

Disjoint Pattern Database Heuristics

by R.E. Korf and A. Felner

Tsan-sheng Hsu

徐讚昇

tshsu@iis.sinica.edu.tw

<http://www.iis.sinica.edu.tw/~tshsu>

Abstract

- Introducing a new form of heuristic called **pattern databases**.
 - Compute the cost of solving individual subgoals independently.
 - If the subgoals are disjoint, then we can use the sum of costs of the subgoals as a new and better admissible cost function.
 - ▷ *A way to get a new and better heuristic function by composing known heuristic functions.*
 - Make use of the fact that computers can memorize lots of patterns.
 - Solutions to pre-stored patterns can be pre-computed.
 - Speed up factor of over 2000 compared to previous results in 1985.

Definitions

- $n^2 - 1$ puzzle problem:
 - The numbers 1 through $n^2 - 1$ are arranged in a n by n square with one empty cell.
 - ▷ *Let $N = n^2 - 1$.*
 - Slide the tiles to a given goal position.
- 15 puzzle:
 - May be invented in 1874 and was popular in 1880.
 - It looks like one can rearrange an arbitrary state into a given goal state.
 - Publicized and published by Sam Loyd in January 1896.
 - ▷ *A prize of US\$ 1000 was offered to solve one “impossible”, but seems to be feasible case.*
- Generalizations:
 - $n \cdot m - 1$ puzzle.
 - Puzzles of different shapes.

15 puzzle

■ Rules:

- 15 tiles in a 4*4 square with numbers from 1 to 15.
- One empty cell.
- A tile can be slid horizontally or vertically into an empty cell.
- From an initial position, slide the tiles into a goal position.

■ Examples:

- Initial position:

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

- Goal position:

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

15 Puzzle — State Space

- State space is divided into subsets of even and odd permutations [Johnson & Story 1879].
 - Treat a board into a permutation by appending the rows from left to right and from top to bottom.
 - f_1 is number of inversions in a permutation $\pi_1\pi_2\cdots\pi_N$ where an inversion is a distinct pair $\pi_i > \pi_j$ such that $i < j$.
 - ▷ Let $inv(i, j) = 1$ if $\pi_i > \pi_j$ and $i < j$; otherwise, it is 0.
 - ▷ $f_1 = \sum_{\forall i, j} inv(i, j)$.
 - f_2 is the row number of the empty cell.
 - $f = f_1 + f_2$.
 - Even parity: one whose f value is even.
 - Odd parity: one whose f value is odd.
 - Slide a tile never change the parity.
- Note: the above statement may not be true for other values of n and for other shapes.

Proof: Sketch

- Slide a tile horizontally does not change the parity.
- Slide a tile vertically:
 - Change the parity of f_2 , i.e., row number of the empty cell.
 - Change the value of f_1 , i.e., the number of inversions by
 - ▷ +3
 - ▷ +1
 - ▷ -1
 - ▷ -3
 - Example: when “a” is slided down
 - ▷ *only the relative order of “a”, “b”, “c” and “d” are changed*
 - ▷ *analyze the 4 cases according to the rank of “a” in “a”, “b”, “c” and “d”.*

*	*	*	*
*	a	b	c
d		*	*
*	*	*	*

Core of past algorithms

- Using DEC 2060 a 1-MIPS machine: solves several random instances of the 15 puzzle problem within 30 CPU minutes in 1985.
- Using Iterative-deepening A^* .
- Using the Manhattan distance heuristic as an estimation of the remaining cost.
 - Suppose a tile is currently at (i, j) and its goal is at (i', j') , then
 - ▷ *the Manhattan distance for this tile is $|i - i'| + |j - j'|$.*
 - The Manhattan distance between a board and a goal board is the sum of the Manhattan distance of all the tiles.
- Manhattan distance is a lower bound on the number of slides needed to reach the goal position.
 - It is admissible.

Non-additive pattern databases

- **Intuition: do not measure the distance of one tile at a time.**
 - Pattern database: measure the collective distance of a pattern, i.e., a group of tiles, at a time.
- **Complications.**
 - The tiles get in each other's way.
 - Sliding a tile to reach its goal destination may make the other tiles that are already in their destinations to move away.
 - A form of interaction is called **linear conflict**:
 - ▷ *To flip two adjacent tiles needs more than 2 moves.*
 - ▷ *In addition, sliding tiles other than the two adjacent tiles to be flipped is also needed in order to flip them.*

Fringe

- A **fringe** is the arrangement of a subset of tiles, and may include the empty cell, by treating tiles not selected don't-care.
 - Don't-cared tiles are indistinguishable within themselves.
 - The subset of tiles selected is called a **pattern**.

- Example:

*	*	4	
*	8	*	12
*	13	*	15
*	*	14	*

- "*" means don't-care.
 - There are $16!/8! = 518,918,400$ possible fringe arrangements which is called the **pattern size**.
- The **goal fringe** arrangement for the selected subset of tiles:

*	*	*	4
*	*	*	8
*	*	*	12
13	14	15	

Solving a fringe arrangement

- For each fringe arrangement, pre-compute the **minimum** number of moves needed to make it into the goal fringe arrangement.
 - This is called the **fringe number** for the given fringe arrangement.
 - There are many possible ways to solve this problem since the pattern size is small enough to fit into the main memory.
 - ▷ *Sample solution 1: Using the original Manhattan distance heuristic to solve this smaller problem.*
 - ▷ *Sample solution 2: BFS.*

Comments on pattern size

■ Pro's.

- Pattern with a larger size is better in terms of having a larger fringe number.
- A larger fringe number usually means better estimation, i.e., closer to the goal fringe arrangement.

■ Con's.

- Pattern with a larger size means consuming lots of memory to memorize these arrangements.
- Pattern with a larger size also means consuming lots of time in constructing these arrangements.

▷ *Depends on your resource, pick the right pattern size.*

Usage of fringe numbers (1/2)

■ Divide and conquer.

- Reduce a 15-puzzle problem into a 8-puzzle one.

- Solution =

- ▷ First reach a goal fringe arrangement consisted of the first row and column.
- ▷ Then solve the 8-puzzle problem without using the fringe tiles.
- ▷ Finally Combining these two partial solutions to form a solution for the 15-puzzle problem.

- May not be optimal.

*	*	*	4
13	*	3	*
*	9	5	*
*	2	*	1

⇒

1	2	3	4
5	*	*	*
9	*	*	*
13	*	*	*

■ Divide and conquer may not be working because often times you cannot combine two sub-solutions to form the final optimal solution easily.

- In solving the second half, you may affect tiles that have reached the goal destinations in the first half.
- The two partial solutions may not be disjoint.

Usage of fringe numbers (2/2)

- New heuristic function $h()$ for IDA*: using the fringe number as the new lower bound estimation.
 - The fringe number is a lower bound on the remaining cost.
 - ▷ *It is admissible.*
- How to find better patterns for fringes?
 - Large pattern require more space to store and more time to compute.
 - Can we combine smaller patterns to form bigger patterns?
 - ▷ *They are not disjoint.*
 - ▷ *May be overlapping physically.*
 - ▷ *May be overlapping in solutions.*

More than one patterns

- Can have many different patterns that may have some overlaps:

*	*	3	*
*	*	7	*
9	10	11	12
*	*	15	

1	2	3	4
5	*	*	*
9	*	*	*
13	*	*	

- Cannot use the divide and conquer approach anymore for some of the patterns.
- If you have many different pattern databases P_1, P_2, P_3, \dots
 - The heuristics or patterns may not be disjoint.
 - ▷ *Solving tiles in one pattern may help/hurt solving tiles in another pattern even if they have no common cells.*
 - The heuristic function we can use is

$$h(P_1, P_2, P_3, \dots) = \max\{h(P_1), h(P_2), h(P_3), \dots\}.$$

Problems with multiple patterns (1/2)

- If you have many different pattern databases P_1, P_2, P_3, \dots
 - It is better to have
 - ▷ $h(P_1, P_2, P_3, \dots) = h(P_1) + h(P_2) + h(P_3) + \dots$,instead of
 - ▷ $h(P_1, P_2, P_3, \dots) = \max\{h(P_1), h(P_2), h(P_3), \dots\}$.
 - A larger $h()$ means a better performance for A^* .
- Key problem: how to make sure $h()$ is admissible?

Problems with multiple patterns (2/2)

- **Why not making the heuristics and the patterns disjoint?**
 - Though patterns are disjoint, their costs are not disjoint.
 - ▷ *Some moves are counted more than once.*
 - If the patterns are not disjoint, then we cannot add them together.
 - ▷ *Divide the board into several disjoint regions.*
- **Q: Why we add the Manhattan distance of all tiles together to form a heuristic function?**
 - We add 15 1-cell patterns together to form a better heuristic function.
 - What are the property of these patterns that can be added together?

Key observations (1/2)

- Partition the board into disjoint regions.
 - Using the tiles in a region of the goal arrangement as a pattern.

- Examples:

- | | | | |
|---|---|---|---|
| A | A | A | A |
| A | A | A | A |
| B | B | B | B |
| B | B | B | B |

- | | | | |
|---|---|---|---|
| A | A | B | B |
| A | A | B | B |
| A | A | B | B |
| A | A | B | B |

- Can also divide the board into more than 2 disjoint patterns.

- | | | | |
|---|---|---|---|
| A | A | A | B |
| A | A | B | B |
| C | A | C | B |
| C | C | C | B |

Key observations (2/2)

- For each region, solve the problem optimally and then count the moves **that are made only by tiles in this region**.
 - The “fringe” number for an arrangement is the minimum number of slides made on tiles in this region.
 - It is now possible to add fringe numbers of all disjoint regions together to form a composite fringe number.
 - ▷ *Q: How to prove this?*
- For the Manhattan distance heuristic:
 - Each pattern is a tile.
 - They are disjoint.
 - ▷ *They only count the number of slides made by each tile.*
 - Thus they can be added together to form a heuristic function.

Disjoint patterns

- A heuristic function $f()$ is **disjoint** with respect to two patterns P_1 and P_2 if
 - P_1 and P_2 have no common cells.
 - The solutions corresponding to $f(P_1)$ and $f(P_2)$ do not interfere each other.
- If they are disjoint, then $f(P_1) + f(P_2)$ is admissible if both $f(P_1)$ and $f(P_2)$ are admissible.
 - Q: How to prove this?

Revised fringe number

- Fringe number: for each fringe arrangement, the **minimum** number of moves needed to make it into the goal fringe arrangement.
 - Given a fringe arrangement H , let $f(H)$ be its fringe number.
- Revised fringe number: for each fringe arrangement F during the course of making a sequence of moves to the goal fringe arrangement, the **minimum** number of fringe-only moves in the sequence of moves.
 - Given a fringe arrangement H , let $f'(H)$ be its revised fringe number.
- Given two patterns P_1 and P_2 without overlapping cells, then
 - $f(P_1)$ and $f'(P_1)$ are both admissible.
 - $f(P_2)$ and $f'(P_2)$ are both admissible.
 - $f(P_1) + f(P_2)$ is not admissible.
 - $f'(P_1) + f'(P_2)$ is admissible.
- Note: the Manhattan distance of a 1-cell pattern is a lower bound of its revised fringe number.

Comments

- A special form of divide and conquer with additional properties.
- Spaces required by patterns must be within the main memory.
- Each pattern must be able to be solved optimally by “primitive” methods.
- It is better to put near-by tiles together to better deal with the conflicting problem.
- It is now possible to design a better admissible heuristic function f by composing two simple admissible heuristic functions f_1 and f_2 .
 - Let f'_1 be the function that does not count moves of tiles not in its region when computing f_1 .
 - ▷ $f'_1(x) \leq f_1(x)$
 - Let f'_2 be the function that does not count moves of tiles not in its region when computing f_2 .
 - ▷ $f'_2(x) \leq f_2(x)$
 - Let $f = f'_1 + f'_2$.
 - ▷ *Hopefully, $f(x) > f_1(x)$ and $f(x) > f_2(x)$.*

Performance

- Running on a 440-MHZ Sun Ultra 10 workstation.
 - SPECint = 1.0 (1 MIPS) in 1985.
 - SPECint = 17.9 in 2002.
- Solves the 15 puzzle problem that is more than 2,000 times faster than the previous result by using the Manhattan distance heuristic.
- Solves the 24-puzzle problem
 - An average of two days per problem instance.
 - Generates 2,110,000 nodes per second.
 - The average solution length was 100.78 moves.
 - The maximum solution length was 114 moves.
 - Prediction: using the Manhattan distance heuristic, it would take an average of about 50,000 years to solve a problem instance.
 - ▷ *The average Manhattan distance is 76.078 moves.*
 - ▷ *The average value for the disjoint database heuristic is 81.607 moves, which gives a tighter bound.*

Other heuristics

- The main drawback of disjoint heuristics is that they do not capture interactions between tiles in different regions.
- 2-tile pattern database:
 - For each pair of tiles, and for each pair of possible locations, compute the optimal solution for this pair of tiles to move to their destinations.
 - ▷ This is called *pairwise distance*.
 - ▷ For an $n^2 - 1$ puzzle, we have $O(n^4)$ different combinations.
 - ▷ For $n = 4$, $n^4 = 256$.
 - ▷ For $n = 5$, $n^4 = 625$.
 - For a given board, partition the board into a collection of 2-tiles so that the sum of cost is maximized.
 - ▷ This can be done using a maximum weighted perfect matching.
 - ▷ Build a complete graph with the tiles being the vertices.
 - ▷ The edge cost is the pairwise distance between these two tiles.
 - ▷ Try to find a perfect matching with the sum of edge costs being the largest possible.
 - ▷ Algorithm runs in $O(n(m + n \log n))$ is known where n is the number of vertices and m is the number of edges.

Comments

- The Manhattan distance is a partition into 1-tile patterns.
- For 2-tile patterns:
 - Faster approximation algorithms for finding maximum perfect matchings on complete graphs are known.
 - The cost for exhaustive enumeration is



$$\binom{16}{2} \binom{14}{2} \cdots \binom{4}{2} \binom{2}{2} / 8!$$

▷ = $16! / (2^8 \cdot 8!) = 2,027,025$

- Can also build 3-tile databases, but the corresponding 3-D matching problem for partitioning is NP-C.
- Requires much less memory than that of the the fringe method.
- Some kinds of bootstrapping: solving smaller problems using primitive methods, and then using these results to solve larger problems.

What else can be done?

- Looks like some kinds of two-stage search.
 - First stage searching means building pre-computed results, e.g., patterns.
 - Second stage searching meets the pre-computed results if found.
- Better way of partitioning.
- Is it possible to generalize this result to other problem domains?
- How to decide the amount of time used in searching and the amount of time used in retrieving pre-computed knowledge?
 - Memorize vs Compute

References and further readings

- Wm. Woolsey Johnson and William E. Story. Notes on the "15" puzzle. *American Journal of Mathematics*, 2(4):397–404, December 1879.
- R. E. Korf. Depth-first iterative-deepening: An optimal admissible tree search. *Artificial Intelligence*, 27:97–109, 1985.
- J. Culberson and J. Schaeffer. Pattern databases. *Computational Intelligence*, 14(3):318–334, 1998.
- * R. E. Korf and A. Felner. Disjoint pattern database heuristics. *Artificial Intelligence*, 134:9–22, 2002.