

Introduction

CISC 7404 - Decision Making

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• Python numerical programming

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- Deep learning

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- Statistics and probability

Python numerical programming:

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You should know:

• Python loops, lists, dicts, etc

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Assignments in jax/equinox, similar to torch - final project in torch

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If you do not know numerical programming, **you must learn immediately**: https://numpy.org/doc/stable/user/quickstart.html

Prerequisites Deep learning:

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You should know:

• How to construct a neural network in torch

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If you do not, review the deep learning slides: https://github.com/smorad/um_cisc_7026

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• Random variables (X)

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Question: What does P(X = x) mean?

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Answer: Probability of random variable X taking on a value of x

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Question: What does P(X = x) mean?

Answer: Probability of random variable X taking on a value of x

If you did not know this, you should review!

Grading

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• Quizzes 30%

- Quizzes 30%
- Assignments 30%

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- Final Project 30%

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- Assignments 30%
- Final Project 30%
- Participation 10%

Grading Quizzes:

Quizzes:

• I will tell you week before exam

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- Expect 3 quizzes

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Example 1: Quiz 1: 70%, Quiz 2: 80%, Quiz 3: 60%

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Final quiz score: (70 + 80) / 2 = 75%

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Example 2: Quiz 1: 90%, Quiz 2: (sick) 0%, Quiz 3: 70%

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Example 2: Quiz 1: 90%, Quiz 2: (sick) 0%, Quiz 3: 70%

Final quiz score: (90 + 70) / 2 = 80%

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Question: What if you are sick for two quizzes?

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- I will tell you week before exam
- Expect 3 quizzes
- I will drop your lowest quiz score

Example 1: Quiz 1: 70%, Quiz 2: 80%, Quiz 3: 60%

Final quiz score: (70 + 80) / 2 = 75%

Example 2: Quiz 1: 90%, Quiz 2: (sick) 0%, Quiz 3: 70%

Final quiz score: (90 + 70) / 2 = 80%

Question: What if you are sick for two quizzes? Only one quiz dropped, other quiz is zero

Grading Assignments:

Assignments:

• Programming

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- Programming
- Expect 2-3 assignments

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- We will use Google Colab: https://colab.research.google.com

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Late assignment policy:

• 0-1 day late (-15%)

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- 2-3 days late (-50%)

Assignments:

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- Expect 2-3 assignments
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- 0-1 day late (-15%)
- 1-2 days late (-30%)
- 2-3 days late (-50%)
- 3+ days late (-100%)

Grading Final Project:

Final Project:



Final Project:

Honor of Kings



 Research project based on Tencent platform

Final Project:



- Research project based on Tencent platform
- Train agents to play each other

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- Implement RL algorithm, improve it, write up analysis

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- Research project based on Tencent platform
- Train agents to play each other
- Implement RL algorithm, improve it, write up analysis
- More information later

Grading Participation:

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Significance

Despite active learning being recognized as a superior method of instruction in the classroom, a major recent survey found that most college STEM instructors still choose traditional teaching methods. This article addresses the long-standing question of why students and faculty remain resistant to active learning. Comparing passive lectures with active learning using a randomized experimental approach and identical course materials, we find that students in the active classroom learn more, but they feel like they learn less. We show that this negative correlation is caused in part by the increased cognitive effort required during active learning. Faculty who adopt active learning are encouraged to intervene and address this misperception, and we describe a successful example of such an intervention.

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- Individual participation

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Syllabus lists textbook chapter for each lecture

Github: https://github.com/smorad/um_cisc_7404

Question: What is cheating?

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Answer:

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• Copying assignment or exam from another student

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- Having notes, laptop, or phone during quiz/exam

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Answer:

- Copying assignment or exam from another student
- Having notes, laptop, or phone during quiz/exam
- Submitting LLM output for assignments

I don't like cheating

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All assignments and final project will use turnitin.com

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Turnitin detects copying between students and LLM use

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I gave many zeros in last term's deep learning course

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This term, I will report any cheating to the head of department

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It is not worth cheating, do your best and you will get partial credit

Secret: After you graduate nobody will care about your grade/degree!

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For AI jobs, you will do 5 hours of in-person interviews

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Your degree will be useless without the knowledge

I want you to learn the material so you succeed in life

Machine Learning Engineer Interview

16 May 2024 •••



Anonymous interview candidate

No offer — Neutral experience X Difficult interview

Application

Linterviewed at Baidu in 16/5/2024

Interview

it has 5 rounds of interviews, each of them is very long. I only managed to get to the second round with technical interview, and I failed. It's not a pleasent experience

Interview questions [1]

Question 1

describle tree algorithm and wirte in python

Answer question \rightarrow



Helpful



Machine Learning Engineer Interview

Aug 27, 2023 •••



Beijing, Beijing

─ Declined offer ✓ Positive experience X Difficult interview

Application

I applied online. I interviewed at ByteDance (Beijing, Beijing) in 8/27/2023

Interview

machine learning knowledge, code. leetcode (medium, hard with python), AUC definition, method on dealing with long tailed data and causal inference, machine learning question and deep learning method such as attention mechaism.

Interview questions [1]

Question 1

machine learning knowledge, code. leetcode, AUC definition, method on dealin...

read more

Answer question \rightarrow

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Ok: LLM, why does my Q function return large values?

Cheating: LLM, implement the policy gradient algorithm in pytorch

Lecture Topics

• Basics

- Basics
- Modern Methods

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- Active Research

Basics:

• Bandits

- Bandits
- Decision Processes

- Bandits
- Decision Processes
- Value Iteration

- Bandits
- Decision Processes
- Value Iteration
- Policy Gradient

- Bandits
- Decision Processes
- Value Iteration
- Policy Gradient
- Actor Critic

Modern Methods:

• Advantage Actor Critic

- Advantage Actor Critic
- Trust Region Policy Optimization

- Advantage Actor Critic
- Trust Region Policy Optimization
- Proximal Policy Optimization

- Advantage Actor Critic
- Trust Region Policy Optimization
- Proximal Policy Optimization
- Deep Q Learning

- Advantage Actor Critic
- Trust Region Policy Optimization
- Proximal Policy Optimization
- Deep Q Learning
- Deep Deterministic Policy Gradient

- Advantage Actor Critic
- Trust Region Policy Optimization
- Proximal Policy Optimization
- Deep Q Learning
- Deep Deterministic Policy Gradient
- Soft Actor Critic

- Advantage Actor Critic
- Trust Region Policy Optimization
- Proximal Policy Optimization
- Deep Q Learning
- Deep Deterministic Policy Gradient
- Soft Actor Critic
- Imitation learning

Active Research:

• Memory

- Memory
- Offline RL

- Memory
- Offline RL
- RL and Search

- Memory
- Offline RL
- RL and Search
- World Models

- Memory
- Offline RL
- RL and Search
- World Models
- RL from Human Feedback

In this course, we will focus primarily on reinforcement learning

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But reinforcement learning is a method, not a problem

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The problem is decision making

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But reinforcement learning is a method, not a problem

The problem is decision making

In this course, we will learn how to make good decisions

Question: What is decision making?

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It depends, each field has their own definition

Philosophy

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- Philosophy
- Cognitive science

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- Machine learning

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- Philosophy
- Cognitive science
- Economics
- Machine learning

Answer: Given information, make a choice that impacts the world

Question: Why should we care about decision making?

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Everything in life is a decision

• Do I eat dumplings or noodles?

Question: Why should we care about decision making?

- Do I eat dumplings or noodles?
- What time should I leave for class?

Question: Why should we care about decision making?

- Do I eat dumplings or noodles?
- What time should I leave for class?
- Should I go to school or find a job?

Question: Why should we care about decision making?

- Do I eat dumplings or noodles?
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- Should I date this person?

Question: Why should we care about decision making?

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- Where should I live?

Question: Why should we care about decision making?

- Do I eat dumplings or noodles?
- What time should I leave for class?
- Should I go to school or find a job?
- Should I date this person?
- Where should I live?
- What should we use taxes for?

Humans are decision making machines – it is all we do!

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We can represent life as a series of decisions

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What we do defines who we are

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What we do defines who we are

"All we have to decide is what to do with the time that is given to us"

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To study decision making is to study ourselves

Humans are decision making machines – it is all we do!

We can represent life as a series of decisions

What we do defines who we are

"All we have to decide is what to do with the time that is given to us"

To study decision making is to study ourselves

If we learn to make better decisions, we can lead better lives

In this course, we focus on **optimal** decision making

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Make the best possible decision, given the information we have

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We will find methods that **guarantee** optimal decision making

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With an optimal decision making machine, you can create:

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With an optimal decision making machine, you can create:

• Best possible doctor (which medicine to give?)

In this course, we focus on optimal decision making

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With these methods, we can create optimal decision making machines

With an optimal decision making machine, you can create:

- Best possible doctor (which medicine to give?)
- Best possible lawyer (what to argue?)

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Make the best possible decision, given the information we have

We will find methods that **guarantee** optimal decision making

With these methods, we can create optimal decision making machines

With an optimal decision making machine, you can create:

- Best possible doctor (which medicine to give?)
- Best possible lawyer (what to argue?)
- Best possible scientist (what to research?)

In this course, we focus on optimal decision making

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We will find methods that guarantee optimal decision making

With these methods, we can create optimal decision making machines

With an optimal decision making machine, you can create:

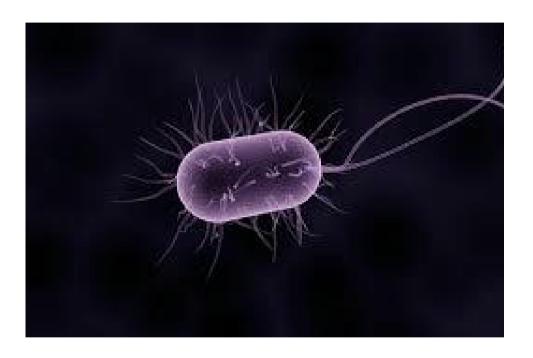
- Best possible doctor (which medicine to give?)
- Best possible lawyer (what to argue?)
- Best possible scientist (what to research?)

If the machine understands why it makes decisions, it is conscious

Let us discuss the history of decision making to better understand it

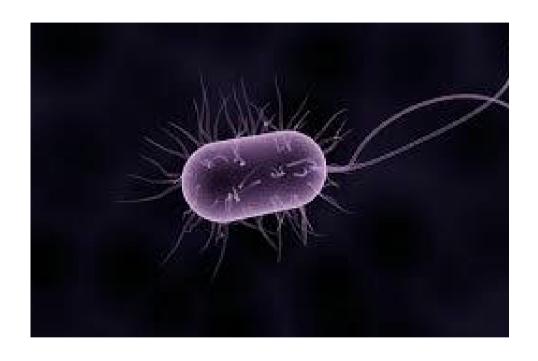
Question: Who was the first to apply decision making algorithms?

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3.5 GYA: Single cell organism

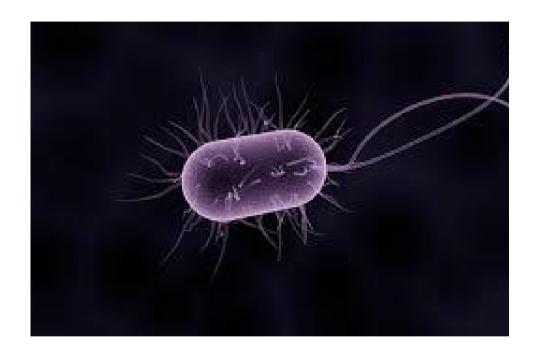
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3.5 GYA: Single cell organism

Decides to move away from danger and move towards food

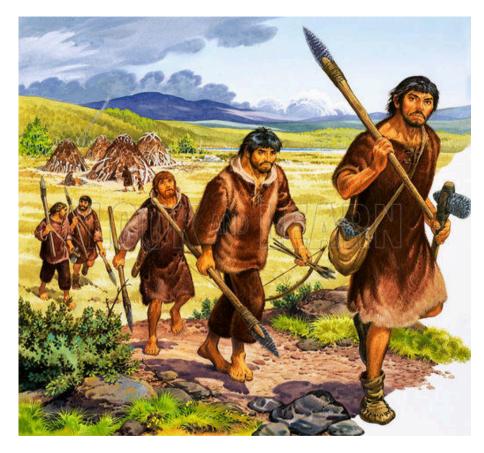
Question: Who was the first to apply decision making algorithms?



3.5 GYA: Single cell organism

Decides to move away from danger and move towards food

Decision making is necessary for life





200 kYA: Humanoid huntergatherers develop more complex decision making capabilities



200 kYA: Humanoid huntergatherers develop more complex decision making capabilities

Sequence of decisions to make fire



200 kYA: Humanoid huntergatherers develop more complex decision making capabilities

Sequence of decisions to make fire

Sequence of decisions to plant crops



500 BCE: Humans begin to study decision making



500 BCE: Humans begin to study decision making

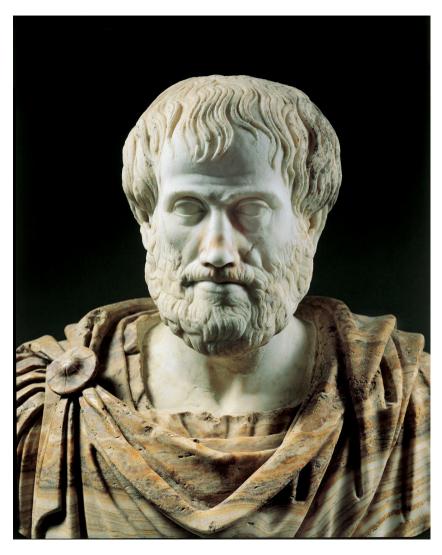
Sun Tzu studies and writes about various forms of decision making



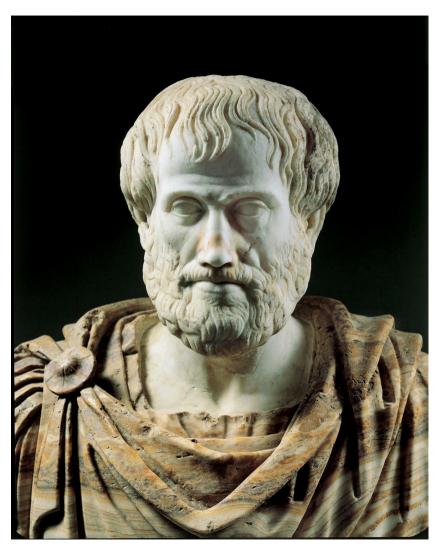
500 BCE: Humans begin to study decision making

Sun Tzu studies and writes about various forms of decision making

E.g., zero sum games: "Attack where he is unprepared; appear where you are not expected."

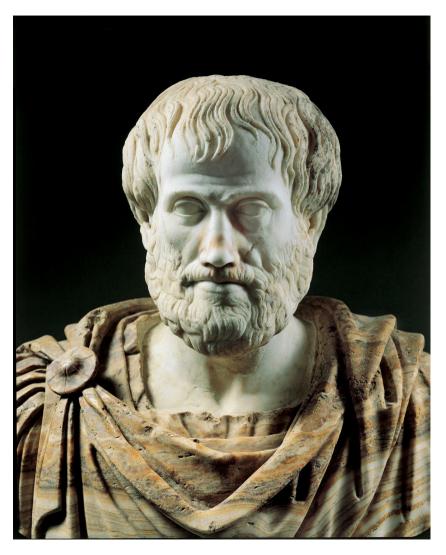


400 BCE: Aristotle creates the earliest recorded framework for decision making



400 BCE: Aristotle creates the earliest recorded framework for decision making

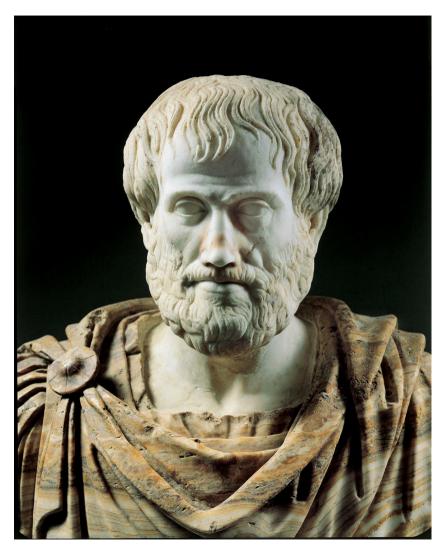
Syllogistic logic and deductive reasoning from axioms



400 BCE: Aristotle creates the earliest recorded framework for decision making

Syllogistic logic and deductive reasoning from axioms

Axiom 1: All philosophers prioritize knowledge over leisure

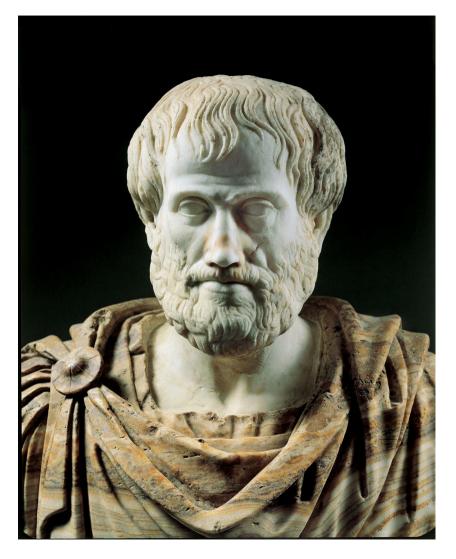


400 BCE: Aristotle creates the earliest recorded framework for decision making

Syllogistic logic and deductive reasoning from axioms

Axiom 1: All philosophers prioritize knowledge over leisure

Axiom 2: I am a philosopher



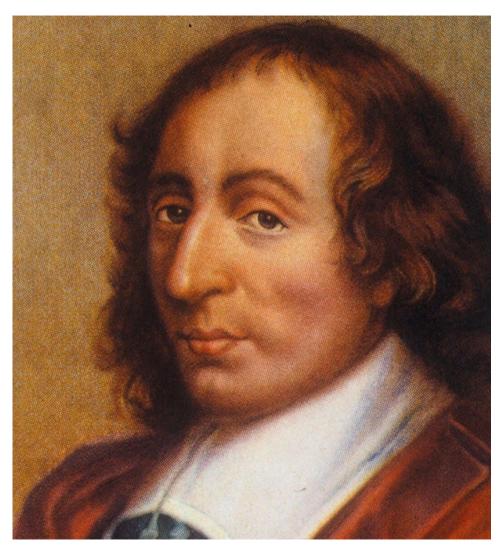
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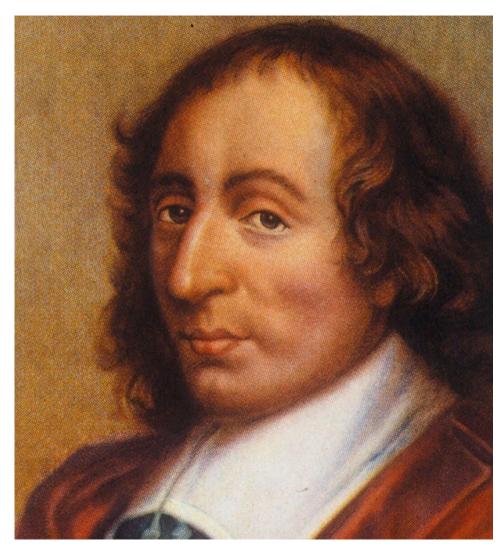
Axiom 1: All philosophers prioritize knowledge over leisure

Axiom 2: I am a philosopher

Decision: I must attend lecture instead of the party

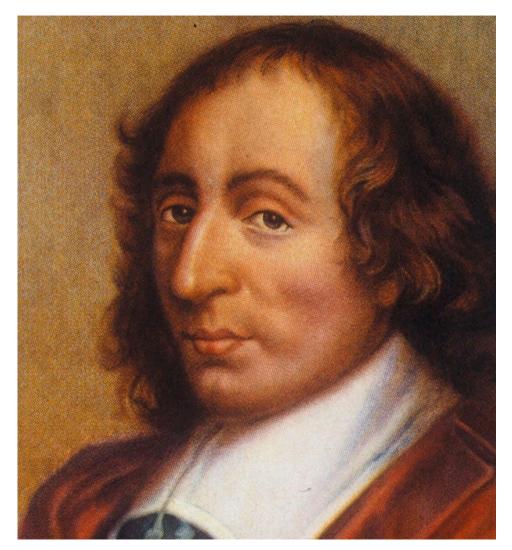


1654: Pascal formalizes decision making under uncertainty with "Pascal's Wager"



1654: Pascal formalizes decision making under uncertainty with "Pascal's Wager"

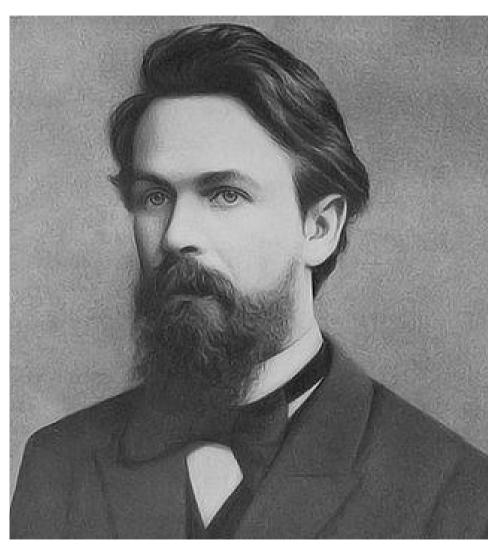
Premise: You are in bed, about to die. Should you believe in God?



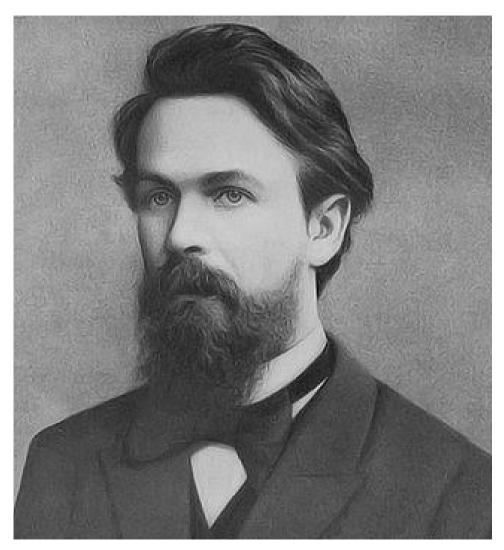
1654: Pascal formalizes decision making under uncertainty with "Pascal's Wager"

Premise: You are in bed, about to die. Should you believe in God?

	Believe	Do not
		believe
God exists	Good	Bad
God does	Neutral	Neutral
not exist		

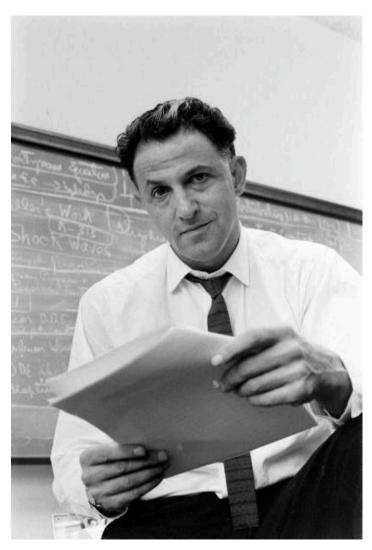


1906: Markov discovers Markov processes

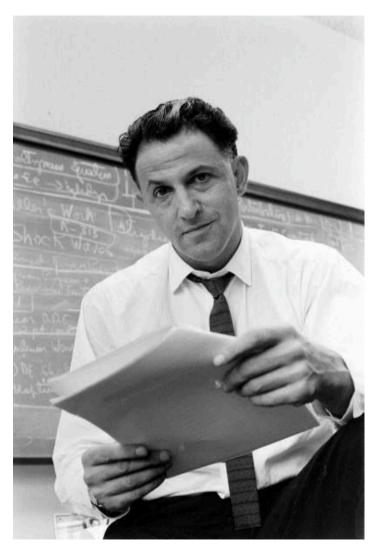


1906: Markov discovers Markov processes

Modern decision making relies on Markov processes



1953: Bellman discovers dynamic programming



1953: Bellman discovers dynamic programming

Gives us the **Bellman equation**, the basis for optimal decision making



1983: Sutton solves the Bellman equation using neural networks



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Combines reinforcement learning and neural networks



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He is still alive and might answer your emails



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We use his textbook: *An Introduction to Reinforcement Learning*

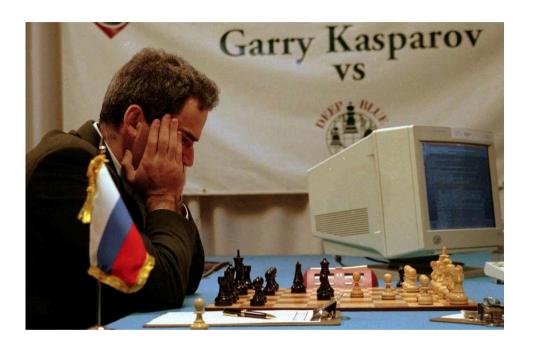


1997: DeepBlue beats world champion Kasparov at chess



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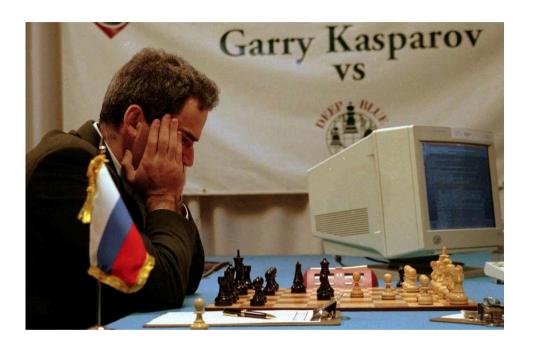
People start to pay attention to decision making machines



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Chess AIs play each other because humans are too easy



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https://www.youtube.com/watch? v=KF6sLCeBj0s



2016: AlphaGo beats world champion Sedol at Go



2016: AlphaGo beats world champion Sedol at Go

https://www.youtube.com/watch? v=tXlM99xPQC8



2018: OpenAI Five beats world champions at Dota2



2018: OpenAI Five beats world champions at Dota2

https://www.youtube.com/watch? v=eHipy_j29Xw



2020-2024: GPT-3, GPT-4 trained using reinforcement learning

2025?

We will formally define decision making and reinforcement learning later in the course

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For now, I want to clarify decision making in the context of machine learning

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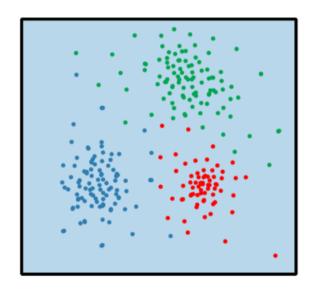
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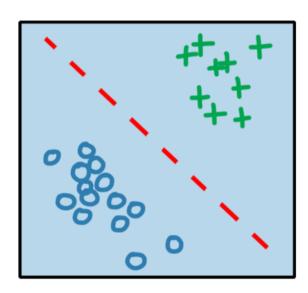
How does decision making differ from regular deep learning?

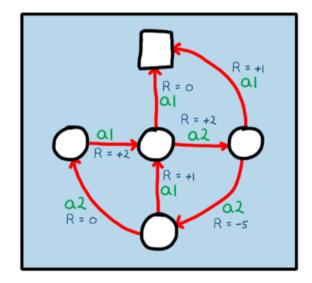
machine learning

unsupervised learning supervised learning

reinforcement learning







In deep learning, we usually know the answer

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$$f(\boldsymbol{x}, \boldsymbol{\theta}) = \boldsymbol{y}$$

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In deep learning, we usually know the answer

$$f(\boldsymbol{x}, \boldsymbol{\theta}) = \boldsymbol{y}$$

In decision making, we often do not know the answer!

$$f(\boldsymbol{x}, \boldsymbol{\theta}) = ?$$

What does this mean?

Example: You train a model f to play chess

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$$f: X \times \Theta \mapsto Y$$

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 $X \in \text{Position of pieces on the board}$

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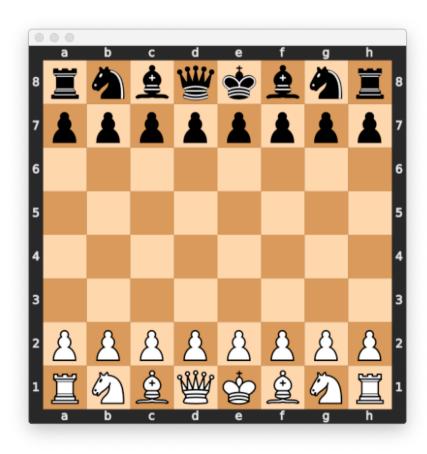
$$f: X \times \Theta \mapsto Y$$

 $X \in \text{Position of pieces on the board}$

 $Y \in$ Where to move piece

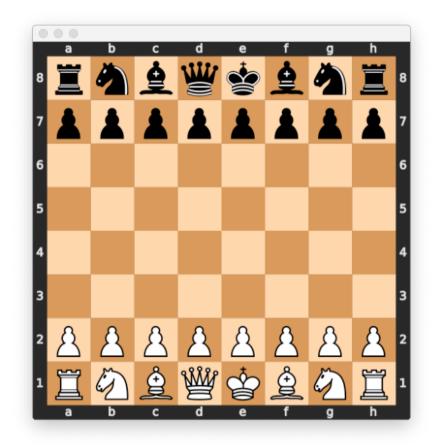
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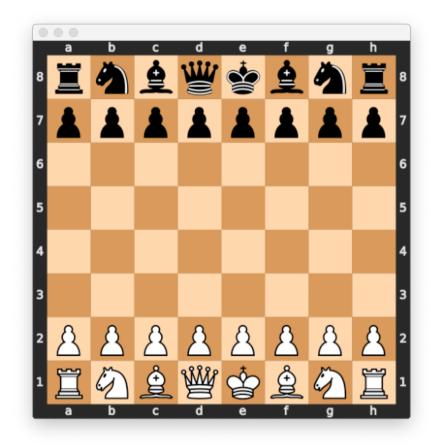


What is the correct answer?



What is the correct answer?

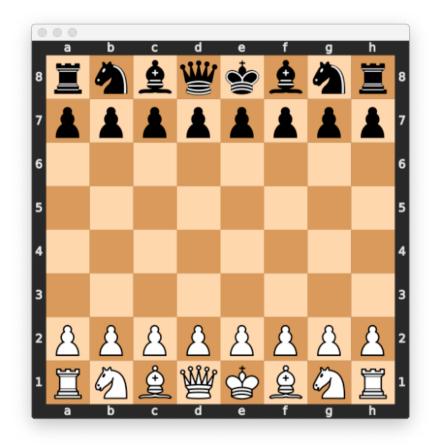
We do not know the answer



How can we learn a model without an answer?



An answer gives us just one move



An answer gives us just one move We

We need many moves to win

Decision making can give us the best **sequence** of moves to:

• Win a game of chess

- Win a game of chess
- Drive a customer to the store

- Win a game of chess
- Drive a customer to the store
- Cook a tasty meal

- Win a game of chess
- Drive a customer to the store
- Cook a tasty meal
- Treat a sick patient

- Win a game of chess
- Drive a customer to the store
- Cook a tasty meal
- Treat a sick patient
- Prevent climate change

- Win a game of chess
- Drive a customer to the store
- Cook a tasty meal
- Treat a sick patient
- Prevent climate change
- Reduce human suffering

Decision making can give us the best **sequence** of moves to:

- Win a game of chess
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We do not know the correct moves

Decision making can give us the best **sequence** of moves to:

- Win a game of chess
- Drive a customer to the store
- Cook a tasty meal
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We do not know the correct moves

But with decision making, we can find them!

Questions?

• Review prerequisites

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 - Especially probability

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 - ► Read Chapter 2 before next lecture