

# 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

## Target Audience

### Ideally:

- You know what classical planning is.  
*keywords:* STRIPS, SAS<sup>+</sup>
- 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

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## Dissecting the Title

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

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

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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**

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## Dissecting the Title

### 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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## Dissecting the Title

### 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.

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

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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

$$x + 2y + z \leq 10$$

$$x - z \leq 0$$

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

↪ unique optimal solution:

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

# 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**  $\Pi_1, \dots, \Pi_n$  of planning task  $\Pi$
  - each has its own **operator cost function**  $cost_i$ ; (otherwise identical to  $\Pi$ )
  - 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