

## Problem Spaces & Search

·CSE 573

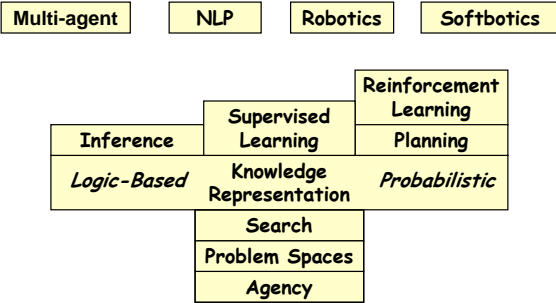
## Logistics

- Mailing list
- Reading
  - Ch 4.2, Ch 6
- Mini-Project I
  - Partners
- Game Playing?

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## 573 Topics



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## Weak Methods

- "In the knowledge lies the power..."
  - [Feigenbaum]
- But what if you have very little knowledge????

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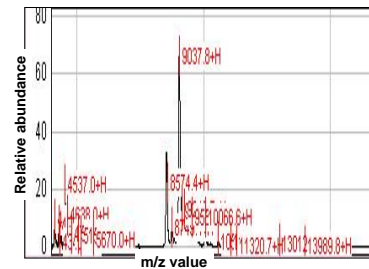
## Generate & Test

- As weak as it gets...
- Works on semi-decidable problems!

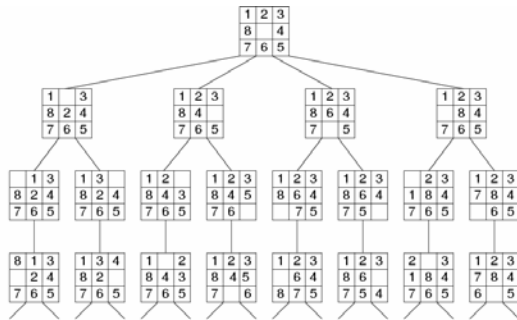
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## Mass Spectrometry



### Example: Fragment of 8-Puzzle Problem Space



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### Search thru a Problem Space / State Space

#### • Input:

- Set of states
- Operators [and costs]
- Start state
- Goal state [test]

#### • Output:

- Path: start  $\Rightarrow$  a state satisfying goal test
- [May require shortest path]

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### Example: Route Planning



#### • Input:

- Set of states
- Operators [and costs]
- Start state
- Goal state (test)

#### • Output:

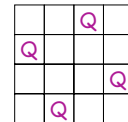
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### Example: N Queens

#### • Input:

- Set of states
- Operators [and costs]
- Start state
- Goal state (test)

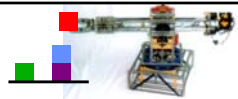


#### • Output

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### Ex: Blocks World



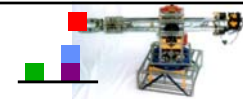
#### • Input:

- Set of states
  - Partially specified plans
- Operators [and costs]
  - Plan modification operators
- Start state
  - The null plan (no actions)
- Goal state (test)
  - A plan which provably achieves
  - The desired world configuration

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### Multiple Problem Spaces



#### • Real World

- States of the world (e.g. block configurations)
- Actions (take one world-state to another)

#### Robot's Head

##### • Problem Space 1

- PS states =
  - models of world states
- Operators =
  - models of actions

##### • Problem Space 2

- PS states =
  - partially spec. plan
- Operators =
  - plan modificat'n ops

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## Classifying Search

- **GUESSING ("Tree Search")**
  - Guess how to extend a partial solution to a problem.
  - Generates a tree of (partial) solutions.
  - The leaves of the tree are either "failures" or represent complete solutions
- **SIMPLIFYING ("Inference")**
  - Infer new, stronger constraints by combining one or more constraints (without any "guessing")
  - Example:
 
$$\begin{array}{l} X+2Y = 3 \\ X+Y = 1 \\ \text{therefore } Y = 2 \end{array}$$
- **WANDERING ("Markov chain")**
  - Perform a (biased) random walk through the space of (partial or total) solutions

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## • Guessing - State Space Search

1. BFS
2. DFS
3. Iterative Deepening
4. Bidirectional
5. Best-first search
6. A\*
7. Game tree
8. Davis-Putnam (logic)
9. Cutset conditioning (probability)

## • Simplification - Constraint Propagation

1. Forward Checking
2. Path Consistency (Waltz labeling, temporal algebra)
3. Resolution
4. "Bucket Algorithm"

## • Wandering - Randomized Search

1. Hillclimbing
2. Simulated annealing
3. Walksat
4. Monte-Carlo Methods

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## Search Strategies v2

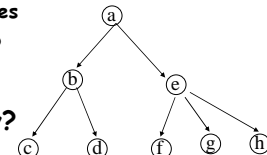
- **Blind Search**
  - Depth first search
  - Breadth first search
  - Iterative deepening search
  - Iterative broadening search
- **Informed Search**
- **Constraint Satisfaction**
- **Adversary Search**

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## Depth First Search

- **Maintain stack of nodes to visit**
- **Evaluation**
  - **Complete?**  
Not for infinite spaces
  - **Time Complexity?**  
 $O(b^d)$
  - **Space Complexity?**  
 $O(bd)$



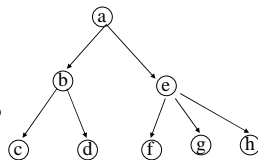
$d = \text{max depth of tree}$

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## Breadth First Search

- **Maintain queue of nodes to visit**
- **Evaluation**
  - **Complete?**  
Yes (if branching factor is finite)
  - **Time Complexity?**  
 $O(b^{d+1})$
  - **Space Complexity?**  
 $O(b^{d+1})$



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## Memory a Limitation?

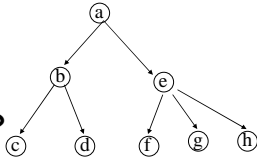
- **Suppose:**
  - 2 GHz CPU
  - 1 GB main memory
  - 100 instructions / expansion
  - 5 bytes / node
- 200,000 expansions / sec
- Memory filled in 100 sec ... < 2 minutes

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## Iterative Deepening Search

- DFS with limit; incrementally grow limit
- Evaluation
  - Complete? Yes
  - Time Complexity?  $O(b^d)$
  - Space Complexity?  $O(d)$



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## Cost of Iterative Deepening

b	ratio ID to DFS
2	3
3	2
5	1.5
10	1.2
25	1.08
100	1.02

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## Speed

Assuming 10M nodes/sec & sufficient memory

	BFS	Iter. Deep.
	Nodes	Nodes
• 8 Puzzle	$10^5$	$10^5$
• 2x2x2 Rubik's	$10^6$	$10^6$
• 15 Puzzle	$10^{13}$	$10^{17}$
• 3x3x3 Rubik's	$10^{19}$	$10^{20}$
• 24 Puzzle	$10^{25}$	$10^{37}$

Time: .01 sec, .2 sec, 6 days, 68k yrs, 12B yrs, .01 sec, .2 sec, 20k yrs, 574k yrs,  $10^{23}$  yrs

1Mx, 8x

Why the difference?  
Rubik has higher branch factor  
15 puzzle has greater depth

# of duplicates

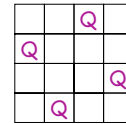
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Adapted from Richard Korf presentation

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## When to Use Iterative Deepening

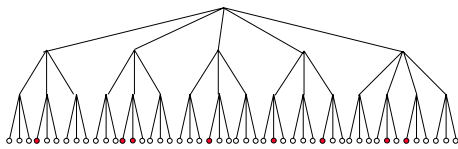
- N Queens?



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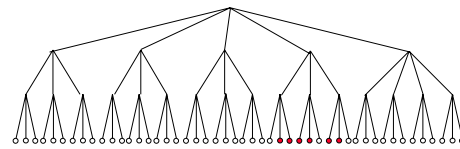
## Search Space with Uniform Structure



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## Search Space with Clustered Structure

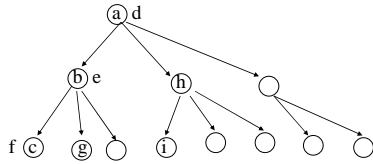


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## Iterative Broadening Search

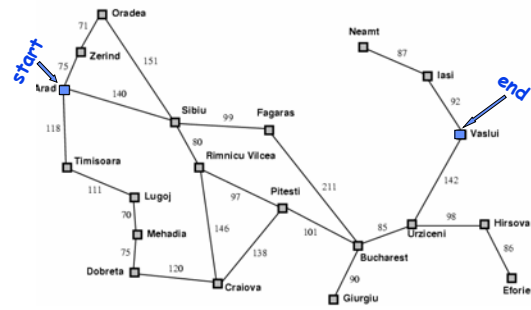
- What if know solutions lay at depth N?
- No sense in doing iterative *deepening*
  - Increment sum of sibling indicies



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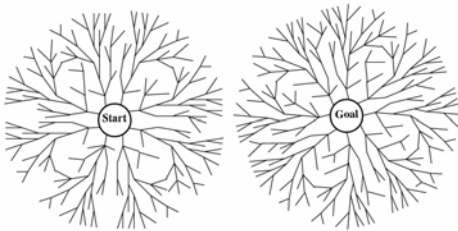
## Forwards vs. Backwards



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## vs. Bidirectional



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## Problem

- All these methods are slow (blind)
- Solution
  - add guidance ("heuristic estimate")
  - "informed search"

... coming next ...

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## Recap: Search thru a Problem Space / State Space

- Input:
  - Set of states
  - Operators [and costs]
  - Start state
  - Goal state [test]
- Output:
  - Path: start  $\Rightarrow$  a state satisfying goal test
  - [May require shortest path]

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## Cryptarithmic

- Input:
 

SEND
+ MORE
-----
MONEY

  - Set of states
  - Operators [and costs]
  - Start state
  - Goal state (test)
- Output:

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## Concept Learning

Labeled Training Examples

<p1,blond,32,mc,ok>  
 <p2,red,47,visa,ok>  
 <p3,blond,23,cash,ter>  
 <p4,...>

Output:  $f: \langle pn... \rangle \rightarrow \{ok, ter\}$

- **Input:**
  - Set of states
  - Operators [and costs]
  - Start state
  - Goal state (test)
- **Output:**

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## Symbolic Integration

• E.g.  $\int x^2 e^x dx = e^x(x^2 - 2x + 2) + C$

•

**Operators:**

Integration by parts  
 Integration by substitution  
 ...

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## Search Strategies v2

- **Blind Search**
- **Informed Search**
  - Best-first search
  - A\* search
  - Iterative deepening A\* search
  - Local search
- **Constraint Satisfaction**
- **Adversary Search**

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## Best-first Search

- Generalization of breadth first search
- Priority queue of nodes to be explored
- Cost function  $f(n)$  applied to each node

Add initial state to priority queue  
 While queue not empty  
   Node = head(queue)  
   If goal?(node) then return node  
   Add children of node to queue

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## Old Friends

- **Breadth first = best first**
  - With  $f(n) = \text{depth}(n)$
- **Dijkstra's Algorithm = best first**
  - With  $f(n) = g(n)$ , i.e. the sum of edge costs from start to n
  - Space bound (stores all generated nodes)

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## A\* Search

- **Hart, Nilsson & Rafael 1968**
  - Best first search with  $f(n) = g(n) + h(n)$
  - Where  $g(n) =$  sum of costs from start to n
  - $h(n) =$  estimate of lowest cost path  $n \rightarrow$  goal  
 $h(\text{goal}) = 0$
  - If  $h(n)$  is **admissible** (and **monotonic**) then A\* is optimal

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