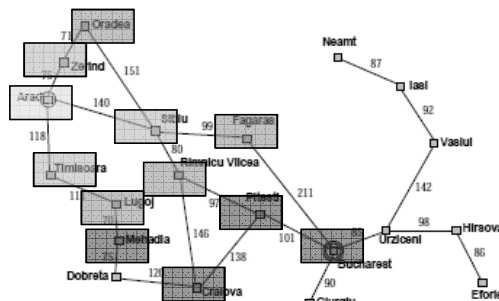


Iterative Deepening

- Problem with choosing depth bound; incomplete or admits poor solutions
- Iterative deepening a variation which is complete and find best solution
- Basic idea is:
 - do d.l.s. for depth $n = 0$; if solution found, return it;
 - otherwise do d.l.s. for depth $n = n + 1$; if solution found, return it, etc;
 - So we repeat d.l.s. for all depths until solution found.
- Useful if the search space is large and the maximum depth of the solution is not known.

Example: Romania Problem

D=0 D=1 D=2 D=3



General Algorithm for Iterative Deepening

```

depth limit = 0;
repeat
{ result = depth_limited_search
(max depth = depth limit;
agenda = initial node; );
if result contains goal then
return result;
depth limit = depth limit + 1;}
until false; /* i.e., forever */

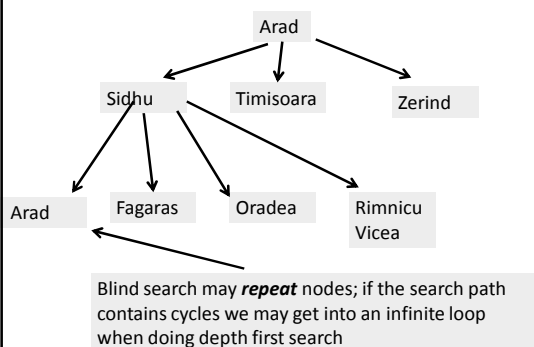
```

- Calls d.l.s. as subroutine.

IDS Properties

- Note that in iterative deepening, we re-generate nodes on the fly.
- Each time we do call on depth limited search for depth d , we need to regenerate the tree to depth $d - 1$.
- Trade off time for memory.
- In general we might take a little more time, but we save a lot of memory.
 - Example: Suppose $b = 10$ and $d = 5$.
 - Breadth first search would require examining 111, 111 nodes, with memory requirement of 100, 000 nodes.
 - Iterative deepening for same problem: 123, 456 nodes to be searched, with memory requirement only 50 nodes.
 - Takes 11% longer in this case, but savings on memory are immense

The Search Tree



Avoiding Repeated States

- There are three ways to deal with this (in order of increasing effectiveness and computational overhead):
 - do not return to the state you have just come from;
 - do not create paths with cycles in them;
 - do not generate any state that was ever generated before.
- Note there is a trade off between the cost of extra search and the cost of checking for repeated states.

Branching

- In analyses branching is often assumed to be uniform
- But in practice this is often not so
- This can make a big difference to the search space

Blocks World

Move any *clear* block to another *clear* block

Move any block to the table

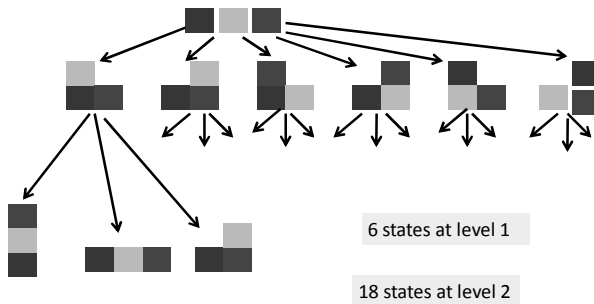
Initial



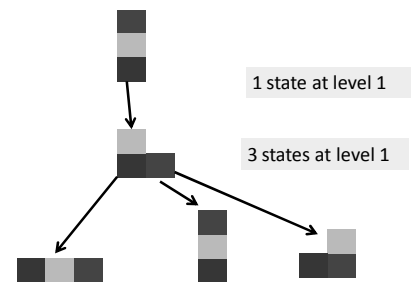
Goal



Search Space for Initial to Goal



Search Space for Goal to Initial



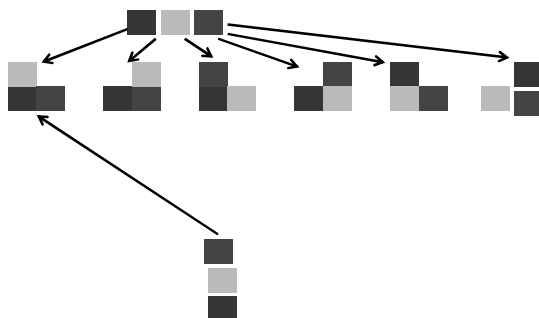
Goal vs Data driven search

- We can chose to search from initial state to the goal (data driven)
- Or from the goal to the initial state (goal driven)
- The branching may be very different which will affect the search
- Goal driven search is very often very much more efficient (few paths reach the goal)
- Often used in expert systems (and Prolog)

Bi directional search

- If we are unsure the searching from both ends may be the best

Bidirectional Search



Bi-directional Search: Good

- Much more efficient.
- Rather than doing one search of bd , we do *two* $bd/2$ searches.
 - Example:
 - Suppose $b = 10$, $d = 6$.
 - Breadth first search will examine $10^6 = 1,000,000$ nodes.
 - Bidirectional search will examine $2 \times 10^3 = 2,000$ nodes.
- Can combine different search strategies in different directions.

Bi-directional Search: Bad

- Must be able to generate predecessors of states.
- There must be an efficient way to check whether each new node appears in the other search.
- For large d , is still impractical!
- For two bi-directional breadth-first searches, with branching factor b and depth of the solution d we have memory requirement of $bd/2$ for each search.

Today

- More advanced problem solving techniques:
 - Depth limited search
 - Iterative deepening
 - Bi-directional search
 - Avoiding repeated states
- These improved on basic techniques like breadth-first and depth-first search.
- However, they still aren't always powerful enough to give solutions for realistic problems.
- Are there more improvements we can make?
- Next Time
 - Heuristic Search Techniques