

# Get me out of here: Determining optimal policies

SEMINAR: CYBER-PHYSICAL SYSTEMS

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# Outline

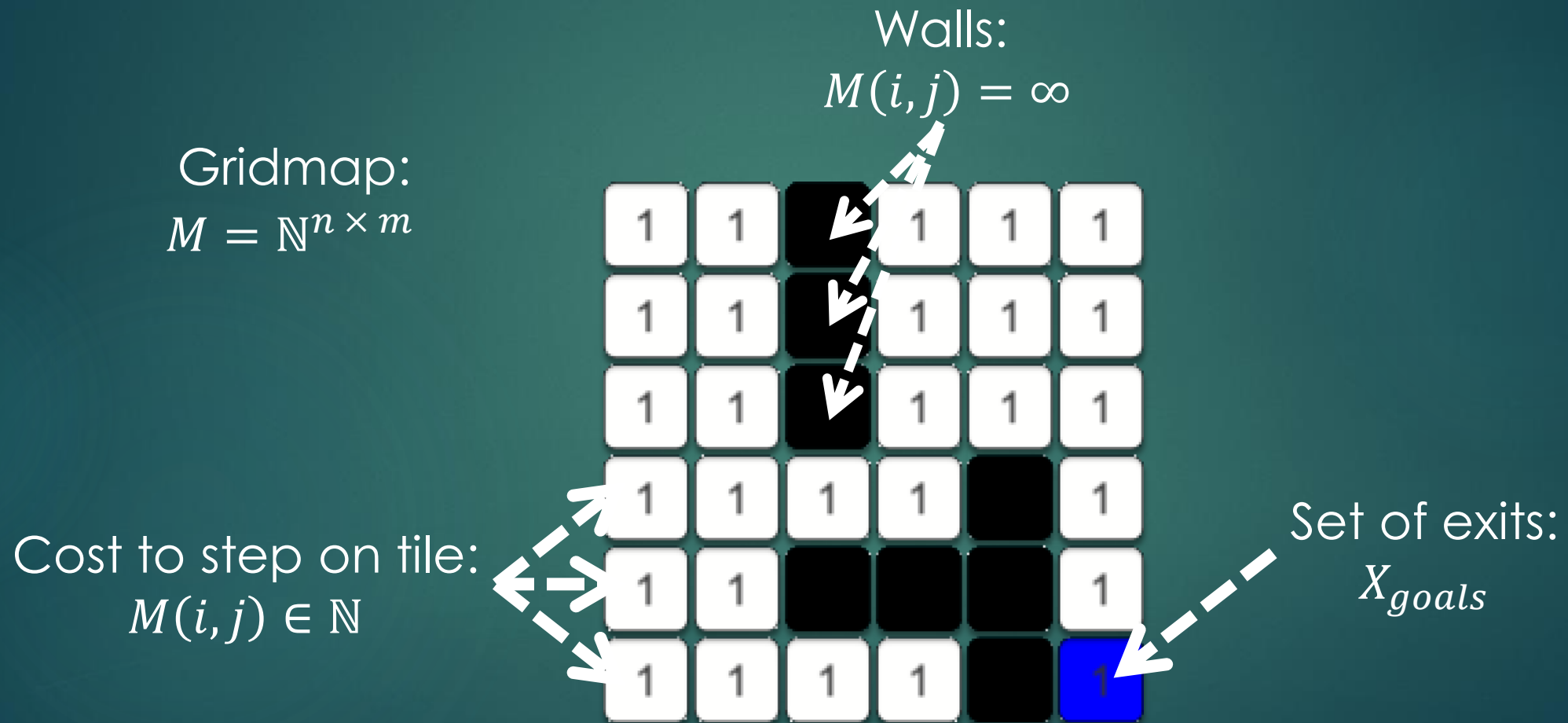
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- ▶ Background and Problem Statement
- ▶ Fundamentals of Dynamic Programming
- ▶ Solving the Dynamic Programming equations
- ▶ Related work and use cases in robot motion planning

# Background and Problem Statement

# Background

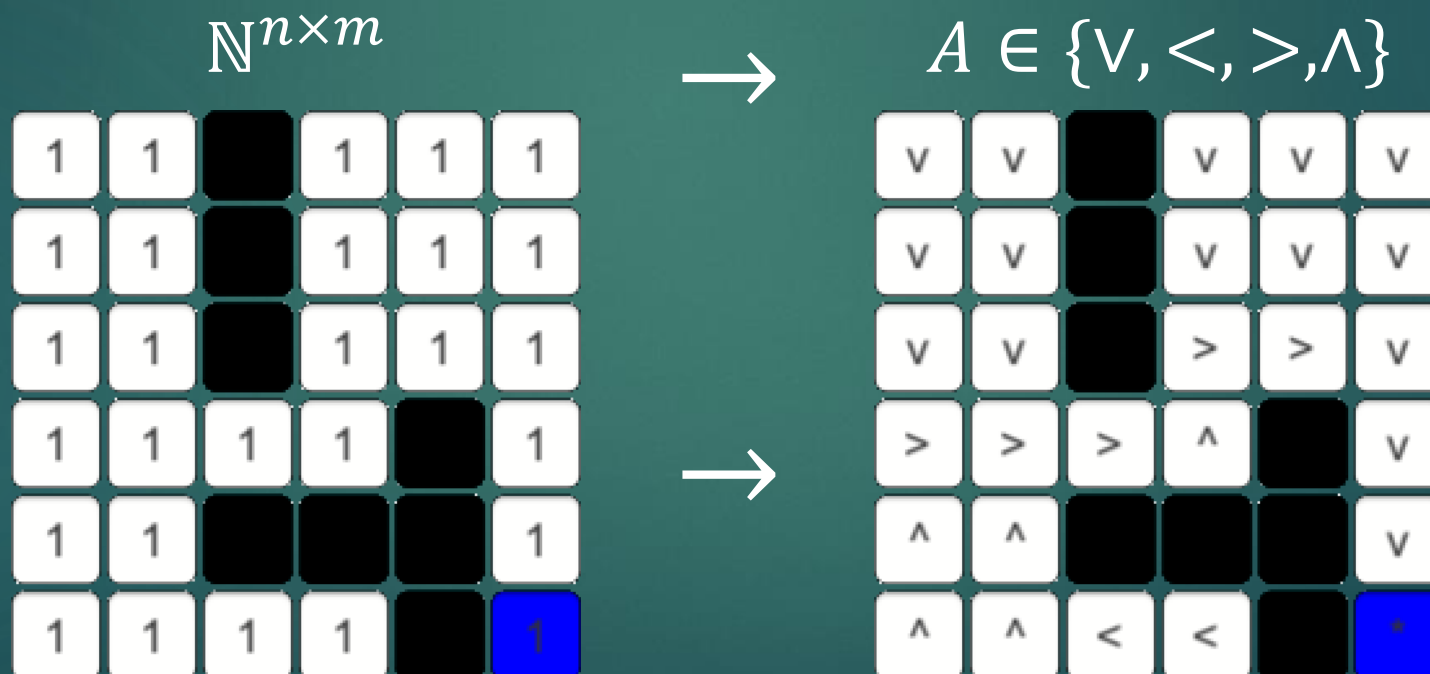
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# Problem Statement

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- Determine optimal policies



# Fundamentals of Dynamic Programming

# What is Dynamic Programming

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- ▶ Method used in mathematical optimization
- ▶ Transforms complex problems into sequences of simpler subproblems
- ▶ General framework for a broad variety of problems
- ▶ Base concept: Sequential decision making problem

# The principle of optimality

- ▶ A **policy** is a sequence of decisions
- ▶ An **optimal policy** is a sequence of decisions that has the best outcome w.r.t. a predefined criterion



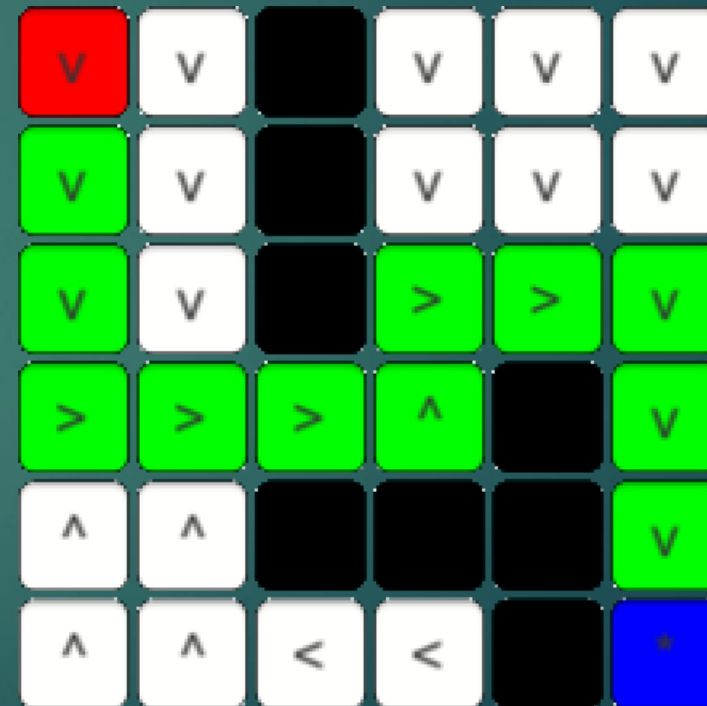
# The principle of optimality

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A policy



An optimal policy



# Characteristics of problems solvable by Dynamic Programming

FUNDAMENTALS OF DYNAMIC PROGRAMMING

# Overlapping Subproblems

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- ▶ Space of subproblems is small
- ▶ Solutions reused many times
- ▶ Recursive algorithm solves same subproblems over and over

# Overlapping Subproblems

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V	V		V	V	V
V	V		V	V	V
V	V		>	>	V
>	>	>	^		V
^	^				V
^	^	<	<		*

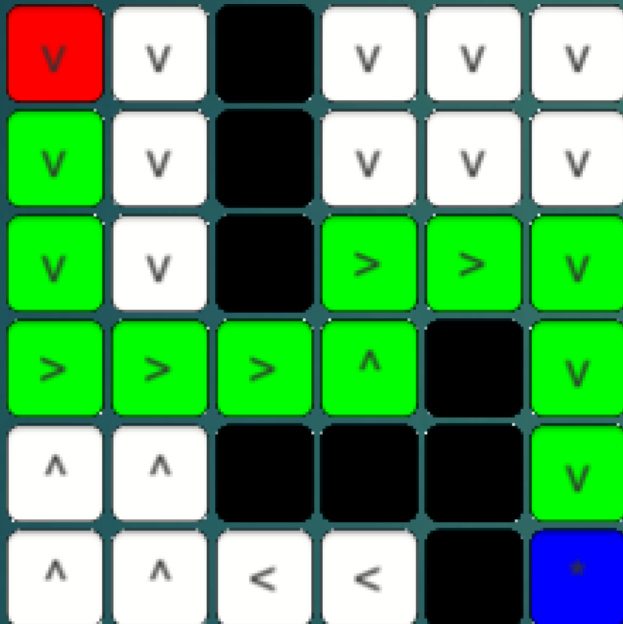
V	V		V	V	V
V	V		V	V	V
V	V		>	>	V
>	>	>	^		V
^	^				V
^	^	<	<		*

V	V		V	V	V
V	V		V	V	V
V	V		>	>	V
>	>	>	^		V
^	^				V
^	^	<	<		*

# Optimal Substructure

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- ▶ Solution for whole problem by combination of solutions for subproblems



$S = red \rightarrow blue$



$S = red \rightarrow x \cup x \rightarrow blue$



$S = red \rightarrow x \cup x \rightarrow y \cup y \rightarrow blue$

# How to apply Dynamic Programming in Computer Science

FUNDAMENTALS OF DYNAMIC PROGRAMMING

# Memoization

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- ▶ Break problem down into set of subproblems
- ▶ Store results in a table
- ▶ Use cached result if subproblem reoccures

# Bellman equation

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- ▶ Find a connection between one step and the next one
- ▶  $V(x) = \max_{a \in \Gamma(x)} \{Payoff(x, a) + V(Transformation(x, a))\}$
- ▶ Richard Bellman introduced backwards induction
  - ▶ Begin at step  $N \rightarrow$  Derive  $(N - 1) \rightarrow \dots \rightarrow$  Initial step



# Solving the Dynamic Programming equations

# Finding the value function

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- ▶ Goal is to find value function  $v(i, j) \in \mathbb{N}$
- ▶ Cost function  $c(i, j) = \begin{cases} \infty, & \text{if } (i, j) \notin M \\ M(i, j) & \text{else} \end{cases}$
- ▶ Trivial cases  $v(i, j) = \begin{cases} \infty, & \text{if } c(i, j) = \infty \\ 0, & \text{if } (i, j) \in X_{goal} \end{cases}$

# Finding the value function

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- ▶ Backwards iteration:  $(N, N - 1, \dots, 0)$
- ▶ Store results in memoization table  $V = \mathbb{N}^{n \times m}$

- ▶ 
$$\left. \begin{array}{ll} v(i-1, j) & (left) \\ v(i+1, j) & (right) \\ v(i, j-1) & (up) \\ v(i, j+1) & (down) \end{array} \right\} = v(i, j) + c(i, j)$$

# Finding optimal policies

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- ▶ Memoization table  $V$  holding the values of all tiles
- ▶ Find shortest paths  $\Leftrightarrow$  Sequence with lowest sum of values

$$\text{▶ } a(i, j) = \begin{cases} \textit{up}, & \text{if } v(i, j - 1) = \min_{(x, y)} \{v(x, y)\} \\ \textit{right}, & \text{if } v(i + 1, j) = \min_{(x, y)} \{v(x, y)\} \\ \textit{down}, & \text{if } v(i, j + 1) = \min_{(x, y)} \{v(x, y)\} \\ \textit{left}, & \text{if } v(i - 1, j) = \min_{(x, y)} \{v(x, y)\} \end{cases}$$

- ▶ With  $(x, y) \in \{(i + 1, j), (i - 1, j), (i, j + 1), (i, j - 1)\}$

# Related work and use cases in robot motion planning

# Extending the 2-dimensional algorithms onto N-dimensional problems

- ▶ Define a state as a node in a graph
- ▶ Define the transitions between states as edges in the graph
- ▶ Apply 2-dimensional algorithm onto graph to solve  
N-dimensional problem
- ▶ But curse of dimensions

# Autonomous vehicles

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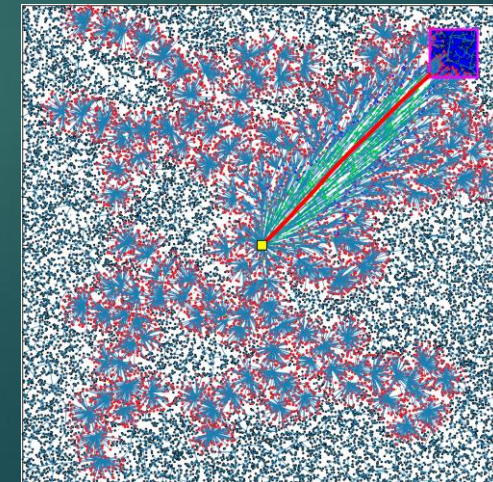
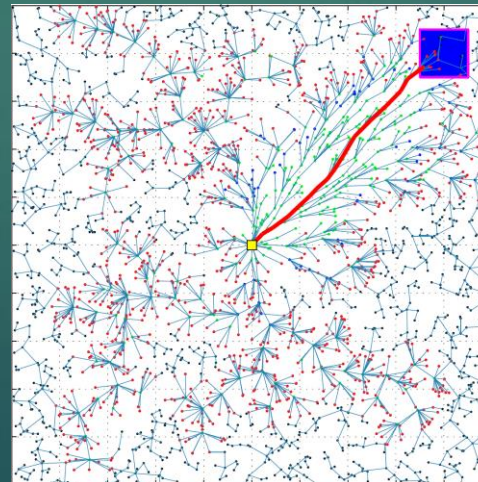
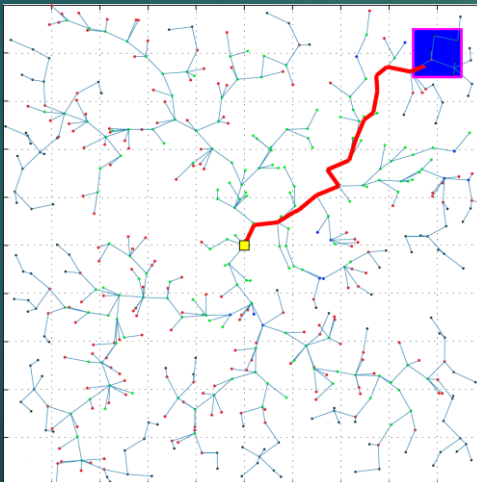
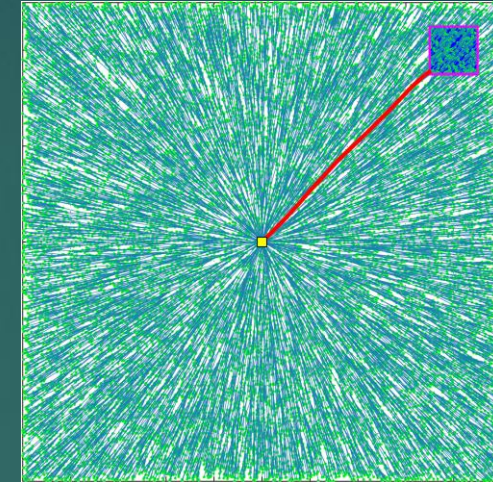
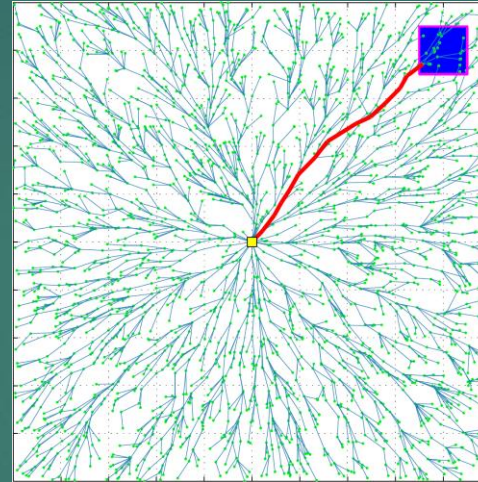
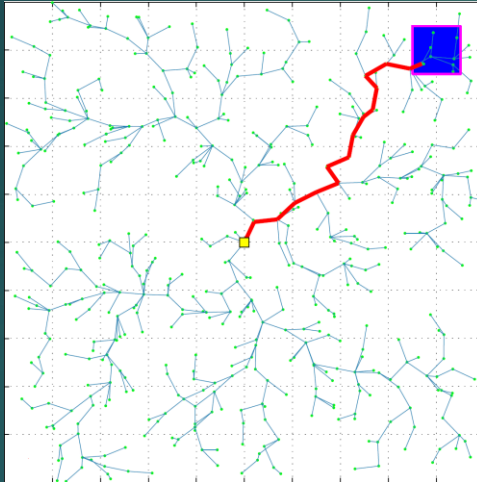
- ▶ Cars etc. are just one type of robots
- ▶ Many research in last years
- ▶ Best action for every position is calculated
- ▶ Also applicable for vacuum robots...



# Rapidly exploring random trees

## Dynamic Programming

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Some more examples!

# Conclusion

- Multistage decision problems can be solved by  
Dynamic Programming
- Overlapping subproblems and optimal substructure
- Memoization and the Bellman equation
- Especially for robot motion planning and path finding