

# Planning and Optimization

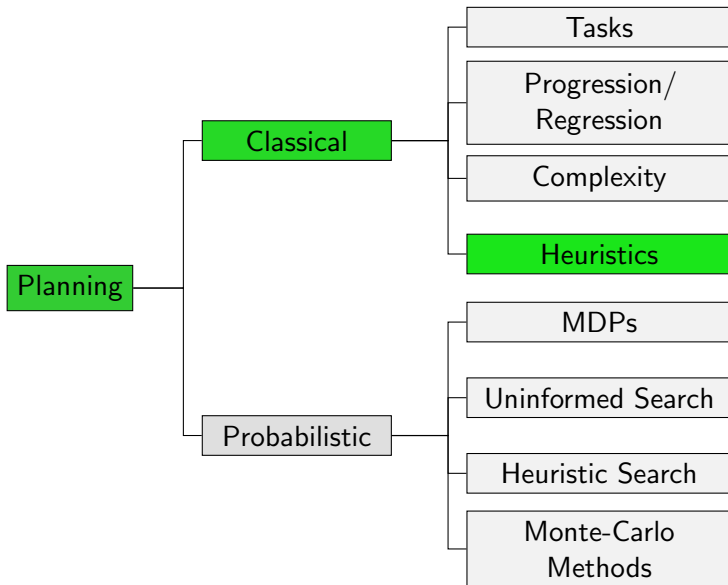
## C3. Delete Relaxation: Hardness of Optimal Planning & AND/OR Graphs

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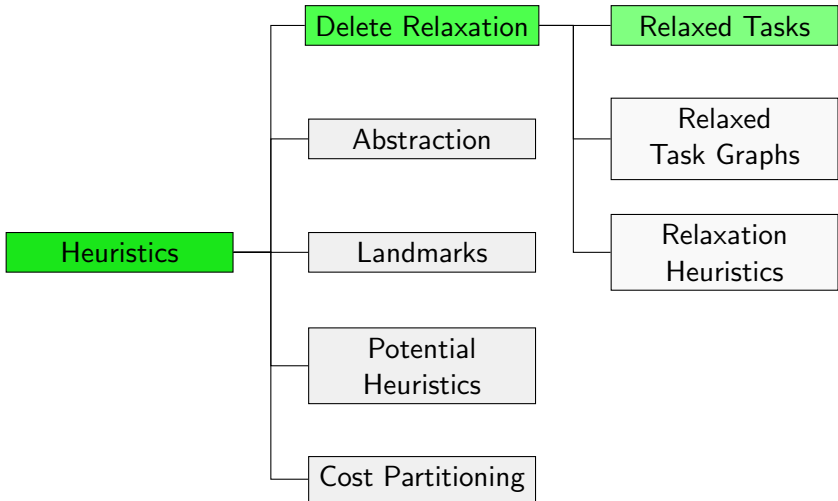
Universität Basel

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# Content of this Course



# Content of this Course: Heuristics



# The Story So Far

- A general way to come up with heuristics is to solve a **simplified** version of the real problem.
- **delete relaxation**: given a task in positive normal form, discard all delete effects
- A simple greedy algorithm **solves relaxed tasks efficiently** but usually generates plans of **poor quality**.

# The Story So Far

- A general way to come up with heuristics is to solve a **simplified** version of the real problem.
- **delete relaxation**: given a task in positive normal form, discard all delete effects
- A simple greedy algorithm **solves relaxed tasks efficiently** but usually generates plans of **poor quality**.

How hard is it to find **optimal** plans?

# Optimal Relaxed Plans

# The Set Cover Problem

To obtain an admissible heuristic, we must compute optimal relaxed plans. Can we do this efficiently?

This question is related to the following problem:

## Problem (Set Cover)

*Given:* a finite set  $U$ , a collection of subsets  $C = \{C_1, \dots, C_n\}$  with  $C_i \subseteq U$  for all  $i \in \{1, \dots, n\}$ , and a natural number  $K$ .

*Question:* Is there a set cover of size at most  $K$ , i.e., a subcollection  $S = \{S_1, \dots, S_m\} \subseteq C$  with  $S_1 \cup \dots \cup S_m = U$  and  $m \leq K$ ?

The following is a classical result from complexity theory:

## Theorem (Karp 1972)

*The set cover problem is NP-complete.*

# Complexity of Optimal Relaxed Planning (1)

## Theorem (Complexity of Optimal Relaxed Planning)

*The BCPLANEX problem restricted to delete-relaxed planning tasks is NP-complete.*

## Proof.

For **membership in NP**, guess a plan and verify.

It is sufficient to check plans of length at most  $|V|$  where  $V$  is the set of state variables, so this can be done in nondeterministic polynomial time.

For **hardness**, we reduce from the set cover problem. ...



## Complexity of Optimal Relaxed Planning (2)

### Proof (continued).

Given a set cover instance  $\langle U, C, K \rangle$ , we generate the following relaxed planning task  $\Pi^+ = \langle V, I, O^+, \gamma \rangle$ :

- $V = U$
- $I = \{v \mapsto \mathbf{F} \mid v \in V\}$
- $O^+ = \{\langle T, \bigwedge_{v \in C_i} v, 1 \rangle \mid C_i \in C\}$
- $\gamma = \bigwedge_{v \in U} v$

If  $S$  is a set cover, the corresponding operators form a plan. Conversely, each plan induces a set cover by taking the subsets corresponding to the operators. There exists a plan of cost at most  $K$  iff there exists a set cover of size  $K$ .

Moreover,  $\Pi^+$  can be generated from the set cover instance in polynomial time, so this is a polynomial reduction. □

# AND/OR Graphs

## Using Relaxations in Practice

How can we use relaxations for heuristic planning in practice?

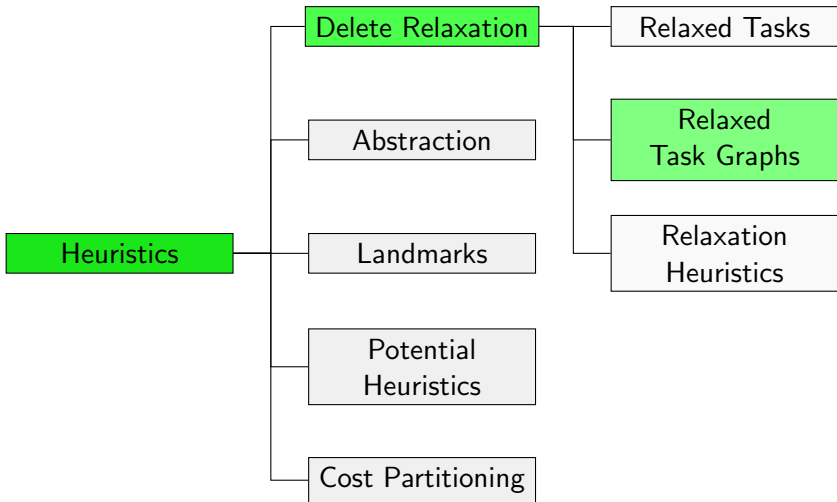
Different possibilities:

- Implement an **optimal planner** for relaxed planning tasks and use its solution costs as estimates, even though optimal relaxed planning is NP-hard.  
↪  **$h^+$  heuristic**
- Do not actually solve the relaxed planning task, but compute an approximation of its solution cost.  
↪  **$h^{\max}$  heuristic,  $h^{\text{add}}$  heuristic,  $h^{\text{LM-cut}}$  heuristic**
- Compute a solution for relaxed planning tasks which is not necessarily optimal, but “reasonable”.  
↪  **$h^{\text{FF}}$  heuristic**

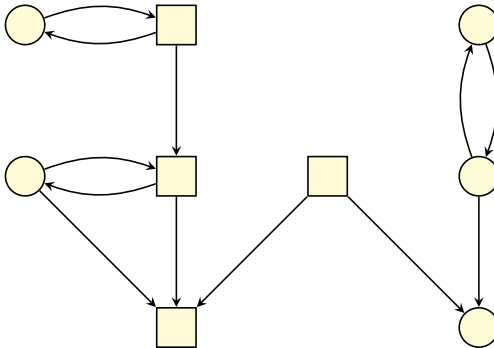
# AND/OR Graphs: Motivation

- Most relaxation heuristics we will consider can be understood in terms of computations on graphical structures called **AND/OR graphs**.
- We now introduce AND/OR graphs and study some of their major properties.
- In the next chapter, we will relate AND/OR graphs to relaxed planning tasks.

# Content of this Course: Heuristics



# AND/OR Graph Example



# AND/OR Graphs

## Definition (AND/OR Graph)

An **AND/OR graph**  $\langle N, A, type \rangle$  is a directed graph  $\langle N, A \rangle$  with a node label function  $type : N \rightarrow \{\wedge, \vee\}$  partitioning nodes into

- **AND nodes** ( $type(v) = \wedge$ ) and
- **OR nodes** ( $type(v) = \vee$ ).

We write  $succ(n)$  for the successors of node  $n \in N$ , i.e.,  
 $succ(n) = \{n' \in N \mid \langle n, n' \rangle \in A\}$ .

**Note:** We draw AND nodes as squares and OR nodes as circles.

# AND/OR Graph Valuations

## Definition (Consistent Valuations of AND/OR Graphs)

Let  $G$  be an AND/OR graph with nodes  $N$ .

A **valuation** or **truth assignment** of  $G$  is a valuation  $\alpha : N \rightarrow \{\mathbf{T}, \mathbf{F}\}$ , treating the nodes as propositional variables.

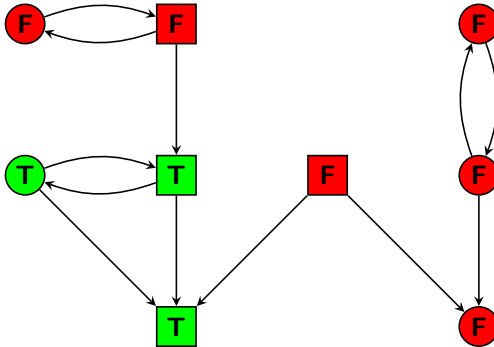
We say that  $\alpha$  is **consistent** if

- for all AND nodes  $n \in N$ :  $\alpha \models n$  iff  $\alpha \models \bigwedge_{n' \in \text{succ}(n)} n'$ .
- for all OR nodes  $n \in N$ :  $\alpha \models n$  iff  $\alpha \models \bigvee_{n' \in \text{succ}(n)} n'$ .

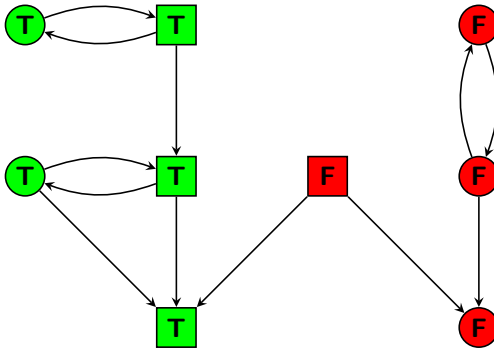
Note that  $\bigwedge_{n' \in \emptyset} n' = \top$  and  $\bigvee_{n' \in \emptyset} n' = \perp$ .



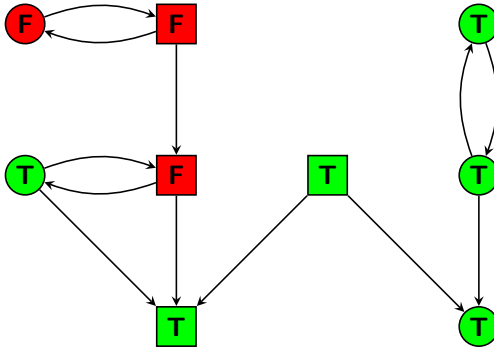
# Example: A Consistent Valuation



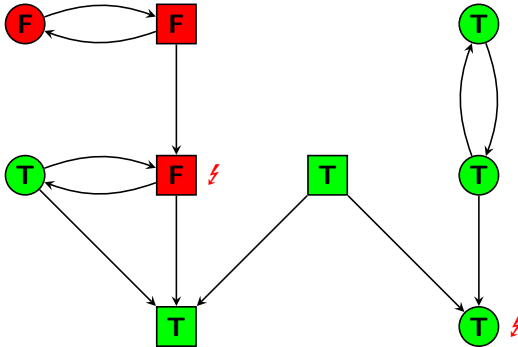
# Example: Another Consistent Valuation



# Example: An Inconsistent Valuation



# Example: An Inconsistent Valuation



## How Do We Find Consistent Valuations?

If we want to use valuations of AND/OR graphs algorithmically, a number of questions arise:

- Do consistent valuations **exist** for every AND/OR graph?
- Are they **unique**?
- If not, how are different consistent valuations **related**?
- Can consistent valuations be **computed efficiently**?

Our example shows that the answer to the second question is “no”. In the rest of this chapter, we address the remaining questions.

# Forced Nodes

## Forced Nodes

### Definition (Forced True/False Nodes)

Let  $G$  be an AND/OR graph.

A node  $n$  of  $G$  is called **forced true**  
if  $\alpha(n) = \mathbf{T}$  for all consistent valuations  $\alpha$  of  $G$ .

A node  $n$  of  $G$  is called **forced false**  
if  $\alpha(n) = \mathbf{F}$  for all consistent valuations  $\alpha$  of  $G$ .

How can we efficiently determine that nodes are forced true/false?

↪ We begin by looking at some simple rules.

## Rules for Forced True Nodes

### Proposition (Rules for Forced True Nodes)

Let  $n$  be a node in an AND/OR graph.

**Rule T-( $\wedge$ ):** If  $n$  is an AND node and *all* of its successors are forced true, then  $n$  is forced true.

**Rule T-( $\vee$ ):** If  $n$  is an OR node and *at least one* of its successors is forced true, then  $n$  is forced true.

### Remarks:

- These are “if, then” rules.  
Would they also be correct as “if and only if” rules?
- For the first rule, it is easy to see that the answer is “yes”.
- For the second rule, this is not so easy. (Why not?)



## Rules for Forced False Nodes

### Proposition (Rules for Forced False Nodes)

Let  $n$  be a node in an AND/OR graph.

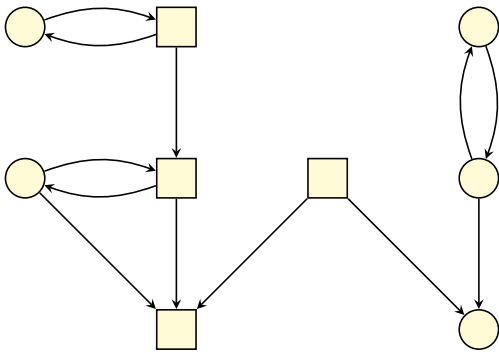
**Rule F-( $\wedge$ ):** If  $n$  is an AND node and *at least one* of its successors is forced false, then  $n$  is forced false.

**Rule F-( $\vee$ ):** If  $n$  is an OR node and *all* of its successors are forced false, then  $n$  is forced false.

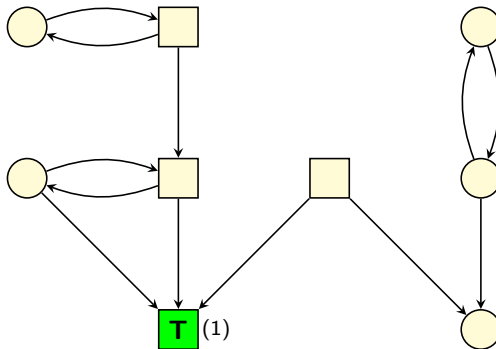
#### Remarks:

- Analogous comments as in the case of forced true nodes apply.
- This time, it is the first rule for which it is not obvious if a corresponding “if and only if” rule would be correct.

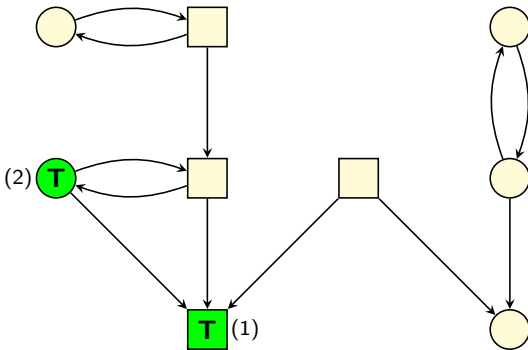
# Example: Applying the Rules for Forced Nodes



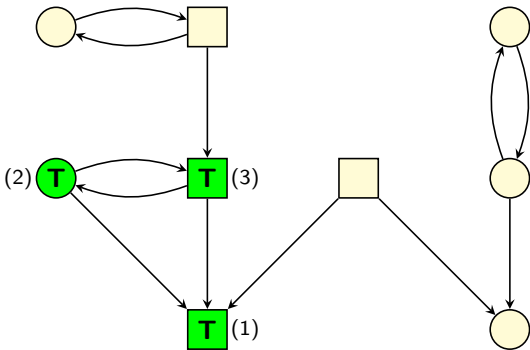
# Example: Applying the Rules for Forced Nodes



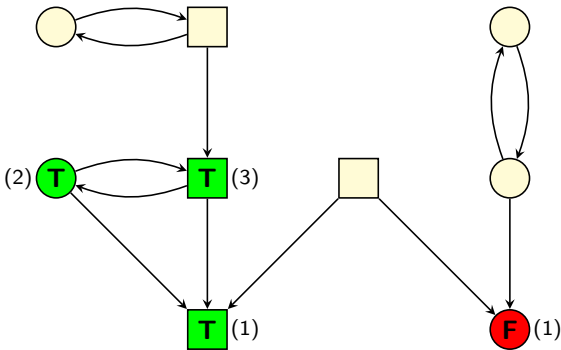
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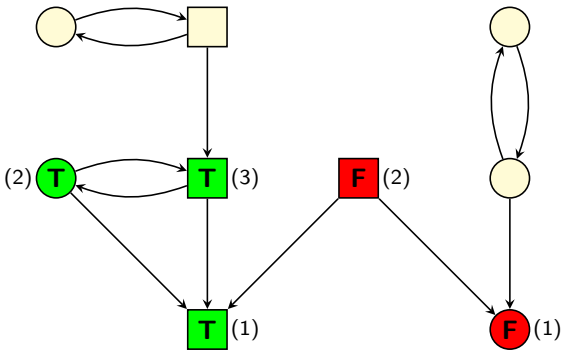
# Example: Applying the Rules for Forced Nodes



# Example: Applying the Rules for Forced Nodes



# Example: Applying the Rules for Forced Nodes



# Completeness of Rules for Forced Nodes

## Theorem

*If  $n$  is a node in an AND/OR graph that is forced true, then this can be derived by a sequence of applications of Rule **T**-( $\wedge$ ) and Rule **T**-( $\vee$ ).*

## Theorem

*If  $n$  is a node in an AND/OR graph that is forced false, then this can be derived by a sequence of applications of Rule **F**-( $\wedge$ ) and Rule **F**-( $\vee$ ).*

We prove the result for **forced true** nodes.

The result for forced false nodes can be proved analogously.



# Completeness of Rules for Forced Nodes: Proof (1)

## Proof.

- Let  $\alpha$  be a valuation where  $\alpha(n) = \mathbf{T}$  iff there exists a sequence  $\rho_n$  of applications of Rules  $\mathbf{T}-(\wedge)$  and Rule  $\mathbf{T}-(\vee)$  that derives that  $n$  is forced true.
- Because the rules are monotonic, there exists a sequence  $\rho$  of rule applications that derives that  $n$  is forced true for **all**  $n \in on(\alpha)$ . (Just concatenate all  $\rho_n$  to form  $\rho$ .)
- By the correctness of the rules, we know that all nodes reached by  $\rho$  are forced true. It remains to show that none of the nodes **not** reached by  $\rho$  is forced true.
- We prove this by showing that  **$\alpha$  is consistent**, and hence no nodes with  $\alpha(n) = \mathbf{F}$  can be forced true.

## Completeness of Rules for Forced Nodes: Proof (2)

### Proof (continued).

#### Case 1: nodes $n$ with $\alpha(n) = \mathbf{T}$

- In this case,  $\rho$  must have reached  $n$  in one of the derivation steps. Consider this derivation step.
- If  $n$  is an AND node,  $\rho$  must have reached all successors of  $n$  in previous steps, and hence  $\alpha(n') = \mathbf{T}$  for all successors  $n'$ .
- If  $n$  is an OR node,  $\rho$  must have reached at least one successor of  $n$  in a previous step, and hence  $\alpha(n') = \mathbf{T}$  for at least one successor  $n'$ .
- In both cases,  $\alpha$  is consistent for node  $n$ .

## Completeness of Rules for Forced Nodes: Proof (3)

### Proof (continued).

#### Case 2: nodes $n$ with $\alpha(n) = \mathbf{F}$

- In this case, by definition of  $\alpha$  no sequence of derivation steps reaches  $n$ . In particular,  $\rho$  does not reach  $n$ .
- If  $n$  is an AND node, there must exist some  $n' \in \text{succ}(n)$  which  $\rho$  does not reach. Otherwise,  $\rho$  could be extended using Rule **T**-( $\wedge$ ) to reach  $n$ . Hence,  $\alpha(n') = \mathbf{F}$  for some  $n' \in \text{succ}(n)$ .
- If  $n$  is an OR node, there cannot exist any  $n' \in \text{succ}(n)$  which  $\rho$  reaches. Otherwise,  $\rho$  could be extended using Rule **T**-( $\vee$ ) to reach  $n$ . Hence,  $\alpha(n') = \mathbf{F}$  for all  $n' \in \text{succ}(n)$ .
- In both cases,  $\alpha$  is consistent for node  $n$ .



## Remarks on Forced Nodes

### Notes:

- The theorem shows that we can compute all forced nodes by applying the rules repeatedly until a fixed point is reached.
- In particular, this also shows that the order of rule application does not matter: we always end up with the same result.
- In an efficient implementation, the sets of forced nodes can be computed in linear time in the size of the AND/OR graph.
- The proof of the theorem also shows that every AND/OR graph has a consistent valuation, as we explicitly construct one in the proof.

# Most/Least Conservative Valuations

# Most and Least Conservative Valuation

## Definition (Most and Least Conservative Valuation)

Let  $G$  be an AND/OR graph with nodes  $N$ .

The **most conservative valuation**  $\alpha_{\text{mcv}}^G : N \rightarrow \{\mathbf{T}, \mathbf{F}\}$  and the **least conservative valuation**  $\alpha_{\text{lcv}}^G : N \rightarrow \{\mathbf{T}, \mathbf{F}\}$  of  $G$  are defined as:

$$\alpha_{\text{mcv}}^G(n) = \begin{cases} \mathbf{T} & \text{if } n \text{ is forced true} \\ \mathbf{F} & \text{otherwise} \end{cases}$$
$$\alpha_{\text{lcv}}^G(n) = \begin{cases} \mathbf{F} & \text{if } n \text{ is forced false} \\ \mathbf{T} & \text{otherwise} \end{cases}$$

**Note:**  $\alpha_{\text{mcv}}^G$  is the valuation constructed in the previous proof.

# Properties of Most/Least Conservative Valuations

## Theorem (Properties of Most/Least Conservative Valuations)

Let  $G$  be an AND/OR graph. Then:

- 1  $\alpha_{\text{mcv}}^G$  is consistent.
- 2  $\alpha_{\text{lcv}}^G$  is consistent.
- 3 For all consistent valuations  $\alpha$  of  $G$ ,  
 $on(\alpha_{\text{mcv}}^G) \subseteq on(\alpha) \subseteq on(\alpha_{\text{lcv}}^G)$ .

# Properties of MCV/LCV: Proof

## Proof.

Part 1. was shown in the preceding proof. We showed that the valuation  $\alpha$  considered in this proof is consistent and satisfies  $\alpha(n) = \mathbf{T}$  iff  $n$  is forced true, which implies  $\alpha = \alpha_{\text{mcv}}^G$ .

The proof of Part 2. is analogous, using the rules for forced false nodes instead of forced true nodes.

Part 3 follows directly from the definitions of forced nodes,  $\alpha_{\text{mcv}}^G$  and  $\alpha_{\text{lcv}}^G$ . □



## Properties of MCV/LCV: Consequences

This theorem answers our remaining questions about the existence, uniqueness, structure and computation of consistent valuations:

- Consistent valuations always exist and can be efficiently computed.
- All consistent valuations lie between the most and least conservative one.
- There is a unique consistent valuation iff  $\alpha_{\text{mcv}}^G = \alpha_{\text{lcv}}^G$ , or equivalently iff each node is forced true or forced false.

# Summary

# Summary I

- For an informative heuristic, we would ideally want to find **optimal relaxed plans**.
- The solution cost of an optimal relaxed plan is the estimate of the  $h^+$  heuristic.
- However, the bounded-cost plan existence problem for relaxed planning tasks is **NP-complete**.
- Other relaxation heuristics can be understood in terms of computations on **AND/OR graphs**.

## Summary II

- **AND/OR graphs** are directed graphs with **AND nodes** and **OR nodes**.
- We can assign **truth values** to AND/OR graph nodes.
- Such valuations are called **consistent** if they match the intuitive meaning of “AND” and “OR”.
- Consistent valuations always exist.
- Consistent valuations can be computed efficiently.
- All consistent valuations fall between two extremes:
  - the **most conservative valuation**, where only nodes that are **forced to be true** are true
  - the **least conservative valuation**, where all nodes that are **not forced to be false** are true