



CSCE 580: Introduction to Al

CSCE 581: Trusted Al

Lecture 8: Search – Informed, Heuristics

PROF. BIPLAV SRIVASTAVA, AI INSTITUTE 19<sup>TH</sup> SEP 2023

Carolinian Creed: "I will practice personal and academic integrity."

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#### Organization of Lecture 8

- Introduction Segment
  - Recap of Lecture 7
  - Review of Quiz1
- Main Segment
  - Informed Search
  - Heuristics and Properties
  - Designing Heuristics
- Concluding Segment
  - Course Project Discussion
  - About Next Lecture Lecture 8
  - Ask me anything

#### Introduction Section

#### Announcement: Modified Office Timing

Instructor's new timing for office hour - available at Room 515 of the AI Institute, 1112 Greene St (5th Floor):

- \* 1-2pm (Mondays)
- \* 3-4pm (Tuesdays)

If this does not work, we can schedule a separate mutually convenient time.

#### Recap of Lecture 7

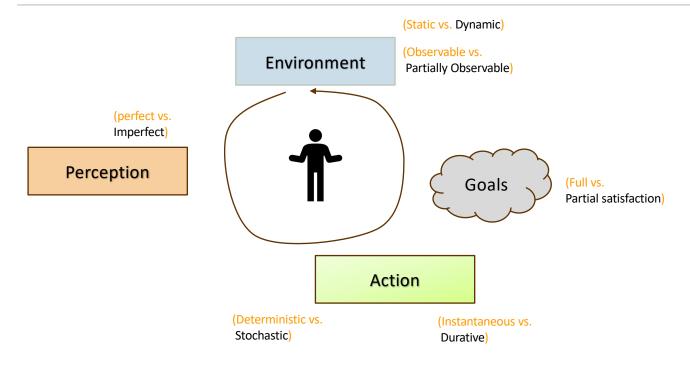
- Goal-directed problem solving agents
- How to formulate problem formulations
- Search concepts
  - Problems of controlled robot navigation, 8-tile, N-queens
- Search strategies
- Quiz 1

#### Review of Quiz 1

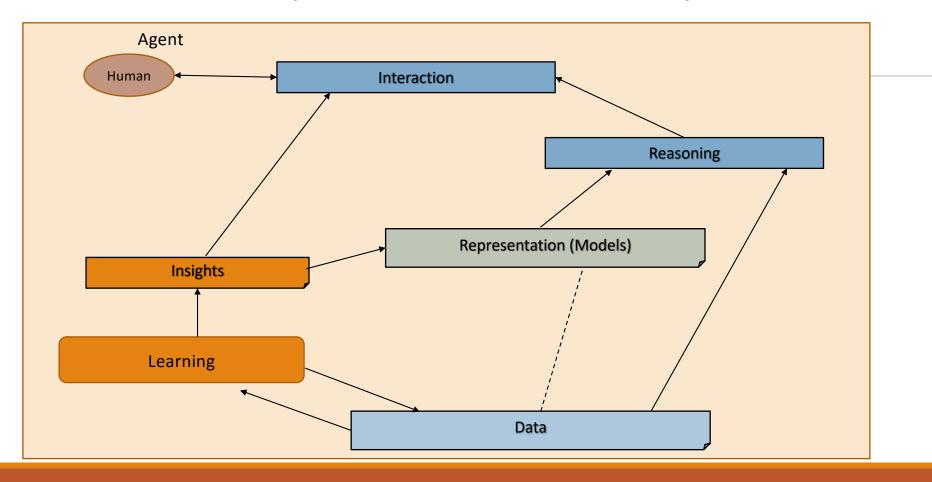
# States Initial state Goal state Actions Transition model Action cost

• Transition model: tells what it means to take actions -- how to move from one state to another. So, you will have to come up with a state representation, then decide on actions, and then define function(s) to state what it means to take actions.

## Intelligent Agent Model



#### Relationship Between Main Al Topics



## Where We Are in the Course

#### CSCE 580/581 - In This Course

- Week 1: Introduction, Aim: Chatbot / Intelligence Agent
- Weeks 2-3: Data: Formats, Representation and the Trust Problem
- Week 4-5: Search, Heuristics Decision Making
- Week 6: Constraints, Optimization Decision Making
- Week 7: Classical Machine Learning Decision Making, Explanation
- Week 8: Machine Learning Classification
- Week 9: Machine Learning Classification Trust Issues and

#### Mitigation Methods

- Topic 10: Learning neural network, deep learning, Adversarial attacks
- Week 11: Large Language Models Representation, Issues
- Topic 12: Markov Decision Processes, Hidden Markov models Decision making
- Topic 13: Planning, Reinforcement Learning Sequential decision making
- Week 14: <u>AI for Real World: Tools, Emerging Standards and Laws;</u>
   Safe AI/ Chatbots

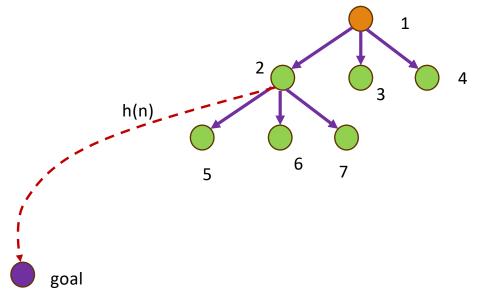
#### Main Section

## Informed Search – Greedy best-first

Uses domain/problem specific hints to guide search

$$f(n) = h(n)$$

- f: <u>estimated</u> cost of best path via n to goal
- •h: <u>estimated</u> cost to goal from n // h is also called heuristic function



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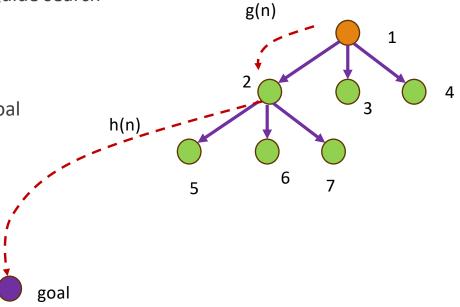
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#### Informed Search – A\* search

Uses domain/problem specific hints to guide search

$$f(n) = g(n) + h(n)$$

- f: estimated cost of best path via n to goal
- g: cost of best path to n
- h: estimated cost to goal from n

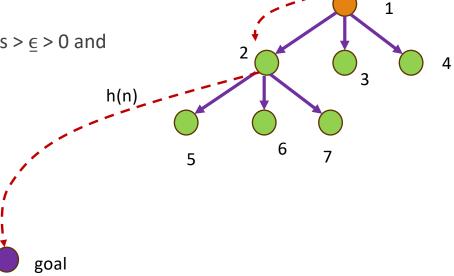


#### Properties of A\* search

f(n) = g(n) + h(n)

• A\* is complete, assuming actions costs  $> \le > 0$  and state space has solution or is finite

 h(n) is admissible, i.e., never overestimates true cost to reach goal

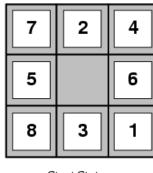


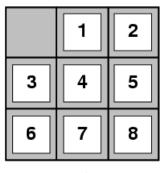
g(n)

#### Finding Heuristics Function

- h1: number of misplaced tiles (excluding blank)
  - H1(start) = 8
- h2: sum of the distance of tiles from goal (excluding blank)
  - h2(start) = (3 + 1 + 2) + (2 + 3) + (2 + 2 + 3)







Start State

Goal State

• True cost: 26

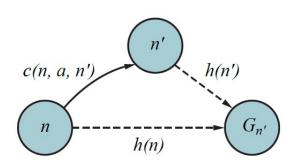
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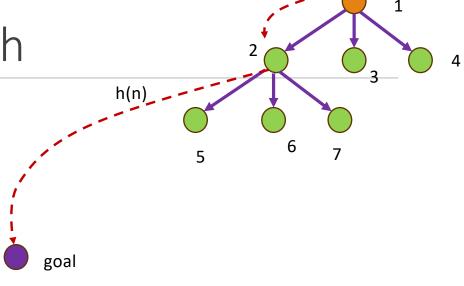
Russell & Norvig, Al: A Modern Approach

## Properties of A\* search

$$f(n) = g(n) + h(n)$$

 A heuristic is consistent if h(n) <= c(n, a, n') + h(n')</li>





g(n)

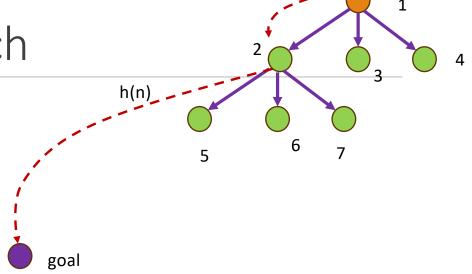
Adapted from:

Russell & Norvig, AI: A Modern Approach

#### Properties of A\* search

$$f(n) = g(n) + h(n)$$

- A\* with consistent heuristic is optimally efficient
- A\*
  - Any algo using search path and same heuristics at A\* will atleast expand these nodes
  - Prunes (removes) search nodes that are not necessary for finding optimal solution



g(n)

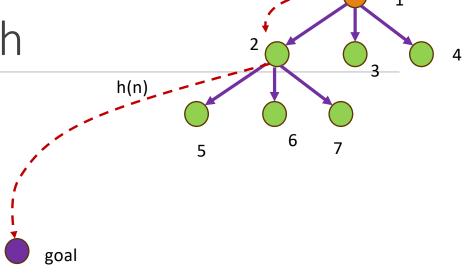
Adapted from:

Russell & Norvig, AI: A Modern Approach

## Type: Satisficing Search

$$f(n) = g(n) + W * h(n)$$

 If heuristic is inadmissible, A\* may find just any solution



g(n)

Adapted from:

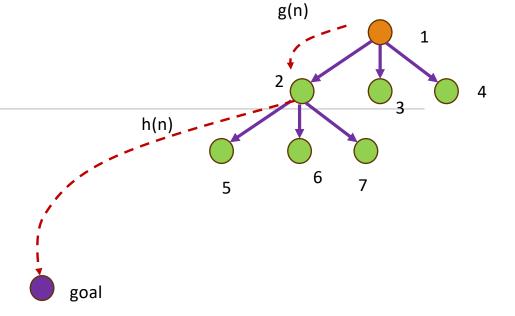
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#### Type: Beam Search

$$f(n) = g(n) + h(n)$$

- Keep only k (a parameter) nodes with the best f-score in frontier
- Incomplete and sub-optimal, but space efficient



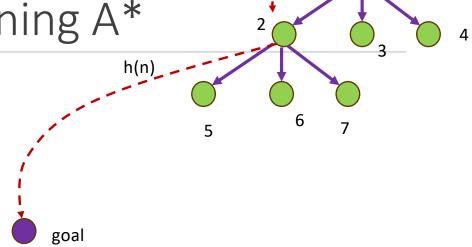
Adapted from:

Russell & Norvig, AI: A Modern Approach

## Type: Iterative Deepening A\*

$$f(n) = g(n) + h(n)$$

- Similar to Iterative Deepening Depth search, but for f-score. Optimizes memory usage.
- In each iteration, search until find a node with f-score exceeding threshold; use the node's f-score as the new threshold
- Iterative search takes more time than A\*



g(n)

Adapted from:

Russell & Norvig, AI: A Modern Approach

## Informed Search Types

A* search	f(n) = g(n) + h(n)	(W = 1)
Uniform-cost search	f(n) = g(n)	(W = 0)
Greedy best-first search	f(n) = h(n)	(W = infinite)
Weighted A* search	f(n) = g(n) + W * h(n)	(1 < W < infinite)

#### Notes:

Uniform-cost => uninformed Weighted A\* => satisficing

#### Impact of Heuristics Function

d	Search Cost (nodes generated)			Effective Branching Factor		
a	BFS	$A^*(h_1)$	$A^*(h_2)$	BFS	$A^*(h_1)$	$A^*(h_2)$
6 8 10 12 14 16 18 20 22 24 26 28	128	24	19	2.01	1.42	1.34
8	368	48	31	1.91	1.40	1.30
10	1033	116	48	1.85	1.43	1.27
12	2672	279	84	1.80	1.45	1.28
14	6783	678	174	1.77	1.47	1.31
16	17270	1683	364	1.74	1.48	1.32
18	41558	4102	751	1.72	1.49	1.34
20	91493	9905	1318	1.69	1.50	1.34
22	175921	22955	2548	1.66	1.50	1.34
24	290082	53039	5733	1.62	1.50	1.36
26	395355	110372	10080	1.58	1.50	1.35
28	463234	202565	22055	1.53	1.49	1.36

**Figure 3.26** Comparison of the search costs and effective branching factors for 8-puzzle problems using breadth-first search,  $A^*$  with  $h_1$  (misplaced tiles), and  $A^*$  with  $h_2$  (Manhattan distance). Data are averaged over 100 puzzles for each solution length d from 6 to 28.

Reduces effective Bbranching factor!

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#### Choosing From a Choice of (Admissible) Heuristics

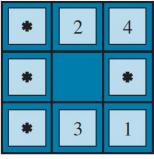
- Choose dominating heuristics
  - For n, h2(n) >= h1(n)
- If not dominating, choose maximum
  - $h(n) = max \{h1(n), h2(n), ..., h_k(n)\}$

#### Creating Heuristics Automatically

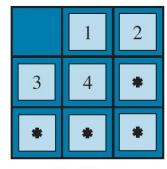
- From relaxed problems
  - Formulate a relaxed problem
  - Solve relaxed problem
  - Use solution length as heuristics for original problem (Relaxed problem heuristics)
- From sub-problems
  - Formulate a sub-problem
  - Solve relaxed sub-problem
  - Store solution of sub-problem
  - Compute admissible heuristic h\_DB for each node by looking up sub-problem and its solution cost

(Pattern databases)

- Learn heuristics
  - From data: past solutions, relaxed problems, ...
  - Predict heuristic value







Goal State

Adapted from:

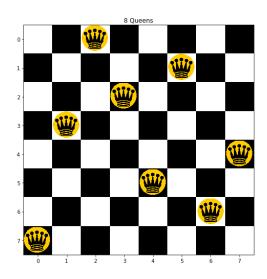
Russell & Norvig, AI: A Modern Approach

## Coding Example

- 8-Puzzle code notebook
  - <a href="https://github.com/biplav-s/course-ai-tai-f23/blob/main/sample-code/Class6-To-Class9-search.md">https://github.com/biplav-s/course-ai-tai-f23/blob/main/sample-code/Class6-To-Class9-search.md</a>

#### Discussion: Relaxed Problems

- For N-Queens
- Pancake problem
- Many more https://www.movingai.com/SAS/index.html



Adapted from:

Russell & Norvig, AI: A Modern Approach

#### Discussion: Rubic's Cube

- Search and deep-learning
  - Demo video: Solving with search and distance-based heuristics <a href="https://youtu.be/YQZ2sj-x5js">https://youtu.be/YQZ2sj-x5js</a>
  - Live demo: Solving with A\*search and deep learning-based heuristics (DeepCube-A) https://deepcube.igb.uci.edu/

#### **Uninformed** Search Strategies

Search strategies use only the information available in the problem definition. They do not use a measure of distance to goal (*uninformed*).

- Breadth-first search
- Uniform-cost search
- Depth-first search
- Depth-limited search
- Iterative deepening search
- Bidirectional search

**Consideration:** type of queue used for the fringe of the search tree (collection of tree nodes that have been generated but not yet expanded)

#### Adapted from:

- 1. Russell & Norvig, AI: A Modern Approach
- 2. Bart Selman's CS 4700 Course

#### Informed Search – A\* search

- Best-first
- A\*
- Weighted A\*
- Beam search [Incomplete]
- •Iterative-deepening A\* [Incomplete]

A* search	f(n) = g(n) + h(n)	(W = 1)
Uniform-cost search	f(n) = g(n)	(W = 0)
Greedy best-first search	f(n) = h(n)	(W = infinite)
Weighted A* search	f(n) = g(n) + W * h(n)	(1 < W < infinite)

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## Course Project

#### Project Discussion: What Problem Fascinates You?

- Data
  - Water
  - Finance
  - •
- Analytics
  - Search, Optimization, Learning, Planning, ...
- Application
  - Building chatbot
- Users
  - Diverse demographics
  - Diverse abilities
  - Multiple human languages

#### **Project execution in sprints**

- Sprint 1: (Sep 12 Oct 5)
  - Solving: Choose a decision problem, identify data, work on solution methods
  - Human interaction: Develop a basic chatbot (no AI), no problem focus
- Sprint 2: (Oct 10 Nov 9)
  - Solving: Evaluate your solution on problem
  - Human interaction: Integrated your choice of chatbot (rule-based or learning-based) and methods
- Sprint 3: (Nov 14 30)
  - Evaluation: Comparison of your solver chatbot with an LLMbased alternative, like ChatGPT

#### Project Discussion: Dates and Deliverables

#### Project execution in sprints

- Sprint 1: (Sep 12 Oct 5)
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- Sprint 3: (Nov 14 30)
  - Evaluation: Comparison of your solver chatbot with an LLMbased alternative, like ChatGPT

- Oct 12, 2023
  - Project checkpoint
  - In-class presentation
- Nov 30, 2023
  - Project report due
- Dec 5 / 7, 2023
- In-class presentation

#### Skeleton: A Basic Chatbot

- Run in an infinite loop until the user wants to quit
- Handle any user response
  - User can quit by typing "Quit" or "quit" or just "q"
  - User can enter any other text and the program has to handle it. The program should write back what the user entered and say – "I do not know this information".
- Handle known user query types // Depends on your project
  - "Tell me about N-queens", "What is N?"
  - "Solve for N=4?"
  - "Why is this a solution?"
- Handle <u>chitchat</u> // Support at least 5, extensible from a file
  - "Hi" => "Hello"
  - ...
- Store session details in a file

#### **Illustrative Project**

- **1. Title**: Solve and explain solving of n-queens puzzle
- **2. Key idea**: Show students how a course project will look like
- 3. Who will care when done: students of the course, prospective Al students and teachers
- **4. Data need**: n: the size of game; interaction
- **5. Methods**: search
- **6. Evaluation**: correctness of solution, quality of explanation, appropriateness of chat
- **7. Users**: with and without Al background; with and without chess background
- 8. Trust issue: user may not believe in the solution, may find interaction offensive (why queens, not kings? ...)

#### Project Discussion: Illustration

- Create a private Github repository called "CSCE58x-Fall2023-<studentname>-Repo". Share with Instructor (biplav-s) and TA (kausik-l)
- Create Google folder called "CSCE58x-Fall2023-<studentname>-SharedInfo". Share with Instructor (prof.biplav@gmail.com) and TA (lakkarajukausik90@gmail.com)
- 3. Create a Google doc in your Google repo called "Project Plan" and have the following by next class (Sep 5, 2023)

- 1. Title: Solve and explain solving of n-queens puzzle
- 2. Key idea: Show students how a course project will look like
- **3.** Who will care when done: students of the course, prospective AI students and teachers
- **4. Data need**: n: the size of game; interaction
- 5. Methods: search
- **6. Evaluation**: correctness of solution, quality of explanation, appropriateness of chat
- **7. Users**: with and without AI background; with and without chess background
- **8. Trust issue**: user may not believe in the solution, may find interaction offensive (why queens, not kings? ...)

#### Project Illustration: N-Queens

- •Sprint 1: (Sep 12 Oct 5)
  - Solving: Choose a decision problem, identify data, work on solution methods
    - Method 1: Random solution
    - Method 2: Search BFS
    - Method 3: Search ...
  - Human interaction: Develop a basic chatbot (no AI) as outlined
  - Deliverable
    - Code structure in Github
      - ./data
      - ./code
      - ./docs
      - ./test
    - Presentation: Make sprint presentation on Oct 12, 2023

#### Reference: Project Rubric

- Project results 60%
  - Working system ? 30%
  - Evaluation with results superior to baseline? 20%
  - Considered related work? 10%
- Project efforts 40%
  - Project report 20%
  - Project presentation (updates, final) 20%
- Bonus
  - Challenge level of problem 10%
  - Instructor discretion 10%
- Penalty
  - Lack of timeliness as per announced policy (right) up to 30%

#### Milestones and Penalties

- •Oct 12, 2023
  - Project checkpoint
  - In-class presentation
  - Penalty: presentation not ready by Oct 10, 2023 [-10%]
- Nov 30, 2023
  - Project report due
  - Project report not ready by date [-10%]
- Dec 5 / 7, 2023
  - In-class presentation
  - Project presentations not ready by Dec 4, 2023 [-10%]

#### Lecture 8: Summary

- We talked about
  - Informed Search
  - Heuristics and Properties
  - Designing Heuristics

## **Concluding Section**

#### About Next Lecture – Lecture 9

#### Lecture 9: Local Search

- Searching in large spaces
  - Hill climbing
  - Simulated Annealing
  - Genetic programming

Class 10: Adversarial games and search