

# Seminar: Search and Optimization

## 2. Search Problems

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September 20, 2012

# Classical Search Problems

# Informal Description

(Classical) search problems are one of the “easiest” and most important classes of AI problems.

Task of an agent:

- starting from an initial state
- apply actions
- to reach a goal state

Measure of performance: Minimize cost of actions

# Motivating Example: 15-Puzzle

9	2	12	6
5	7	14	13
3		1	11
15	4	10	8



1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

More examples later on

# Classical Assumptions

“Classical” assumptions:

- only one agent in the environment (**single agent**)
- always knows the complete world state (**full observability**)
- only the agent can change the state (**static**)
- finite amount of possible states/actions (**discrete**)
- actions change the state **deterministically**

↔ each assumption can be generalized  
(not the focus of this seminar)

We omit “**classical**” in the following.

# Formalization

# State Spaces

To talk about algorithms for search problems we need a **formal definition**.

## Definition (State Space)

A **state space** (or **transition system**) is a 6-tuple  $\mathcal{S} = \langle S, A, cost, T, s_0, S_\star \rangle$  where

- $S$  finite set of **states**
- $A$  finite set of **actions**
- $cost : A \rightarrow \mathbb{R}_0^+$  **action costs**
- $T \subseteq S \times A \times S$  **transition relation**;  
**deterministic in  $\langle s, a \rangle$**  (see next slide)
- $s_0 \in S$  **initial state**
- $S_\star \subseteq S$  set of **goal states**

# State Spaces: Transitions, Determinism

## Definition (Transition, deterministic)

Let  $\mathcal{S} = \langle S, A, cost, T, s_0, S_* \rangle$  be a state space.

The triples  $\langle s, a, s' \rangle \in T$  are called **transitions**.

We say  $\mathcal{S}$  **has the transition**  $\langle s, a, s' \rangle$  if  $\langle s, a, s' \rangle \in T$  and write  $s \xrightarrow{a} s'$  ( $s \rightarrow s'$ , if we do not care about  $a$ ).

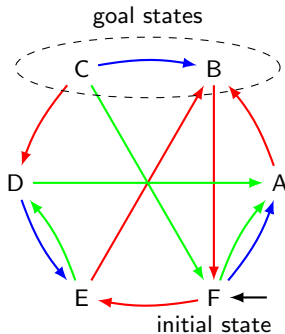
Transitions are **deterministic** in  $\langle s, a \rangle$ :  $s \xrightarrow{a} s_1$  and  $s \xrightarrow{a} s_2$  with  $s_1 \neq s_2$  is not allowed.



# State Space: Example

State spaces are often visualized as **directed graphs**.

- **states**: nodes
- **transitions**: labeled edges  
(here: colors instead of labels)
- **initial state**: node marked with arrow
- **goal states**: marked  
(here: with ellipse)
- **actions**: edge labels
- **action costs**: given separately (or implicit = 1)
- **paths** to goal states correspond to **solutions**
- **shortest paths** correspond to **optimal solutions**



# State Spaces: Terminology

We use common terminology from graph theory.

## Definition (predecessor, successor, applicable action)

Let  $\mathcal{S} = \langle S, A, cost, T, s_0, S_\star \rangle$  be a state space.

Let  $s, s' \in S$  be states with  $s \rightarrow s'$ .

- $s$  is a **predecessor** of  $s'$
- $s'$  is a **successor** of  $s$

If we have  $s \xrightarrow{a} s'$ , action  $a$  is **applicable** in  $s$ .

# State Spaces: Terminology

We use common terminology from graph theory.

## Definition (path)

Let  $\mathcal{S} = \langle S, A, cost, T, s_0, S_* \rangle$  be a state space.

Let  $s^{(0)}, \dots, s^{(n)} \in S$  be states and  $\pi_1, \dots, \pi_n \in A$  actions, with  $s^{(0)} \xrightarrow{\pi_1} s^{(1)}, \dots, s^{(n-1)} \xrightarrow{\pi_n} s^{(n)}$ .

- $\pi = \langle \pi_1, \dots, \pi_n \rangle$  is a **path** from  $s^{(0)}$  to  $s^{(n)}$
- **length** of the path:  $|\pi| = n$
- **cost** of the path:  $cost(\pi) = \sum_{i=1}^n cost(\pi_i)$

## Note:

- paths with length 0 are allowed
- sometimes the state sequence  $\langle s^{(0)}, \dots, s^{(n)} \rangle$  or the sequence  $\langle s^{(0)}, \pi_1, s^{(1)}, \dots, s^{(n-1)}, \pi_n, s^{(n)} \rangle$  are also called **path**

# State Spaces: Terminology

Additional terminology:

**Definition (solution, optimal, solvable, reachable, dead end)**

Let  $\mathcal{S} = \langle S, A, cost, T, s_0, S_\star \rangle$  be a state space.

- A path from a state  $s \in S$  to a state  $s_\star \in S_\star$  is a **solution for/of  $s$** .
- A solution for  $s_0$  is a **solution for/of  $\mathcal{S}$** .
- **Optimal solutions** (for  $s$ ) have minimal cost among all solutions (for  $s$ ).
- State space  $\mathcal{S}$  is **solvable** if a solution for  $\mathcal{S}$  exists.
- State  $s$  is **reachable** if there is a path from  $s_0$  to  $s$ .
- State  $s$  is a **dead end** if no solution for  $s$  exists.

# Representation of State Spaces

# Representation of State Spaces

How to get the state space into the computer?

## 1 As an explicit graph:

Nodes (states) and edges (transitions) represented explicitly, e. g. as **adjacency lists** or as **adjacency matrix**

- **impossible** for **large** problems (needs too much space)
- **Dijkstra** for small problems:  $O(|S| \log |S| + |T|)$

## 2 As a declarative description:

- **compact** description as input  
     $\rightsquigarrow$  state space **exponentially larger** than input
- algorithms work **directly on compact description**  
    (e. g. reformulation, simplification of problem)

# Representation of State Spaces

How to get the state space into the computer?

- ③ **As a black box: abstract interface** for state spaces (used here)

## abstract interface for state spaces

State space  $\mathcal{S} = \langle S, A, cost, T, s_0, S_\star \rangle$  as black box:

- **init()**: creates initial state  
Returns: the state  $s_0$
- **is-goal( $s$ )**: tests if state  $s$  is goal state  
Returns: **true** if  $s \in S_\star$ ; **false** otherwise
- **succ( $s$ )**: lists all applicable actions and successors of  $s$   
Returns: List of tuples  $\langle a, s' \rangle$  with  $s \xrightarrow{a} s'$
- **cost( $a$ )**: determines action cost of action  $a$   
Returns: the non-negative number  $cost(a)$

# Examples



# Examples for Search Problems

- “Toy problems”: combinatorial puzzles (Rubik’s Cube, 15-puzzle, Towers of Hanoi, ...)
- Scheduling, e. g. in factories
- Query optimization in databases
- NPCs in computer games
- Code optimization in compilers
- Verification of soft- and hardware
- Sequence alignment in bio-informatics
- Route planning (e. g. Google Maps)
- ...

Thousands of practical examples!

# Example 1: Blocks world

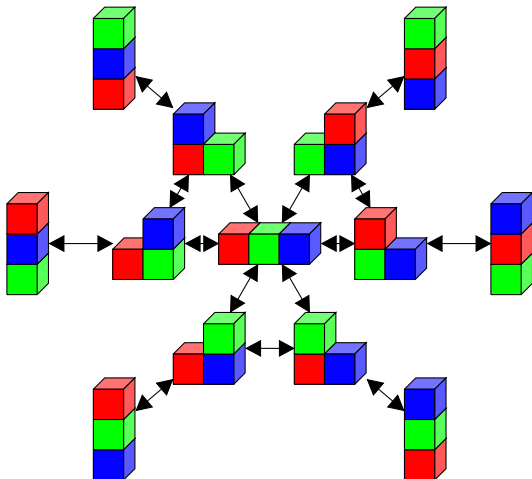
- The **Blocks world** is a traditional example problem in AI.

## Task: blocks world

- Some **colored blocks** are on a table.
- They can be **stacked to towers** but only one block may be moved at a time.
- Our task is to reach a given goal configuration.

# Blocks World with Three Blocks

(action names not shown;  
initial state and goal states can be chosen for each problem)





# Blocks World: Formal Definition

State space  $\langle S, A, cost, T, s_0, S_* \rangle$  blocks world with  $n$  Blocks

State space: blocks world

States  $S$ :

Partitioning of  $\{1, 2, \dots, n\}$  into non-empty (ordered) sequences

Examples:  $\{\langle 1, 2 \rangle, \langle 3 \rangle\} \sim$   ,  $\{\langle 1, 2, 3 \rangle\} \sim$  

Initial state  $s_0$  and goal state  $S_*$ :

different choices possible, e. g.:

- $s_0 = \{\langle 1, 3 \rangle, \langle 2 \rangle\}$
- $S_* = \{\{\langle 3, 2, 1 \rangle\}\}$

# Blocks World: Formal Definition

State space  $\langle S, A, cost, T, s_0, S_* \rangle$  blocks world with  $n$  Blocks

## State space: blocks world

### Actions $A$ :

- $\{move_{b,b'} \mid b, b' \in \{1, \dots, n\} \text{ with } b \neq b'\}$ 
  - Move block  $b$  on top of block  $b'$ .
  - Both have to be topmost block of a tower.
- $\{tatable_b \mid b \in \{1, \dots, n\}\}$ 
  - Move block  $b$  on the table ( $\rightsquigarrow$  creates new tower).
  - Has to be topmost block of a tower.

### Action costs $cost$ :

$cost(a) = 1$  for all actions  $a$

# Blocks World: Formal Definition

State space  $\langle S, A, cost, T, s_0, S_* \rangle$  blocks world with  $n$  Blocks

State space: blocks world

Transitions:

Example for action  $a = move_{2,4}$ :

Transition  $s \xrightarrow{a} s'$  exists if and only if

- $s = \{ \langle b_1, \dots, b_k, 2 \rangle, \langle c_1, \dots, c_m, 4 \rangle \} \cup X$  and
- in case  $k > 0$ :  $s' = \{ \langle b_1, \dots, b_k \rangle, \langle c_1, \dots, c_m, 4, 2 \rangle \} \cup X$
- in case  $k = 0$ :  $s' = \{ \langle c_1, \dots, c_m, 4, 2 \rangle \} \cup X$

# Blocks World: Properties

Blocks	States	Blocks	States
1	1	10	58941091
2	3	11	824073141
3	13	12	12470162233
4	73	13	202976401213
5	501	14	3535017524403
6	4051	15	65573803186921
7	37633	16	1290434218669921
8	394353	17	26846616451246353
9	4596553	18	588633468315403843

- For every given initial state and goal state with  $n$  blocks simple algorithms can find **solutions** in  $O(n)$  time. ([How?](#))
- Finding **optimal solutions** is **NP-complete** (for a compact problem representation).

## Example 2: Logistics

### Task: logistics

- Given: **Cities** with locations, **objects to be delivered**
- Goal: Transport objects to destination locations

### Actions: logistics

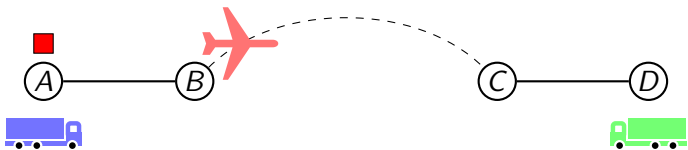
- Objects can be **loaded** and **unloaded** to trucks and airplanes.
- Trucks can **drive** between locations in a city.
- Airplanes can **fly** between airports.

### Complexity of Logistics

- Finding suboptimal solutions is polynomial.
- Finding **optimal solutions** is **NP-hard**.



# Logistics: Example



Goal: Transport red package from location  $A$  to location  $D$ .

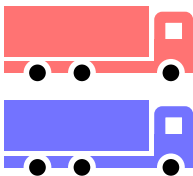
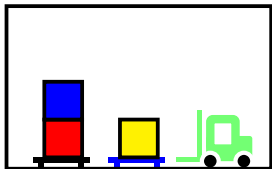
- 1 load package in blue truck, drive to  $B$ , unload package
- 2 load package in airplane, fly to  $C$ , unload package
- 3 drive green truck to  $C$ , load package, drive to  $D$ , unload package

## Example 3: Depot

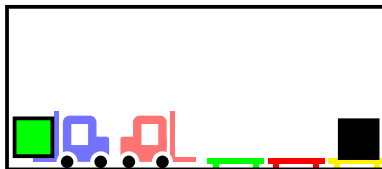
- Warehouse logistics
  - **transport crates** between depots and distributors
  - limited number of **pallets** in each place
- Within each warehouse
  - **like blocks world**
  - multiple forklifts possible
- Between warehouses
  - **similar to logistics**
  - crates only transported with trucks

# Depot: Example

Depot 1



Distributor 1



# Depot: Properties

## Task: Depot

Satisfy goal properties, given an initial configuration of places, crates, and vehicles.

Different goals possible:

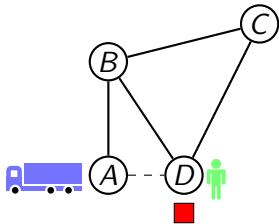
- enable **access** to a crate
- **transport** crates to Distributor
- **rearrange** crates
- combinations

## Complexity of depot

- Can include blocks world subtask.
- $\rightsquigarrow$  Finding **optimal solutions** is also **NP-hard**

## Example 4: Driverlog

- Another **package delivery** problem
- Path planning for **drivers** and **trucks**
- Given
  - map of **streets** (—) and **footpaths** (- - -)
  - **initial locations** of packages, trucks and drivers



# Driverlog

## Task: Driverlog

- Deliver packages to goal locations.
- Trucks and drivers can also have goal locations.

## Actions: Driverlog

- Drivers can **walk** on footpaths.
- Drivers can **board** and **leave** trucks.
- Trucks with a driver can **drive** on streets.
- Packages can be **loaded** and **unloaded** into trucks.

## Complexity of Driverlog

- Finding suboptimal solutions is polynomial.
- Finding **optimal solutions** is **NP-hard**.

## Example 5: Scanalyzer

- Business application (LemnaTec)
- Logistics for **smart greenhouses**
  - automated greenhouses with **integrated imaging facilities**
  - plants on **conveyor belts**

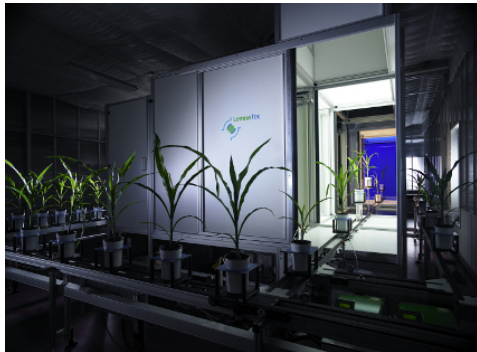


Image credit: LemnaTec

# Scanalyzer

## Difficulty

- Confined space
- Conveyor belts packed to capacity
- Conveyor belts only move in one direction
- **Moving one plant moves others as well**

## Task: Scanalyzer

- Given a **layout** of conveyor belts
- Transport **all plants** through the **imaging chamber**
- Return every plant to its **original position**



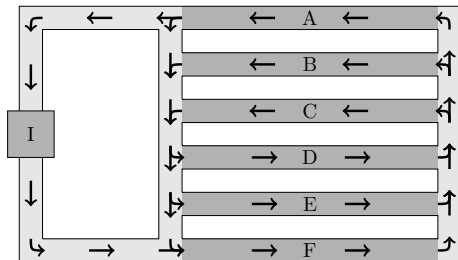
# Scanalyzer: Actions

## Actions: Scanalyzer

- Depend on the **layout**
- **Rotate** plant batches on two conveyor belts
- **Rotate** while **routing** through the **imaging chamber**

## Complexity of Scanalyzer

- Depends on the layout
- Polynomial for simple, **symmetric** layouts



# Example 6: Sokoban

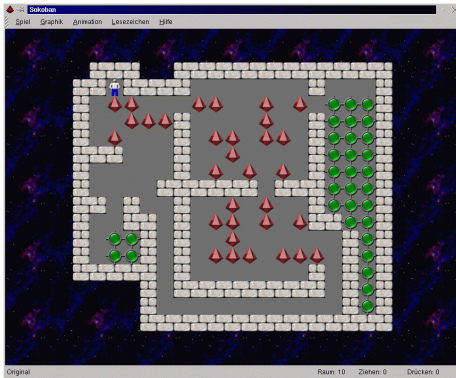


Image credit: KDE (KSokoban)

- Single player game
- Agent can push objects
- Goal: All objects are at destination locations

# Sokoban

## More Detailed Problem Description

- Given: Grid of locations, some locations contain objects
- Agent can **push** objects to free and adjacent locations
  - For example, to push an object to the right, the agent has to be located left to the object.
- Objects **cannot be pulled**

## Complexity of Sokoban

- **PSPACE-complete**
- Particularly: Many **dead-end states** (e. g., objects in corners)

## Example 7: Woodworking

- **Scheduling** problem
- Use different **tools** to create parts with the correct
  - **size** (here: one dimensional)
  - **color**
  - **material** (pine, oak, mahogany, ...)
  - **surface** (smooth, rough, ...)
  - **treatment** (varnished, glazed, untreated, ...)
- Different tools can be used **in parallel**
- **Minimize time** to finish all parts

# Woodworking

## Available Tools

- **Saws** and **high-speed saws**
  - cut boards to size
  - **dead ends** possible by wrong cut
  - high-speed saws cut faster but need set-up time
- **Grinders** and **planers**
  - remove existing color and treatment
  - grinder leaves smoother surface
  - planer removes more material
- **Glazers, immersion varnishers** and **spray varnishers**
  - change color and treatment
  - color has to be available for this machine

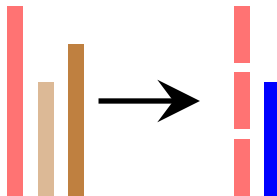


Image Credit: GoRapid



# Woodworking: Example

- Initial state (available boards/tools)
  - 10m oak (red, glazed, smooth)
  - 6m pine (natural, rough)
  - 8m pine (natural, smooth)
  - one of each tool
- Goal state (desired parts)
  - 3x 3m oak (red)
  - 6m pine (blue, smooth)
- Solution (optimality depends on action durations)
  - use high-speed saw for red part
  - grind and spray varnish 6m board while sawing red part
  - What if no grinder is available?
  - What if only one saw is available?



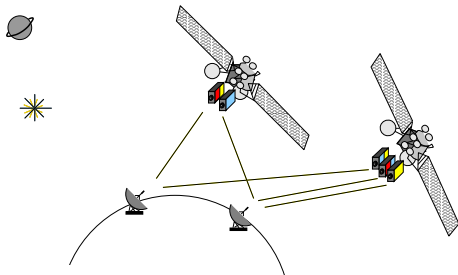
## Example 8: Satellite

- Space application
- Collect image data with a number of **satellites**
  - Can be **turned** to ground stations, stars or phenomena
  - Equipped with **instruments**, each supporting certain **modes**
  - Only power for **one instrument at a time**
  - After switching them on, instruments must be **calibrated** on a calibration target before taking images.
- **Goal:** Take images of stars or phenomena in certain modes and have some satellites pointing to specified directions.



Image credit: eutelsat

# Satellite: Example

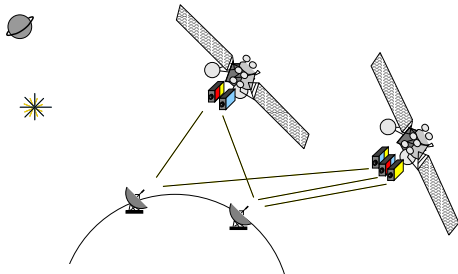




# Satellite: Example

## Goal images

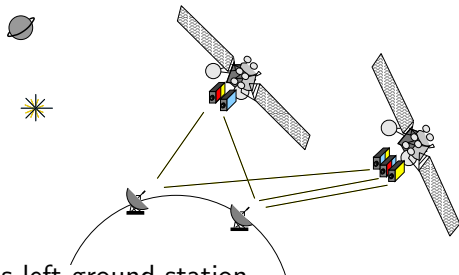
- star in red mode
- planet in yellow mode



# Satellite: Example

## Goal images

- star in red mode
- planet in yellow mode



- 1 Turn left satellite towards left ground station
- 2 Switch red-yellow instrument on
- 3 Calibrate red-yellow instrument on ground station
- 4 Turn left satellite towards star
- 5 Take image of star with calibrated instrument in red mode
- 6 Turn left satellite towards planet
- 7 Take image of planet in yellow mode

# Satellite: Properties



Image credit: DLR

## Complexity of Satellite

- We can find **some** plan in **polynomial time**.
- Finding an **optimal** plan is **NP-hard**.

# Example 9: Rovers

- Route planning and task distribution
- Multiple rovers with different capabilities
- Collect samples and take pictures of landmarks
- Transmit pictures and analysis results to lander

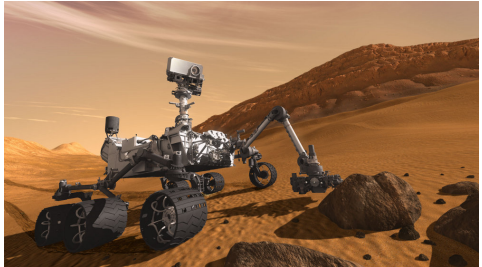


Image credit: NASA

# Rovers

## Rover capabilities

- Movement
  - different road map for each rover
- Rock/soil analysis tools
  - optional
  - limited storage capacity
- Cameras
  - optional
  - different modes (high res, color, ...)
  - have to be calibrated first
  - line of sight needed for calibration and taking pictures
- Transmission
  - only possible if lander is visible

# Rovers

## Task: Rovers

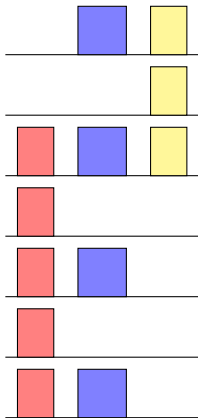
- Given a set of **rovers** with their **equipment** and **road maps**
- Collect all **designated samples and pictures**
- **Transmit results** back to lander

## Complexity of Rovers

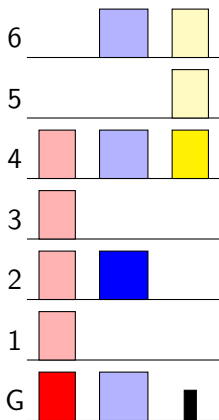
- Finding suboptimal solutions is polynomial.
- Finding **optimal solutions** is **NP-hard**.

# Example 10: Elevators

- transport passengers with lifts
- two types of lifts
  - different capacity
  - different cost models (modelling the energy consumption)
  - different reachability of floors
  - **slow**: capacity 2  
moving costs  $5 + \# \text{floors}$
  - **fast**: capacity 3  
moving costs  $1 + 3\# \text{floors}$
- (un-)boarding passengers is free



# Elevators: Example

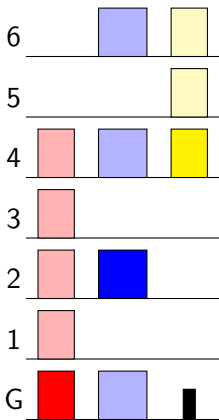


Goal:

Passenger on floor 6



# Elevators: Example

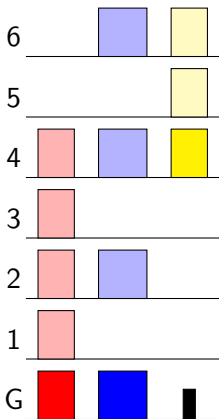


Possible plan:

Goal:

Passenger on floor 6

# Elevators: Example



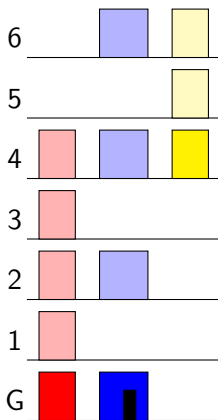
Goal:

Passenger on floor 6

Possible plan:

- blue lift **moves** to ground floor [7]

# Elevators: Example



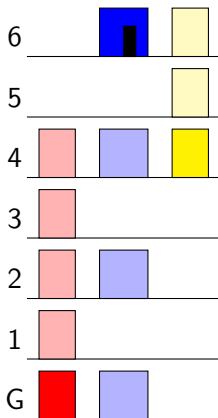
Goal:

Passenger on floor 6

Possible plan:

- blue lift **moves** to ground floor [7]
- passenger **boards** blue lift [0]

# Elevators: Example



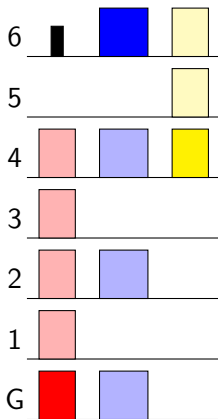
Goal:

Passenger on floor 6

Possible plan:

- blue lift **moves** to ground floor [7]
- passenger **boards** blue lift [0]
- blue lift **moves** to floor 6 [19]

# Elevators: Example



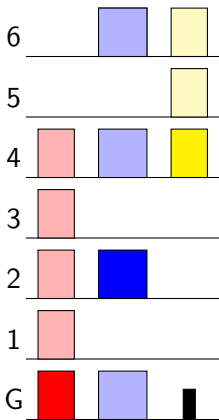
Goal:

Passenger on floor 6

Possible plan (cost 26):

- blue lift **moves** to ground floor [7]
- passenger **boards** blue lift [0]
- blue lift **moves** to floor 6 [19]
- passenger **leaves** blue lift [0]

# Elevators: Example

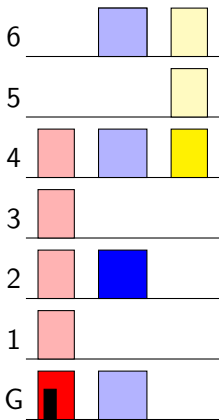


Alternative plan:

Goal:

Passenger on floor 6

# Elevators: Example



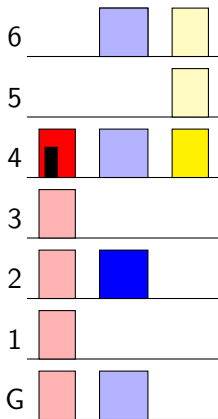
Goal:

Passenger on floor 6

Alternative plan:

- passenger boards red lift [0]

# Elevators: Example



Goal:

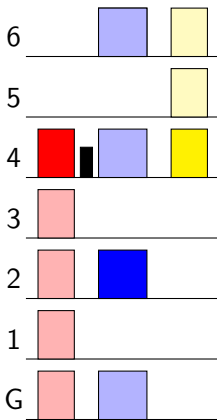
Passenger on floor 6

Alternative plan:

- passenger **boards** red lift [0]
- red lift **moves** to floor 4 [9]



# Elevators: Example



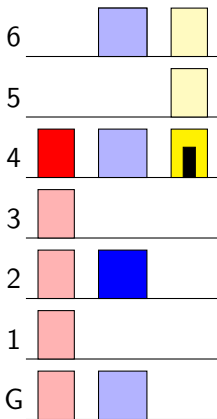
Goal:

Passenger on floor 6

Alternative plan:

- passenger **boards** red lift [0]
- red lift **moves** to floor 4 [9]
- passenger **leaves** red lift [0]

# Elevators: Example



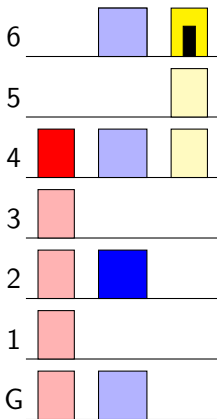
Goal:

Passenger on floor 6

Alternative plan:

- passenger **boards** red lift [0]
- red lift **moves** to floor 4 [9]
- passenger **leaves** red lift [0]
- passenger **boards** yellow lift [0]

# Elevators: Example



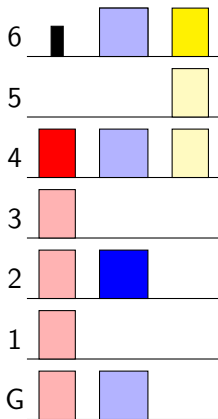
Goal:

Passenger on floor 6

Alternative plan:

- passenger **boards** red lift [0]
- red lift **moves** to floor 4 [9]
- passenger **leaves** red lift [0]
- passenger **boards** yellow lift [0]
- yellow lift **moves** to floor 6 [7]

# Elevators: Example



Goal:

Passenger on floor 6

Alternative plan (cost 16):

- passenger **boards** red lift [0]
- red lift **moves** to floor 4 [9]
- passenger **leaves** red lift [0]
- passenger **boards** yellow lift [0]
- yellow lift **moves** to floor 6 [7]
- passenger **leaves** yellow lift [0]