

Pattern Databases

Culberson and Schaeffer, 1998

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Introduction

- Motivation:
 - A* search algorithm of fundamental importance in AI
 - Try to find “good” heuristics
- Contribution: introduction of **pattern databases**
 - New approach for enhancing single-agent search
 - Evaluated on the 15-Puzzle

Outline

- 1 Introduction
- 2 The 15-Puzzle
- 3 Pattern Databases
- 4 Using Symmetry
- 5 Experiments
- 6 Summary

The 15-Puzzle

- Specification:
 - Set of 15 **labeled tiles**, labeled 1 to 15, and an **empty tile**, labeled 0
 - 4×4 **square grid** with positions labeled from 0 to 15
 - **State**: permutation of $(0 \dots 15)$ mapping locations to tiles
 - **Move**: swaps the empty tile with one of its neighbors, denoted by the direction the empty tile moves, i.e. $move \in \{l, r, u, d\}$

	1	2	3
4	5	6	7
8	9	10	11
12	13	14	15

Goal state τ

1	2		3
4	9	6	7
8	10	5	11
12	13	14	15

State p induced by *rrdlldruu*

The 15-Puzzle ctd.

- More definitions:

- **Distance** between two states: minimum number of moves required to produce one from the other
- **Cost** of state s : distance from s to τ
- **Path** from state p to state q : move sequence transforming p into q

- Notation:

- p is obtained from τ by the move sequence *rrdlldruu*:

$$\begin{pmatrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 \\ 1 & 2 & 0 & 3 & 4 & 9 & 6 & 7 & 8 & 10 & 5 & 11 & 12 & 13 & 14 & 15 \end{pmatrix}$$

- More compact: (1 2 0 3 4 9 6 7 8 10 5 11 12 13 14 15)

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Pattern Databases

- Definitions:
 - **Pattern**: partial specification of a permutation (or state), i.e. some locations are occupied by unspecified tiles, called **blanks**
 - **Target pattern**: partial specification of τ
 - **Pattern database (PDB)**: set of all patterns that can be obtained by permutations of a target pattern
 - **Cost** associated with each pattern in a PDB: distance to the target pattern

Pattern Databases ctd.

Lemma

For any state s , for any PDB, the cost of the pattern induced by s obtained from the PDB is a lower bound on the cost of s .

- Using the lower bound yields an **admissible heuristic**

Example

- Two example target patterns:

			3
			7
			11
12	13	14	15

Fringe target pattern

8	9	10	
12	13	14	15

Corner target pattern

- Consider state s :

1	2	3	7
4		5	6
8	9	10	11
12	13	14	15

- Cost of s in the fringe: 6
- Cost of s in the corner: 2

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Symmetries

- Symmetries:
 - Use symmetries to identify “equivalent states”
 - Here: only consider reflections
- The mirror symmetry D :
 - Reflection across the main diagonal (0, 15)
 - Maps the puzzle onto an equivalent puzzle with the empty cell in the goal state remaining in the top left corner
 - Path P has an equivalent mirror path P' obtained by replacing moves: $l \leftarrow u, u \leftarrow l, r \leftarrow d, d \leftarrow r$

Computing the Mirror Symmetry

- Computing the mirror state p' of p (without computation of the mirror path!):
 - Apply state p to the mirror D , i.e. compute $p \circ D$
 - “Undo” the mirror operation, i.e. apply D^{-1} on the result
 - Overall computation: $p' = D^{-1} \circ p \circ D$

Lemma

Any upper or lower bound on the cost of a state p or its mirror p' applies to both p and p' .

Using the Mirror Symmetry

- Using the mirror symmetry:
 - PDBs only need to store information about a state and not its mirror (need a canonic definition of state/mirror state)
 - Save nearly **50% of space** in PDBs (some states are self-reflections)
 - Fringe PDB symmetric w.r.t. the mirror, thus $cost(p) = cost(p')$
 - Same **not true** for corner PDB

Other symmetries

- Using other reflections:
 - More **involved** because empty tile not reflected on itself in the goal state
 - Possibly yields lower bounds for mirrored states different from the original one
 - **Combining several reflections**: lookup different reflections of state s in the same PDB and use the maximal retrieved value

Summary symmetries

- PDBs:
 - Use the mirror symmetry
 - Only store states and not their mirrors (reduction of one half)
- Other symmetries:
 - Given a state, compute several reflections
 - Use different reflections to retrieve additional information from PDBs

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Experiments

- Setup for IDA*:
 - Manhattan Distance heuristic (MD)
 - MD and fringe PDB (FR)
 - MD and corner PDB (CO)
 - MD and a combination of FR and CO (FC)

Experiment	Total nodes	Tree size (%)	Improvement
MD	36,302,808,031	100.00	1
MD&FR	105,067,478	0.29	346
MD&CO	83,125,633	0.23	437
MD&FC	34,987,894	0.10	1038

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Summary

- Conclusion:
 - New approach for **computation of lower bounds**: PDBs
 - Domain specific implementation for 15-Puzzle
 - **Using symmetry** of the problem allows for great performance improvements
- Outlook:
 - **Generalize** idea to non-permutation problems and general search problems
 - Find an algorithmic way to **determine good (target) patterns** other than fringe and corner

Using the Mirror Symmetry

	1	2	3
4	5	6	7
8	9	10	11
12	13	14	15

 \xrightarrow{D}

	4	8	12
1	5	9	13
2	6	10	14
3	7	11	15

 $\downarrow p$

4	1	2	3
8	6	10	7
	9	5	11
12	13	14	15

 \xleftarrow{D}

1	4	8	12
2	9	10	13
	6	5	14
3	7	11	15

Combination of FR and CO

- Combination of FR and CO:
 - PDBs often produced complementary results
 - Taking the maximum of both required too much memory
 - Split the PDBs instead and look up a value in a PDB depending on the position of the empty tile

F	C	F	C
C	F	C	F
F	C	F	C
C	F	C	F