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CSCI-564

CONSTRAINT PROCESSING AND HEURISTIC SEARCH

LECTURE 13 – STATE SPACE PRUNING (CONTINUED)

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Recap

- Pruning is a technique to **ignore parts of the search tree** (and thus reduce the branching factor) to save runtime and memory.
- Pruning requires runtime and memory. We need to ensure that **the costs are outweighed by the corresponding savings**.
- Pruning exploit the **expert knowledge of the domain**.
 - Regularities.
- The pruning can be **static or dynamic**.





State Space Pruning

- Static and dynamic pruning give optimal solutions.
- The pruning algorithm needs to verify **before pruning** that the branch is **not leading to an optimal solution**.
- In large state space, pruning techniques does not reduce the time complexity enough.

What can we do?





Nonadmissible State Space Pruning

- We can **sacrifice the optimality** of the solution.
 - Sometimes a good, but quick solution is better.
- Example (GPS).
 - On small distances, you can calculate the optimal solution very fast.
 - But calculating the optimal path between Antigonish and San Francisco can be very long.
 - However, you only want a good solution not the optimal.
 - What is 1 hour on a 3-day travel.





Nonadmissible State Space Pruning

- The pruning technique that sacrifice the optimality are **nonadmissible**.
- We will see two techniques:
 - Macro problem solving
 - Relevance cut





Macro Problem Solving

- The idea is to **group a sequence of actions** into a new action.
 - Ex: **4x turn 90°** can be grouped into **turn 360°**.
- The **problem solver (algorithm)** can apply multiple **primitive operators** at once.
- Where is the pruning?
 - Requires **fewer decisions**.
 - **Choices inside a branch are ignored**.





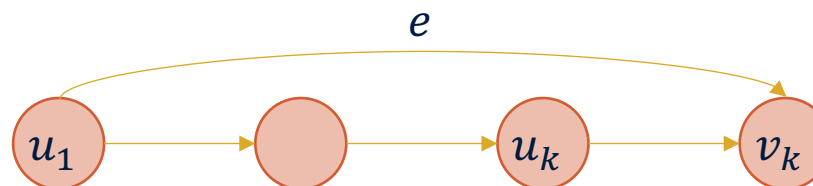
Macro Problem Solving

- Is there a catch?
 - If the **substitutions operators are too generous** (grouping to many primitive operators) the **goal might not be found**.
- We need to ensure that the **goal is still reachable**.



Macro Problem Solving

- **Definition (Macro Operator):**
 - A **macro operator** (macro) is a fixed sequence of elementary operators executed together.
 - For a problem graph with node set V and an edge set E .
 - A macro refers to an additional edge $e = (u, v)$ in $V \times V$ for which there are edges $e_1 = (u_1, v_1), \dots, e_k = (u_k, v_k) \in E$ with $u = u_1, v = v_k$ and $v_i = u_{i+1}$ for all $1 \leq i \leq k - 1$.
- In other words, the path (u_1, \dots, u_k, v_k) between u and v is shortcut by introducing e .





Macro Problem Solving

- Macros turn an **unweighted graph** into a **weighted graph**.
- Why?
 - Macros can have different lengths.
 - We need to know the weight of a macro to find the **best solution**.
- The weight of the macro is the **accumulated weight of the original edges**:
 - $w(u, v) = \sum_{i=1}^k w(u_i, v_i)$





Macro Problem Solving

- Inserting edges **does not affect the reachability** status of nodes.
- If there is no alternative in the choice of successors.
 - $Succ(u_i) = \{v_i\}$
 - Macros can **substitute the original edges without loss of information**.
- Example:
 - Maze areas with width of one (tunnel).





Macro Problem Solving

- If there are more paths between a node?
- To preserve the optimality of an underlying search algorithm.
 - We take the shortest path $w(u, v) = \delta(u, v)$.
- These macros are called admissible.





Macro Problem Solving

- How can we **create the macros**?
 - The *All-Pairs Shortest Paths* algorithm of Floyd-Warshall is one way.
 - At the end of the algorithm, all two nodes are connected.
 - The original edges are no longer needed to determine the shortest path.
 - It keeps the optimality of the search.
- So we can find the optimal solution with macros?





Macro Problem Solving

- True, for small problems.
- For larger problems computing *All-Pairs Shortest Paths* is infeasible.
- If we accept feasible solutions:
 - We can use inadmissible macros.
 - Delete edges after some admissible macros have been introduced.
- The importance of macros is that they can be determined before the search.
 - It's called Macro learning.





Macro Problem Solving

- How to use **inadmissible macros**:
 - Inserting them with a weight $w(u, v)$ **smaller than the optimum** $\delta(u, v)$.
 - The macros will be used with higher priority.
 - Or we can **restrict the search to macros only**.
 - Only possible if the **goal stay reachable**.
- Creating inadmissible macros depends on the problem.





Macro Problem Solving

- **Example:**
 - We decompose the problem in **subgoals**.
 - For each subgoals a set of macros is defined that **transform a state into the next subgoal**.





Eight-Puzzle

- Actions are labeled by the direction in **which the blank is moving**.
- We create a table:
 - The entry in row r and in the column c contains a macro.
 - The macro is the sequence to position the tile in position r to the position c .
 - After execution the tiles in position 1 to $r - 1$ remain correctly placed.

6	1	3
8	4	7
2	5	

Starting state

1	2	3
8		4
7	6	5

Goal state





Eight-Puzzle

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
4	L	RULD	RULD	LURRD			
5	UL	DRUL	RDLU	RULD	RDLU		
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU

6	1	3
8	4	7
2	5	

c=0, r=5



6	1	3
8	4	
2	5	7



6	1	3
8		4
2	5	7

1	2	3
8		4
7	6	5

Goal state

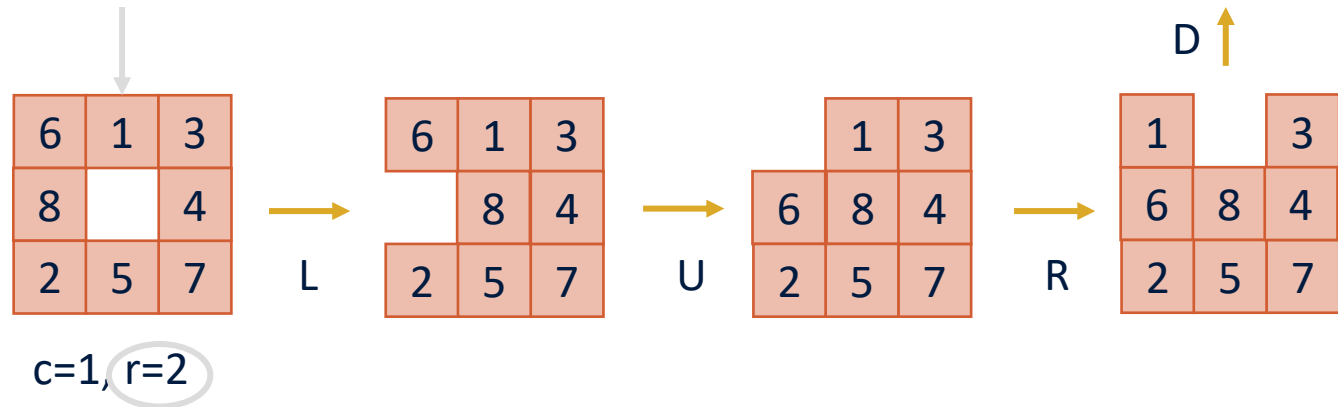




Eight-Puzzle

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
4	L	RULD	RULD	LURRD			
5	UL	DRUL	RDLU	RULD	RDLU		
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU

The tiles in the correct position didn't move.



1	8	3
6		4
2	5	7

D ↑

1		3
6	8	4
2	5	7

1	2	3
8		4
7	6	5

Do you see a pattern?

Goal state

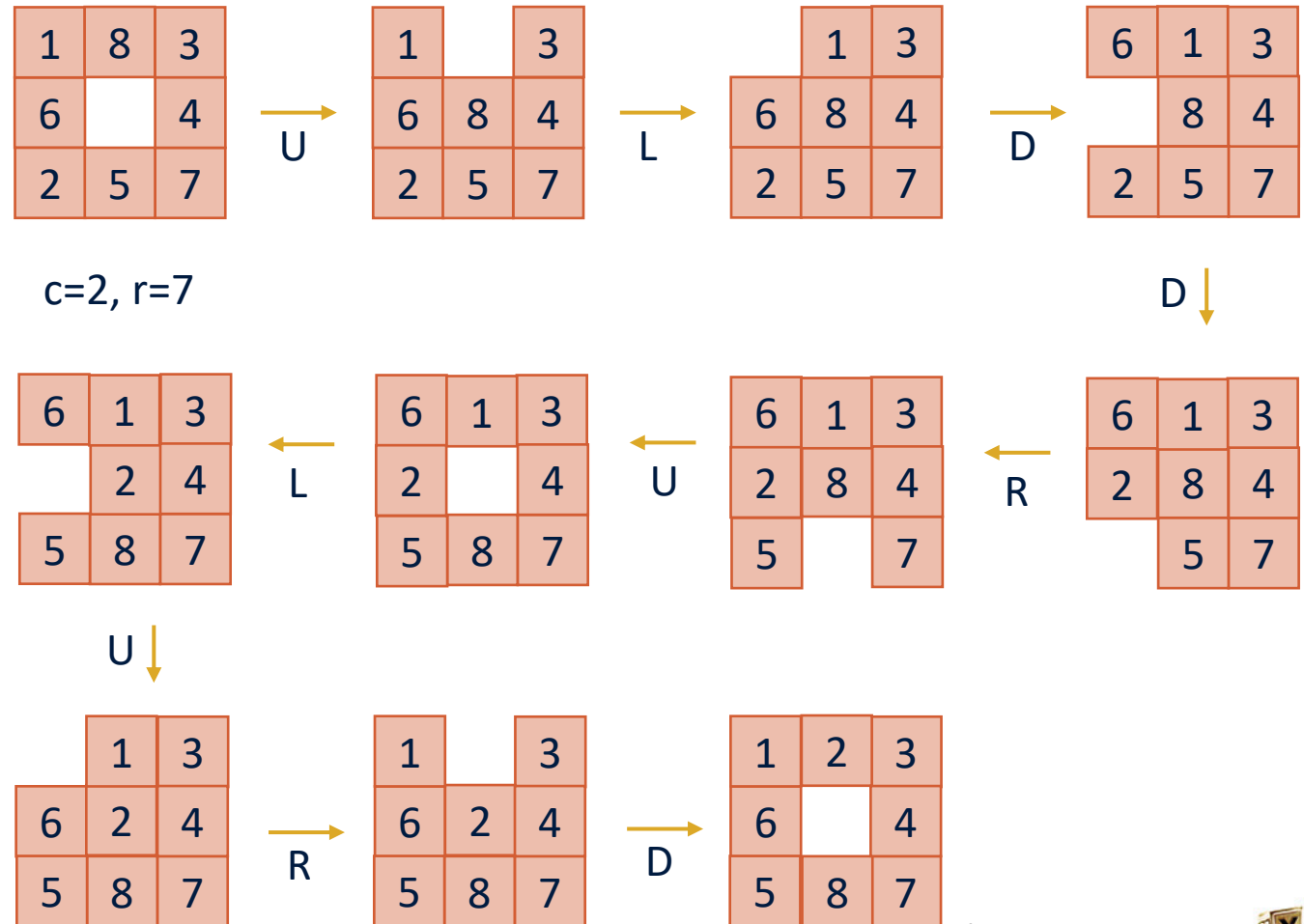




Eight-Puzzle

	0	1	2	3	4	5	6
--	---	---	---	---	---	---	---

0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
		LURD					
4	L	RULD	RULD	LURRD			
		LURD		LULDR			
5	UL	DRUL	RDLU	RULD	RDLU		
		DLUR	RULD	RDLU			
		ULDR		URDL			
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
		ULDR		RDRU		DLURU	
				LLDR		LLDR	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
		ULDR	ULURD	URDRU	DRUL	URRDLU	
			LLDR	LLDR			
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU
			UULD	ULLDR		LDRU	
						RDLU	





Eight-Puzzle

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
		LURD					
4	L	RULD	RULD	LURRD			
		LURD		LULDR			
5	UL	DRUL	RDLU	RULD	RDLU		
		DLUR	RULD	RDLU			
		ULDR		URDL			
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
		ULDR		RDRU		DLURU	
				LLDR		LLDR	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
		ULDR	ULURD	URDRU	DRUL	URRDLU	
			LLDR				
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU
			UULD	ULLDR		LDRU	
						RDLU	

1	2	3
6		4
5	8	7

The tiles 3 and 4 are in the correct positions.
We can skip it.





Eight-Puzzle

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
		LURD					
4	L	RULD	RULD	LURRD			
		LURD		LULDR			
5	UL	DRUL	RDLU	RULD	RDLU		
		DLUR	RULD	RDLU			
		ULDR		URDL			
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
		ULDR		RDRU		DLURU	
				LLDR		LLDR	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
		ULDR	ULURD	URDRU	DRUL	URRDLU	
			LLDR				
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU
			UULD	ULLDR		LDRU	
						RDLU	

1	2	3
6		4
5	8	7

DRULDL
 →
 URRDLU

1	2	3
7		4
6	8	5

c=5, r=7

The tiles in the correct positions don't move.





Eight-Puzzle

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
		LURD					
4	L	RULD	RULD	LURRD			
		LURD		LULDR			
5	UL	DRUL	RDLU	RULD	RDLU		
		DLUR	RULD	RDLU			
		ULDR		URDL			
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
		ULDR		RDRU		DLURU	
				LLDR		LLDR	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
		ULDR	ULURD	URDRU	DRUL	URRDLU	
				LLDR			
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU
			UULD	ULLDR		LDRU	
						RDLU	

1	2	3
7		4
6	8	5

c=6, r=7

DLUR →

1	2	3
8		4
7	6	5

It's done.

Was it optimal?





Eight-Puzzle

- In the **worst-case scenario**:
 - We sum the string size maxima in the columns.
 - $2+12+10+14+8+14+4=64$
- The **average solution length**:
 - We calculate the arithmetic means.
 - $12/9+52/8+40/7+58/6+22/5+38/4+8/3=39.78$
- Considering that you're using the macro table!

	0	1	2	3	4	5	6
0							
1	DR						
2	D	LURD					
3	DL	URDL	URDL				
		LURD					
4	L	RULD	RULD	LURRD			
		LURD		LULDR			
5	UL	DRUL	RDLU	RULD	RDLU		
		DLUR	RULD	RDLU			
		ULDR		URDL			
6	U	DLUR	DRUULD	DLUU	DRUL	LURRD	
		ULDR		RDRU		DLURU	
				LLDR		LLDR	
7	UR	LDRU	ULDDR	LDRUL	DLUR	DRULDL	DLUR
		ULDR	ULURD	URDRU	DRUL	URRDLU	
				LLDR			
8	R	ULDR	LDRR	LURDR	LDRRUL	DRUL	LDRU
			UULD	ULLDR		LDRU	
						RDLU	





Eight-Puzzle

- How can we construct a macro table?
 - The **most efficient** way is using **DFS or BFS**.
 - **Starting from each goal state** to every other states.
- Depending on the problem the search effort can be important.





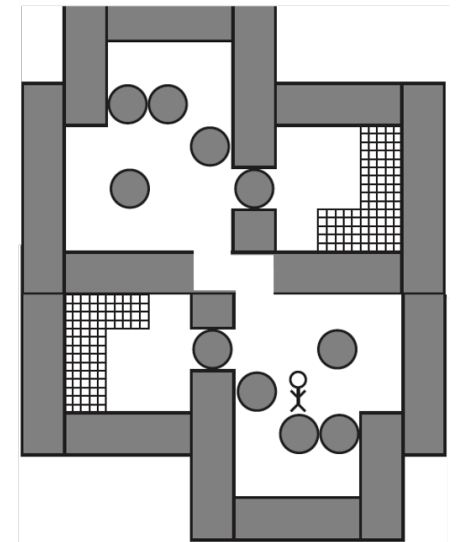
Relevance Cuts

- Humans can **navigate through large state spaces** due to an ability to use meta-level reasoning.
- Meta-level strategy (reasoning) **distinguish between relevant and irrelevant actions**.
 - **Divide a problem into several subgoals**, then solve the subgoals one after the other.
- Standard search algorithm like A* **always consider all possible moves available**.



Relevance Cuts

- **Example:**
 - In mirror-symmetrical Sokoban.
 - It is obvious that each half can be **solved independently**.
 - Algorithm like A* will explore strategy that humans would never consider.
 - Switching back and forth between the two subproblems.





Relevance Cuts

- **Relevance cuts:**
 - Attempt to restrict the way the algorithm **chooses the next action**.
 - The idea is to prevent the program from trying **all possible move sequences**.
 - It introduces the notion of **influence**.
- Moves that don't influence each other are called **distant moves**.





Relevance Cuts

- A move can be cut off:
 - If within **the last m moves more than k distant moves** were made.
 - This cut will discourage arbitrary switches between non-related areas of the maze.
 - Or a move that is **distant with respect to the previous move, but not distant to a move in the past m moves.**
 - This will not allow switches back into an area previously worked on and abandoned just briefly.





Relevance Cuts

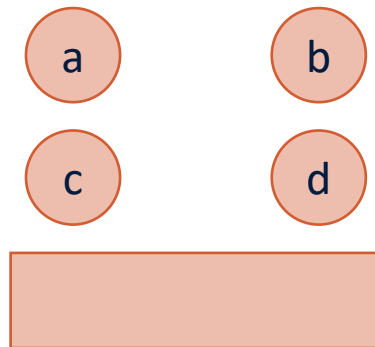
- The definition of distant moves **depends on the problem domain.**
- For the Sokoban:
 - Create a **measure for influence.**
 - **Compute a table** for the influence of each square on each other.
 - The influence relation reflects the **number of paths between the squares.**
 - The more alternatives exists, the less influence.





Relevance Cuts

- In this example:
 - a and b influence each other less than c and d .
 - Squares on the optimal place should have a stronger influence than others.
 - a influences c more than c influences a .
 - Neighboring squares that are connected by a possible ball push are more influencing than if only the man can move between them





Relevance cuts

- Given an influence table, a move M2 is regarded as distant from a previous move M1, if its from-square influences M1's from-square by less than some threshold, θ .





Nonadmissible State Space Pruning

- **Macro problem solving** prunes actions in favor of a few action sequences (called **macros**), which not only **decreases the branching factor but also the search depth**.
 - We applied it on the Eight-Puzzle where the macros bring one tile after the other into place without disturbing the tiles in the correct position.
- **Relevance cuts** prune actions in a state that are considered unimportant because they do not contribute to the subgoal currently pursued.
 - **Actions that do not influence each other** are called **distant actions**.
 - Relevance cuts can prune an action if more than a certain number of distant actions have been executed recently
 - We used Sokoban to illustrate relevance cuts.

