

Planning and Optimization

E1. Constraints: Introduction

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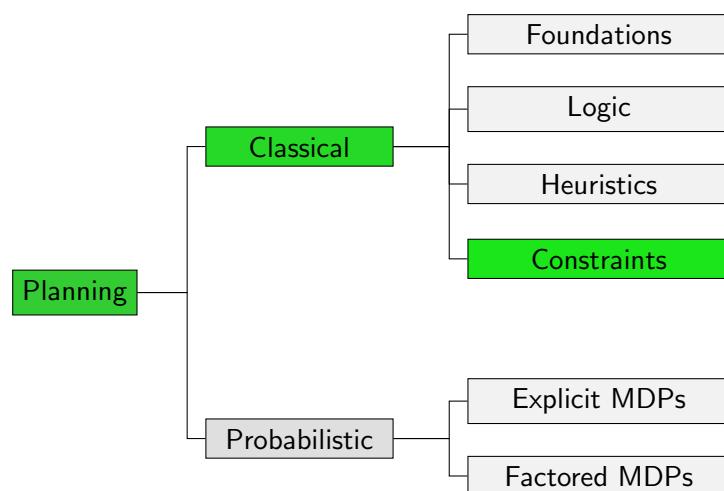
November 16, 2020 — E1. Constraints: Introduction

E1.1 Constraint-based Heuristics

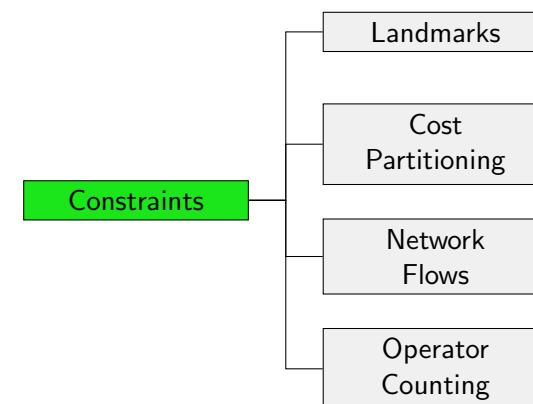
E1.2 Multiple Heuristics

E1.3 Summary

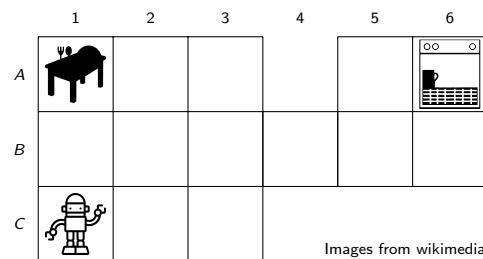
Content of this Course



Content of this Course: Constraints



E1.1 Constraint-based Heuristics



Constraints: Example

Coming Up with Heuristics in a Principled Way

General Procedure for Obtaining a Heuristic

Solve a simplified version of the problem.

Major ideas for heuristics in the planning literature:

- ▶ delete relaxation
- ▶ abstraction
- ▶ **landmarks**
- ▶ critical paths
- ▶ **network flows**
- ▶ **potential heuristic**

Landmarks, network flows and potential heuristics are based on **constraints** that can be specified for a planning task.

Constraints: Example

Example

Consider a FDR planning task $\langle V, I, O, \gamma \rangle$ with

- ▶ $V = \{robot-at, dishes-at\}$ with
 - ▶ $\text{dom}(robot-at) = \{A1, \dots, C3, B4, A5, \dots, B6\}$
 - ▶ $\text{dom}(dishes-at) = \{\text{Table}, \text{Robot}, \text{Dishwasher}\}$
- ▶ $I = \{robot-at \mapsto C1, dishes-at \mapsto \text{Table}\}$
- ▶ operators
 - ▶ move- $x-y$ to move from cell x to adjacent cell y
 - ▶ pickup dishes, and
 - ▶ load dishes into the dishwasher.
- ▶ $\gamma = (robot-at = B6) \wedge (dishes-at = \text{Dishwasher})$

Constraints

Some heuristics exploit **constraints** that describe something that holds in every solution of the task.

For instance, every solution is such that

- ▶ a variable takes some value in at least one visited state.
(a **fact landmark** constraint)

Fact Landmarks: Example

Which values do *robot-at* and *dishes-at* take in every solution?

	1	2	3	4	5	6
A						
B						
C						



- ▶ *robot-at* = C1, *dishes-at* = Table (initial state)
- ▶ *robot-at* = B6, *dishes-at* = Dishwasher (goal state)
- ▶ *robot-at* = A1, *robot-at* = B3, *robot-at* = B4, *robot-at* = B5, *robot-at* = A6, *dishes-at* = Robot

Constraints

Some heuristics exploit **constraints** that describe something that holds in every solution of the task.

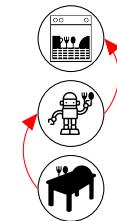
For instance, every solution is such that

- ▶ a variable takes some value in at least one visited state.
(a **fact landmark** constraint)
- ▶ an action must be applied.
(an action landmark constraint)

Action Landmarks: Example

Which actions must be applied in every solution?

	1	2	3	4	5	6
A						
B						
C						



- ▶ pickup
- ▶ load
- ▶ move-B3-B4
- ▶ move-B4-B5

Constraints

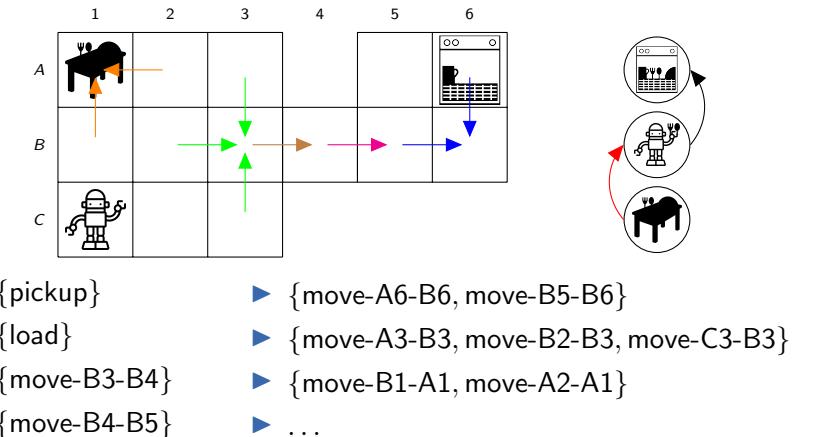
Some heuristics exploit **constraints** that describe something that holds in every solution of the task.

For instance, every solution is such that

- ▶ a variable takes some **value** in at least one visited state. (a **fact landmark** constraint)
- ▶ an action must be applied. (an **action landmark** constraint)
- ▶ at least one action from a set of actions must be applied. (a **disjunctive action landmark** constraint)

Disjunctive Action Landmarks: Example

Which set of actions is such that at least one must be applied?



Constraints

Some heuristics exploit **constraints** that describe something that holds in every solution of the task.

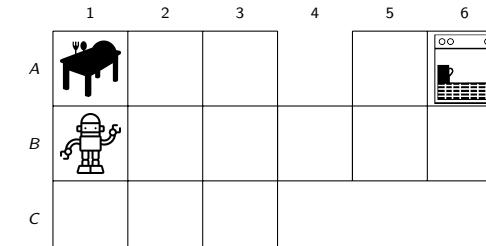
For instance, every solution is such that

- ▶ a variable takes some value in at least one visited state. (a **fact landmark** constraint)
- ▶ at least one action from a set of actions must be applied. (a **disjunctive action landmark** constraint)
- ▶ fact consumption and production is "balanced". (a **network flow** constraint)

Network Flow: Example

Consider the fact $\text{robot-at} = B1$.

How often are actions used that enter this cell?



Answer: as often as actions that leave this cell

If Count_o denotes how often operator o is applied, we have:

$$\begin{aligned} \text{Count}_{\text{move-A1-B1}} + \text{Count}_{\text{move-B2-B1}} + \text{Count}_{\text{move-C1-B1}} = \\ \text{Count}_{\text{move-B1-A1}} + \text{Count}_{\text{move-B1-B2}} + \text{Count}_{\text{move-B1-C1}} \end{aligned}$$

E1.2 Multiple Heuristics

Combining Admissible Heuristics Admissibly

Major ideas to combine heuristics admissibly:

- ▶ maximize
- ▶ canonical heuristic (for abstractions)
- ▶ minimum hitting set (for landmarks)
- ▶ cost partitioning
- ▶ operator counting

Often computed as solution to a (integer) linear program.

Combining Heuristics Admissibly: Example

Example

Consider an FDR planning task $\langle V, I, \{o_1, o_2, o_3, o_4\}, \gamma \rangle$ with $V = \{v_1, v_2, v_3\}$ with $\text{dom}(v_1) = \{A, B\}$ and $\text{dom}(v_2) = \text{dom}(v_3) = \{A, B, C\}$, $I = \{v_1 \mapsto A, v_2 \mapsto A, v_3 \mapsto A\}$,

$$o_1 = \langle v_1 = A, v_1 := B, 1 \rangle$$

$$o_2 = \langle v_2 = A \wedge v_3 = A, v_2 := B \wedge v_3 := B, 1 \rangle$$

$$o_3 = \langle v_2 = B, v_2 := C, 1 \rangle$$

$$o_4 = \langle v_3 = B, v_3 := C, 1 \rangle$$

and $\gamma = (v_1 = B) \wedge (v_2 = C) \wedge (v_3 = C)$.

Let \mathcal{C} be the pattern collection that contains all atomic projections. What is the canonical heuristic function $h^{\mathcal{C}}$?

Answer: Let $h_i := h^{v_i}$. Then $h^{\mathcal{C}} = \max \{h_1 + h_2, h_1 + h_3\}$.

Reminder: Orthogonality and Additivity

Why can we add h_1 and h_2 (h_1 and h_3) admissibly?

Theorem (Additivity for Orthogonal Abstractions)

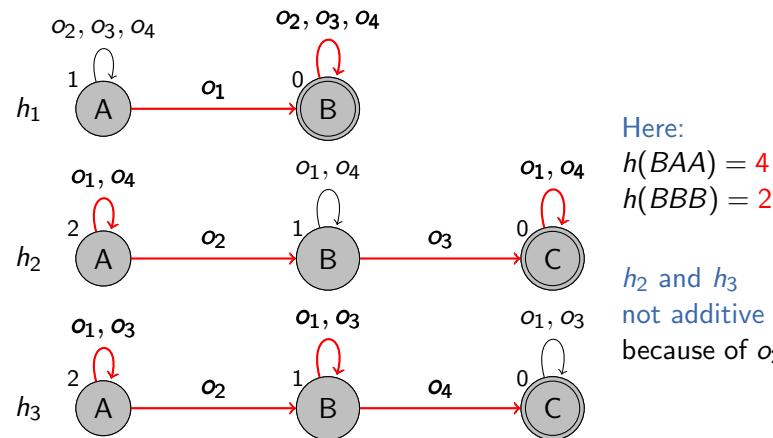
Let $h^{\alpha_1}, \dots, h^{\alpha_n}$ be abstraction heuristics of the same transition system such that α_i and α_j are orthogonal for all $i \neq j$.

Then $\sum_{i=1}^n h^{\alpha_i}$ is a safe, goal-aware, admissible and consistent heuristic for Π .

Consistency proof exploits that every concrete transition induces state-changing transition in at most one abstraction.

Combining Heuristics Admissibly: Example

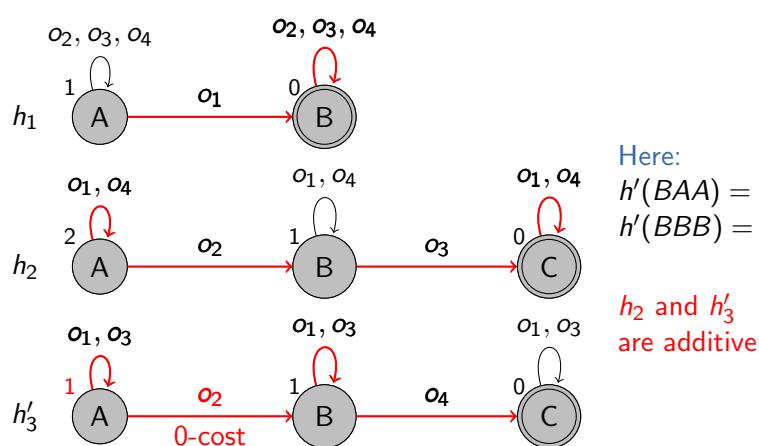
Let $h = h_1 + h_2 + h_3$. Where is consistency violated?



Consider solution $\langle o_1, o_2, o_3, o_4 \rangle$

Combining Heuristics Admissibly: Example

Let $h' = h_1 + h_2 + h'_3$, where $h'_3 = h^{v_3}$ assuming $cost_3(o_2) = 0$.



Consider solution $\langle o_1, o_2, o_3, o_4 \rangle$

Inconsistency of h_2 and h_3

The reason that h_2 and h_3 are not additive is because the cost of o_2 is considered in both.

Is there anything we can do about this?

Solution: We can ignore the cost of o_2 in one heuristic by setting its cost to 0 (e.g., $cost_3(o_2) = 0$).

Cost partitioning

Using the cost of every operator only in one heuristic is called a **zero-one cost partitioning**.

More generally, heuristics are additive if all operator costs are distributed in a way that the sum of the individual costs is no larger than the cost of the operator.

This can also be expressed as a constraint, the **cost partitioning constraint**:

$$\sum_{i=1}^n cost_i(o) \leq cost(o) \text{ for all } o \in O$$

(more details later)

E1.3 Summary

Summary

- ▶ Landmarks and network flows are **constraints** that describe something that holds in every solution of the task.
- ▶ Heuristics can be summed up admissibly if the **cost partitioning constraint** is satisfied.