

LP-based Heuristics for Cost-optimal Classical Planning

1. Introduction and Overview

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About This Tutorial

About Us



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Questions? Don't be shy to talk to us and/or email!

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Target Audience

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Ideally:

- You know what classical planning is.
keywords: STRIPS, SAS⁺
- You know what planning as heuristic search is.
keywords: A*, admissible heuristic, consistent heuristic
- You are familiar with major concepts of planning heuristics.
keywords: abstraction, landmarks, delete relaxation

Please ask questions at any time!

Tutorial Topic

Dissecting the Title

LP-based Heuristics for Cost-optimal Classical Planning

Tutorial Topic

Dissecting the Title

LP-based Heuristics for Cost-optimal **Classical Planning**

- Find path from initial to goal state
in declaratively specified state space

Tutorial Topic

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LP-based Heuristics for **Cost-optimal** Classical Planning

- Find path from initial to goal state in declaratively specified state space
- **with minimal total cost**

Tutorial Topic

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LP-based **Heuristics** for Cost-optimal Classical Planning

- Find path from initial to goal state in declaratively specified state space
- with minimal total cost
- using heuristic search algorithms

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LP-based Heuristics for Cost-optimal Classical Planning

- Find path from initial to goal state in declaratively specified state space
- with minimal total cost
- using heuristic search algorithms
- with cost estimates based on linear programming.

Tutorial Topic

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LP-based Heuristics for Cost-optimal Classical Planning

- Find path from initial to goal state in declaratively specified state space
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- with cost estimates based on linear programming.

Background: Linear Programs

Linear Programs and Integer Programs

Linear Program

A **linear program (LP)** consists of:

- a finite set of **real-valued variables** V
- a finite set of **linear inequalities** (constraints) over V
- an **objective function**, which is a linear combination of V
- which should be **minimized** or **maximized**.

Integer program (IP): ditto, but with **integer-valued** variables

Linear Program: Example

Example:

maximize $2x - 3y + z$ subject to

$$\begin{aligned} x + 2y + z &\leq 10 \\ x - z &\leq 0 \end{aligned}$$

$$x \geq 0, \quad y \geq 0, \quad z \geq 0$$

~~> unique optimal solution:

$$x = 5, y = 0, z = 5 \text{ (objective value 15)}$$

Solving Linear Programs and Integer Programs

Complexity:

- LP solving is a **polynomial-time** problem.
- Finding solutions for IPs is **NP-complete**.

Common idea:

- Approximate IP solution with corresponding LP
(**LP relaxation**).

Some LP Theory: Duality

Some LP theory: Every LP has an alternative view (its **dual**).

- roughly: variables and constraints swap roles
- dual of maximization LP is minimization LP and vice versa
- same objective value if one exists
- dual of dual: original LP

Three Key Ideas in This Tutorial

Cost Partitioning

Idea 1: Cost Partitioning

- create **copies** Π_1, \dots, Π_n of planning task Π
- each has its own **operator cost function** $cost$;
(otherwise identical to Π)
- for all o : require $cost_1(o) + \dots + cost_n(o) \leq cost(o)$

↝ sum of solution costs in copies is **admissible heuristic**:

$$h_{\Pi_1}^* + \dots + h_{\Pi_n}^* \leq h_{\Pi}^*$$

Motivation:

- method for obtaining additive admissible heuristics
- very general and powerful

Operator Counting Constraints

Idea 2: Operator Counting Constraints

- **linear constraints** whose variables denote **number of occurrences** of a given operator
- must be satisfied by every plan that solves the task

Examples:

- $Y_{o_1} + Y_{o_2} \geq 1$ "must use o_1 or o_2 at least once"
- $Y_{o_1} - Y_{o_3} \leq 0$ "cannot use o_1 more often than o_3 "

Motivation:

- declarative way to **represent knowledge** about solutions
- allows **reasoning about solutions** to derive heuristic estimates

Potential Heuristics

Idea 3: Potential Heuristics

Heuristic design as an optimization problem:

- Define simple numerical **state features** f_1, \dots, f_n .
- Consider heuristics that are **linear combinations** of features:

$$h(s) = w_1 f_1(s) + \dots + w_n f_n(s)$$

with weights (**potentials**) $w_i \in \mathbb{R}$

- Find potentials for which h is admissible and well-informed.

Motivation:

- **declarative approach** to heuristic design
- heuristic **very fast to compute** if features are

Connections

Three unrelated ideas?

- No! It turns out they are closely connected.

Tutorial Structure

- 1 Introduction and Overview
- 2 Cost Partitioning
- 3 Operator Counting
- 4 Potential Heuristics