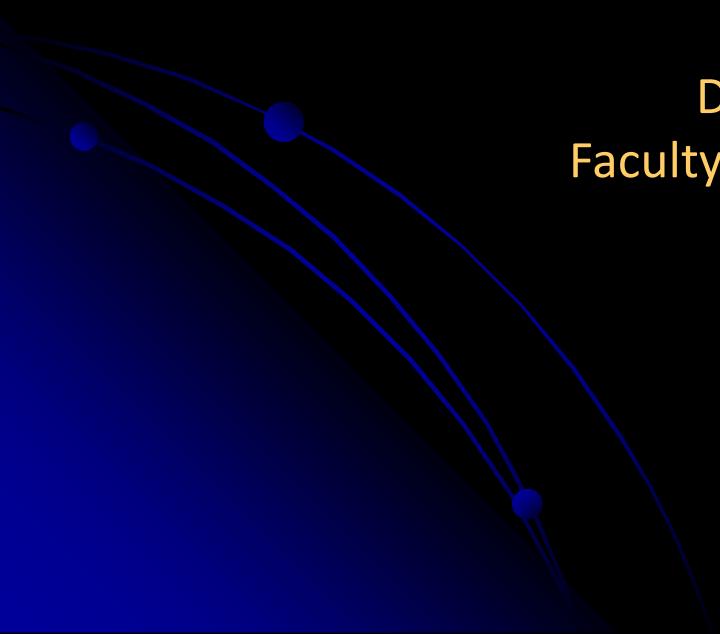


# Configuration variables

Kombinatorická optimalizace (NI-KOP)  
Combinatorial Optimization (NIE-KOP)  
2024/2025

Dept. of Digital Design  
Faculty of Information Technology  
CTU in Prague



# Configuration

## ➤ Recall...

- Given by the problem
  - no matter what algorithm is used
- Describes the state of the search
- Brute force tries all configurations
- For constructive problems, the configuration typically equals to the output

# Configuration variables

## ➤ Recall...

- Given by the problem
    - no matter what algorithm is used
    - ... *but for some exceptions*
    - ... *and their implementation effectiveness may differ*
  - They encode the configuration
  - Important:
    - finite number
    - finite and discrete domains
- ⇒ It must be possible to evaluate all their values by brute force

# Configuration variables

## ➤ Recall...

- For constructive problems, they are typically equal to output variables
- Every solution must have a corresponding configuration

## ➤ The variables must

- Represent any configuration which is a solution,
- Together with input variables, they allow to compute the constraints (to verify whether the configuration is a valid solution),
- Together with input variables, allow to evaluate the optimization criterion for optimization problems.

## ➤ The variables should

- Support any required operation effectively
  - e.g. enumeration for a brute force search
- Not represent a number of configurations that cannot be solutions

# Examples

## ➤ Sources:

- A compendium of NP optimization problems
  - <http://www.csc.kth.se/~viggo/wwwcompendium/>
- Wikipedia
  - [https://en.wikipedia.org/wiki/List\\_of\\_NP-complete\\_problems](https://en.wikipedia.org/wiki/List_of_NP-complete_problems)

Note that these are optimization problems

- but this doesn't matter – for decision (or enumeration, etc.) problems, the configuration variables will be the same

# Maximum Knapsack

## ➤ Problem definition

- Instance:
  - $M$  – knapsack capacity
  - $U$  – a set of items, for each  $u \in U$ :
    - $w(u)$  – a weight
    - $c(u)$  – a cost

### ● Solution:

- A subset  $U' \subseteq U$ , so that

$$\sum_{u \in U'} w(u) \leq M$$

### ● Optimization criterion (measure)

$$\sum_{u \in U'} c(u) \quad - \text{to be maximized}$$

## ➤ Configuration

- The subset  $U' \subseteq U$   
(which items are in the knapsack)

## ➤ Configuration vars.

- Bit vector
- Set
- Unordered list
- ...

# Maximum Knapsack

## ➤ Configuration

- Subset of  $U$   
(which items are in the knapsack)

## ➤ Configuration vars.

- Bit vector
- Set
- Unordered list
- ...

Brute force enumerates all their values



- Can they represent any solution?
- Can we check the constraints?
- Can we evaluate the optimization criterion?
- How can we enumerate them?
- How many possibilities?
- Which configurations are valid solutions?

# Maximum Knapsack

## ➤ Configuration

- Subset of  $U$   
(which items are in the knapsack)

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- Bit vector
- Set
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- ...

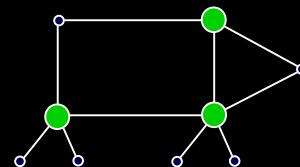
Incompetent configuration variables:  
**the thing which is in the knapsack**

- Can they represent any solution? NO – solutions with more things are missing

# Minimum Vertex Cover

## ➤ Problem definition

- Instance:
  - Graph  $G = (V, E)$
- Solution:
  - A vertex cover for  $G$ , i.e., a subset  $V' \subseteq V$ , such that, for each edge  $(u, v) \in E$  at least one of  $u$  and  $v$  belongs to  $V'$
- Optimization criterion (measure)
  - Cardinality of the vertex cover, i.e.,  $|V'|$ . Should be minimized.



## ➤ Configuration

- The subset  $V' \subseteq V$   
(which vertices are in the cover)

## ➤ Configuration vars.

- Bit vector
- Set
- Unordered list
- Subgraph
- ...

# Minimum Vertex Cover

## ➤ Configuration

- Subset of  $V$   
(which vertices are in the cover)

## ➤ Configuration vars.

- Bit vector
- Set
- Unordered list
- Subgraph
- ...

Brute force enumerates all their values



- Can they represent any solution?
- Can we check the constraints?
- Can we evaluate the optimization criterion?
- How can we enumerate them?
- How many possibilities?
- Which configurations are valid solutions?

# Travelling Salesperson Problem

## ➤ Problem definition

- Instance:
  - Set  $C$  of  $m$  cities, distances  $d(c_i, c_j) \in \mathbb{N}$  for each pair  $c_i, c_j \in C$
- Solution:
  - A tour of  $C$ , i.e., a permutation  $\pi: [1..m] \rightarrow [1..m]$
- Optimization criterion (measure)
  - The length of the tour, i.e., the summary distance of the cities in the permutation  $\pi$ . Should be minimized.

## ➤ Configuration

- The permutation  $\pi$  (i.e., the tour)

## ➤ Configuration vars.

- Ordered list of cities
- A permuted vector
- ...

# Travelling Salesperson Problem

## ➤ Configuration

- The permutation  $\pi$   
(i.e., the tour)

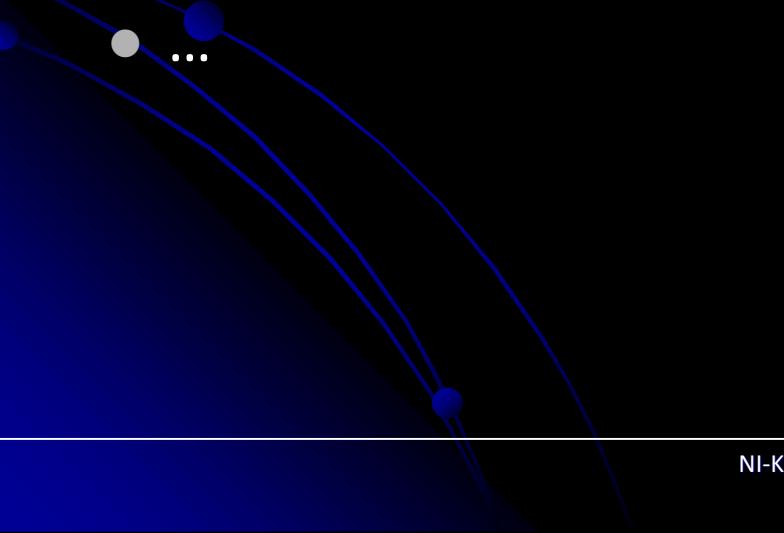
## ➤ Configuration vars.

- Ordered list of cities
- Permuted vector

Brute force enumerates all their values



- Can they represent any solution?
- Can we check the constraints?
- Can we evaluate the optimization criterion?
- How can we enumerate them?
- How many possibilities?
- Which configurations are valid solutions?



# Travelling Salesperson Problem

## ➤ Configuration

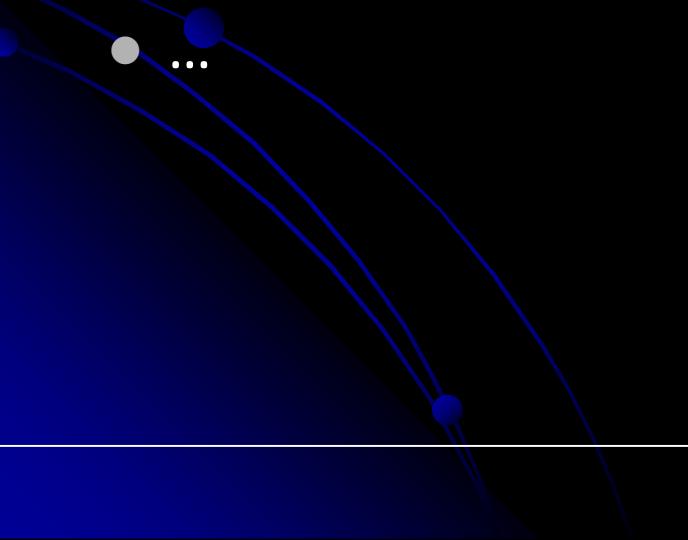
- The permutation  $\pi$   
(i.e., the tour)

## ➤ Configuration vars.

- Ordered list of cities
- Permuted vector

An ordered list of cities can contain:

- Less cities than  $m$
- More cities than  $m$
- One city twice
- The implementation of configuration variables is correct but inefficient
- A brute force search takes much longer than necessary



# Minimum Bin Packing

## ➤ Problem definition

- Instance:
  - Finite set  $U$  of items, a size  $s(u) \in \mathbb{Z}^+$  for each  $u \in U$
  - A positive integer bin capacity  $B$
- Solution:
  - A partition of  $U$  into disjoint sets  $U_1, U_2, \dots, U_m$ , such that the sum of the sizes of items in each  $U_i$  is  $B$  or less
  - Optimization criterion (measure)
    - The number of used bins, i.e., the number of disjoint sets,  $m$ . Should be minimized.

## ➤ Configuration

- The **partition of  $U$** , i.e. the placement of items into the bins

## ➤ Configuration vars.

1. For each item, in which bin it is placed
2. For each bin, what items it contains

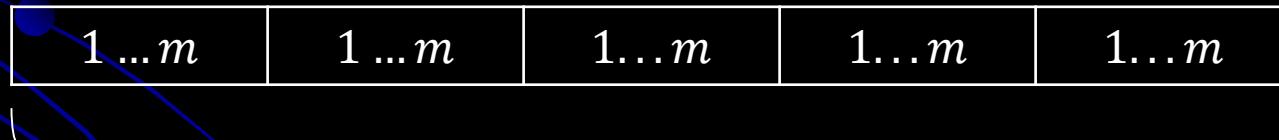
# Minimum Bin Packing

## ➤ Problem definition

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- Solution: A partition of  $U$  into disjoint sets  $U_1, U_2, \dots, U_m$ , such that the sum of the sizes of items in each  $U_i$  is  $B$  or less

## ➤ Configuration vars.

1. For each item, in which bin it is placed



⇒ array of  $|U|$  integers

How can we enumerate them?  
How many possibilities?  
Which configurations are valid  
solutions?

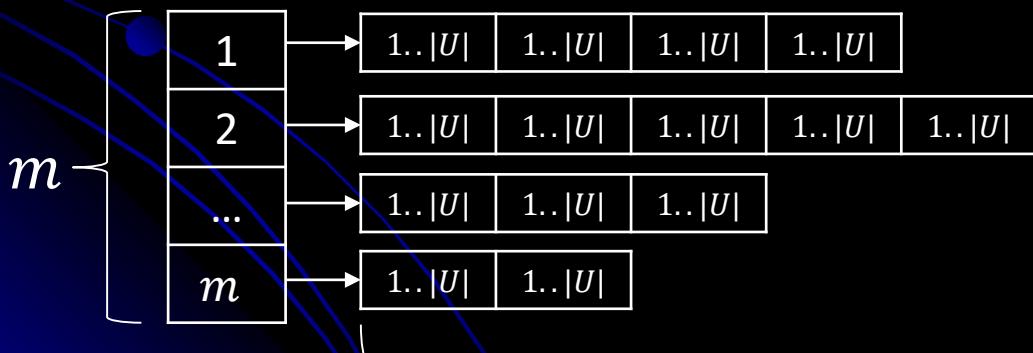
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## ➤ Configuration vars.

2. For each bin, what items it contains



⇒  $m$ -dimensional array of  $|U|$   
unordered lists (sets) of integers

How can we enumerate them?  
How many possibilities?  
Which configurations are valid solutions?



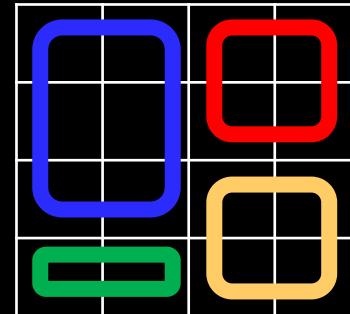
# Minimum Rectangle Tiling

## ➤ Problem definition

- Instance:
  - An  $n \times n$  array  $A$  of non-negative numbers, positive integer  $p$ .
- Solution:
  - A partition of  $A$  into  $p$  non-overlapping rectangular subarrays.
- Optimization criterion (measure)
  - The maximum weight of any rectangle in the partition. The weight of a rectangle is the sum of its elements. To be minimized.

## ➤ Configuration

- The **partition of  $A$** , i.e. the rectangular subarrays



## ➤ Configuration vars.

How to encode the rectangular subarrays?

# Minimum Rectangle Tiling

## ➤ Problem definition

- Instance: An  $n \times n$  array  $A$  of non-negative numbers, positive integer  $p$ .
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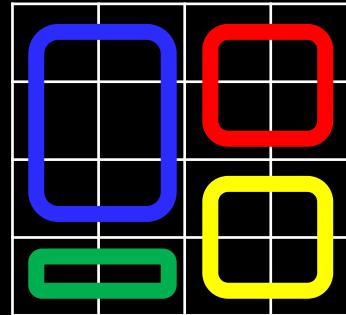
## ➤ Configuration

- The rectangular subarrays

## ➤ Configuration vars.

1. Set of rectangles specified by coordinates

$\{((0, 0), (1, 2)); ((2, 0), (3, 1));$   
 $((0, 3), (1, 3)); ((2, 2), (3, 3))\}$



How can we enumerate them?  
How many possibilities?  
Which configurations are valid solutions?



# Minimum Rectangle Tiling

## ➤ Problem definition

- Instance: An  $n \times n$  array  $A$  of non-negative numbers, positive integer  $p$ .
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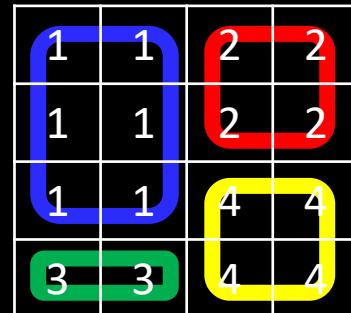
## ➤ Configuration

- The rectangular subarrays

## ➤ Configuration vars.

2. Which cell belongs to which rectangle

$\Rightarrow n \times n$  array of integers



How can we enumerate them?  
How many possibilities?  
Which configurations are valid solutions?



# Minimum Consistent Finite Automaton

## ➤ Problem definition

- Instance:
  - Two finite sets  $P, N$  of binary strings.

### ● Solution:

- A deterministic finite automaton  $(Q, \{0, 1\}, \delta, q_0, F)$  such that strings in  $P$  are accepted and strings in  $N$ .

- Optimization criteria:
  - The number of states in the automaton. Smaller is better.

## ➤ Configuration

- The automaton

## ➤ Configuration vars.

How to encode an automaton?

Transition table:

$Q_t$	$Q_{t+1}/0$	$Q_{t+1}/1$	$Q_t \in F$
$q_0$	$q_1$	$q_2$	y/n
$q_1$	$q_3$	$q_2$	y/n
$q_2$	...	...	y/n
...	...	...	y/n
$q_n$	...	...	y/n

How can we enumerate them?  
How many possibilities?  
Which configurations are valid solutions?



# Minimum Graph Motion Planning

## ➤ Problem definition

- **Instance:**

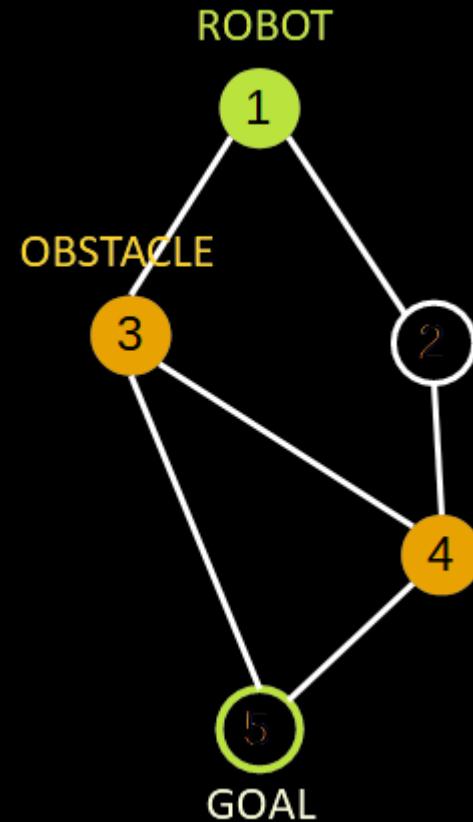
- Graph  $G = (V, E)$ , start vertex  $s \in V$ , where the robot is initially placed, goal vertex  $t \in V$ , and a subset  $W \subseteq V$  of vertices where the obstacles are initially placed.

- **Solution:**

- A motion scheme for the robot and the obstacles. In each time step either the robot or one obstacle may be moved to a neighboring vertex that is not occupied by the robot or an obstacle.

- **Optimization criterion (measure)**

- The number of steps until the robot has been moved to the goal vertex  $t$ .



# Minimum Graph Motion Planning

## ➤ Problem definition

- Instance:

- Graph  $G = (V, E)$ , start vertex  $s \in V$ , where the robot is initially placed, goal vertex  $t \in V$ , and a subset  $W \subseteq V$  of vertices where the obstacles are initially placed.

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- Optimization criterion (measure)

- The number of steps until the robot has been moved to the goal vertex  $t$ .

## ➤ Configuration

- The motion scheme

## ➤ Configuration vars.

- Sequence of moves of the robot/obstacles

How can we enumerate them?

How many possibilities?

Which configurations are valid solutions?



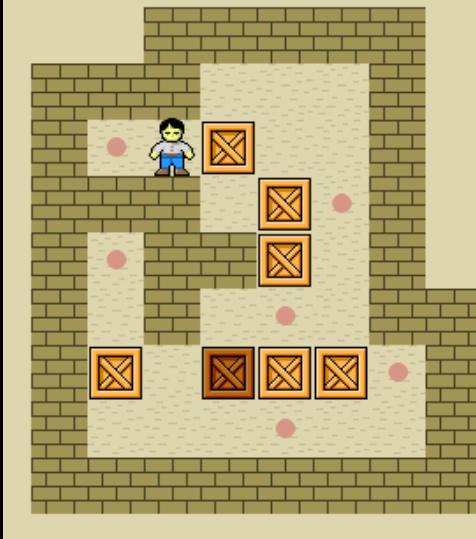
# Sokoban

## ➤ What is Sokoban?

- A game from 1980
- But also an important combinatorial problem
  - It is proven to be PSPACE-complete

## ➤ Problem definition (informal)

- Instance:
  - A maze (warehouse) with boxes and a warehouseman
    - The warehouseman can
      - move to unoccupied places in the maze
      - push the boxes to unoccupied places in the maze
  - Target places of boxes
- Solution
  - A motion scheme for the warehouseman
- Optimization criterion (measure)
  - The number of warehouseman steps



# Sokoban

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## ➤ What does it mean?

- One of the hardest problems solvable in polynomial space  
↓
- $\text{SAT} \leq_P \text{Sokoban}$
- $\text{QBF}_k \leq_P \text{Sokoban}$
- ~~$\text{Sokoban} \geq_P \text{SAT}$~~

# Sokoban

## > Problem definition (informal)

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  - A maze (warehouse) with boxes and a warehouseman
  - The warehouseman can
    - move to unoccupied places in the maze
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- Solution
  - A motion scheme for the warehouseman
- Optimization criterion (measure)
  - The number of warehouseman steps

## > Configuration

- A motion scheme for the warehouseman

## > Configuration vars.

- Sequence of moves of the warehouseman
- Or sequence of valid moves of boxes
  - It is “shorter”
  - The movement of the warehouseman can be computed (shortest path in a graph)

# Minimum Equivalent Boolean Formula

## ➤ Problem definition

- Instance:

- Boolean formula consisting of literals, operators, parentheses. E.g.,

$$F = ((a + \bar{b}) \cdot c + b \cdot (a + \bar{c})) \cdot \bar{b}$$

- Solution:

- Equivalent Boolean formula consisting of literals, operators, parentheses. E.g.,

$$F = \bar{b}c$$

- Optimization criterion (measure)

- The number of literals. Should be minimized.

- Note

- This problem is more complex than NP.  
Why?

## ➤ Configuration

- The formula

## ➤ Configuration vars.

How to encode a formula?

- String

How can we enumerate them?

How many possibilities?

Which configurations are valid solutions?



# Rough problems classification

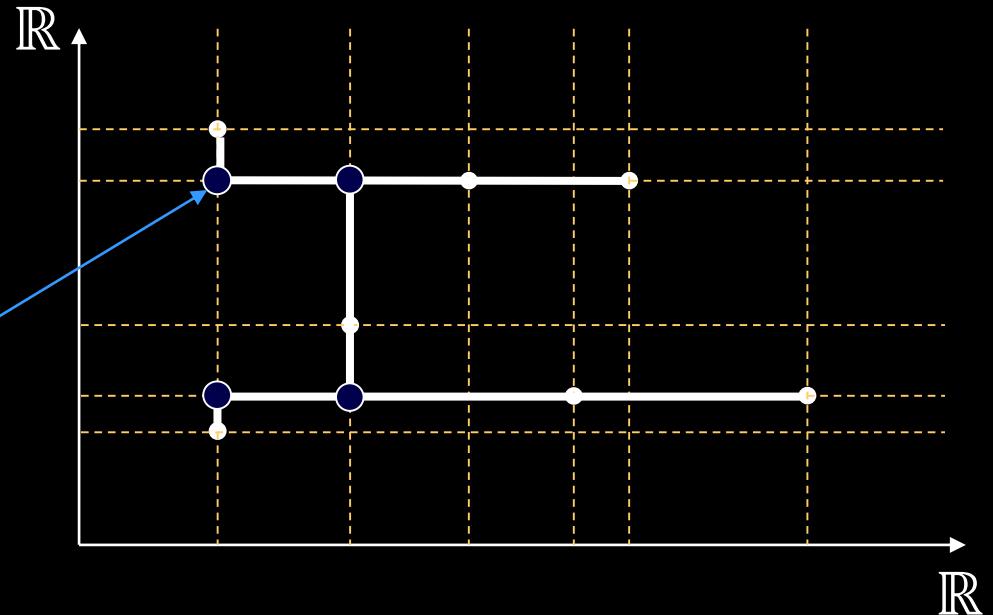
## ➤ What have we found?

- Optimum subset problems
  - Knapsack
  - Vertex cover
  - SAT (0/1 evaluation)
- Optimum permutation
  - TSP
- Optimum assignment
  - Minimum bin packing
- Optimum sequence
  - Minimum graph motion planning
- Other
  - Minimum consistent automaton
  - ...

# Minimum Steiner tree in Manhattan metric

## ➤ Problem definition

- **Instance:**
  - A set  $P \subseteq \mathbb{R} \times \mathbb{R}$  points in a plane
- **Solution:**
  - A finite set of Steiner points  $Q \subseteq \mathbb{R} \times \mathbb{R}$
- **Optimization criterion**
  - The total weight of the minimum spanning tree for the vertex set  $P \cup Q$ , where the weight of an edge  $\langle (x_1, y_1), (x_2, y_2) \rangle$  is the length in Manhattan metric, i.e.,  
$$|x_1 - x_2| + |y_1 - y_2|$$



- ⇒ Discrete coordinates, even though in real numbers
- ⇒ Discretization
- ⇒ Still a combinatorial problem

# The Buckets problem

## ➤ Problem definition

- There are  $n$  buckets, a tap, and a sink.  
Given:
  - Capacities of the buckets
  - Initial amounts of water in each bucket
  - Target amounts of water in each bucket
- In each step you can:
  - Fill any bucket up to the brim (to its maximal capacity)
  - Empty any bucket
  - Pour the water from one bucket to another, either emptying one or fully filling the second bucket
- The goal is to start from the initial state and reach the target amounts of water in the respective buckets
- Optimization criterion
  - The number of operations (pouring) with buckets

# The Buckets problem

## ➤ Discussion

- What is the instance?
  - $n$  buckets
  - Capacities of the buckets
  - Initial amounts of water in each bucket
  - Target amounts of water in each bucket



- What is the configuration?
  - Sequence of operations (pouring)



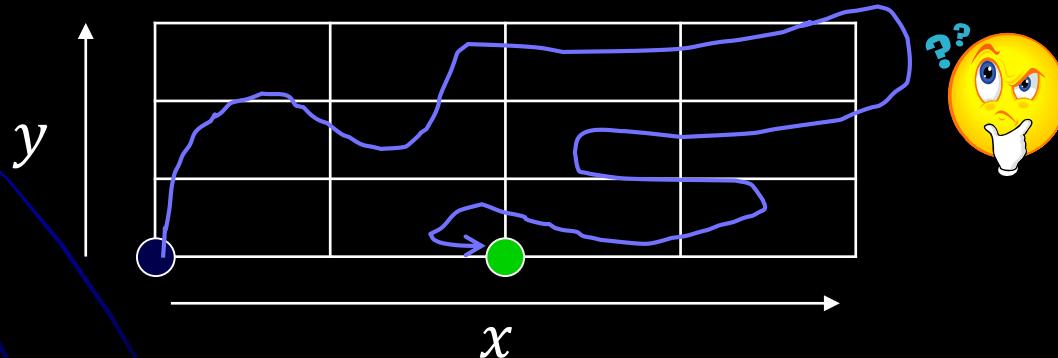
# The Buckets problem

## ➤ Given example instance

- Buckets: 2
- Bucket capacities: (4, 3)
- Initial amounts: (0, 0)
- Target amounts: (2, 0)

## ➤ Whiteboard exercise

Current amounts of water in buckets:  $(x, y)$   
 $x \in \{0, 1, 2, 3, 4\}$   
 $y \in \{0, 1, 2, 3\}$



# The Buckets problem

## ➤ The bucket space

Current amounts of water in buckets:  $(x, y)$   
 $x \in \{0, 1, 2, 3, 4\}$   
 $y \in \{0, 1, 2, 3\}$

- Discrete amounts
  - Finite number of them
  - How many?
- ⇒ Discretization  
⇒ Still a combinatorial problem

