CSC520 - Artificial Intelligence Lecture 1

Dr. Scott N. Gerard

North Carolina State University

Jan 7, 2025

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Agenda

- Introductions
- Overview
- History
- State of the Art
- Benefits and Risks
- Logistics

Introductions

- Instructor
 - Dr. Scott N. Gerard sngerard@ncsu.edu
 - * Many decades at IBM
 - ★ Lecturer, North Carolina State University
 - ★ Chief AI Officer at Knowledge Reactor
- TAs
 - ► Fardin Saad fsaad@ncsu.edu

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- Asimov's Three Laws of Robotics
 - A robot may not injure a human being or, through inaction, allow a human being to come to harm.
 - A robot must obey orders given it by human beings except where such orders would conflict with the First Law.
 - A robot must protect its own existence as long as such protection does not conflict with the First or Second Law.









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Article

Magnetic control of tokamak plasmas through deep reinforcement learning

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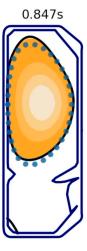
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Tracey-in-Francesco Carpanese-1-1, Timo Ewalds': Rolland Hafner', Abbas Abdolmalesk', Deligo de las Casa's Criaj Donner', Letale Fritz', Cristian Daipert', Andreas Habert', Arbone Merle', Jean-Marc Moret', Jeanes Keeling', Maria Tsimpoukelli', Jackie Koy', Antoine Merle', Jean-Marc Moret', Seb Noury', Federico Pesamosca', David Haru', Olivier Santer', Cristian Sommarira', Stefano Coda', Basil Lovat', Ambrogio Fasoli', Pushmeet Kohli', Koray Kavukcuoglu', Demis Hassabis' à Martin Ridentilles'.

Jonas Degrave^{1,3}, Federico Felici^{2,3,22}, Jonas Buchli^{1,3,22}, Michael Neunert^{1,3}, Brendan

Nuclear fusion using magnetic confinement, in particular in the tokamak configuration, is a promising path towards sustainable energy. A croe challenge is to shape and maintain a high-temperature plasma within the tokamak vessel. This requires high-dimensional, high-frequency, closed-loop control using magnetic actuator coils, further complicated by the diverse requirements across a wide range of plasma configurations. In this work, we introduce a previously undescribed architecture for tokamak magnetic controller design that autonomously learns to command the full set of control coils. This architecture meets control objectives specified at an alph level, at the same time satisfying physical and operational constraints. This approach has unprecedented flexibility and generality in problem specification and yields a notable reduction in design effort to produce new plasma configurations. We successfully produce and control a diverse set of plasma configurations. We successfully produce and control a diverse set of plasma configurations on the Tokamak & Configuration variable³, including elongated,



What is intelligence?

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AI Definitions

- Various definitions of AI can be categorized based on two dimensions
 - ► Comparison with *humans* versus notion of *rationality*

AI Definitions

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 - Comparison with humans versus notion of rationality
 - Internal thought processes versus external behavior

Al Definitions

- Various definitions of AI can be categorized based on two dimensions
 - Comparison with humans versus notion of rationality
 - Internal thought processes versus external behavior

Think like humans	Think rationally
Cognitive science, "machines with minds"	Logical reasoning, "study of mental faculties through the use of computational models"
Act like humans	Act rationally
Turing test, "Machines that perform functions that require intelligence when performed by people"	Agents that do the right thing, "intelligent behavior in artifacts"

Foundations of AI

Philosophy logic, reasoning, rationality

Mathematics formal logic, algorithms,

complexity analysis, probability

Economics decision theory, operations

research, utility

Neuroscience human brain, nervous system

Psychology knowledge-based agent,

human-computer interaction

Computer engineering efficient computers

Linguistics natural language processing, knowledge representation

History of Al

- Inception of AI (1943-1956)
 - ▶ McCulloch and Pitts: Boolean circuit model of brain
 - Turing test

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History of Al

- Inception of AI (1943-1956)
 - McCulloch and Pitts: Boolean circuit model of brain
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- Great expectation (1952-1969)
 - Al programs for games, puzzles, etc. based on microworlds
 - Newell and Simon: General problem solver
 - John McCarthy: Lisp programming language
 - ► Frank Rosenblatt: Perceptrons, perceptron convergence theorem

History of Al

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- Dose of reality (1966-1986)
 - ► Early AI systems failed to scale to the size of real-world problems
 - Knowledge-based expert systems industry boom followed by bust
 - Al Winter

History of AI (Cont.)

- Return of neural networks and statistical approaches (1990-2012)
 - ▶ Back-propagation learning algorithm re-invented
 - Probability and statistical methods to address uncertainity
 - ▶ Hidden markov models, Bayesian networks, reinforcement learning

History of AI (Cont.)

- Return of neural networks and statistical approaches (1990-2012)
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- Recent history (2012-present)
 - Big data, big compute
 - Deep learning
 - Al used extensively
 - Generative AI (GenAI) and Large Language Models (LLM)

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Which of the following can be done at present?

• Play and defeat humans in chess

Which of the following can be done at present?

- Play and defeat humans in chess √
- Play and defeat humans in Go

Which of the following can be done at present?

- ullet Play and defeat humans in chess \checkmark
- ullet Play and defeat humans in Go \checkmark
- Drive a car autonomously

Which of the following can be done at present?

- Play and defeat humans in chess √
- Play and defeat humans in Go √
- Drive a car autonomously √
- Discover and prove a new mathematical theorem

Which of the following can be done at present?

- Play and defeat humans in chess √
- Play and defeat humans in Go √
- Drive a car autonomously √
- Discover and prove a new mathematical theorem ??
- Design and execute a research program in molecular biology

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- Write an intentionally funny story

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- Write an intentionally funny story √
- Buy groceries in a local grocery store

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- Buy groceries in a local grocery store X
- Translate spoken English into spoken Chinese in real time

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- Perform a complex surgical operation

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- Perform a complex surgical operation X
- Unload a dishwasher and put dishes in right places

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- Robotics
 - Self-driving cars
 - ► Factory automation
 - Soccer

- Robotics
 - Self-driving cars
 - Factory automation
 - Soccer
- Natural Language Processing
 - Speech recognition, text to speech
 - Question answering, language translation, information retrieval
 - General purpose language models, story generation

- Game playing
 - Deep blue defeated world chess champion Garry Kasparov in 1997
 - ▶ Alpha Go defeated European champion Fan Hui in 2015

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- Computer Vision
 - Image classification
 - Object recognition, face recognition
 - Image segmentation
 - Generative adversarial networks

- Benefits
 - Decrease dangerous and repetitive work
 - Increase production of goods and services
 - ▶ Accelerate scientific research: cure diseases, climate change solutions

Benefits

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- Increase production of goods and services
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Risks

- ▶ Lethal autonomous weapons can eliminate human targets
- ▶ Mass surveillance using phone, video, emails, etc.
- Biased decision making
- Impact on employment
- ▶ Use of Copyrighted/Intellectual Property materials for ML training
- ► Safety critical applications: self-driving cars, water supply mgmt., etc.
- Hackers using AI for cybersecurity attacks

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- Trustworthy AI
 - Correctness
 - Privacy
 - Bias
 - Explainability
 - Ethics
 - Governance

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Course

What will we cover in this course?

- Fundamentals of Al
 - ▶ Build agents that act rationally i.e. maximize expected utility

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- Fundamentals of Al
 - Build agents that act rationally i.e. maximize expected utility
- Agents, problem solving, search, knowledge, planning, ML basics

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Course

What will we cover in this course?

- Fundamentals of Al
 - ▶ Build agents that act rationally i.e. maximize expected utility
- Agents, problem solving, search, knowledge, planning, ML basics
- This is not a Machine Learning or Deep Learning course!

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- Agents: Problems, problem solving.
- Search: Uninformed, informed, adversarial.
- Optimization: Local search and adaptation.
- Constraints: Solving problems with constraints.
- Knowledge: Logic, knowledge-based agents.
- Planning: Semi-constrained search, scheduling.
- Uncertainty: Probabilistic models, decision-making.
- ML: Machine Learning, deep learning.
- NLP: Basics of NLP in Al.
- CV: Basics of CV in AI.
- Risks and Benefits:

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Generative AI

Large Language Models

- Kind of Neural Network
- Attention/Transformer mechanism

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Large Language Models

- Kind of Neural Network
- Attention/Transformer mechanism
- Not the only Al technology
 - ► They hallucinate
 - Poor reasoners

Generative AI

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Agentic Al

- All agents wrap an LLM
- Not the only Al agent technology

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Agentic Al

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LLMs are not covered in this course

Five Tribes of Machine Learning

Machine Learning Tribes

Domingos, The Master Algorithm

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Five Tribes of Machine Learning

Symbolist Logic

Machine Learning Tribes

Domingos, The Master Algorithm

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Five Tribes of Machine Learning

Symbolist Logic Connectionist Neural Network

Machine Learning Tribes

Five Tribes of Machine Learning

Symbolist Logic

Connectionist Neural Network

Machine Learning Tribes

Bayesian Probability

Domingos, The Master Algorithm

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Five Tribes of Machine Learning

Symbolist Logic Connectionist Neural Network

Machine Learning

Bayesian Tribes

Probability

Evolutionary
Survival of the Fittest

Five Tribes of Machine Learning

Symbolist Logic

Connectionist Neural Network

Machine

Learning Tribes

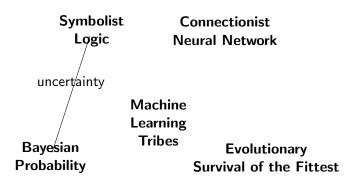
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Survival of the Fittest

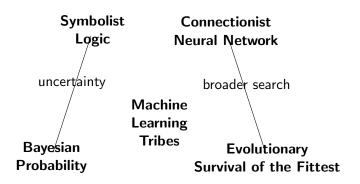
Analogizer Similarity

Five Tribes of Machine Learning



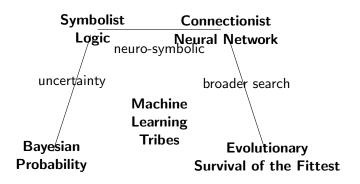
Analogizer Similarity

Five Tribes of Machine Learning



Analogizer Similarity

Five Tribes of Machine Learning



Analogizer Similarity

Domingos, The Master Algorithm

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Course Logistics

Logistics

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