

# Overview of AI

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read: Ch. 1

# What is AI?

- popular view
  - based on movies, sci-fi...
  - Hal from 2001; Data from Star Trek NG; Spielberg's AI; Ex Machina...
- landmark AI systems
  - Deep Blue (chess), AlphaGo
  - IBM Watson (general question-answering)
  - Siri, Alexa, ChatGPT
- real industrial AI applications
  - robotics, autonomous vehicles
  - mail sorting, OCR; face recognition; speech2text, translation...
  - medical diagnosis, chest radiography, fraud detection, legal precedent finding, drug design...
  - stock trading/portfolio risk management/economic modeling/quant funds...

# Intelligence

- What is Intelligence?
  - difficult to define
  - the thing that makes humans unique. what about animals, dogs, dolphins...?
- some properties of intelligent behavior
  - doing more than just reacting to the environment
  - acting without being programmed (autonomous decision-making)
  - having volition/motivation/intent
    - does this require free-will? consciousness? self-awareness?
  - learning from experience, adaptiveness
- What do “IQ” tests measure?
  - there are many dimensions to intelligence, not just factual knowledge or reasoning ability

# Aspects of Intelligence

- *thinking vs. doing* distinction
  - are outcomes all that matter, or is how we “think” about it relevant?
- rationality – doing the “right” or “optimal” thing
  - instead of just “reacting” to the environment
  - implies there are goals/objectives

<b>Thinking ideally (rationally)</b> <ul style="list-style-type: none"><li>• did we weigh all the pros and cons?</li><li>• did we make the right inference?</li><li>• did we do it for the right reasons?</li></ul>	<b>Acting ideally (rationally)</b> <ul style="list-style-type: none"><li>• if we can quantify goals and costs,</li><li>• we can define optimal actions</li><li>• usually NP-hard</li></ul>
<b>Thinking humanly</b> <ul style="list-style-type: none"><li>• can we emulate decisions a human would make?</li><li>• practical choices in face of incomplete info.</li><li>• humans make mistakes, are biased by emotions</li></ul>	<b>Acting humanly</b> <ul style="list-style-type: none"><li>• play chess like a human (grandmaster)</li><li>• think about “good” driving</li><li>• are humans always the best measure?</li></ul>

# Perspectives on AI

- People view the *goals* of AI from different perspectives:
  - **Philosophical Perspective**
    - What is intelligence? Is it even possible for a machine to be intelligent?
  - **Psychological Perspective**
    - AI should be about creating programs that emulate human cognition.
  - **Engineering Perspective**
    - Forget the theoretical questions - this is practical!
    - How do we build more sophisticated systems for solving more complex problems? (like autonomous driving)
    - AI is all about advanced algorithms, complexity reduction, and optimization

# 1. Philosophical Perspective on AI

- attempts to define “correct thought” have a long history
  - Aristotle, Plato...Descartes, Kant, Carnap...Russell, Wittgenstein, Frege...
  - innate vs. empirical knowledge
- syllogisms – laws of thought, or patterns of correct thought
  - Categorical syllogisms: All A are B; x is A; therefore x is B.
    - *Socrates is a man; men are mortal; therefore, Socrates is mortal.*
  - Conditional syllogisms:
    - *Only people with good grades get into college; Kate is in college; Kate must have gotten good grades.*
  - Disjunctive syllogisms:
    - *All living things are plants or animals; Snoopy is not a plant; therefore, Snoopy is an animal.*
  - examples of a fallacy:
    - *\* Insects crawl; Charlotte crawls; therefore, Charlotte is an insect.*
    - [incorrect - she is a spider, which is an arachnid, not an insect]
    - *\* Whenever it rains, the ground gets wet. The ground is wet. Therefore, it must have rained.*
    - [incorrect - there could have been other reasons. This is an example of Abductive reasoning]

# 1. Philosophical Perspective on AI

- grounded understanding – mind/body problem (Descarte)
  - can you have an “isolated” intelligence, or must be connected to sensors?
- consciousness
  - is consciousness necessary?
  - what is it? introspection? self-awareness?

# 1. Philosophical Perspective on AI

- Turing Test
  - proposed by computer scientist Alan Turing (1950, CACM) Can Machines Think?
  - how it works
    - 15 min conversation via terminal (blinded)
    - restricted to a topic/subject, like pets, or gardening (but not current events)
    - human assessors judge whether human or computer
    - statistical criterion – pass if can fool >50% of judges – indistinguishable from humans

# 1. Philosophical Perspective on AI

- Turing Test (proposed by computer scientist Alan Turing)
  - advantages:
    - Integrates reasoning and understanding and communication
    - Empirical evaluation based on performance (doing): doesn't matter if it is "conscious", or how it is implemented internally
    - Comparison to the "average" intelligence of humans (by averaging against multiple decoys, avoiding individual differences in intelligence)
    - Not "subjective", in the sense that it does not depend on a single judge's opinion (averaged over multiple judges)
  - disadvantages:
    - non-conversational aspects of intelligence – recognizing objects visually, tying shoes...
    - does computer have delays answers and make mistakes to emulate humans? does it have to appreciate humor? analogies?

# 1. Philosophical Perspective on AI

- Turing Test (proposed by computer scientist Alan Turing)
  - Loebner competition – annual competition, \$100,000 prize
  - no true winner as of 2019
  - chatbots
    - Eliza – emulates a psychiatrist; pattern-based: “Tell me more about X...” “How did that make you feel?”
    - Eugene Goostman – emulates a 13-yr old Ukrainian boy (which induces people to forgive his conversational mistakes)

# Eliza transcript

- ELIZA: Please tell me what's been bothering you.
- USER: I feel happy to be writing right now, to be typing this as we speak. Lately, I've had writer's block.
- ELIZA: Can you elaborate on that ?
- USER: I only really feel happy when I'm writing.
- ELIZA: How long have you been writing ?
- USER: Always, but I've been writing seriously since I was eighteen.
- ELIZA: What incident are you thinking of ?
- USER: Um.
- ELIZA: What does that suggest to you ?
- USER: I don't think it's enough, to only be happy when you write.
- ELIZA: Does that trouble you ?
- USER: ...Um, because you can't write all the time. And then there's the real world, and you still have to deal with real people.
- ELIZA: How does that make you feel ?

# 1. Philosophical Perspective on AI

- Can “intelligence” be a *computable* function? (posed by Bertrand Russell)
  - think of mapping from “situations” to “actions”
  - (remember – computable functions are expressible as finite combinations of simpler functions using operators like addition, comparison, composition, recursion...)
  - the answer is not clear, but it brings out issues of role of “free will”, etc
- Does intelligence rely on quantum effects? (Roger Penrose)
  - some things are physically unknowable, like joint position and momentum of an electron
  - could the (human) brain be exploiting uncertainty in states of water molecules?

## 2. Psychological Perspective on AI

- How do humans think?
  - Humans are our best examples of intelligent beings
  - Can we write programs that think/act like humans?
- Example: Natural Language Understanding
  - How do humans parse and interpret sentences?
  - How do we turn sentences into meaning?
  - How can we correct for mistakes? (grammatical, spelling...)
  - How do we resolve ambiguities?
    - I saw the man on the hill with the hat.
    - I saw the man on the hill with the tree.
    - I saw the man on the hill with the telescope.
  - How do we fill-in missing/foreign words, or recognize a pun?
  - How do children learn to speak (language acquisition)?
  - When/why do we commit a slip-of-the-tongue, or a faux pas?

Another example:  
Should chess programs be designed to think and choose moves “like a human”?

## 2. Psychological Perspective on AI

- Humans are great at ...
  - making complex decisions in the face of uncertainty (situation assessment; multi-objectives; deadlines, costs, tradeoffs; ethical conflicts...)
  - interpreting perceptual input (recognizing a tiger, or the sound of a tiger), even if there is interference/noise, occlusions, shadows...
  - using context to resolve ambiguities
  - reasoning about other humans' intentions
  - discovering unexpected solutions (“Eureka” effect) (creativity – Douglas Hofstadter)
- But humans also have limitations:
  - humans make mistakes, and are subject to fatigue
  - limits on short-term memory; calculation speed; attention (distraction)
  - decisions can be influenced by emotions
  - biases: recency bias, expectation/confirmation bias
  - long-term memory is malleable, subjective, incomplete
  - decisions are often sub-optimal (satisficing=“good enough”)

## 2. Psychological Perspective on AI

- Paradigms: Behaviorism vs. Information Processing
  - behaviorism: stimulus-response
  - Info Proc: can form internal representations of the world based on sensory input, reason about it, make predictions and decisions
- Cognitive Science
  - concept representations; inter-connections (links) among concepts
  - memory (models of short-term and long-term memory); activation and retrieval
  - focus of attention
  - mental imagery

## 2. Psychological Perspective on AI

- Neuroscience
  - Brains have inherent parallelism
  - Cognitive processes like perception, memory, and reasoning operate in parallel
  - Distributed nature (and plasticity) makes brains fault-tolerant
- The mind compensates for missing info, and constantly tries multiple ways to interpret and explain the data it senses
- Psychological models of Consciousness
  - debate among modern philosophers John Searles, Daniel Dennett
  - can it be explained as an “illusion” of perceiving stream of perceptual states?

## 2. Psychological Perspective on AI

- (Major paradigms for intelligence and problem-solving)
- **Symbolic AI** (John McCarthy; Herb Simon; Alan Newell)
  - **Symbol Systems Hypothesis** – thoughts are like discrete abstract tokens that relate to and generate other tokens (e.g. by reminding or classification or inference)
  - GPS (General Problem Solver) – algebra problems, checkers, question-answering...
  - ACT (Jon Anderson) – psychology-based simulation of memory and rules (production system) that simulates human subject task performance
  - SOAR (Alan Newell) – Universal Subgoaling (when stuck, humans create a new search space to figure out how to get unstuck)

## 2. Psychological Perspective on AI

- (Major paradigms for intelligence and problem-solving)
- **Symbolic AI** (John McCarthy; Herb Simon; Alan Newell)
  - **Symbol Systems Hypothesis**
  - Is this the (only) path to truly intelligent behavior?
  - strengths: tractable for computing (world idealized as discrete states; can simulate reasoning with symbolic rules)
  - weaknesses:
    1. uncertainty/partial matching is difficult to deal with;
    2. need to program a huge network of concepts and rules to do anything productive (“knowledge engineering”);
    3. brittleness

## 2. Psychological Perspective on AI

- (Major paradigms for intelligence and problem-solving)
- **Connectionism** (Marvin Minsky)
  - Perceptrons, Neural Networks, Deep Networks
    - concepts are distributed as weights throughout network
    - output nodes are 'activated' collectively from connections among many neurons
    - weights are tuned through prior experience
  - Is this the (only) path to truly intelligent behavior?
  - strengths:
    - ideal for perceptual and motor-control tasks
    - all you need is training examples
  - weakness:
    - can't "explain" why decisions are made (because concepts are distributed as weights throughout network)

### 3. Engineering Perspective on AI

- what do we need to built programs that solve complex problems?
- need more sophisticated *algorithms* in our toolbox
- performance is everything; it doesn't matter how it is achieved (thinking ideally or like a human is irrelevant; consciousness is not required)
- task-focused intelligence; “wholistic” (integrated) AI is not necessary

# 3. Engineering Perspective on AI

- Core Concepts
  - Search Algorithms
  - Knowledge Representation and Inference
  - Machine Learning

# 3. Engineering Perspective on AI

- Search
  - almost everything in AI reduces to a Search problem (in some space)
    - problem solving, planning, learning, parsing, automated deduction...
  - we will study a variety of Search Algorithms
  - especially Heuristic Search (A\*)

# 3. Engineering Perspective on AI

- Edward Feigenbaum: “In the knowledge is the power.”
- Knowledge-based programming
  - rule bases
  - declarative programming vs. procedural programming: say “what” not “how”
- Knowledge Representation
  - formal logic(s) (propositional/Boolean, predicate/First-Order logic, epistemic logics, temporal logics, default logics, fuzzy logic...)
  - ontologies for qualitative reasoning about events, time, spatial relationships, processes...
  - challenge: how to express uncertainty? probability, confidence factors...

# 3. Engineering Perspective on AI

- Inference Algorithms
  - how can you generate deductions from a knowledge base?
- Is an action or decision a logical consequence of what I know?
- How to generate proofs?
- Is a formula satisfiable? Is there a faster way than try all combinations of variables?
  
- (Automated inference has diverse applications to reasoning about mathematics, proving theorems in number/set/group theory, and computer program verification.)

# 3. Engineering Perspective on AI

- *Expertise* in high-performing humans
  - chess grandmasters; military generals; football coaches; air-traffic controllers; 5-star chefs...
  - typically accumulated from experience (hence, think: knowledge base)
- Expert Systems
  - example: MYCIN – asks questions to diagnose blood diseases and infections
  - commonly used ES shells: PROLOG, CLIPS, JESS
  - can often generate explanations of decisions
- Major limitation of ES: *brittleness*
  - ES often draw inaccurate/silly conclusions outside scope of what they were designed for

# 3. Engineering Perspective on AI

- Machine Learning
  - who says machines can't be adaptive and learn from experience?!
  - optimizing performance for an objective function
  - many learning methods:
    - statistics (regression)
    - concept generalization, features, classification trees
    - rule revision (e.g. adding antecedents, special conditions)
    - clustering, pattern recognition
    - neural networks
    - reinforcement learning: sequences of actions to achieve goals in dynamic environments
  - to develop/build/program/implement a system, you need to *train* it on *examples*

# 3. Engineering Perspective on AI

- Complexity
  - we will see that most “AI” problems are computationally difficult
    - often NP-complete or worse
    - huge search spaces (exponential or infinite)
  - much of AI focuses on using knowledge and heuristics to reduce complexity and find an acceptable solution (often sacrificing completeness/optimalty)
  - this connects AI to *optimization*
    - greedy algorithms; approximation algorithms
    - control theory; operations research