

# Planning and Optimization

## E6. Pattern Databases: Introduction

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November 11, 2024

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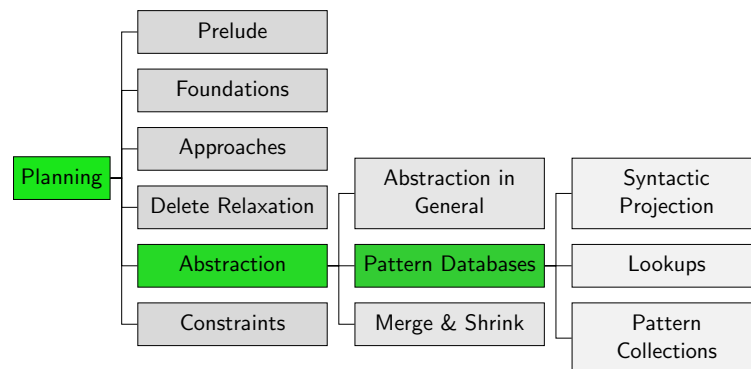
E6.1 Projections and Pattern Database Heuristics

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E6.3 Implementing PDBs: Lookup

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## Content of the Course



# E6.1 Projections and Pattern Database Heuristics

## Pattern Database Heuristics

- ▶ The most commonly used abstraction heuristics in search and planning are **pattern database (PDB) heuristics**.
- ▶ PDB heuristics were originally introduced for the **15-puzzle** (Culberson & Schaeffer, 1996) and for **Rubik's cube** (Korf, 1997).
- ▶ The first use for **domain-independent planning** is due to Edelkamp (2001).
- ▶ Since then, much research has focused on the theoretical properties of pattern databases, how to use pattern databases more effectively, how to find good patterns, etc.
- ▶ Pattern databases are a research area both in planning and in (domain-specific) heuristic search.
- ▶ For many search problems, pattern databases are the **most effective admissible heuristics** currently known.

## Pattern Database Heuristics Informally

### Pattern Databases: Informally

A pattern database heuristic for a planning task is an abstraction heuristic where

- ▶ some aspects of the task are represented in the abstraction **with perfect precision**, while
- ▶ all other aspects of the task are **not represented at all**.

This is achieved by **projecting** the task onto the variables that describe the aspects that are represented.

### Example (15-Puzzle)

- ▶ Choose a subset  $T$  of tiles (the **pattern**).
- ▶ Faithfully represent the locations of  $T$  in the abstraction.
- ▶ Assume that all other tiles and the blank can be anywhere in the abstraction.

## Projections

Formally, pattern database heuristics are abstraction heuristics induced by a particular class of abstractions called **projections**.

### Definition (Projection)

Let  $\Pi$  be an FDR planning task with variables  $V$  and states  $S$ . Let  $P \subseteq V$ , and let  $S'$  be the set of states over  $P$ .

The **projection**  $\pi_P : S \rightarrow S'$  is defined as  $\pi_P(s) := s|_P$ , (where  $s|_P(v) := s(v)$  for all  $v \in P$ ).

We call  $P$  the **pattern** of the projection  $\pi_P$ .

In other words,  $\pi_P$  maps two states  $s_1$  and  $s_2$  to the same abstract state iff they agree on all variables in  $P$ .

## Pattern Database Heuristics

Abstraction heuristics based on projections are called **pattern database (PDB) heuristics**.

### Definition (Pattern Database Heuristic)

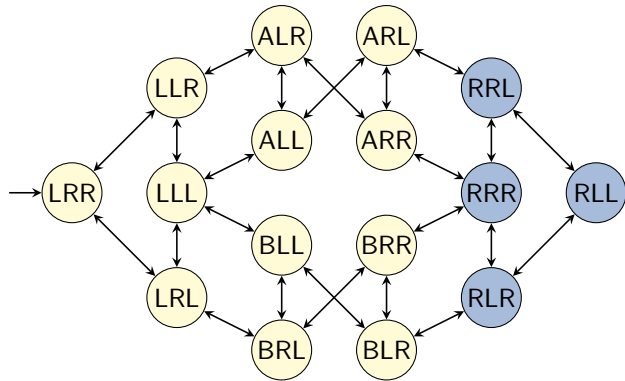
The abstraction heuristic induced by  $\pi_P$  is called a **pattern database heuristic** or **PDB heuristic**.

We write  $h^P$  as a shorthand for  $h^{\pi_P}$ .

Why are they called **pattern database heuristics**?

- ▶ Heuristic values for PDB heuristics are traditionally stored in a 1-dimensional table (array) called a **pattern database (PDB)**. Hence the name "PDB heuristic".
- ▶ The word **pattern database** alludes to **endgame databases** for 2-player games (in particular chess and checkers).

## Example: Transition System

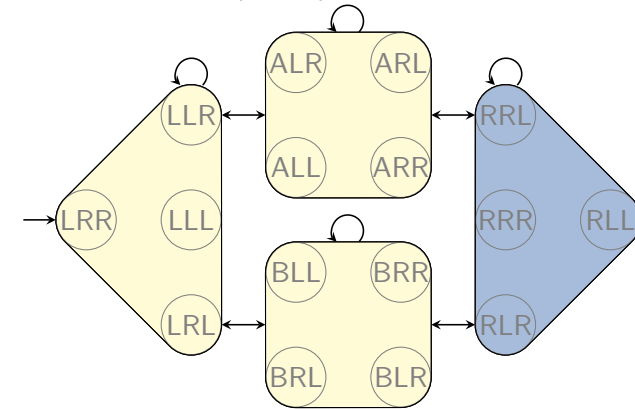


Logistics problem with one package, two trucks, two locations:

- ▶ state variable **package**:  $\{L, R, A, B\}$
- ▶ state variable **truck A**:  $\{L, R\}$
- ▶ state variable **truck B**:  $\{L, R\}$

## Example: Projection (1)

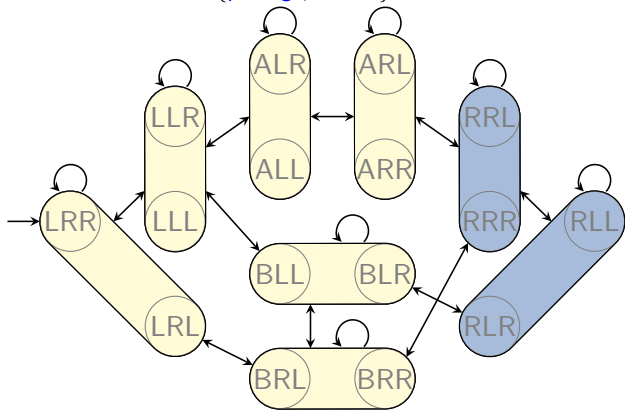
Abstraction induced by  $\pi_{\{\text{package}\}}$ :



$$h^{\{\text{package}\}}(\text{LRR}) = 2$$

## Example: Projection (2)

Abstraction induced by  $\pi_{\{\text{package}, \text{truck A}\}}$ :



$$h^{\{\text{package}, \text{truck A}\}}(\text{LRR}) = 2$$

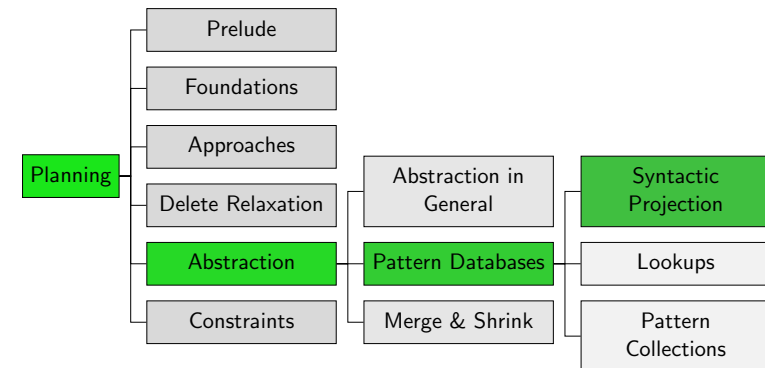
## Pattern Databases: Chapter Overview

In the following, we will discuss:

- ▶ how to **implement** PDB heuristics  
↪ this chapter
- ▶ how to effectively make use of **multiple** PDB heuristics  
↪ Chapter E7
- ▶ how to **find good patterns** for PDB heuristics  
↪ Chapter E8

## E6.2 Implementing PDBs: Precomputation

## Content of the Course



## Pattern Database Implementation

Assume we are given a pattern  $P$  for a planning task  $\Pi$ .  
How do we implement  $h^P$ ?

- 1 In a **precomputation** step, we compute a graph representation for the abstraction  $\mathcal{T}(\Pi)^{\pi_P}$  and compute the abstract goal distance for each abstract state.
- 2 During search, we use the precomputed abstract goal distances in a **lookup** step.

## Precomputation Step

Let  $\Pi$  be a planning task and  $P$  a pattern.

Let  $\mathcal{T} = \mathcal{T}(\Pi)$  and  $\mathcal{T}' = \mathcal{T}^{\pi_P}$ .

- ▶ We want to compute a graph representation of  $\mathcal{T}'$ .
- ▶  $\mathcal{T}'$  is defined through an abstraction of  $\mathcal{T}$ .
  - ▶ For example, each concrete transition induces an abstract transition.
- ▶ However, we cannot **compute**  $\mathcal{T}'$  by iterating over all transitions of  $\mathcal{T}$ .
  - ▶ This would take time  $\Omega(\|\mathcal{T}\|)$ .
  - ▶ This is prohibitively long (or else we could solve the task using uniform-cost search or similar techniques).
- ▶ Hence, we need a way of computing  $\mathcal{T}'$  in time which is **polynomial only in  $\|\Pi\|$  and  $\|P\|$** .

## Syntactic Projections

### Definition (Syntactic Projection)

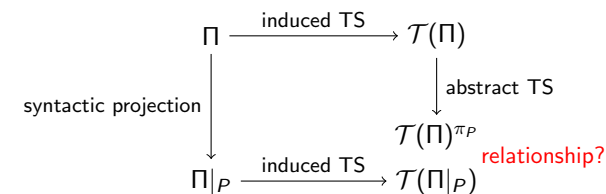
Let  $\Pi = \langle V, I, O, \gamma \rangle$  be an FDR planning task, and let  $P \subseteq V$  be a subset of its variables.

The **syntactic projection**  $\Pi|_P$  of  $\Pi$  to  $P$  is the FDR planning task  $\langle P, I|_P, \{o|_P \mid o \in O\}, \gamma|_P \rangle$ , where

- ▶  $\varphi|_P$  for formula  $\varphi$  is defined as the formula obtained from  $\varphi$  by replacing all atoms  $(v = d)$  with  $v \notin P$  by  $\top$ , and
- ▶  $o|_P$  for operator  $o$  is defined by replacing all formulas  $\varphi$  occurring in the precondition or effect conditions of  $o$  with  $\varphi|_P$  and all atomic effects  $(v := d)$  with  $v \notin P$  with the empty effect  $\top$ .

Put simply,  $\Pi|_P$  throws away all information not pertaining to variables in  $P$ .

## Idea



- ▶  $\Pi|_P$  can be computed in linear time in  $\|\Pi\|$ .
- ▶ If  $\mathcal{T}(\Pi|_P)$  was “equivalent” to  $\mathcal{T}(\Pi)^{\pi_P}$  this would give us an efficient way to compute  $\mathcal{T}(\Pi)^{\pi_P}$ .
- ▶ What do we mean with “equivalent”?
- ▶ Is this actually the case?

## Isomorphic Transition Systems

Isomorphic = equivalent up to renaming

### Definition (Isomorphic Transition Systems)

Let  $\mathcal{T} = \langle S, L, c, T, s_0, S_* \rangle$  and  $\mathcal{T}' = \langle S', L', c', T', s'_0, S'_* \rangle$  be transition systems.

We say that  $\mathcal{T}$  is **isomorphic to**  $\mathcal{T}'$ , in symbols  $\mathcal{T} \sim \mathcal{T}'$ , if there exist bijective functions  $\varphi : S \rightarrow S'$  and  $\lambda : L \rightarrow L'$  such that:

- ▶  $s \xrightarrow{\ell} t \in T$  iff  $\varphi(s) \xrightarrow{\lambda(\ell)} \varphi(t) \in T'$ ,
- ▶  $c'(\lambda(\ell)) = c(\ell)$  for all  $\ell \in L$ ,
- ▶  $\varphi(s_0) = s'_0$ , and
- ▶  $s \in S_*$  iff  $\varphi(s) \in S'_*$ .

$(\sim)$  is an equivalence relation. Two isomorphic transition systems are interchangeable for all practical intents and purposes.

## Equivalence Theorem for Syntactic Projections

### Theorem (Syntactic Projections vs. Projections)

Let  $\Pi$  be a SAS<sup>+</sup> task, and let  $P$  be a pattern for  $\Pi$ . Then  $\mathcal{T}(\Pi)^{\pi_P} \sim \mathcal{T}(\Pi|_P)$ .

### Proof.

$\rightsquigarrow$  exercises □

## PDB Computation

Using the equivalence theorem, we can compute pattern databases for  $SAS^+$  tasks  $\Pi$  and patterns  $P$ :

### Computing Pattern Databases

**def** compute-PDB( $\Pi$ ,  $P$ ):

  Compute  $\Pi' := \Pi|_P$ .

  Compute  $\mathcal{T}' := \mathcal{T}(\Pi')$ .

  Perform a backward uniform-cost search from the goal states of  $\mathcal{T}'$  to compute all abstract goal distances.

*PDB* := a table containing all goal distances in  $\mathcal{T}'$

**return** *PDB*

The algorithm runs in polynomial time and space in terms of  $\|\Pi\| + |PDB|$ .

## Generalizations of the Equivalence Theorem

- ▶ The restriction to  $SAS^+$  tasks is necessary.
- ▶ We can slightly generalize the result if we allow general negation-free formulas, but still forbid conditional effects.
  - ▶ In that case, the weighted graph of  $\mathcal{T}(\Pi)^{\pi_P}$  is isomorphic to a subgraph of the weighted graph of  $\mathcal{T}(\Pi|_P)$ .
  - ▶ This means that we can use  $\mathcal{T}(\Pi|_P)$  to derive an admissible estimate of  $h^P$ .
- ▶ With negations in conditions or with conditional effects, not even this weaker result holds.

## Going Beyond $SAS^+$ Tasks

- ▶ Most practical implementations of PDB heuristics are limited to  $SAS^+$  tasks (or modest generalizations).
  - ▶ One way to avoid the issues with general FDR tasks is to convert them to equivalent  $SAS^+$  tasks.
  - ▶ However, most direct conversions can exponentially increase the task size in the worst case.
- ↪ We will only consider  $SAS^+$  tasks in the chapters dealing with pattern databases.

## E6.3 Implementing PDBs: Lookup



## Lookup Step: Example (2)

▶  $P = \{v_1, v_2\}$  with  $v_1 = \text{package}$ ,  $v_2 = \text{truck A}$ .

▶  $\text{dom}(v_1) = \{L, R, A, B\} \approx \{0, 1, 2, 3\}$

▶  $\text{dom}(v_2) = \{L, R\} \approx \{0, 1\}$

↪  $N_1 = \prod_{j=1}^0 |\text{dom}(v_j)| = 1$ ,  $N_2 = \prod_{j=1}^1 |\text{dom}(v_j)| = 4$

↪  $\text{index}(s) = 1 \cdot s(\text{package}) + 4 \cdot s(\text{truck A})$

Pattern database:

abstract state	LL	RL	AL	BL	LR	RR	AR	BR
index	0	1	2	3	4	5	6	7
value	2	0	2	1	2	0	1	1

## E6.4 Summary

## Summary

- ▶ **Pattern database (PDB) heuristics** are abstraction heuristics based on **projection** to a subset of variables.
- ▶ For  $\text{SAS}^+$  tasks, they can easily be implemented via **syntactic projections** of the task representation.
- ▶ PDBs are **lookup tables** that store heuristic values, indexed by **perfect hash values** for projected states.
- ▶ PDB values can be looked up **very fast**, in time  $O(k)$  for a projection to  $k$  variables.