

Depth-First Iterative-Deepening: An Optimal Admissible Tree Search

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Abstract

- The complexities of various search algorithms are considered in terms of time, space, and cost of the solution paths.
 - Brute-force search
 - ▷ *Breadth-first search (BFS)*
 - ▷ *Depth-first search (DFS)*
 - ▷ *Depth-first Iterative-deepening (DFID)*
 - ▷ *Bi-directional search*
 - Heuristic search: best-first search
 - ▷ A^*
 - ▷ IDA^*
- The issue of storing information in DISK instead of main memory.
- Solving 15-puzzle.

Definitions

- **Node branching factor b** : the number of different new states generated from a state.
 - Average node branching factor.
 - Assumed to be a constant here.
- **Edge branching factor e** : the number of possible new, maybe **duplicated**, states generated from a state.
 - Average node branching factor.
 - Assumed to be a constant here.
- **Depth** of a solution d : the shortest length from the initial state to one of the goal states
 - The depth of the root is 0.
- **A search program finds a goal state starting from the initial state by exploring states in the state space.**
 - Brute-force search
 - Heuristic search

Brute-force search

- A **brute-force search** is a search algorithm that uses information about
 - the initial state,
 - operators on finding the states adjacent to a state,
 - and a test function whether a goal is reached.
- A “pure” brute-force search program.
 - A state maybe re-visited many times.
- An “intelligent” brute-force search algorithm.
 - Make sure a state will be visited a limited number of times.
 - ▷ *Make sure a state will be eventually visited.*

A “pure” brute-force search

- A “**pure**” brute-force search is a brute-force search algorithm that does not care whether a state has been visited before or not.
- Algorithm Brute-force(N_0)
{* do brute-force search from the starting state N_0 *}
 - $current \leftarrow N_0$
 - While true do
 - ▷ *If current is a goal, then return success*
 - ▷ *current \leftarrow a state that can reach current in one step*
- Comments
 - Very easy to code and use very little memory.
 - May take infinite time because there is no guarantee that a state will be eventually visited.

Intelligent brute-force search

- An “**intelligent**” brute-force search algorithm.
 - Assume S is the set of all possible states
 - Use a systematic way to examine each state in S one by one so that
 - ▷ *A state is not examined too many times — does not have too many duplications.*
 - ▷ *It is **efficient** to find an unvisited state in S .*
- Need to know whether a state has been visited before efficiently.
- Some notable algorithms.
 - Breadth-first search (BFS).
 - Depth-first search (DFS) and its variations.
 - Depth-first Iterative deepening (DFID).
 - Bi-directional search.

Breadth-first search (BFS)

- $deeper(N)$: gives the set of all possible states that can be reached from the state N .
 - It takes at least $O(e)$ time to compute $deeper(N)$.
 - The number of distinct elements in $deeper(N)$ is b .
- Algorithm $BFS(N_0)$ **{* do BFS from the starting state N_0 *}**
 - If the starting state N_0 is a goal, then return success
 - Initialize a Queue Q
 - Add N_0 to Q
 - While Q is not empty do
 - ▷ Remove a state N from Q
 - ▷ If one of the states in $deeper(N)$ is goal, then return success
 - ▷ Add states in $deeper(N)$ that have not been visited before to Q ;
 - ▷ Mark these newly added states as visited; if there are duplications in $deeper(N)$, add only once;
 - Return fail

BFS: analysis

■ Space complexity:

- $O(b^d)$

- ▷ *The average number of distinct elements at depth d is b^d .*
- ▷ *We need to store all distinct elements at depth d in the Queue.*
- ▷ *We need to keep a record on visited nodes in order not to re-visit them.*

■ Time complexity:

- $1 * e + b * e + b^2 * e + b^3 * e + \dots + b^{d-1} * e = (b^d - 1) * e / (b - 1) = O(b^{d-1} * e)$,
if b is a constant.

- ▷ *For each element N in the Queue, it takes at least $O(e)$ time to find deeper(N).*
- ▷ *It is always true that $e \geq b$.*

BFS: comments

- Always finds an optimal solution, i.e., one with the smallest possible depth d .
 - Do not need to worry about falling into loops.
- Most critical drawback: huge space requirement.
 - It is tolerable for an algorithm to be 100 times slower, but not so for one that is 100 times larger.

BFS: ideas when there is little memory

- What can be done when you do not have enough main memory?
 - DISK
 - ▷ *Store states that has been visited before into DISK and maintain them as sorted.*
 - ▷ *Store the QUEUE into DISK.*
 - Memory: buffers
 - ▷ *Most recently visited nodes.*
 - ▷ *Candidates of possible newly explored nodes.*
 - Merge the buffer of visited nodes with the one in DISK when memory is full.
 - ▷ *We only need to know when a newly explored node has been visited or not when it is about to be removed from the QUEUE.*
 - ▷ *The decision of whether it has been visited or not can be **delayed**.*
 - Append the buffer of newly explored nodes to the QUEUE the DISK when memory is full or it is empty.

BFS: disk based

■ Algorithm $\text{BFS}_{\text{disk}}(N_0)$

{* do disk based BFS from the starting state N_0 *}

- If the starting state N_0 is a goal, then return success
- Initialize a Queue Q_d using DISK
- Initialize a buffer Q_m of potential states to visit using main memory
- Initialize a sorted list V_d of visited nodes using DISK
- Initialize a buffer V_m of visited nodes using main memory
- Add N_0 to Q_d
- While Q_d and Q_m are not both empty do
 - ▷ If Q_d is empty, then { Sort Q_m ; Write Q_m to Q_d ; Empty Q_m }
 - ▷ Remove a state N from Q_d
 - ▷ Add N to V_m
 - ▷ If V_m is full, then { Sort V_m ; Merge V_m into V_d ; Empty V_m }
 - ▷ If one of the states in $\text{deeper}(N)$ is goal, then return success
 - ▷ Add unvisited states in $\text{deeper}(N)$ to Q_m ; Mark them as visited;
 - ▷ If Q_m is full, then {
Sort Q_m ;
Add states in Q_m that are not in V_d and V_m to Q_d ;
Empty Q_m }
- Return fail

Disk based algorithms (1/3)

- **When data cannot be loaded into the memory, you need to re-invent algorithms even for tasks that may look simple.**
 - **Batched processing.**
 - ▷ *Accumulate tasks and then try to perform these tasks when they need to.*
 - ▷ *Combine tasks into one to save disk I/O time.*
 - ▷ *Order disk accessing patterns.*
- **Main ideas:**
 - **When two files are sorted, it is cost effective to compare the difference of them.**
 - **It is not too slow to read all records of a large file in sequence.**
 - **It is very slow and cost a lot to read every record in a large file in a random order.**

Disk based algorithms (2/3)

- **Implementation of the QUEUE.**
 - QUEUE can be stored in one disk file.
 - Newly explored ones are appended **at the end** of the file.
 - Retrieve the one at the head of the file.
- **A newly explored node will be explored after the current QUEUE is empty.**

Disk based algorithms (3/3)

- How to find out a list of newly explored nodes have been visited or not?
 - Maintain the list of visited nodes on DISK **sorted**.
 - ▷ *When the member buffer is full, sort it.*
 - ▷ *Merge the sorted list of newly visited nodes in buffer into the one stored on DISK.*
 - We can easily compare two sorted lists and find out the intersection or difference of the two.
 - ▷ *We can easily remove the ones that are already visited before once Q_m is sorted.*
 - ▷ *To revert items in Q_m back to its the original BFS order, which is needed for persevering the BFS search order, we need to sort again using the original BFS ordering.*
- Why we can delay the decision of whether a newly explored node has been visited or not?
 - We only need to know when a newly explored node has been visited or not when it is about to be removed from the QUEUE.
 - The decision of whether it has been visited or not can be **delayed**.

Depth-first search (DFS)

- $next(current, N)$: returns the state next to the state “ $current$ ” in $deeper(N)$.
 - Assume states in $deeper(N)$ are given a linear order with dummy first and last elements both being $null$, and assume $current \in deeper(N)$.
 - Assume we can efficiently generate $next(current, N)$ based on “ $current$ ” and N .
- **Algorithm DFS(N_0)** { * do DFS from the starting state N_0 * }
 - Initialize a Stack S
 - Push $(null, N_0)$ to S
 - While S is not empty do
 - ▷ Pop $(current, N)$ from S
 - ▷ $R \leftarrow next(current, N)$
 - ▷ If R is a goal, then return success
 - ▷ If R is $null$, then continue { * searched all children of N * }
 - ▷ Push (R, N) to S
 - ▷ If R is already in S , then continue { * to avoid loops * }
 - ▷ Push $(null, R)$ to S { * search deeper * }
 - ▷ Can introduce some cut-off depth here in order not to go too deep
 - Return fail

DFS: analysis

■ Time complexity:

- $O(e^d)$

▷ *The number of possible branches at depth d is e^d .*

■ Space complexity:

- $O(d)$

▷ *Only need to store the current path in the Stack.*

■ Comments:

- Without a good cut-off depth, it may not be able to find a solution in time.
- May not find an optimal solution at all.
- Heavily depends on the **move ordering**.
 - ▷ *Which one to search first when you have multiple choices for your next move?*
- A node can be searched many times.
 - ▷ *Need to do something, e.g., hashing, to avoid researching too much.*
 - ▷ *Need to balance the effort to memorize and the effort to research.*
- Most critical drawback: huge and unpredictable time complexity.

DFS: when there is little memory

- Need to have a hash table to store the set of visited nodes in order not to visit a node too many times.
 - We need to decide instantly whether a node is visited or not.
 - The decision of whether a node is visited or not cannot be delayed.
 - ▷ *Batch processing is not working here.*
 - ▷ *It takes too much time to handle a disk based hash table.*
- Use data compression and/or bit-operation techniques to store as many visited nodes as possible.
 - Some nodes maybe visit again and again.
 - Need a good heuristic to store the most frequently visited nodes.
 - ▷ *Avoid swapping too often.*

DFS with a depth limit

- Do DFS from the starting state N_0 without exceeding a given depth *limit*.
 - $\text{length}(x, y)$: the number of edges between a shortest path from the node x and the node y .
 - Depth of a node x in a tree = $\text{length}(\text{root}, x)$.
- Algorithm $\text{DFS}_{\text{depth}}(N_0, \text{limit})$
 - Initialize a Stack S
 - Push (null, N_0) to S where N_0 is the initial state
 - While S is not empty do
 - ▷ Pop $(\text{current}, N)$ from S
 - ▷ $R \leftarrow \text{next}(\text{current}, N)$
 - ▷ If R is a goal, then return success
 - ▷ If R is null, then continue $\{ * \text{ searched all children of } N * \}$
 - ▷ If $\text{length}(N_0, R) > \text{limit}$, then continue $\{ * \text{ cut off } * \}$
 - ▷ Push (R, N) to S
 - ▷ If R is already in S , then continue $\{ * \text{ to avoid loops } * \}$
 - ▷ Push (null, R) to S $\{ * \text{ search deeper } * \}$
 - Return fail

Depth-first iterative-deepening (DFID)

- $\text{DFS}_{depth}(N, limit)$: DFS from the starting state N and with a depth cut off at the depth $limit$
- Algorithm $\text{DFID}(N_0, cut_off_depth)$ **{* do DFID from the starting state N_0 with a depth limit cut_off_depth *}**
 - $limit \leftarrow 0$
 - **While** $limit < cut_off_depth$ **do**
 - ▷ *If $\text{DFS}_{depth}(N_0, limit)$ finds a goal state g , then return g as the found goal state*
 - ▷ $limit \leftarrow limit + 1$
 - **Return fail**
- **Space complexity:**
 - $O(d)$

Time complexity of DFID (1/2)

- The branches at depth i are generated $d - i + 1$ times.

- There are e^i branches at depth i .

- Total number of branches visited $M(e, d)$ is

$$\begin{aligned} & (d + 1)e^0 + de^1 + (d - 1)e^2 + \dots + 2e^{d-1} + e^d \\ &= e^d(1 + 2e^{-1} + 3e^{-2} + \dots + (d + 1)e^{-d}) \\ &\leq e^d(1 - 1/e)^{-2} \text{ if } e > 1 \end{aligned}$$

- Analysis:

▷ $(1 - x)^{-2} = 1/(1 - 2x + x^2) = 1 + 2x + 3x^2 + \dots + kx^{k-1} + (k + 1)x^k - kx^{k+1}$.

▷ Hence $1 + 2x + 3x^2 + \dots + kx^{k-1} \leq (1 - x)^{-2}$, if $|x| < 1$.

▷ Since $|x| < 1$,

$$\lim_{k \rightarrow \infty} ((k + 1)x^k - kx^{k+1}) = 0.$$

▷ If k is large enough and $|x| < 1$, then $(1 - x)^{-2} \approx 1 + 2x + 3x^2 + \dots + kx^{k-1}$.

Time complexity of DFID (2/2)

- Let $M(e, d)$ be the total number of branches visited by DFID with an edge branching factor of e and depth d .
- Examples:
 - When $e = 2$, $M(e, d) \leq 4e^d$.
 - When $e = 3$, $M(e, d) \leq 9/4e^d$.
 - When $e = 4$, $M(e, d) \leq 16/9e^d$.
 - When $e = 5$, $M(e, d) \leq 25/16e^d$.
 - $M(e, d) = O(e^d)$ with a small constant factor.

DFID: comments

- No need to worry about a good cut-off depth as in DFS.
- Still need a mechanism to decide instantly whether a node has been visited before or not.
- Good for a tournament situation where each move needs to be made in a limited amount of time.
- Q:
 - ▷ *Does DFID always find an optimal solution?*
 - ▷ *How about BFID?*

DFS with depth limit and direction (1/2)

- Two refined service routines when direction of the search is considered:
 - $\text{DFS}_{dir}(B, G, \text{successor}, i)$: DFS with the set of starting states B , goal states G , successor function and depth limit i .
 - $\text{next}_{dir}(\text{current}, \text{successor}, N)$: returns the state next to the state “current” in $\text{successor}(N)$.
- In the above two routines:
 - successor is *deeper* for forward searching
 - successor is *prev* for backward searching
- Note:
 - Given a state N , $\text{prev}(N)$ gives all states that can reach N in one step.
 - Given a state N , $\text{deeper}(N)$ gives the set of all possible states that can be reached from the state N in one step.

DFS with depth limit and direction (2/2)

- **DFS_{dir}($B, G, successor, i$):** DFS with the set of starting states B , goal states G , *successor* function and depth limit i .
- **Algorithm DFS_{dir}($B, G, successor, limit$)**
 - Initialize a Stack S
 - For each possible starting state t in B do
 - ▷ *Push (null, t) to S*
 - While S is not empty do
 - ▷ *Pop (current, N) from S*
 - ▷ $R \leftarrow next_{dir}(current, successor, N)$
 - ▷ *If R is a goal in G, then return success*
 - ▷ *If R is null, then continue { * searched all children of N * }*
 - ▷ *If length(B, R) > limit, then continue { * cut off * }*
 - ▷ *Push (R, N) to S*
 - ▷ *If R is already in S, then continue { * to avoid loops * }*
 - ▷ *Push (null, R) to S { * search deeper * }*
 - Return fail
- **Note** length(B, x) is the length of a shortest path between the state x and a state in B .

Bi-directional search

- Combined with iterative-deepening.
- $\text{DFS}_{dir}(B, G, successor, i)$: DFS with the set of starting states B , goal states G , $successor$ function and depth limit i .
 - $successor$ is deeper for forward searching
 - $successor$ is $prev$ for backward searching
 - ▷ Given a state S_i , $prev(S_i)$ gives all states that can reach S_i in one step.
- Algorithm $\text{BDS}(N_0, cut_off_depth)$
 - $limit \leftarrow 0$
 - while $limit < cut_off_depth$ do
 - ▷ if $\text{DFS}_{dir}(\{N_0\}, G, deeper, limit)$ returns success, then return success **{* forward searching *}**
else store all states at depth = $limit$ in an area H
 - ▷ if $\text{DFS}_{dir}(G, H, prev, limit)$ returns success, then return success **{* backward searching *}**
 - ▷ if $\text{DFS}_{dir}(G, H, prev, limit + 1)$ returns success, then return success **{* in case the optimal solution is odd-lengthed *}**
 - ▷ $limit \leftarrow limit + 1$
 - return fail
- Backward searching at depth = $limit + 1$ is needed to find odd-lengthed optimal solutions.

Bi-directional search: analysis

- **Time complexity:**
 - $O(e^{d/2})$
- **Space complexity:**
 - $O(e^{d/2})$: needed to store the half-way meeting points H .
- **Comments:**
 - Run well in practice.
 - Depth of the solution is expected to be the same for a normal uni-directional search, however the number of nodes visited is greatly reduced.
 - Pay the price of storing solutions at half depth.
 - **Need to know how to enumerate the set of goals.**
 - Trade off between time and space.
 - ▷ *What can be stored on DISK?*
 - ▷ *What operations can be batched?*
 - **Q:**
 - ▷ *How about using BFS in forward searching?*
 - ▷ *How about using BFS in backward searching?*
 - ▷ *How about using BFS in both directions?*

Heuristic search

- **Heuristics**: criteria, methods, or principles for deciding which among several alternative courses of actions promises to be the most effective in order to achieve some goal [Judea Pearl 1984].
 - Need to be simple and effective in discriminate correctly between good and bad choices.
- A **heuristic search** is a search algorithm that uses information about
 - the initial state,
 - operators on finding the states adjacent to a state,
 - a test function whether a goal is reached, and
 - heuristics to pick the next state to explore.
- A “good” heuristic search algorithm:
 - States that are not likely leading to the goals will not be explored further.
 - ▷ *A state is cut or pruned.*
 - States are explored in an order that are according to their likelihood of leading to the goals → **good move ordering**.

Heuristic search: A^*

- Combining DFID with best-first heuristic search such as A^* .
- A^* search: branch and bound with a lower-bound estimation.
- Algorithm $A^*(N_0)$
 - Initialize a Priority Queue PQ to store partial paths with the key the cost of this path.
 - ▷ Initially, store only a path with the starting node N_0 only.
 - ▷ Paths in PQ are sorted according to their current cost plus a lower bound on the remaining distances.
 - While PQ is not empty do
 - ▷ Remove a path P with the **least cost** from PQ
 - ▷ 11: If the goal is found, then return success
 - ▷ 12: Find extended paths from P by extending one step
 - ▷ Insert all generated paths to PQ
 - ▷ Update PQ
 - ▷ If two paths reach a common node then keep only one with the **least cost**
 - Return fail

A* algorithm

■ Cost function:

- Given a path P ,
 - ▷ let $g(P)$ be the current cost of P ;
 - ▷ let $h(P)$ be the estimation of remaining, or **heuristic** cost of P ;
 - ▷ $f(P) = g(P) + h(P)$ is the cost function.
- How to find a good $h()$ is the key of an A* algorithm?
- It is known that if $h()$ never overestimates the actual cost to the goal (this is called **admissible**), then A* always finds an optimal solution.
 - ▷ Q: How to prove this?

■ Checking of the termination condition:

- We need check for a goal only when a path is popped from the PQ , i.e., at Line 11.
- We cannot check for a goal when a path is generated and inserted into the PQ , i.e., at Line 12.
 - ▷ We will not be able find the optimal solution if we do the checking at Line 12.

A* algorithm: Comments

- When a path is inserted, check for whether it has reached some nodes that have been visited before.
 - ▷ *It may take a huge space and a clever algorithm to implement an efficient Priority Queue.*
 - ▷ *It may need a clever data structure to efficiently check for possible duplications.*
- Cost function:
 - Need an lower bound estimation that is as large as possible.
 - Can design the cost function so that A* emulates the behavior of other search routines.
- It consumes a lot of memory to record the set of visited nodes.
- It also consume a lot of memory to store PQ.
- Q:
 - ▷ *What disk based techniques can be used?*
 - ▷ *Why do we need a non-trivial $h(P)$ that is admissible?*
 - ▷ *How to design an admissible cost function?*

DFS with a threshold

- $\text{DFS}_{\text{cost}}(N, f, \text{threshold})$ is a version of DFS with a starting state N and a cost function f that cuts off a path when its cost is more than a given *threshold*.
 - $\text{DFS}_{\text{depth}}(N, \text{cut_of_f_depth})$ is a special version of $\text{DFS}_{\text{cost}}(N, f, \text{threshold})$.
- **Algorithm $\text{DFS}_{\text{cost}}(N_0, f, \text{threshold})$**
 - Initialize a Stack S
 - Push (null, N_0) to S where N_0 is the initial state
 - While S is not empty do
 - ▷ Pop $(\text{current}, N)$ from S
 - ▷ If current is a goal, then return success **{* Goal is found! *}**
 - ▷ $R \leftarrow \text{next}(\text{current}, N)$ **{* pick a good move ordering here *}**
 - ▷ If $R = \text{null}$, then continue; **{* searched all children of N *}**
 - ▷ If $f(P) > \text{threshold}$, then continue where P is the current path **{* cut off *}**
 - ▷ Push (R, N) to S
 - ▷ If R is already in S , then continue **{* to avoid loops *}**
 - ▷ Push (null, R) to S **{* search deeper *}**
 - Return fail

How to pick a good move ordering (1/2)

- Instead of just using $next(current, N)$ to find the next unvisited neighbors of N with the information of the last visited node being $current$, we do the followings.
 - Use a routine to order the neighbors of N so that it is always the case the neighbors are visited from low cost to high cost.
 - Let this routine be $next1(current, N)$.
 - Note we still need dummy first and last elements being $null$.

How to pick a good move ordering (2/2)

- **Algorithm DFS1_{cost}($N_0, f, threshold$)**
 - Initialize a Stack S
 - Push $(null, N_0)$ to S where N_0 is the initial state
 - While S is not empty do
 - ▷ Pop $(current, N)$ from S
 - ▷ If $current$ is a goal, then return success
 - ▷ $R \leftarrow next1(current, N)$
 - ▷ If $R = null$, then continue; **{* searched all children of N *}**
 - ▷ Let P be the path from N_0 to R
 - ▷ If $f(P) > threshold$, then continue **{* cut off *}**
 - ▷ Push (R, N) to S
 - ▷ If R is already in S , then continue **{* to avoid loops *}**
 - ▷ Push $(null, R)$ to S **{* search deeper *}**
 - Return fail

How to incooperate ideas from A^*

- Instead of using a stack in DFS_{cost} , use a priority queue.
- Algorithm $\text{DFS2}_{cost}(N_0, f, threshold)$
 - Initialize a priority queue PQ
 - Insert $(null, N_0)$ to PQ where N_0 is the initial state
 - While PQ is not empty do
 - ▷ Remove $(current, N)$ with the least cost $f(P)$ for the path P from N_0 to N from PQ
 - ▷ If $current$ is a goal, then return success
 - ▷ $R \leftarrow next1(current, N)$
 - ▷ If $R = null$, then continue; **{* searched all children of N *}**
 - ▷ Let P be the path from N_0 to R
 - ▷ If $f(P) > threshold$, then continue **{* cut off *}**
 - ▷ Insert (R, N) to PQ
 - ▷ If R is already in PQ , then continue **{* to avoid loops *}**
 - ▷ Insert $(null, R)$ to PQ **{* search deeper *}**
 - Return fail
- It may be costly to maintain a priority queue as in the case of A^* .

$$\text{IDA}^* = \text{DFID} + \text{A}^*$$

- $\text{DFS}_{\text{cost}}(N, f, \text{threshold})$ is a version of DFS with a starting state N and a cost function f that cuts off a path when its cost is more than a given threshold .
- IDA^* : iterative-deepening A^*
- Algorithm $\text{IDA}^*(N_0, \text{threshold})$
 - $\text{threshold} \leftarrow h(\text{null})$
 - While threshold is reasonable do
 - ▷ $\text{DFS}_{\text{cost}}(N_0, g + h(), \text{threshold})$
{ Can use $\text{DFS1}_{\text{cost}}$ or $\text{DFS2}_{\text{cost}}$ here *}*
 - ▷ If the goal is found,
then return success
 - ▷ $\text{threshold} \leftarrow$ the least $g(P) + h(P)$ cost among all paths P being cut
 - Return fail

IDA*: comments

- IDA* does not need to use a priority queue as in the case of A*.
 - IDA* is optimal in terms of solution cost, time, and space over the class of admissible best-first searches on a tree.
- Issues in updating *threshold*.
 - Increase too little: re-search too often.
 - Increase too large: cut off too little.
 - Q: How to guarantee optimal solutions are not cut?
 - ▷ *It can be proved, as in the case of A*, that given an admissible cost function, IDA* will find an optimal solution, i.e., one with the least cost, if one exists.*
- Cost function is the knowledge used in searching.
- Combine knowledge and search!
- Need to balance the amount of time spent in realizing knowledge and the time used in searching.

15 puzzle (1/2)

■ Introduction of the game:

- 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 (2/2)

- **Total number of positions:** $16! = 20,922,789,888,000 \leq 2.1 * 10^{13}$.
 - It is feasible, in terms of computation time, to enumerate all possible positions, since 2007.
 - ▷ *Can use DFS or DFID now.*
 - ▷ *Need to avoid falling into loops or re-visit a node too many times.*
 - It is still too large to store all possible positions in main memory now (2012).
 - ▷ *Cannot use BFS efficiently even now.*
 - ▷ *Maybe difficult to find an optimal solution.*
 - ▷ *Maybe able to use disk based BFS.*

Solving 15 puzzles

- Using DEC 2060 a 1-MIPS machine: solved the 15 puzzle problem within 30 CPU minutes for all testing positions, generating over 1.5 million nodes per minute.
 - The average solution length was 53 moves.
 - The maximum was 66 moves.
 - IDA* generated more nodes than A*, but ran faster due to less overhead per node.
- Heuristics used:
 - $g(P)$: the number of moves made so far.
 - $h(P)$: the Manhattan distance between the current board and the goal position.
 - ▷ 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 position and a goal position is the sum of the Manhattan distance of every tile.
 - ▷ $h(P)$ is admissible.

What else can be done?

- **Bi-directional search and IDA*?**
 - How to design a good and non-trivial heuristic function?
- **How to get a better move ordering in DFS?**
- **Balancing in resource allocation:**
 - The efforts to memorize past results versus the amount of efforts to search again.
 - The efforts to compute a better heuristic, i.e., the cost function.
 - The amount of resources spent in implementing a better heuristic and the amount of resources spent in searching.
- **Can these techniques be applied to two-person games?**

References and further readings

- Judea Pearl. **Heuristics: Intelligent search strategies for computer problem solving.** Assison-Wesley, 1984.
- * R. E. Korf. **Depth-first iterative-deepening: An optimal admissible tree search.** *Artificial Intelligence*, 27:97–109, 1985.
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