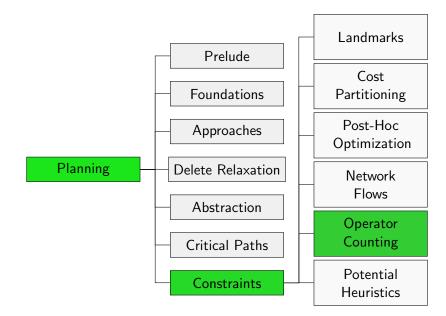
Planning and Optimization G11. Operator Counting

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Content of this Course



Introduction

Introduction

In the previous chapter, we used flow constraints to describe how often operators must be used in each plan.

Example (Flow Constraints)

Let Π be a planning problem with operators $\{o_{red}, o_{green}, o_{blue}\}$. The flow constraint for some atom a is the constraint

$$1 + Count_{Ogreen} = Count_{Ored}$$
 if

a is true in the initial state

ogreen produces a

a is false in the goal

Ored consumes a

In natural language, the flow constraint expresses that

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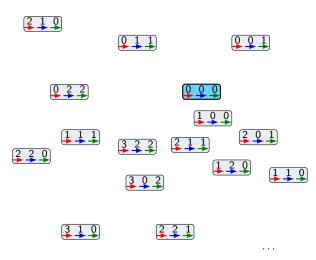
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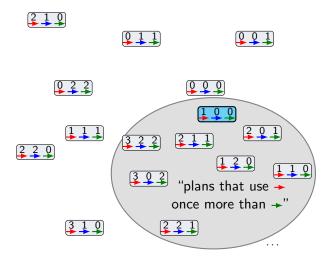
In natural language, the flow constraint expresses that every plan uses o_{red} once more than o_{green} .

Let us now observe how each flow constraint alters the operator count solution space.



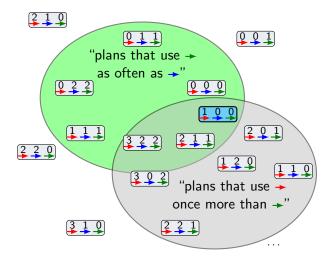
Introduction

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Operator-counting Framework

Operator Counting

Operator counting

- generalizes this idea to a framework that allows to admissibly combine different heuristics.
- uses linear constraints . . .
- that describe number of occurrences of an operator . . .
- ... and must be satisfied by every plan.
- provides declarative way to describe knowledge about solutions.
- allows reasoning about solutions to derive heuristic estimates.

Operator-counting Constraint

Definition (Operator-counting Constraints)

Let Π be a planning task with operators O and let s be a state. Let V be the set of integer variables $Count_O$ for each $o \in O$.

A linear inequality over $\mathcal V$ is called an operator-counting constraint for s if for every plan π for s setting each Count_o to the number of occurrences of o in π is a feasible variable assignment.

Operator-counting Heuristics

Definition (Operator-counting IP/LP Heuristic)

The operator-counting integer program IP_C for a set C of operator-counting constraints for state s is

Minimize
$$\sum_{o \in O} cost(o) \cdot Count_o$$
 subject to C and $Count_o \ge 0$ for all $o \in O$,

where O is the set of operators.

The IP heuristic h_C^{IP} is the objective value of IP_C, the LP heuristic h_C^{LP} is the objective value of its LP-relaxation. If the IP/LP is infeasible, the heuristic estimate is ∞ .

Operator-counting Constraints

- Adding more constraints can only remove feasible solutions.
- Fewer feasible solutions can only increase the objective value.
- Higher objective value means better informed heuristic
- ⇒ Have we already seen other operator-counting constraints?

Reminder: Minimum Hitting Set for Landmarks

Variables

Non-negative variable Applied for each operator o

Objective

Minimize $\sum_{o} cost(o) \cdot Applied_{o}$

Subject to

 $\sum_{o \in L} \mathsf{Applied}_o \ge 1 \text{ for all landmarks } L$

Operator Counting with Disjunctive Action Landmarks

Variables

Non-negative variable Count_o for each operator o

Objective

Minimize $\sum_{o} cost(o) \cdot Count_{o}$

Subject to

 $\sum_{o \in L} \mathsf{Count}_o \ge 1 \text{ for all landmarks } L$

Reminder: Post-hoc Optimization Heuristic

For set of abstractions $\{\alpha_1, \ldots, \alpha_n\}$:

Variables

Non-negative variables X_o for all operators $o \in O$ X_o is cost incurred by operator o

Objective

Minimize $\sum_{o \in O} X_o$

Subject to

$$\sum_{o \in O: o \text{ relev. for } \alpha} X_o \ge h^{\alpha}(s) \quad \text{for } \alpha \in \{\alpha_1, \dots, \alpha_n\}$$
$$X_o > 0 \qquad \text{for all } o \in O$$

For set of abstractions $\{\alpha_1, \ldots, \alpha_n\}$:

Variables

Non-negative variables $Count_o$ for all operators $o \in O$ $Count_o \cdot cost(o)$ is cost incurred by operator o

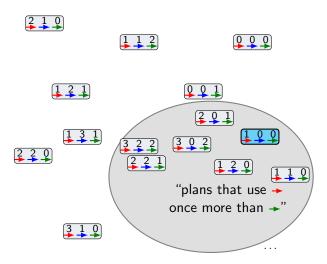
Objective

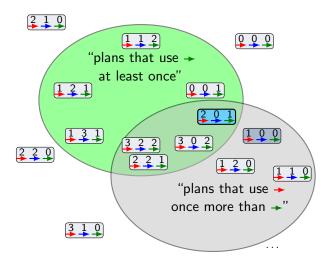
Minimize $\sum_{o \in O} cost(o) \cdot Count_o$

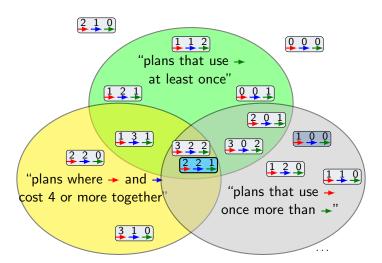
Subject to

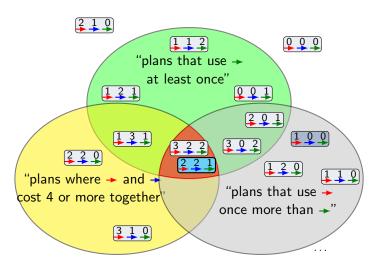
$$\sum\nolimits_{o \in \textit{O}:o \text{ relev. for } \alpha} \textit{cost}(o) \cdot \mathsf{Count}_o \geq \textit{h}^{\alpha}(s) \quad \text{for } \alpha \in \{\alpha_1, \dots, \alpha_n\}$$

$$\textit{cost}(o) \cdot \mathsf{Count}_o \geq 0 \qquad \quad \text{for all } o \in \textit{O}$$









Further Examples?

- The definition of operator-counting constraints can be extended to groups of constraints and auxiliary variables.
- With this extended definition we could also cover. more heuristics, e.g., the perfect relaxation heuristic h^+

Properties

Admissibility

Theorem (Operator-counting Heuristics are Admissible)

The IP and the LP heuristic are admissible.

Proof.

Let C be a set of operator-counting constraints for state s and π be an optimal plan for s. The number of operator occurrences of π are a feasible solution for C. As the IP/LP minimizes the total plan cost, the objective value cannot exceed the cost of π and is therefore an admissible estimate.

Properties

Dominance

$\mathsf{Theorem}$

Let C and C' be sets of operator-counting constraints for s and let $C \subseteq C'$. Then $IP_C \le IP_{C'}$ and $LP_C \le LP_{C'}$.

Proof.

Every feasible solution of C' is also feasible for C. As the LP/IP is a minimization problem, the objective value subject to C can therefore not be larger than the one subject to C'.

Adding more constraints can only improve the heuristic estimate.

Heuristic Combination

Operator counting as heuristic combination

- Multiple operator-counting heuristics can be combined by computing h_C^{LP}/h_C^{IP} for the union of their constraints.
- This is an admissible combination.
 - Never worse than maximum of individual heuristics
 - Sometimes even better than their sum
- We already know a way of admissibly combining heuristics: cost partitioning.
 - ⇒ How are they related?

Connection to Cost Partitioning

Theorem

Let C_1, \ldots, C_n be sets of operator-counting constraints for s and $C = \bigcup_{i=1}^n C_i$. Then h_C^{LP} is the optimal general cost partitioning over the heuristics $h_{C_i}^{LP}$.

Proof Sketch.

In LP_C, add variables Countⁱ_o and constraints Count_o = Countⁱ_o for all operators o and $1 \le i \le n$. Then replace Count_o by Countⁱ_o in C_i .

Dualizing the resulting LP shows that $h_{\mathcal{C}}^{\mathsf{LP}}$ computes a cost partitioning. Dualizing the component heuristics of that cost partitioning shows that they are $h_{\mathcal{C}_i}^{\mathsf{LP}}$.

Comparison to Optimal Cost Partitioning

- some heuristics are more compact if expressed as operator counting
- some heuristics cannot be expressed as operator counting
- operator counting IP even better than optimal cost partitioning
- Cost partitioning maximizes, so heuristics must be encoded perfectly to guarantee admissibility. Operator counting minimizes, so missing information just makes the heuristic weaker.

Summary

Summary

- Many heuristics can be formulated in terms of operator-counting constraints.
- The operator counting heuristic framework allows to combine the constraints and to reason on the entire encoded declarative knowledge.
- The heuristic estimate for the combined constraints can be better than the one of the best ingredient heuristic but never worse.
- Operator counting is equivalent to optimal general cost partitioning over individual constraints.