		
					3		5	2
3			1		5	9		7
6		5		3		8		9
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7	1	9	5	4	8	3	2	6
8	3	2	6	1	9	5	7	4
1	6	3	2	5	7	4	9	8
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9	2	4	3	8	1	7	6	5

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For an arbitrary  $n^2 \times n^2$  grid:

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
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Total number of cells in the grid:

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For an arbitrary  $n^2 \times n^2$  grid:

Total number of cells in the grid:  $n^4$

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
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6	4	5	7	3	2	8	1	9
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For an arbitrary  $n^2 \times n^2$  grid:

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Total time to fill in the grid:  $O(n^4)$

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Total number of rows, columns, and boxes to check:  $O(n^2)$

Total time required to check each row/column/box:

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
6	4	5	7	3	2	8	1	9
7	1	9	5	4	8	3	2	6
8	3	2	6	1	9	5	7	4
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Total number of rows, columns, and boxes to check:  $O(n^2)$

Total time required to check each row/column/box:  $O(n^2)$

Total runtime:

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
6	4	5	7	3	2	8	1	9
7	1	9	5	4	8	3	2	6
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Total number of cells in the grid:  $n^4$

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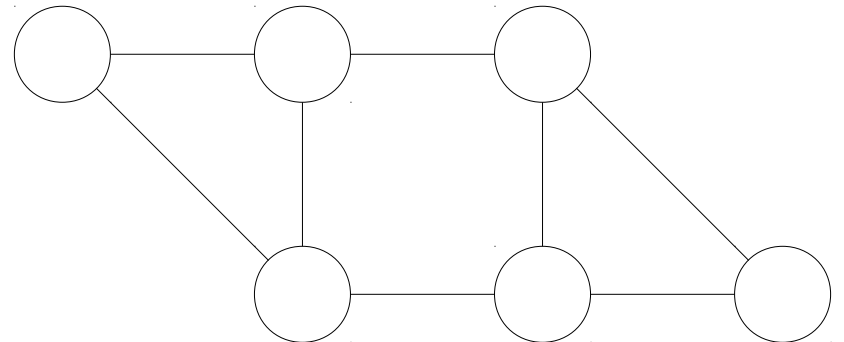
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Total runtime:  $O(n^4)$

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
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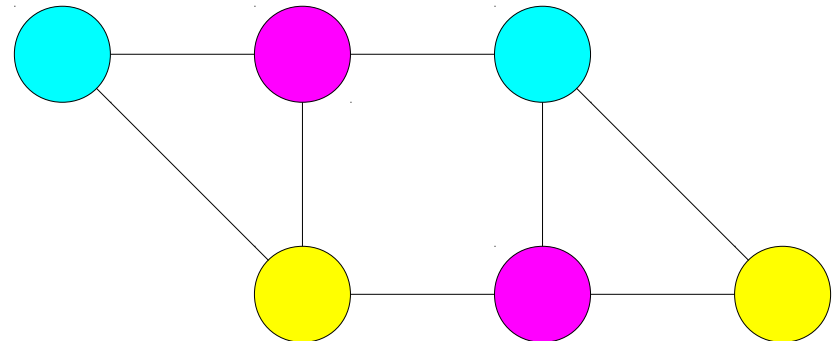
# A Problem in NP

- A **graph coloring** is a way of assigning colors to nodes in an undirected graph such that no two nodes joined by an edge have the same color.
  - Applications in compilers, cell phone towers, etc.
- Question: Can graph  $G$  be colored with at most  $k$  colors?



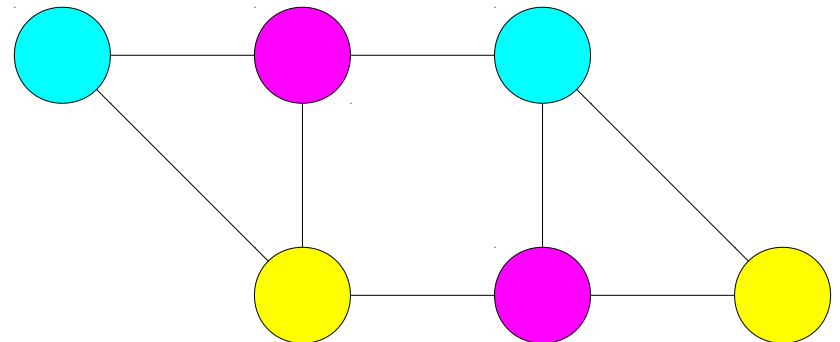
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  - Applications in compilers, cell phone towers, etc.
- Question: Can graph  $G$  be colored with at most  $k$  colors?
- $M =$  “On input  $\langle G, k \rangle$ :
  - **Nondeterministically** guess a  $k$ -coloring of the nodes of  $G$ .
  - **Deterministically** check whether it is legal.
  - If so, accept; if not, reject.”



# Other Problems in **NP**

- **Subset sum:**

Given a set  $S$  of natural numbers and a target number  $n$ , is there a subset of  $S$  that sums to  $n$ ?

- **Longest path:**

- Given a graph  $G$ , a pair of nodes  $u$  and  $v$ , and a number  $k$ , is there a simple path from  $u$  to  $v$  of length at least  $k$ ?

- **Job scheduling:**

- Given a set of jobs  $J$ , a number of workers  $k$ , and a time limit  $t$ , can the  $k$  workers, working in parallel complete all jobs in  $J$  within time  $t$ ?

# Problems and Languages

- Abstract question: does a Sudoku grid have a solution?
- Formalized as a language:

**$SUDOKU = \{ \langle S \rangle \mid S \text{ is a solvable Sudoku grid.} \}$**

- In other words:

$S$  is solvable iff  $\langle S \rangle \in SUDOKU$

# Problems and Languages

- Abstract question: can a graph be colored with  $k$  colors?
- Formalized as a language:

**COLOR = {  $\langle G, k \rangle$  |  $G$  is an undirected graph,  $k \in \mathbb{N}$ , and  $G$  is  $k$ -colorable. }**

- In other words:

$G$  is  $k$ -colorable iff  $\langle G, k \rangle \in \text{COLOR}$



# A General Pattern

- The NTMs we have seen so far always follow this pattern:
  - $M =$  “On input  $w$ :
    - **Nondeterministically** guess some object.
    - **Deterministically** check whether this was the right guess.
    - If so, accept; otherwise, reject.”
- Intuition: The NTM is searching for some *proof* that  $w$  belongs to some language  $L$ .
  - If  $w \in L$ , it can guess the proof.
  - If  $w \notin L$ , it will never guess the proof.

# An Intuition for **NP**

- Intuitively, a language  $L$  is in **NP** iff there is an easy way of proving strings in  $L$  actually belong to  $L$ .
- If  $w \in L$ , there is some information that can easily be used to convince someone that  $w \in L$ .

# A Problem in NP

		7		6		1		
					3		5	2
3			1		5	9		7
6		5		3		8		9
	1						2	
8		2		1		5		4
1		3	2		7			8
5	7		4					
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# A Problem in NP

2	5	7	9	6	4	1	8	3
4	9	1	8	7	3	6	5	2
3	8	6	1	2	5	9	4	7
6	4	5	7	3	2	8	1	9
7	1	9	5	4	8	3	2	6
8	3	2	6	1	9	5	7	4
1	6	3	2	5	7	4	9	8
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# A Problem in **NP**

9	3	11	4	2	13	5	6	1	12	7	8	0	10
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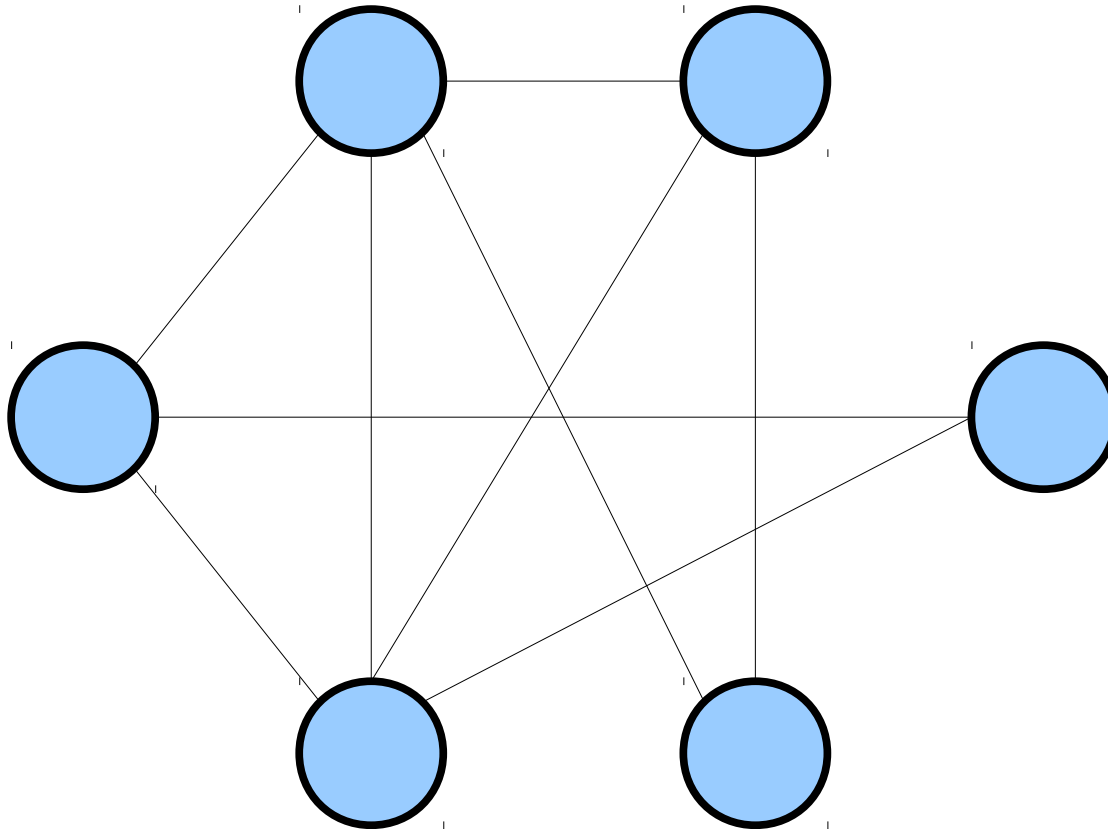
Is there an ascending subsequence of  
length at least 7?

# A Problem in **NP**

9	3	11	4	2	13	5	6	1	12	7	8	0	10
---	---	----	---	---	----	---	---	---	----	---	---	---	----

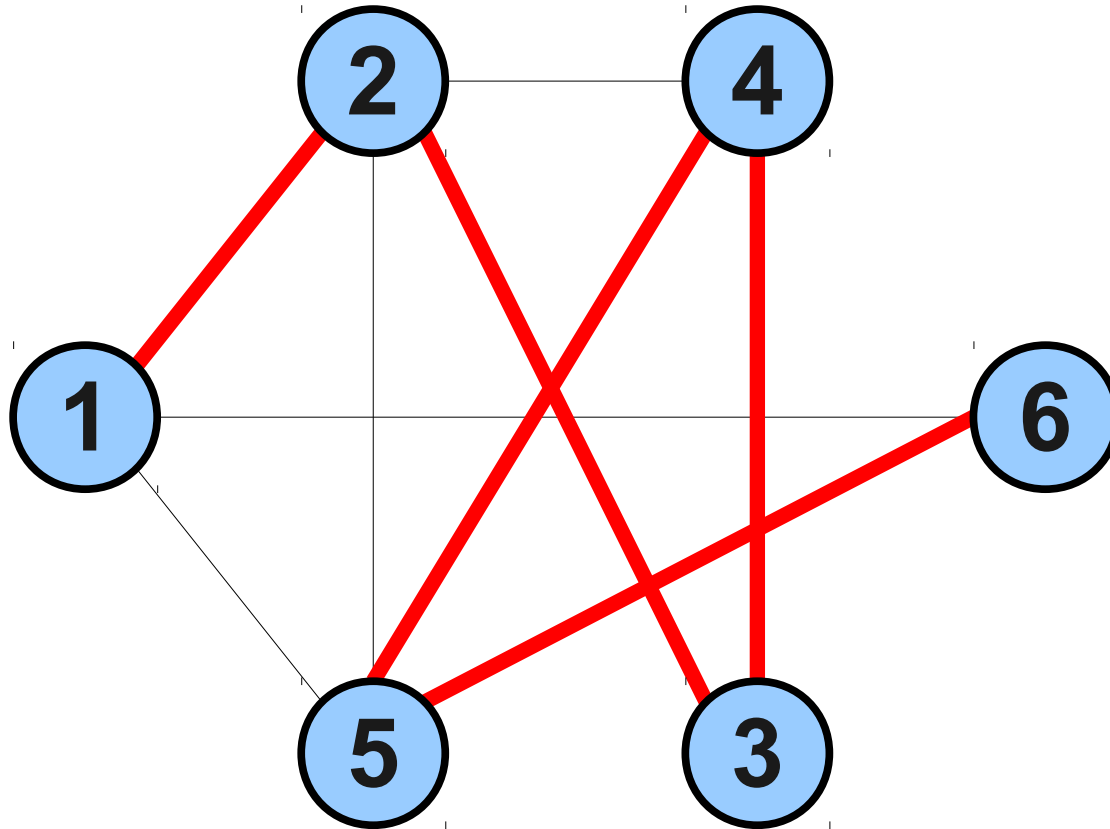
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# A Problem in **NP**



Is there a simple path that goes through every node exactly once?

# A Problem in **NP**



Is there a simple path that goes through every node exactly once?



# Another View of **NP**

- **Theorem:**  $L \in \mathbf{NP}$  iff there is a *deterministic* TM  $V$  with the following properties:
  - $w \in L$  iff there is some  $c \in \Sigma^*$  such that  $V$  accepts  $\langle w, c \rangle$ .
  - $V$  runs in time polynomial in  $|w|$ .

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  - $V$  runs in time polynomial in  $|w|$ .
- **Intuition:** Think about how you would convince someone what a string  $w$  belongs to an **NP** language  $L$ .
  - If  $w \in L$ , there is some information you can provide to easily convince someone that  $w \in L$ .
  - If  $w \notin L$ , then no information you provide can convince someone that  $w \in L$ .

# Another View of **NP**

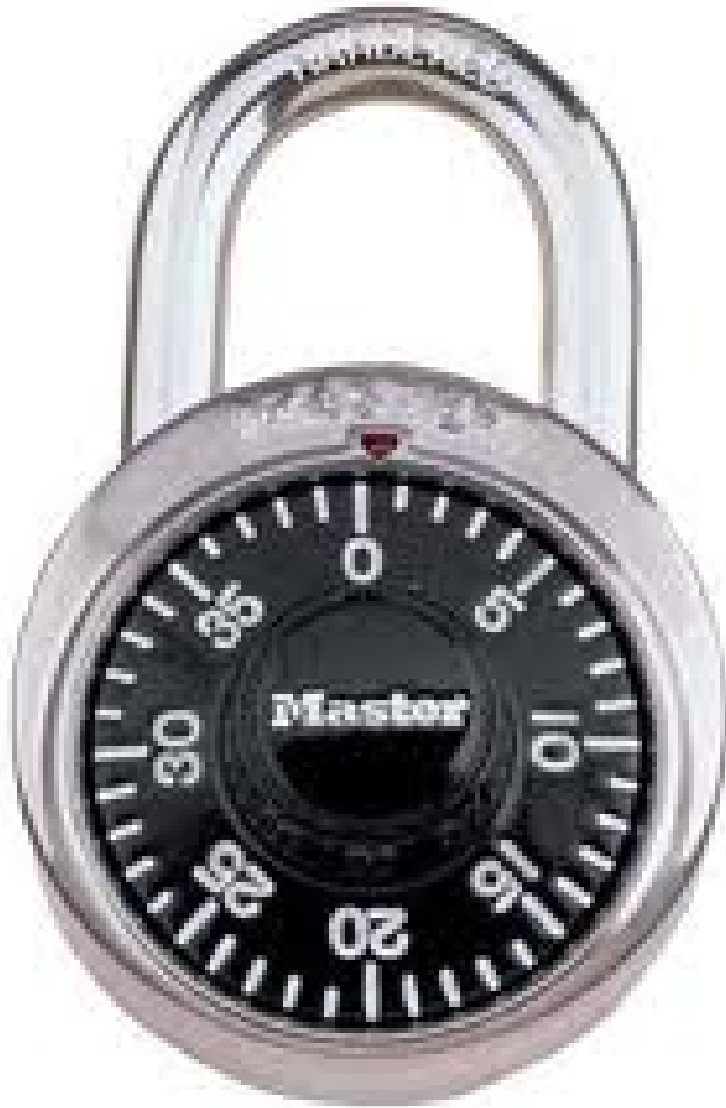
- **Theorem:**  $L \in \mathbf{NP}$  iff there is a *deterministic* TM  $V$  with the following properties:
  - $w \in L$  iff there is some  $c \in \Sigma^*$  such that  $V$  accepts  $\langle w, c \rangle$ .
  - $V$  runs in time polynomial in  $|w|$ .
- Some terminology:
  - A TM  $V$  with the above property is called a **polynomial-time verifier for  $L$** .
  - The string  $c$  is called a **certificate** for  $w$ .
  - You can think of  $V$  as checking the certificate that proves  $w \in L$ .

# An Efficiently Verifiable Puzzle

# An Efficiently Verifiable Puzzle



# An Efficiently Verifiable Puzzle



**Question: Can this lock be opened?**

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- **Theorem:**  $L \in \mathbf{NP}$  iff there is a *deterministic* TM  $V$  with the following properties:
  - $w \in L$  iff there is some  $c \in \Sigma^*$  such that  $V$  accepts  $\langle w, c \rangle$ .
  - $V$  runs in time polynomial in  $|w|$ .
- Important properties of  $V$ :
  - If  $V$  accepts  $\langle w, c \rangle$ , then we're guaranteed  $w \in L$ .
  - If  $V$  does not accept  $\langle w, c \rangle$ , then either
    - $w \in L$ , but you gave the wrong  $c$ , or
    - $w \notin L$ , so no possible  $c$  will work.

# Another View of **NP**

- **Theorem:**  $L \in \mathbf{NP}$  iff there is a *deterministic* TM  $V$  with the following properties:
  - $w \in L$  iff there is some  $c \in \Sigma^*$  such that  $V$  accepts  $\langle w, c \rangle$ .
  - $V$  runs in time polynomial in  $|w|$ .
- Important observations:
  - $\mathcal{L}(V)$  is **not** the language  $L$ .
  - $L$  is the set of strings in the language, while  $\mathcal{L}(V)$  is a set of strings in the language paired with certificates.
  - $V$  **must** be deterministic.



# Another View of **NP**

- **Theorem:**  $L \in \mathbf{NP}$  iff there is a *deterministic* TM  $V$  with the following properties:
  - $w \in L$  iff there is some  $c \in \Sigma^*$  such that  $V$  accepts  $\langle w, c \rangle$ .
  - $V$  runs in time polynomial in  $|w|$ .
- **Proof sketch:**
  - If there is a verifier  $V$  for  $L$ , we can build a poly-time NTM for  $L$  by nondeterministically guessing a certificate  $c$ , then running  $V$  on  $w$ .
  - If there is a poly-time NTM for  $L$ , we can build a verifier for it. The certificate is the sequence of choices the NTM should make, and  $V$  checks that this sequence accepts.

# A Problem in NP

- Does a Sudoku grid have a solution?
  - $M =$  “On input  $\langle S, A \rangle$ , an encoding of a Sudoku puzzle and an alleged solution to it:
    - **Deterministically** check whether  $A$  is a solution to  $S$ .
    - If so, accept; if not, reject.”

		7		6		1		
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  - Applications in compilers, cell phone towers, etc.
- Question: Can  $G$  be colored with at most  $k$  colors?
- $M =$  “On input  $\langle\langle G, k \rangle, C \rangle$ , where  $C$  is an alleged coloring:
  - **Deterministically** check whether  $C$  is a legal  $k$ -coloring of  $G$ .
  - If so, accept; if not, reject.”

