





Introduction to Artificial Intelligence

Faculty of DS & AI Autumn semester, 2025

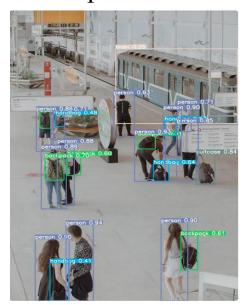
Trong-Nghia Nguyen

nghiant@neu.edu.vn



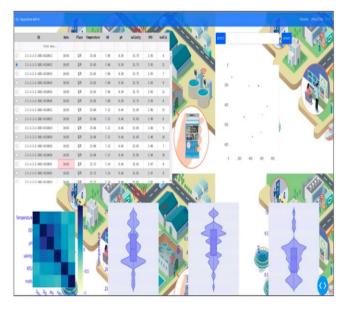
About

- Bio
 - 02.2025: PhD., Chonnam National Univ., Korea.
 - Contact: <u>trongnghia7171@gmail.com</u> or <u>nghiant@neu.edu.vn</u>
- Research Interest
 - Computer vision



Person Re-identification

• ML/AI



Smart Aquaculture

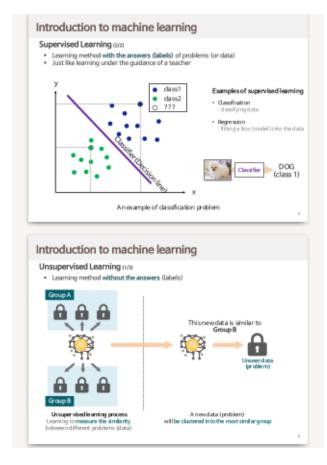
• Time-series analysis



Clinical Emergency
Medicine

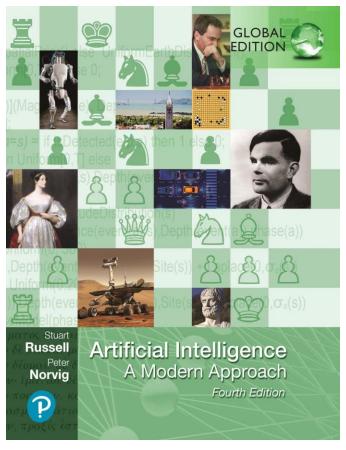
Textbooks

• Main



PPT

Supplementary



E-book

• This class is...

VIE ENG

70% English

Lecture Overview

- Course Introduction
 - Objectives & assessment methods
- What is AI

 Definitions & approaches
- History of AI
 From 1950 to present
- Current AI Applications

 Current AI Applications
- Summary & Next steps
 Summary & Next Steps

Learning Objectives

I	Artificial Intelligence	
1		1
_	11 What Is AI?	1
	1.2 The Foundations of Artificial Intelligence	2
	1.3 The History of Artificial Intelligence	3
	1.4 The State of the Art	4
	1.5 Risks and Benefits of AI	4
	Summary	5
	Bibliographical and Historical Notes	5
2	Intelligent Agents	5
	2.1 Agents and Environments	5
	2.2 Good Behavior: The Concept of Rationality	5
	2.3 The Nature of Environments	6
	2.4 The Structure of Agents	6
	Summary	7
	Bibliographical and Historical Notes	7
П	Problem-solving	
3	Solving Problems by Searching	8
	3.1 Problem-Solving Agents	8
	3.2 Example Problems	8
	3.3 Search Algorithms	8
	3.4 Uninformed Search Strategies	9
	3.5 Informed (Heuristic) Search Strategies	10
	3.6 Heuristic Functions	11
	Summary	12
	Bibliographical and Historical Notes	12
4	Search in Complex Environments	12
	4.1 Local Search and Optimization Problems	12
	4.2 Local Search in Continuous Spaces	13
	4.3 Search with Nondeterministic Actions	14
	4.4 Search in Partially Observable Environments	14
	4.5 Online Search Agents and Unknown Environments	15
	Summary	15
	Bibliographical and Historical Notes	16
5	Constraint Satisfaction Problems	16
	5.1 Defining Constraint Satisfaction Problems	16
	5.2 Constraint Propagation: Inference in CSPs	16

Learning Objectives

10	Know		334
	10.1	Ontological Engineering	332
	10.2	Categories and Objects	335
	10.3	Events	340
	10.4	Mental Objects and Modal Logic	344
	10.5	Reasoning Systems for Categories	347
	10.6		351
	Summ	ary	355
		•	356
11	Auton	nated Planning	362
	11.1	Definition of Classical Planning	362
	11.2	Algorithms for Classical Planning	366
	11.3	Heuristics for Planning	371
	11.4	Hierarchical Planning	374
	11.5	Planning and Acting in Nondeterministic Domains	383
	11.6	Time, Schedules, and Resources	392
	11.7	Analysis of Planning Approaches	396
	Summ	ary	397
	Biblio	graphical and Historical Notes	398
IV	Unc	certain knowledge and reasoning	
12	Quan	tifying Uncertainty	403
_	12.1	Acting under Uncertainty	403
	12.2	Basic Probability Notation	406
	12.3	Inference Using Full Joint Distributions	413
	12.4	Independence	415
	12.5	Bayes' Rule and Its Use	417
	12.6	Naive Bayes Models	420
	12.7	The Wumpus World Revisited	422
	Summ	ary	425
	Biblio	graphical and Historical Notes	426
13	Proba	abilistic Reasoning	430
	13.1	Representing Knowledge in an Uncertain Domain	430
	13.2	The Semantics of Bayesian Networks	432
	13.3		445
	13.4	Approximate Inference for Bayesian Networks	453
	13.5	11	467
	Summ		471
			472
14	Proba	abilistic Reasoning over Time	479
	14.1		479
	14.2	•	483
		*	

18	Prob	abilistic Programming										6
	18.1	Relational Probability Models	 									. 6
	18.2	Open-Universe Probability Models	 									. 6
	18.3	Keeping Track of a Complex World	 									. 6
	18.4	Programs as Probability Models	 									. 6
	Sumn	nary	 									. 6
	Biblio	ographical and Historical Notes	 									. 6
v	Mac	chine Learning										
19	Lear	ning from Examples										6
	19.1	Forms of Learning	 									. 6

		•	Contents
	19.2	Supervised Learning	. 671
	19.3	Learning Decision Trees	
	19.4	Model Selection and Optimization	. 683
	19.5	The Theory of Learning	. 690
	19.6	Linear Regression and Classification	. 694
	19.7	Nonparametric Models	. 704
	19.8	Ensemble Learning	. 714
	19.9	Developing Machine Learning Systems	. 722
	Sumn	nary	. 732
	Biblio	ographical and Historical Notes	. 733
20	Know	dedge in Learning	739
	20.1	A Logical Formulation of Learning	
	20.2	Knowledge in Learning	
	20.3	Explanation-Based Learning	
	20.4	Learning Using Relevance Information	
	20.5	Inductive Logic Programming	
	Sumn	nary	
		ographical and Historical Notes	
21		ning Probabilistic Models	772
	21.1	Statistical Learning	
	21.2	Learning with Complete Data	
	21.3	Learning with Hidden Variables: The EM Algorithm	
		nary	
	Biblio	ographical and Historical Notes	. 798
22	Deep	Learning	801

	19.2	Supervised Learning	671
	19.3	Learning Decision Trees	675
	19.4	Model Selection and Optimization	683
	19.5	The Theory of Learning	690
	19.6	Linear Regression and Classification	694
	19.7	Nonparametric Models	704
	19.8	Ensemble Learning	714
	19.9	Developing Machine Learning Systems	722
	Sumn	ary	732
	Biblio	graphical and Historical Notes	733
20	Know	ledge in Learning	739
	20.1	A Logical Formulation of Learning	739
	20.2	Knowledge in Learning	747
	20.3	Explanation-Based Learning	750
	20.4	Learning Using Relevance Information	754
	20.5	Inductive Logic Programming	758
	Sumn	ary	767
	Biblio	graphical and Historical Notes	768
21		ing Probabilistic Models	772
	21.1	Statistical Learning	772
	21.2	Learning with Complete Data	775
	21.3	Learning with Hidden Variables: The EM Algorithm	788
		ary	797
		ary graphical and Historical Notes	797 798
22	Biblic	graphical and Historical Notes	
22	Biblio Deep	graphical and Historical Notes	798
22	Biblio Deep 22.1	graphical and Historical Notes	798 801 802
22	Deep 22.1 22.2	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning	798 801 802 807
22	Deep 22.1 22.2 22.3	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks	798 801 802 807 811
22	Deep 22.1 22.2 22.3 22.4	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms	798 801 802 807 811 816
22	Deep 22.1 22.2 22.3 22.4 22.5	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization	798 801 802 807 811 816 819
22	Deep 22.1 22.2 22.3 22.4 22.5 22.6	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks	798 801 802 807 811 816 819 823
22	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning	798 801 802 807 811 816 819 823 826
22	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications	798 801 802 807 811 816 819 823 826 833
22	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ	graphical and Historical Notes Learning Simple Feedferward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications	798 801 802 807 811 816 819 823 826 833 835
22	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications	798 801 802 807 811 816 819 823 826 833
	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblio	graphical and Historical Notes Learning Simple Feedferward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications	798 801 802 807 811 816 819 823 826 833 835
	Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblio	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications ary graphical and Historical Notes	798 801 802 807 811 816 819 823 826 833 835 836
	Deep 22 1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblio Reinf	graphical and Historical Notes Learning Simple Feedferward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications Larry graphical and Historical Notes Dreement Learning Learning from Rewards	798 801 802 807 811 816 819 823 826 833 835 836
	Deep 22 1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblio Reinf 23.1	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications arry graphical and Historical Notes orcement Learning	798 801 802 807 811 816 819 823 826 833 835 836 840
	Biblic Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblic Reinf 23.1 23.2	graphical and Historical Notes Learning Simple Feederward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications ary graphical and Historical Notes prement Learning Learning from Rewards Passive Reinforcement Learning	798 801 802 807 811 816 819 823 826 833 835 836 840 842
	Biblic Deep 22 1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblic Reinf 23.1 23.2 23.3	graphical and Historical Notes Learning Simple Feedforward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications ary graphical and Historical Notes Dreement Learning Learning from Rewards Passive Reinforcement Learning Active Reinforcement Learning Generalization in Reinforcement Learning Generalization in Reinforcement Learning	798 801 802 807 811 816 819 823 826 833 835 836 840 842 848
	Biblio Deep 22.1 22.2 22.3 22.4 22.5 22.6 22.7 22.8 Summ Biblio Reinf 23.1 23.2 23.3 23.4	graphical and Historical Notes Learning Simple Feederward Networks Computation Graphs for Deep Learning Convolutional Networks Learning Algorithms Generalization Recurrent Neural Networks Unsupervised Learning and Transfer Learning Applications ary graphical and Historical Notes prement Learning Learning from Rewards Passive Reinforcement Learning Active Reinforcement Learning	798 801 802 807 811 816 819 823 826 833 835 836 840 842 848 854

Assessment Methods

• Class Participation: 30%

Homework, group presentation, etc.

• Midterm Exam: 30%

Could be project (flexible)

• Final Exam: 40%

Focus knowledge

Lecture Overview

Course Introduction

Objectives & assessment methods

- What is AI

 Definitions & approaches
- History of AI
 From 1950 to present
- Current AI Applications

 Current AI Applications
- Summary & Next steps
 Summary & Next Steps

What is Artificial Intelligence?

"Artificial Intelligence is the study of how to make computers do things at which, at the moment, people are better."

- Elaine Rich, 1983

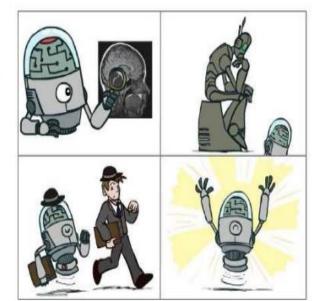
Artificial Intelligence is the field of study that focuses on creating computer systems capable of performing tasks that typically require human intelligence.

Core Concepts:

- · Learning from experience
- · Solving complex problems
- · Pattern recognition
- · Intelligent decision making

Four Categories of AI

Think like people



Think rationally

Act like people

Act rationally

Acting Humanly

Systems that act like humans

Turing Test: Systems that behave indistinguishably from humans

Examples: Chatbots, virtual assistants

Thinking Humanly

Systems that think like humans

Cognitive Modeling: Modeling the human thought process

Examples: Artificial neural networks

Thinking Rationally

Systems that think rationally

Laws of Thought: Using logic and reasoning

Examples: Expert systems, logic programming

Acting Rationally

Systems that act rationally

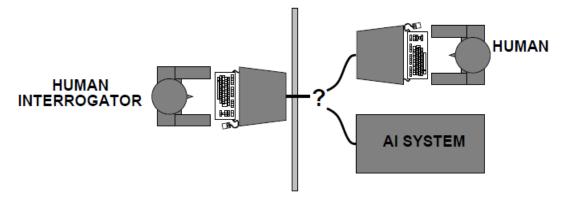
Rational Agents: Acting optimally in an environment

Examples: Autonomous vehicles, game-playing Al

(1) Acting Humanly – Turing Test

Alan Turing(1950) "Computing machinery and intelligence"

- ♦ "Can machines think?" → "Can machines behave intelligently?"
- ♦ Operational test for intelligent behavior: the Imitation Game



- Predicted that by 2000, a machine might have a 30% chance of fooling a lay person for 5 minutes
- Suggested major components of AI: knowledge, reasoning, language understanding, learning
- ♦ Total Turing test: computer vision, robotics

(2) Thinking Humanly – Cognitive Science

Cognitive Revolutions in 1960s

Get inside: introspection, psychological experiments, brain imaging

Requires scientific theories of internal activities of the brain

- What level of abstraction? "Knowledge" or "circuits"?
- How to validate? Requires
 - 1) Predicting and testing behavior of human subjects (top-down)
 - or 2) Direct identification from neurological data (bottom-up)

Both approaches (roughly, Cognitive Science and Cognitive Neuroscience) are now distinct from AI

Both share with Al the following characteristic: the available theories do not explain anything resembling human-level general intelligence. Hence, all three fields share one principal direction!

(3) Thinking Rationally – Laws of Though

Aristotle Syllogism – "Right Thinking"

Several Greek schools developed various forms of logic:

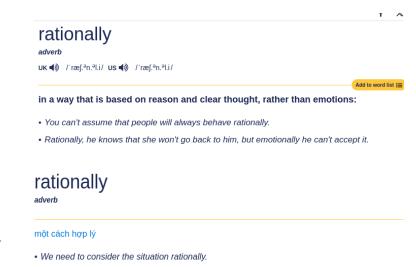
notation and rules of derivation for thoughts;

may or may not have proceeded to the idea of mechanization

Logicist tradition withi Al hopes to build on such programs to create intelligent systems.

Problems:

- 1) Not all intelligent behavior is mediated by logical deliberation
- 2) What is the purpose of thinking? What thoughts **should** I have out of all the thoughts (logical or otherwise) that I **could** have?

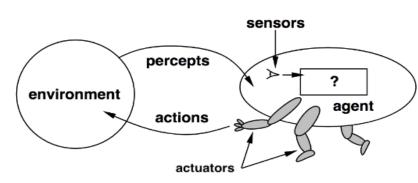


(4) Acting Rationally – Intelligent Agent

Rational Behavior – doing the right thing

The right thing: that which is expected to maximize goal achievement, given the available information

Doesn't necessarily involve thinking—e.g., blinking reflex—but thinking should be in the service of rational action



Rational Agent

An agent is a function from percept histories to actions:

$$f: \mathcal{P}^* \to \mathcal{A}$$

For any given class of environments and tasks, we seek the agent (or class of agents) with the best performance

Lecture Overview

- Course Introduction
 - Objectives & assessment methods
- What is AI

 Definitions & approaches
- History of AI
 From 1950 to present
- Current AI Applications

 Current AI Applications
- Summary & Next steps
 Summary & Next Steps

The Foundations of AI

Philosophy logic, methods of reasoning

mind as physical system

foundations of learning, language, rationality

Mathematics formal representation and proof

algorithms, computation, (un)decidability, (in)tractability

probability

Psychology adaptation

phenomena of perception and motor control

experimental techniques (psychophysics, etc.)

Economics formal theory of rational decisions

Linguistics knowledge representation

grammar

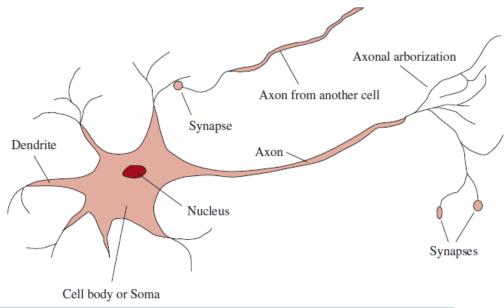
Neuroscience plastic physical substrate for mental activity

Control theory homeostatic systems, stability

simple optimal agent designs

The Foundations of AI

Neuroscience vs Deep Learning



	Supercomputer	Personal Computer	Human Brain
Computational units	10 ⁶ GPUs + CPUs	8 CPU cores	10 ⁶ columns
Ct	10 ¹⁵ transistors	10 ¹⁰ transistors	10 ¹¹ neurons
Storage units	10 ¹⁶ bytes RAM 10 ¹⁷ bytes disk	10 ¹⁰ bytes RAM 10 ¹² bytes disk	10 ¹¹ neurons 10 ¹⁴ synapses
Cycle time	$10^{-9} { m sec}$	$10^{-9} { m sec}$	10^{-3} sec^{-1}
Operations/sec	10^{18}	10^{10}	10^{17}

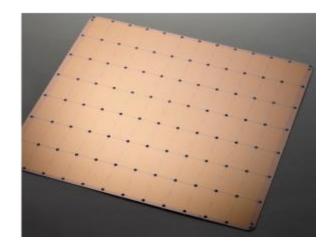
The Foundations of AI

Computer Engineering

- Moore's law: performance doubled every 18 months (1960~2012)
- GPU, TPU, WSE: performance doubled every 100 days (2012~)
 - > a full day training in 2014 \rightarrow 2 minutes training in 2018







- Quantum Computing
 - > far greater accelerations

The gestation (1943~1956)

- 1943, MeCulloch & Pitts, model of neurons
- 1946, Hebb, Hebbian learning for updating the connection strengths
- 1950, Alan Turing, Turing Test, machine learning, reinforcement learning
- 1956, McCarthy et al., Dartmouth College, 2-months Workshop
 - > birth of 'Artificial Intelligence'





Al@50 conference (2005) :

Trenchard More, John McCarthy, Marvin Minsky, Oliver SelfridgeRay Solomonoff

Early enthusiasm, great expectations (1952~1969)

- Logic Theroist, General Problem Solver, Playing checkers
- McCarthy: LISP(1958), Time sharing(1959), Advice Taker
- Minsky: Microworld problmes, algebra story problems, blocks world
- Rosenblatt: Perceptron(1962)
- Block et al., Perceptron Convergence Theorem (1962)

A dose of reality (1966~1973)

- Friedberg, Machine evolution (genetic programming)
- Sputnix(1966), Translate Russian scientific paper
- Minsky(1969), Perceptron, limitation of perceptrons

Expert Systems (1969~1986)

- DNEDRAL, MYCIN
- Prolog(1972)
- Minsky, Frames and OOP
- Japan(1981), 5th Generation Computer, 10-year project for a prolog machine
- McDermott(1982), R1, the first successful commercial expert system

MYCIN

Have you obtained positive cultures?

Yes.

What type of infection is it?

Primary bacteremia.

When did the symptomps first appear?

May 5

recommend gentamycinn using a doze of

The return of neural networks (1986~present)

- Rummelhart and McClelland (1986), Back-propagation learning algorithm
- Multi-layer Perceptrons
- complements the symbolic approaches

Probabilistic reasoning and machine learning (1987~present)

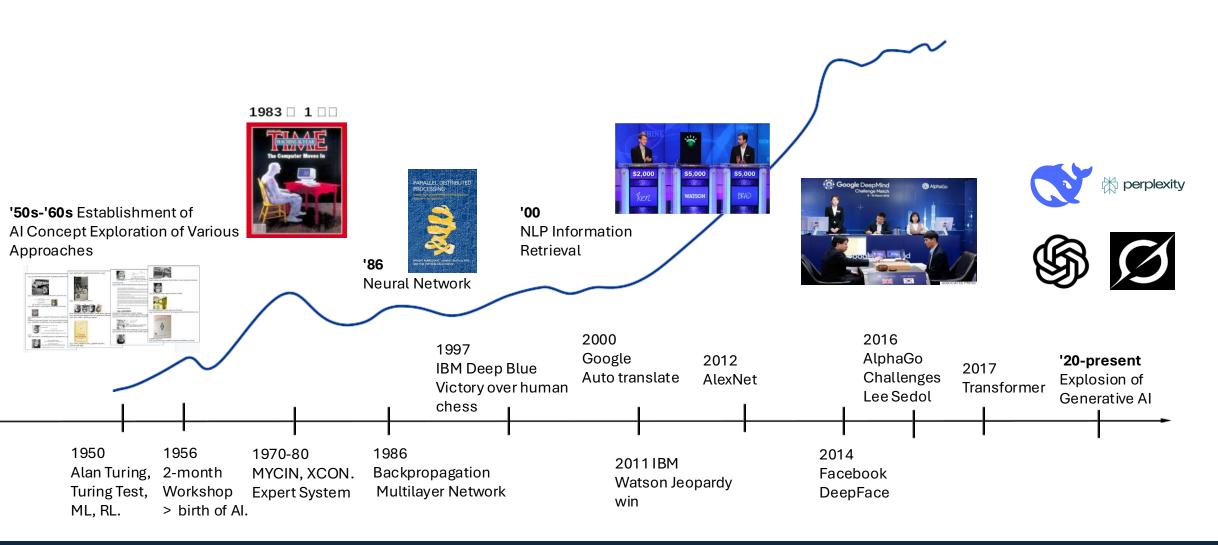
- Hidden Markov models, 1980s
- Bayesian networks, J. Pearl

Big data (2001~present)

- Advances in computing power and IoT's with WWW → Big data
- IBM Watson, winner of Jeopardy (2011)

Deep learning (2011~present)

- Convolutional Neural Networks, LeCun(1995)
- Winner of ImageNet challenge J. Hinton(2012)
- AlphaGo, Silver et al. (2016)
- GPU, TPU, FPGA



Lecture Overview

- Course Introduction
 - Objectives & assessment methods
- What is AI

 Definitions & approaches
- History of AI
 From 1950 to present
- Current AI Applications

 Current AI Applications
- Summary & Next steps
 Summary & Next Steps



Generative Al

GPT-4o, Gemini 2.5, advanced multimodal creation



Agentic Al

Autonomous Al agents for complex workflows



Autonomous Systems

Self-driving cars, robotics, drones



Healthcare Al

>90% diagnostic accuracy, drug discovery



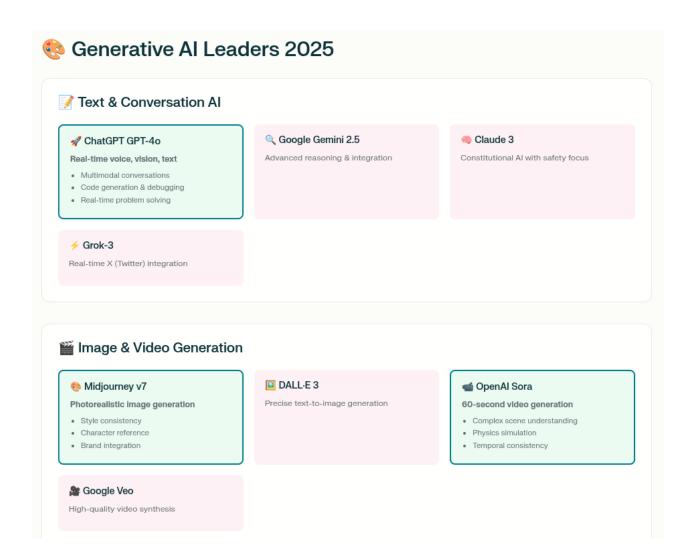
Al for Productivity

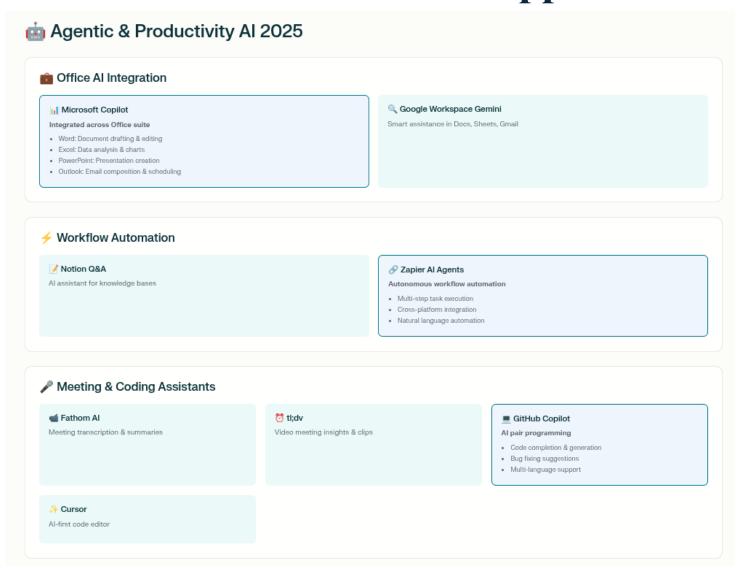
Microsoft Copilot, Al assistants everywhere

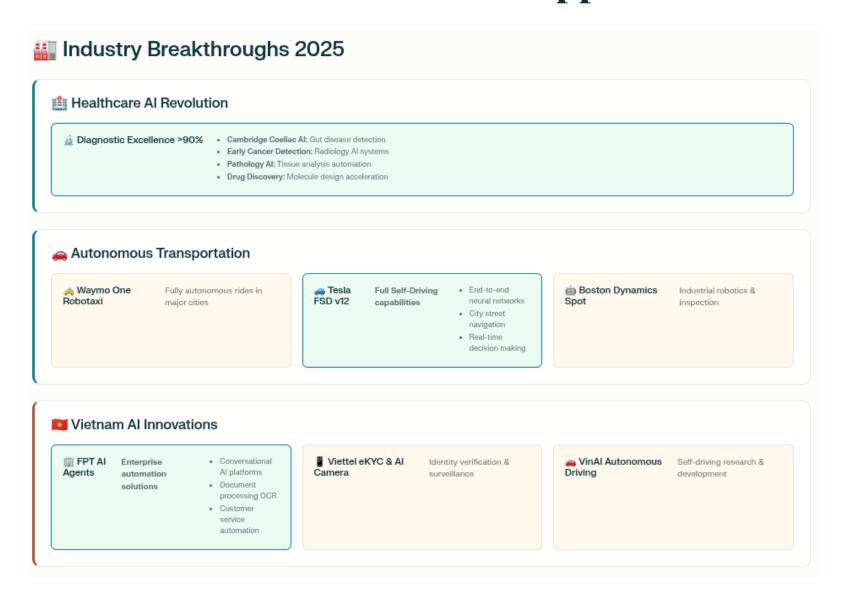


Vietnam Spotlight

FPT AI, VinAI, local innovations







Group Discussion: "Al Around Us"

Instructions:

- 1. Form groups of 4-5 students
- 2. Choose one discussion topic
- 3. Discuss for 10 minutes
- 4. Present key insights (3 minutes per group)

ig Generative Al Impact

- · How has ChatGPT/Al changed your daily life?
- · Which Al tool do you find most valuable?
- · What are the risks of Al dependency?

Future of Human-Al Collaboration

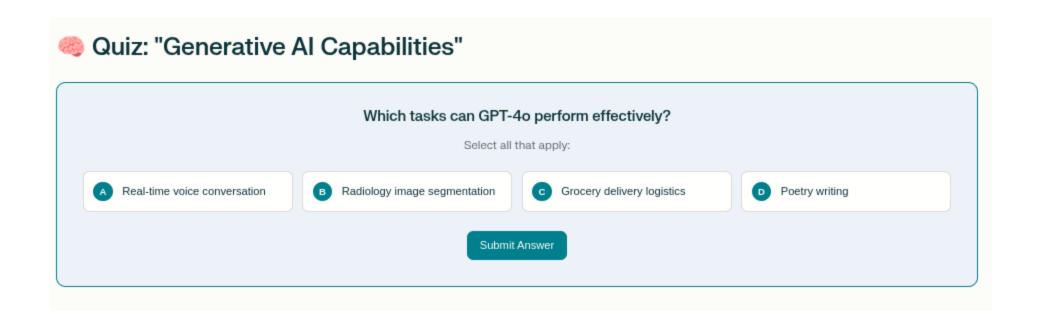
- · Which jobs will Al enhance vs replace?
- · How to stay relevant in an Al world?
- · What skills should students focus on?

Al in Vietnamese Context

- What Vietnamese Al products do you use?
- · How can Vietnam compete globally in Al?
- · What Al careers exist in Vietnam?

Al Ethics & Society

- Should Al replace human teachers/doctors?
- · How to prevent AI bias and misinformation?
- · What rules should govern Al development?



Intro to AI Trong-Nghia Nguyen

31

Lecture Overview

- Course Introduction
 - Objectives & assessment methods
- What is AI

 Definitions & approaches
- History of AI
 From 1950 to present
- Current AI Applications

 Current AI Applications
- Summary & Next steps
 Summary & Next Steps

Summary of key points



Al Definition

Four approaches: Acting/Thinking Humanly vs Rationally



Historical Evolution

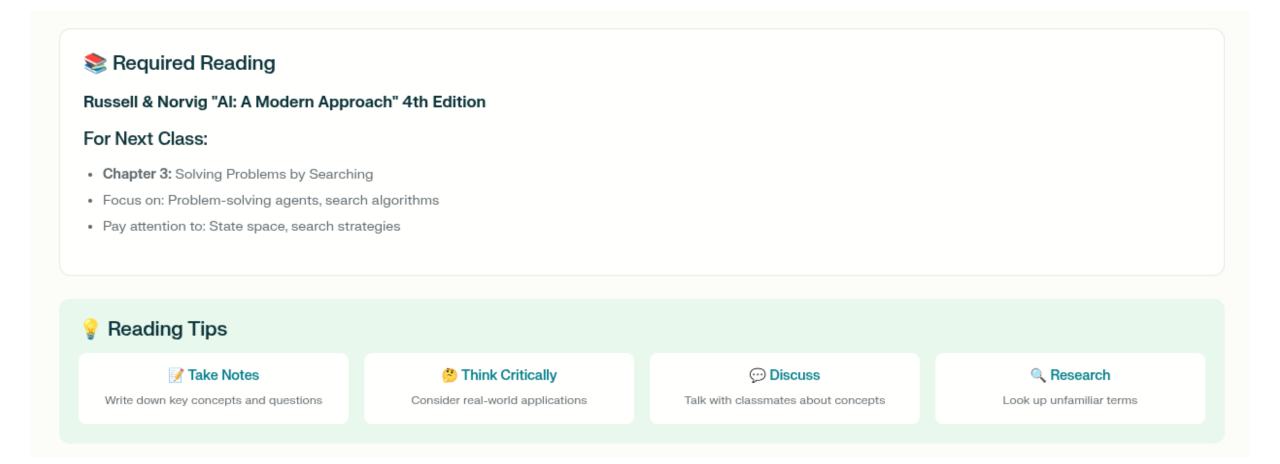
From symbolic AI to modern generative AI breakthroughs



2025 Al Landscape

GPT-4o, autonomous systems, 90%+ diagnostic accuracy

Reading Assignment



Intro to AI Trong-Nghia Nguyen

34

Thank you!

You're now ready to explore the exciting world of AI! Next Lecture: Search and Problem Solving