# Heuristic Search with Pre-Computed Databases

Tsan-sheng Hsu

徐讚昇

tshsu@iis.sinica.edu.tw

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

### **Abstract**

- Use pre-computed partial results to improve the efficiency of heuristic search.
- 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.
  - This year-2002 result has a speed up factor of over the 2,000 compared to a year-1985 previous result.

### **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.

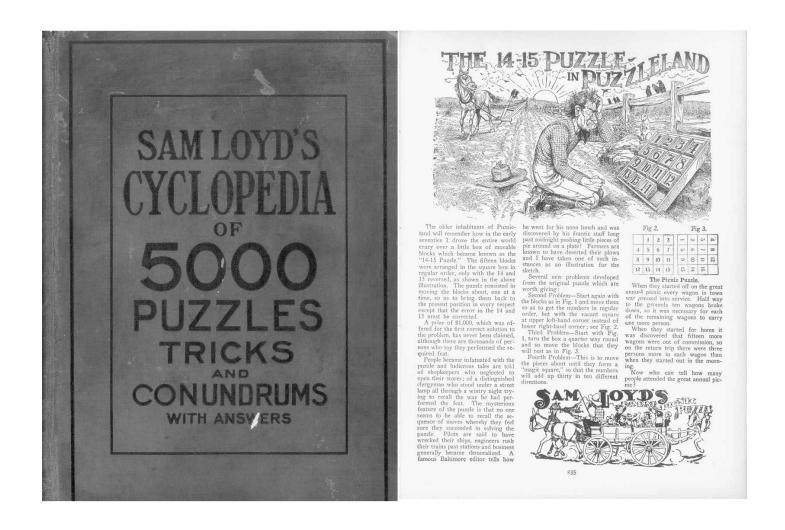
#### 15 puzzle:

- 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 Lloyd in January 1896.
  - ▶ A prize of US\$ 1,000 was offered to solve one "impossible", but seems to be feasible case.
  - ▶ Note: average wage per hour for a worker is US\$0.3.
  - ▶ Page 235, Cyclopedia of Puzzles, 1914, Sam Lloyd

#### Generalizations:

- $n \cdot m 1$  puzzle.
- Puzzles of different shapes.

### Original offer



Page 235, Cyclopedia of Puzzles, 1914, Sam Lloyd http://www.mathpuzzle.com/loyd/

### 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.
  - ▶ Optimal version: using the fewest number of moves.

#### 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 two disjoint subsets of even and odd permutations [Johnson & Story 1879].
  - Treat a board into a permutation by appending non-empty cells in 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.
    - $ightharpoonup f_1 = \sum_{\forall i,j} inv(i,j).$
    - ▶ Example: the permutation 10.8,12.3,7.6,2.1,14.4,11.15,13.9,5 has 9+7+9+2+5+4+1+0+5+0+2+3+2+1+0=50 inversions.
  - $f_2$  is the row number, i.e., 1, 2, 3, or 4, of the empty cell.
  - $f = f_1 + f_2$ .
  - Board parity
    - $\triangleright$  Even parity: one whose f value is even.
    - ▶ Odd parity: one whose f value is odd.

### 15 Puzzle — Properties 1 and 2

- Property 1: The parity of a board is either even or odd.
- Property 2: There exists some boards with even parity and some other boards with odd parity.
  - There is a board with an even parity.
    - ▶ The goal position:

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

- $ightharpoonup f_1 = 0 \text{ and } f_2 = 4.$
- There is a board with an odd parity.

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

- $ightharpoonup f_1 = 1 \text{ and } f_2 = 4.$
- The above two form the cash-prize challenge posed by Sam Lloyd in 1914.

### 15 Puzzle — Properties 3 and 4

- Property 3: Slide a tile never change the parity of a 15-puzzle board.
  - A proof sketch is given in the next slide.
- Property 4: Given two boards with the same parity, we can obtain one from the other by sliding tides.
  - Proof is omitted.
  - Note: it suffices to pick a fixed goal position for the even/odd permutations. Then prove every other permutation of the same parity can be slid into this picked goal position.
    - $\triangleright$  If A can be slid into G, and B can be slid into G, then A can be slid into B, and vice versa.

### **Proof sketch of Property 3**

- 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

 $\triangleright$  -3

- Example: when "a" is slid 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	С
d		*	*
*	*	*	*

### Generalizations (1/2)

- Properties discussed here work for  $n \times m-1$  puzzles where n and m are both at least 2 [Johnson and Story 1879].
  - ullet n is the number of rows and m is the number of columns.
- When n or m is 1, then obviously there is no way to get a different permutation by sliding from a starting permutation. Hence the parity argument does not work.
- The above parity arguments work for n>1, m>1 and m is even.
  - Take the following size-m sub-permutation ended with the empty square:  $a_1, a_2, \ldots, a_m$ .
  - After sliding  $a_1$  down, the corresponding sub-permutation becomes  $a_2, a_3, \ldots, a_m, a_1$ .
  - Let P be the original permutation, and P' be the resulting permutation by sliding down  $a_1$ .
  - When  $a_1$  is the j,  $1 \le j \le m$ , largest in the sub-permutation, then  $f_1(P') = f_1(P) m + 1 + 2 \cdot (j-1) = f_1(P) m 1 + 2 \cdot j$ .
  - $abs(-m-1+2\cdot j)$  is odd when m is even.

### Generalizations (2/2)

- When n > 1, m > 1 and m is odd.
  - $f_1(P)$  is even when m is odd.
  - Let  $f_2'(P) = 0$ .
  - Then the proof still works.
- To have a uniform definition, the proof is more complex, but doable.
  - Revised definitions
    - ▶ Insert the empty square numbered  $n \times m$ , namely largest one in the permutation, into the permutation and call it  $f''_1$ .
    - $\triangleright$   $f_2''$  is the Manhattan distance between the empty square and the right lower corner.
  - Every sliding, horizontal and vertical, changes the parity of  $f_2''$ .
  - Horizontal sliding always increases  $f_1''$  by 1.
  - Vertical sliding also changes the parity of  $f_1''$ .
  - Hence the parity argument also works here.
- Other sizes or types of sliding piece puzzles are challenging and worth researching individually.
  - Ref: Sliding Piece Puzzles, Edward Hordern, 1986, Oxford University Press, ISBN 0-19-853204-0

### 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 by 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.
  - Not good enough in terms of speed and space for solving the 24 puzzle problem.

### 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.

### **Example: Linear conflict**

The sum of Manhattan distance between the board on the left and the goal board on the right is 4.

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

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

However it takes much more than 4 slides to reach the goal.

# Fringe (1/2)

- 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: original

$\boxed{1}$	1   2   4		
3	8	5	12
6	13	7	15
9	10	14	11

fringe

*	*	4	
*	8	*	12
*	13	*	<b>15</b>
 *	The state of the s	1/1	*

- Notations for specifying a pattern.
  - "\*" means don't-care.
  - We need to know the whereabout of the empty cell no matter it is selected or not.
    - ▶ An empty space means a selected empty cell.
    - ▶ "♡" means an unselected empty cell.

# Fringe (2/2)

Example:

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

- In this example, there are 7 selected tiles, including the empty cell.
  - There are 16!/9! = 57,657,600 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.
- Remark: It is not necessary to include everything into the fringe selections to solve a puzzle. In a  $n^2-1$  puzzle, there are  $n^2-1$  tiles and 1 empty square. It suffices to not picking any one in the fringe.
  - Using a generalized pigeon hole principle, it can be argued that once you put  $n^2-1$  selections into the right locations, the one left out is sure to be in the right location.

### 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.
  - ▶ Depend 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.

$\bigcirc$	*	*	4	1	2	3	4
13	*	3	*	5	*	$\Diamond$	*
*	9	5	*	9	*	*	*
*	2	*	1	13	*	*	*

- May not be optimal.
- 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.
    - ▶ *Q*: how to prove 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
	TO	<b>T T</b>	12

	2	3	4
5	*	*	*
9	*	*	*
13	*	*	$\Diamond$

- 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

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

### Problems with multiple patterns (1/2)

- If you have many different pattern databases  $P_1$ ,  $P_2$ ,  $P_3$ , ...
  - It is better to have

```
 h(P_1, P_2, P_3, \ldots) = h(P_1) + h(P_2) + h(P_3) + \cdots,
```

instead of

```
h(P_1, P_2, P_3, \ldots) = \max\{h(P_1), h(P_2), h(P_3), \ldots\}.
```

- A larger h() means a better performance for  $A^*$ .
- Key problem: how to make sure h() is admissible through the composition process?

### Problems with multiple patterns (2/2)

- Making the heuristics and the patterns disjoint and beyond.
  - If the patterns are not disjoint, then we cannot add them together.
    - ▶ Divide the board into several disjoint regions.
  - Though patterns are disjoint, their costs are not disjoint.
    - > Some moves are counted more than once.
- Q: Why can 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 properties of these patterns so that they 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:

Α	Α	Α	Α
Α	Α	Α	Α
В	В	В	В
В	В	В	В

Α	Α	В	В
Α	Α	В	В
Α	Α	В	В
Α	Α	В	В

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

Α	Α	Α	В
Α	Α	В	В
C	Α	С	В
C	C	C	В

# 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.
  - Note: if the empty cell is selected, we do not count the moves of the empty cell.
  - 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

- lacksquare 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.

**Example:** 

1	2	3	4
5	*	*	*
9	*	*	*
13	*	*	$\Diamond$

*	*	*	*
*	6	7	8
*	9	10	11
*	14	15	

- The solutions corresponding to  $f(P_1)$  and  $f(P_2)$  do not interfere each other.
  - ▶ The above example does interfere each other.
- Then  $f(P_1) + f(P_2)$  is admissible if
  - (1) f() is disjoint with respect to  $P_1$  and  $P_2$  and
  - (2) 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 P, let f(P) be its fringe number.
- Revised fringe number: for each fringe arrangement 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 P, let f'(P) 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  $f_1$ .

$$ightharpoonup f_1'(x) \leq f_1(x)$$

• Let  $f_2'$  be the function that does not count moves of tiles not in its region when computing  $f_2$ .

$$f_2'(x) \le f_2(x)$$

• 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.
  - 2,000 \* 17.9 times faster in wall clock time.
- 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.
    - ▶ The improvement of heuristic is only 7.27%, but the speed is 2,000 times faster.

# Other heuristics (1/3)

 One of the main drawbacks of using the disjoint heuristics is that it does not capture interactions between tiles in different regions.

#### 2-tile pattern database:

- For each pair of tiles, and for each pair of possible destinations, compute the optimal solution, i.e., minimum number of all moves made by these 2 tiles, for this pair of tiles to both move to their destinations.
  - ▶ Example: two tiles i and j can be at locations loc(i) and loc(j) and want to move to destinations des(i) and des(j).
  - ▶ Usually, the destination is fixed for a tile in a puzzle.
  - $\triangleright$  The shortest moves to move both tiles i and j to their destinations is called their pairwise distance.

#### Complexities:

- Assume the destinations are fixed in a puzzle, then we need to compute  $O(n^4)$  pairwise distance for an  $n^2-1$  puzzle since the number of possible locations of tile i, loc(i), is  $O(n^2)$ .
- ightharpoonup For n = 4,  $n^4 = 256$ .
- ightharpoonup For n = 5,  $n^4 = 625$ .

# Other heuristics (2/3)

- It is usually the case that the pairwise distance of 2 tiles x and y is much larger than the sum of the Manhattan distances of x and y.
  - The pairwise distance is at least the sum of the Manhattan distances.
  - Q: How to prove this?
- For a given board, partition the board into a collection of 2-tiles so that the sum of cost is maximized.
  - This is called the maximum sum of pairwise distance.
  - For partitioning the board, we mean to find eight 2-titles so that they
    cover all tiles, including the empty cell.
  - This new cost estimation function is admissible.
    - ▶ *Q*: How to prove this?

# Other heuristics (3/3)

- Finding a maximum sum of pairwise distance 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  $O(\sqrt{n} \cdot m)$  time is known where n is the number of vertices and m is the number of edges.
    - ▶ S. Micali and V.V. Vazirani, "An  $O(\sqrt{|V|} \cdot |E|)$  algorithm for finding maximum matching in general graphs", Proc. 21st IEEE Symp. Foundations of Computer Science, pp. 17-27, 1980.
    - ▶ Faster algorithms are known since the input is a complete graph.

### **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

- The above formula gives the number of distinct number of ways to partition 16 titles into 8 pairs.
- Can also build 3-tile databases, but the corresponding 3-D matching problem for partitioning is NP-hard.
- 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

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