



CSCE 580: Introduction to Al

CSCE 581: Trusted Al

Lecture 23 & 24: Making Decisions – Simple and Complex

PROF. BIPLAV SRIVASTAVA, AI INSTITUTE 14<sup>TH</sup> NOV & 16<sup>TH</sup> NOV, 2023

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

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### Organization of Lectures 23 & 24

- Introduction Segment
  - Recap of Lectures 21 and 22
- Main Segment
  - Making Decisions
  - Making simple decisions Maximum Expected Utility (MEU)
  - Making complex decisions Markov Decision Processes (MDPs)
- Concluding Segment
  - Course Project Discussion
  - Quiz 4
  - About Next Lecture Lecture 25
  - Ask me anything

#### Introduction Section

### Recap of Lecture 21 and 22

- Topic discussed
  - Text Processing
  - Language Models (LMs)
  - Learning for LMs with NN
  - Large LMs

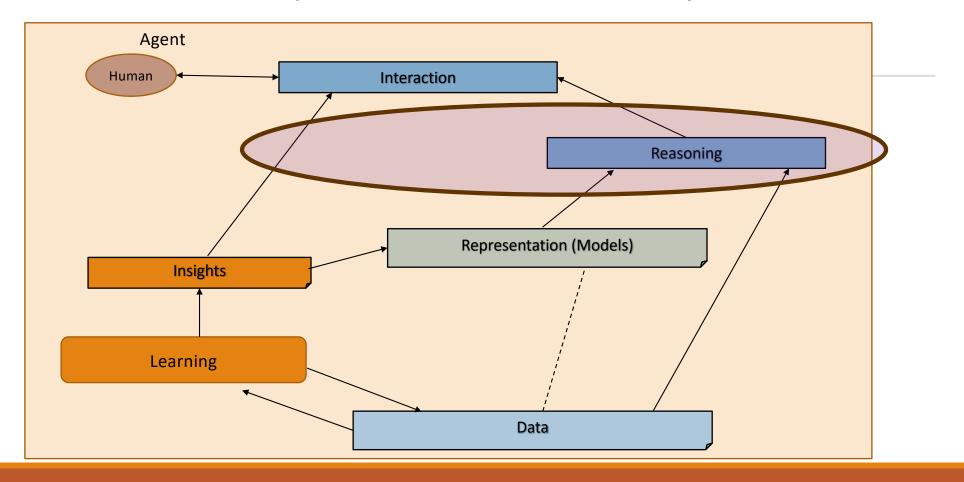
### Graduate Paper Presentation

- Papers between 2021-2023 (last 3 years)
- At top AI venues: AAAI, Neurips, IJCAI, ICML, ICLR, or discuss with instructor
- Guideline on presentation
  - Summary of the paper
  - Critique (+ves/ -ves)
  - Relevance to your and anyone else's project in the class

### 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: Al for Real World: Tools, Emerging Standards and Laws;
   Safe Al/ Chatbots

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#### Main Section

**Credit**: Retrieved from internet

# Making Decisions

#### Real World Decisions

Decision situation: going to airport from home

- Actions:
  - Take own car
  - Take a cab/ limo
  - Take a ride-share
  - Take a bus
  - Hitch-hike
  - Walk

# Students at a College Campus

#### An ideal solution should be:

- free of any errors (Ex: grammatical, calculation, etc.)
- utilize all the information given by the user completely and give a reasonable, practical, and optimal solution.

#### **Example Query:**

I am making a purchase of \$1000 using my credit card. I have a due of \$2000 on my account. My total credit line is \$2,800. Would you recommend I make the purchase now or later in the future?

#### **Ideal Solution:**

Based on the information you have provided, it is not advisable to make the purchase now as you already have a due of \$2000 on your account, which is close to your total credit line of \$2,800. This means you are utilizing a significant portion of your available credit, and adding another \$1000 to your balance would further increase your credit utilization ratio (CUR), which can negatively impact your credit score.

#### The Quality of Everyday Decisions



Major variability due to:

- Emotions
- Biases
- Increasing data volume
- Cognitive ability to process
  - · Decreases under stress and constraints
  - Decreases with age\*

**Source**: <a href="https://www.umassd.edu/fycm/decision-making/process/">https://www.umassd.edu/fycm/decision-making/process/</a>

<sup>\*</sup> Source: A Review of Decision-Making Processes: Weighing the Risks and Benefits of Aging, Mara Mather, https://www.ncbi.nlm.nih.gov/books/NBK83778/

# Evidence #1: Poor Medical Adherence

#### Taking medicines

- 20 -30 % of medication prescriptions are never filled
- ~50 % of medications for chronic disease are not taken as prescribed

#### **Impact**

- causes 125,000 deaths, at least 10 percent of hospitalizations
- <u>Costs the American health care system</u> between \$100 billion and \$289 billion a year.

# Finding relevant guidance is hard, one reason for non-adherence and high costs in health

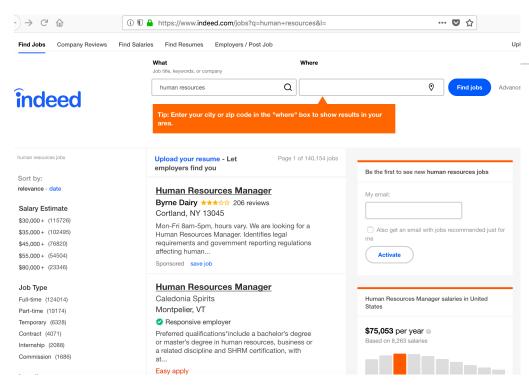
#### Sources:

- Medication Nonadherence, A Diagnosable and Treatable Medical Condition, Zachary A. Marcum, Mary Ann Sevick, Steven M. Handler, <a href="https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3976600/">https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3976600/</a>, 2013.
- https://www.nytimes.com/2017/04/17/well/the-cost-of-not-taking-yourmedicine.html

Example: Hard to understand medicine's information



# Evidence #2: Matching Demand to Supply of Jobs is Inadequate Demand-Supply Gap in Jobs Market [1] and Yet, Low Work Satisfaction/ Engagement [2]



Job search at a portal

- Finding jobs was generally hard around the world (Dec 2019), except for in tight labor markets like US (3.5% unemployment)
- Workforce satisfaction/ engagement was generally low around the world people did not find jobs they were match for [1,2]
- COVID-19 impact [3]:
  - Nearly half of global workforce at risk of losing livelihoods in informal sector
  - 9-12% job loss in the formal sector around the world
  - 14.7% unemployment in US by end of April 2020 [4]
  - Source: Global Skills Trends, Training Needs and Lifelong Learning Strategies for the Future of Work, ILO & OECD Report 2018,
    - http://www.g20.utoronto.ca/2018/g20 global skills trends and III oecd-ilo.pdf
  - Source: For 2016, job satisfaction: US 32%, Global 13%, https://www.gallup.com/workplace/236495/worldwide-employee-engagement-crisis asny
  - https://www.ilo.org/global/about-the-ilo/newsroom/news/WCMS\_743036/lang-en/index.htm
  - 4. https://www.bls.gov/news.release/empsit.nr0.htm

### Decision Imperative: Corona Virus Pandemic

#### Emerging Scenario Around the World\*

- Millions of cases, hundreds of thousands of deaths
- Businesses disrupted, millions going out of business
- Millions loosing jobs
- \* Numbers changing continuously; see reference for details

#### Decisions Need to be Made

- About disease
  - Understand disease
  - Tackle disease
- Understand impact to society: economy, supply chain
- Advise on actions to take
  - Individual
  - Group
  - Societal policy

**Resource**: <a href="https://github.com/biplav-s/covid19-info/wiki/Important-Information-About-COVID19">https://github.com/biplav-s/covid19-info/wiki/Important-Information-About-COVID19</a>

Emerging Landscape: The Problem

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### Before and After: (AI) Decision Support

**Today's tools**: Static, non-interactive, non-contextual, lack explanations

**Future tools**: Dynamic to data, interactive, contextual, explaining with data, anywhere, multi-modal, social (group dependency), societally relevant, ...

Future has potential to improve people's lives, promote well-being and reduce waste

## Simple Decisions

### Setting for a Decision

- An agent has a set of actions available,  $A = \{a_i\}$  and is in a state s
- There may be an uncertainty about current state. So, the agent assigns a probability to current state P(s) to each possible current state.
- When an action is taken, there may be uncertainty about outcome. So, resulting state is:  $P(s' \mid s, a)$
- The probability of reaching state s' after executing a in the current state is:  $P(RESULT(a) = s') = \Sigma_s P(s) P(s' | s, a)$

**Note**: P(RESULT(a) = s') requires perception, learning, knowledge representation and inference

Adapted from:

Russell & Norvig, Al: A Modern Approach

### Making a Simple Decision

- Choose best action based on the desirability of immediate outcome
- Have a utility function U(s) expressing desirability of a state (s)
- Expected utility of an action given the evidence, EU(a), is the average utility value of the outcome, weighted by the probability of that outcome.

$$EU(a) = \Sigma_{s'} P(RESULT(a) = s') U(s')$$

 Principle of maximum expected utility (MEU): rational agent chooses an action which maximizes its maximum expected utility action = argmax<sub>a</sub> EU(a) **Decision situation**: going to airport from home

- Actions:
  - Take own car
- Take a cab/ limo
- Take a ride-share
- Take a bus
- Hitch-hike
- Walk

Adapted from:

Russell & Norvig, AI: A Modern Approach

### Utility Functions: Modeling Preferences

#### Notations

- A > B: agent (decision maker) prefers A over B
- A ~ B: agent (decision maker) is indifferent between A and B
- A ≥ B: agent (decision maker) prefers A over B oris indifferent between A and B

#### Convention

- Outcome of an action is a lottery:  $L = [p_1, S_1; p_2, S_2; ...; p_n, S_n]$
- Utility function U
  - U(A) > U(B), if and only if, A > B
  - U(A) = U(B), if and only if, A ~ B

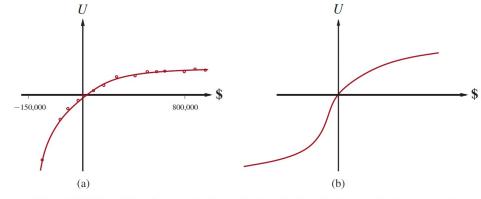
### Example: Choosing a Winning

- Won a game and have to choose
  - Choice 1: Take \$1M
  - Choice 2: Toss coin; Heads => \$2.5 M, Tails => 0
- Expected Monetary Value (EMV)
  - Choice 1: \$1M
  - Choice 2: ½ . \$2.5M + ½ . 0 = \$1.25M
- Expected Utility depends on current money

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## Example: Choosing a Winning

- Won a game and have to choose
  - Choice 1: Take \$1M
  - Choice 2: Toss coin; Heads => \$2.5 M, Tails => 0
- Expected Utility depends on current money(k)
  - EU(Accept) =  $\frac{1}{2}$  U(S<sub>k</sub>) +  $\frac{1}{2}$  U(S<sub>k+\$2.5M</sub>)
  - EU(Decline) = U(S<sub>k+\$1M</sub>)



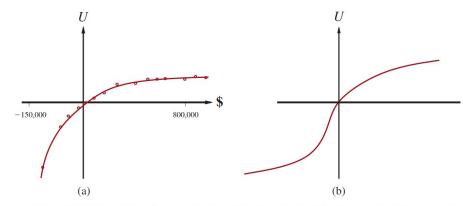
**Figure 16.2** The utility of money. (a) Empirical data for Mr. Beard over a limited range. (b) A typical curve for the full range.

Adapted from/ image credit: Russell & Norvig, AI: A Modern Approach

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### Example: S-Curve, Risk

- S-Curve: Fig 16.2(b)
- utility of a lottery is less than a sure thing
  - U(Lottery) < U(SureThing<sub>EMV(L)</sub>)
  - Risk averse agents: prefer sure payoff than expected monetary value of a gamble
  - Risk seeking agents: (people already in debt)
  - Certainty equivalent of lottery: agent will accept in lieu of a lottery
- According to studies, people will accept \$400 (approx.) in lieu of a gamble than gives \$1,000 half the time and \$0 other
- Insurance premium: difference between EMV of a lottery and its certainty equivalent
  - Risk aversion / positive insurance premium is the basis of insurance industry



**Figure 16.2** The utility of money. (a) Empirical data for Mr. Beard over a limited range. (b) A typical curve for the full range.

Adapted from/ image credit: Russell & Norvig, AI: A Modern Approach

#### Humans STILL Do Now Always Follow Utility Theory

- •Subjects in this experiment are given a choice between lotteries A and B:
  - Comparison scenario 1
    - A: 80% chance of \$4000
    - B: 100% chance of \$3000
  - Comparison scenario 2
    - C: 20% chance of \$4000
    - D: 25% chance of \$3000
- The majority of survey respondents choose B over A and C over D.
  - Comparison scenario 1:
    - A: 0.8 \* 4000 + 0.2 \* 0 = **3200**
    - B: 3000
  - Comparison scenario 2:
    - C: 0. 2\* 4000 + 0.8 \* 0 = **800**
    - D: 0.25 \* 3000 + 0.75 \* 0 = 750

Consistent utility: A over B and C over D.

Source: Russell & Norvig, AI: A Modern Approach

Tversky and Kahneman (1982) experiment

### Multi-Attribute/ Objective Optimization

Decision situation: going to airport from home

- Actions:
  - Take own car
  - Take a cab/ limo
  - Take a ride-share
  - Take a bus
  - Hitch-hike
  - Walk

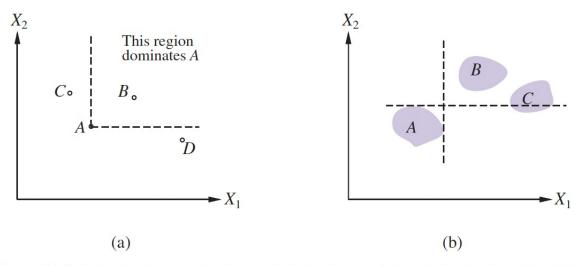
**Attributes**: cost, time, comfort, certainty of arrival time, ...

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### Choosing By Decision Dominance

Two attribute case shown

Choose by dominance



**Figure 16.4** Strict dominance. (a) Deterministic: Option A is strictly dominated by B but not by C or D. (b) Uncertain: A is strictly dominated by B but not by C.

Adapted from/ image credit: Russell & Norvig, AI: A Modern Approach

#### Choosing by Formal Verification of Correctness

Table 1: Different product interaction categories considered, query identifiers, queries posed under each category, variables used in each query with their corresponding chosen values and constraints to consider while answering the user queries.

Product Interac- tions	Query Iden- tifier	Queries	Variables with their values	Constraints
СС	Q1	I am making a purchase of \$1000 using my credit card. My billing cycle is from March 25th to April 24th. Today is March 31st, and I have a due of \$2000 on my account. My total credit line is \$2,800. Would you recommend I make the purchase now or later in the future?	$x_{PA} = 1000, x_{BC} = (March 25th - April 24th),$ $x_{DA} = 2000, x_{CL} = 2800$	$x_{DA} + x_{PA} < x_{CL}$
	Q2	I am making a <b>purchase of \$1000</b> using my credit card. My <b>billing cycle is from March 25th to April 24th.</b> Today is March 31st, and I have a <b>due of \$2000</b> on  my account. My total <b>credit line is \$3,800</b> . Would you recommend I make the  purchase now or later in the future?	$x_{PA} = 1000, x_{BC} = \text{(March 25th - April 24th)},$ $x_{DA} = 2000, x_{CL} = 3800$	
	Q3	I get 5% cashback if I buy furniture using my credit card. I am buying furniture worth \$1000 using my credit card. My billing cycle is from March 25th to April 24th. Today is March 31st, and I have a due of \$2000 on my account. My total credit line is \$2,800. Would you recommend I make the purchase now or later in the future?	$x_{CP} = 5\%$ , $x_{PA} = 1000$ , $x_{BC} = (March 25th - April 24th)$ , $x_{DA} = 2000$ , $x_{CL} = 2800$	
	Q4	I get 5% cashback if I buy furniture using my credit card. I am buying furniture worth \$1000 using my credit card. My billing cycle is from March 25th to April 24th. Today is March 31st, and I have a due of \$2000 on my account. My total credit line is \$3,800. Would you recommend I make the purchase now or later in the future?	$x_{CP} = 5\%, x_{PA} = 1000, x_{BC} = (March 25th - April 24th), x_{DA} = 2000, x_{CL} = 3800$	
CC (AAVE)	Q5	I be makin' a <b>purchase of \$1000</b> usin' i's credit card. I's <b>billin' cycle be from march 25th to april 24th</b> . Today be march 31ts, and i done a <b>due of \$2000</b> on i's account. I's total <b>credit line be \$2,800</b> . Would you recommend i make de purchase now o lateh in de future?	$x_{PA} = 1000, x_{BC} = \text{(March 25th - April 24th)},$ $x_{DA} = 2000, x_{CL} = 2800$	

**Source**: Can LLMs be Good Financial Advisors?: An Initial Study in Personal Decision Making for Optimized Outcomes, <a href="https://arxiv.org/abs/2307.07422">https://arxiv.org/abs/2307.07422</a>

## Complex Decisions

### Complex Decisions

- Making a sequence of decisions
- Making a single decision but with
  - Environment changing
  - Actions not being deterministic
  - Perception not being perfect
  - ...

### Making a Sequence of Decisions

Decision situation: driving to airport from home

- Actions:
  - Take a LEFT at first intersection
  - ENTER a highway
  - GETOUT a highway at EXIT-X
  - Turn RIGHT at intersection
  - PARK in Premium lot

• ..

### Optimal Decision

- What is it? There is no absolute answer. In AI, there is the concept of a rational agent.
- Acting rationally: acting such that one can achieve one's goals given one's beliefs (and information)
  - But what are one's goals
  - Are the goals always of achievement?

#### Some options

- Perfect rationality: maximize expected utility at every time instant
  - Given the available information; can be computationally expensive
  - "Doing the right thing"
- Bounded optimality: do as well as possible given computational resources
  - · Expected utility as high as any other agent with similar resources
- · Calculative rationality: eventually returns what would have been the rational choice

#### What Is It?

- As a working principle
  - Bounded or Calculative Rationality
- In observable and deterministic scenarios
  - Maximize utility: (benefit cost)
- In scenarios with uncertainty and/ or unobservable
  - Maximize expected utility: (benefit cost)

### Example Situation – Course Selection

- A person wants to pass an academic program in two majors: A and B
- There are three subjects: A, B and C, each with three levels (\*1, \*2, \*3). There are thus 9 courses: A1, A2, A3, B1, B2, B3, C1, C2, C3
- To graduate, at least one course at beginner (\*1) level is needed in major(s) of choice(s), and two courses at intermediate levels (\*2) are needed
- Optimality considerations in the problem
  - Least courses, fastest time to graduate, class size, friends attending together, ...
- Answer questions
  - Q1: How many minimum courses does the person have to take?
  - Q2: Can a person graduate in 2 majors studying 3 courses only?
  - ...

### Algorithms for Optimality

- Problem specific methods
  - Path finding
  - Linear programming
  - Constraint satisfaction and optimization
- General Purposed methods for optimality in search

#### Synthetic Example: Grid World

#### A maze-like problem

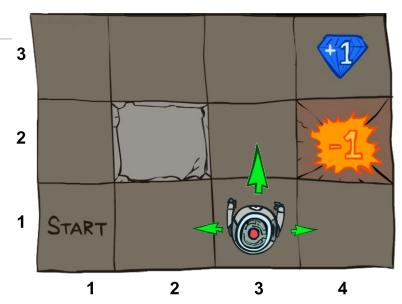
- The agent lives in a grid
- Walls block the agent's path

#### Noisy movement: actions do not always go as planned

- 80% of the time, the action North takes the agent North (if there is no wall there)
- 10% of the time, North takes the agent West; 10% East
- If there is a wall in the direction the agent would have been taken, the agent stays put

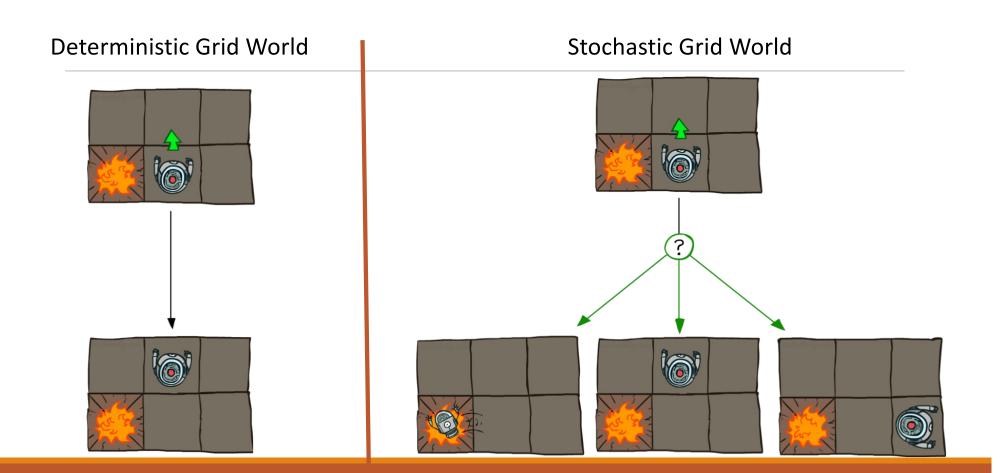
#### The agent receives rewards each time step

- Small "living" reward each step (can be negative)
- Big rewards come at the end (good or bad)



Slide adapted from: Dan Klein and Pieter Abbeel's Al lecture Original example in Russell & Norvig's Al book

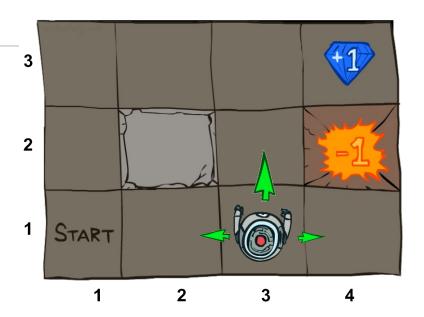
### Grid World Actions



### Markov Decision Processes

### An MDP is defined by:

- ∘ A set of states s ∈ S
- A set of actions a ∈ A
- A transition function T(s, a, s')
  - Probability that a from s leads to s', i.e., P(s' | s, a)
  - Also called the model or the dynamics
- A reward function R(s, a, s')
  - Sometimes just R(s) or R(s')
- A start state
- Maybe a terminal state



MDPs are non-deterministic search problems

Slide adapted from: Dan Klein and Pieter Abbeel's Al lecture Original example in Russell & Norvig's Al book

[Demo – gridworld manual intro (L8D1)]

## Markovian Assumption

"Markov" generally means that given the present state, the future and the past are independent

For Markov decision processes, "Markov" means action outcomes depend only on the current state

$$P(S_{t+1} = s' | S_t = s_t, A_t = a_t, S_{t-1} = s_{t-1}, A_{t-1}, \dots S_0 = s_0)$$
=

$$P(S_{t+1} = s' | S_t = s_t, A_t = a_t)$$



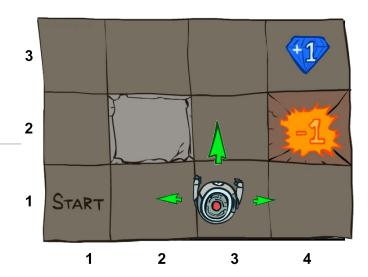
Andrey Markov (1856-1922)

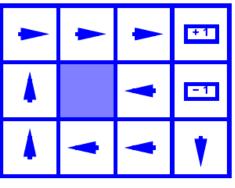
### Output: Policies

 In deterministic single-agent search problems, we have a plan, or sequence of actions, from start to a goal



- A policy  $\pi$  gives an action for each state
- An optimal policy is one that maximizes expected utility if followed





R(s) = -0.01

Slide adapted from: Dan Klein and Pieter Abbeel's Al lecture

### Exercise and Code

- MDP Solution Methods
  - From Book: AI A Modern Approach, <a href="https://github.com/aimacode/aima-python/blob/master/mdp.ipynb">https://github.com/aimacode/aima-python/blob/master/mdp.ipynb</a>

Source: Russell & Norvig, AI: A Modern Approach

# 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

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### Project Discussion: Dates and Deliverables

#### 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)
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  - 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

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

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### 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
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## 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 - NEW

#### Project report – 60%

- Project description: problem, related work, approach, evaluation – 40%
- Working system demo/ video 10%
  - Well organized Github with code (./data, ./code, ./docs, ./test) 10%

#### Project presentation – 40%

Evaluation by peers, instructor and TA

#### Bonus

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%]

### **Evaluation of Presentation**

- 1. An online form will be available during presentation
- 2. During a presentation, three students will be assigned to review along with instructor and TA
- 3. They will enter following survey questions:
  - 1. Their name
  - 2. Presentation number
  - 3. How useful is the system will you use it? [1-5 scale]
  - 4. How well have you understood the project from the presentation? [1-5 scale]
- Top and bottom scores will be removed. Average of remaining three will be used for final presentation marks

# Lecture 23 & 24: Summary

- We talked about
  - Making Decisions
  - Simple Decisions
  - Complex Decisions

# **Concluding Section**

## Quiz 4

- •November 14-21, 2023
  - Data Science
  - NN/ Keras with non-text media

## About Next Lecture – Lecture 25, 26

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## Lecture 25-26: Sequential Decisions

CSCE 580, 581 - FALL 2023

- Planning
- Reinforcement Learning