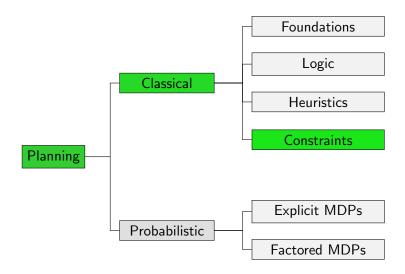
Planning and Optimization E1. Constraints: Introduction

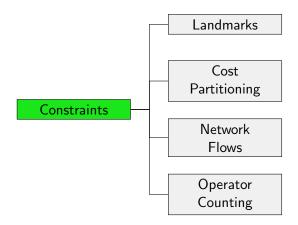
Malte Helmert and Gabriele Röger

Universität Basel

Content of this Course



Content of this Course: Constraints



Constraint-based Heuristics

Coming Up with Heuristics in a Principled Way

General Procedure for Obtaining a Heuristic

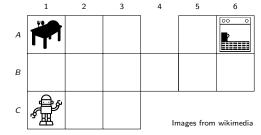
Solve a simplified version of the problem.

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



Constraints: Example

Example

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

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

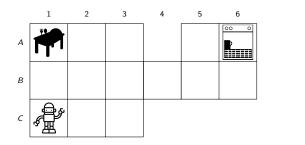
Constraints

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)

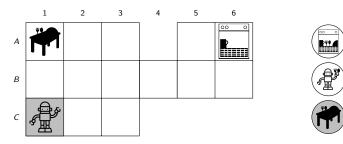
Which values do robot-at and dishes-at take in every solution?





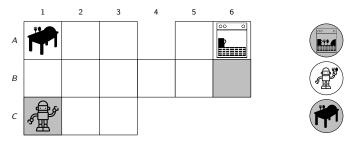


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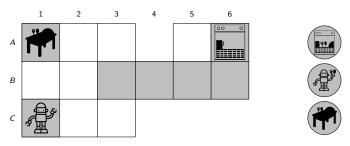
■ robot-at = C1, dishes-at = Table (initial state)

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- robot-at = C1, dishes-at = Table (initial state)
- robot-at = B6, dishes-at = Dishwasher (goal state)

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



- 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

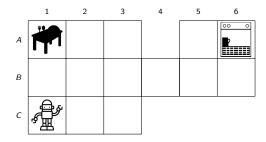
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For instance, every solution is such that

- 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?



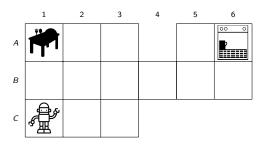


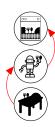




Action Landmarks: Example

Which actions must be applied in every solution?

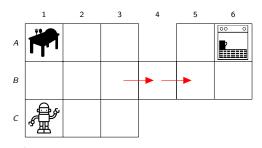


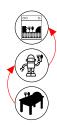


- pickup
- load

Action Landmarks: Example

Which actions must be applied in every solution?





- 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

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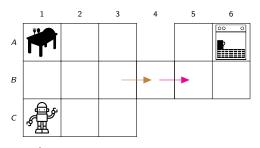
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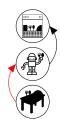
For instance, every solution is such that

- 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)

Disjunctive Action Landmarks: Example

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

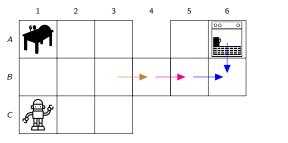


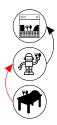


- {pickup}
- {load}
- {move-B3-B4}
- {move-B4-B5}

Disjunctive Action Landmarks: Example

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





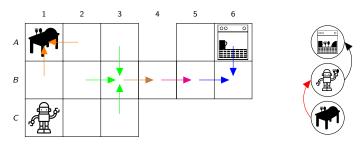
{pickup}

• {move-A6-B6, move-B5-B6}

- {load}
- {move-B3-B4}
- {move-B4-B5}

Disjunctive Action Landmarks: Example

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



- {pickup}
- {load}
- {move-B3-B4}
- {move-B4-B5}

- {move-A6-B6, move-B5-B6}
- {move-A3-B3, move-B2-B3, move-C3-B3}
- {move-B1-A1, move-A2-A1}
- **.** . . .

Constraints

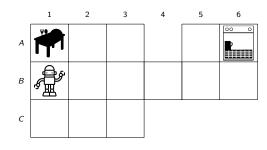
Some heuristics exploit constraints that describe something that holds in every solution of the task.

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)

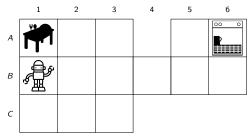
Network Flow: Example

Consider the fact robot-at = B1. How often are actions used that enter this cell?



Network Flow: Example

Consider the fact 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:

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

Multiple Heuristics

Combining Admissible Heuristics Admissibly

Major ideas to combine heuristics admissibly:

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

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

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 $dom(v_1) = \{A, B\}$ and $dom(v_2) = dom(v_3) = \{A, B, C\}$, $I = \{v_1 \mapsto A, v_2 \mapsto A, v_3 \mapsto A\}$,

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

 $o_2 = \langle v_2 = A \land v_3 = A, v_2 := B \land v_3 := B, 1 \rangle$
 $o_3 = \langle v_2 = B, v_2 := C, 1 \rangle$
 $o_4 = \langle v_3 = B, v_3 := C, 1 \rangle$

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

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

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

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

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

Reminder: Orthogonality and Additivity

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

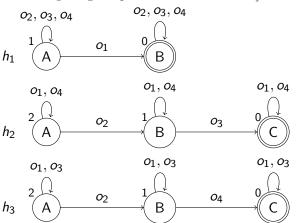
Theorem (Additivity for Orthogonal Abstractions)

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

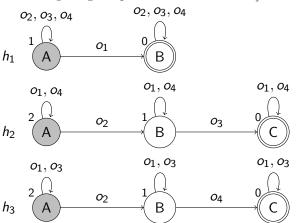
Then $\sum_{i=1}^{n} h^{\alpha_i}$ is a safe, goal-aware, admissible and consistent heuristic for Π .

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

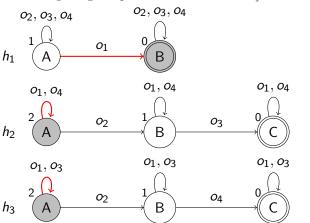
Let $h = h_1 + h_2 + h_3$. Where is consistency violated?

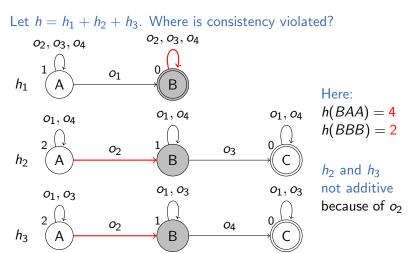


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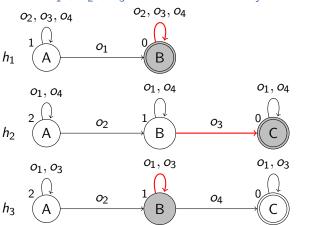


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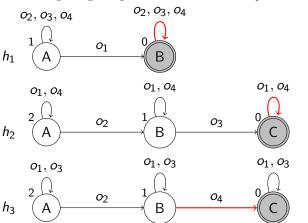




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Inconsistency of h_2 and h_3

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?

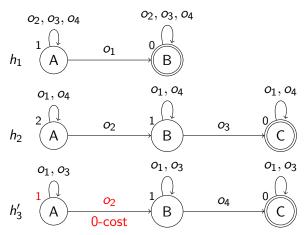
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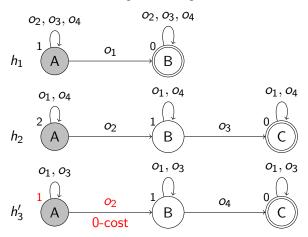
Is there anything we can do about this?

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

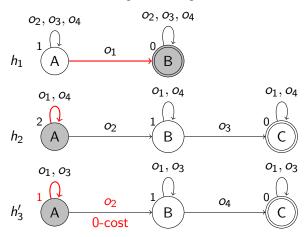
Let $h' = h_1 + h_2 + h'_3$, where $h'_3 = h^{v_3}$ assuming $cost_3(o_2) = 0$.



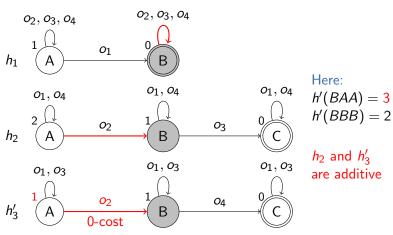
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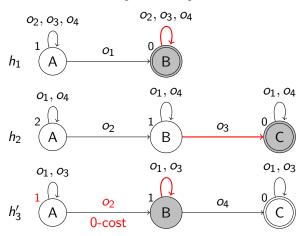
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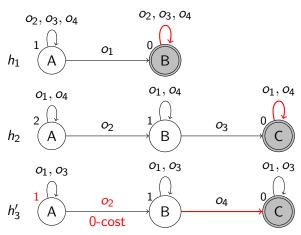
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Cost partitioning

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

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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)

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.