

What is Intelligence?	Artificial Intelligence
 Webster says: The capacity to acquire and apply knowledge. The faculty of thought and reason. 	Studies how to achieve intelligent behavior through computational means. This makes AI a branch of Computer Science
 What features/abilities do humans (animals/animate objects) have that you think are indicative or characteristic of intelligence? 	Why do we think that intelligence can be captured through computation? Modeling the processing that our brains do as computation has proved to be successful. Hence, human intelligence can arguably be best modeled
 Abstract concepts, mathematics, language, problem solving, memory, logical reasoning, planning ahead, emotions, morality, ability to learn/adapt, etc 	as a computational process.
SC384, University of Toronto 5	CSC384, University of Toronto
Classical Test of (Human) Intelligence	Human Intelligence
 The Turing Test: A human interrogator. Communicates with a hidden subject that is either a computer system or a human. If the human interrogator cannot reliably decide whether or not the subject is a computer, the computer is said to have passed the Turing test. 	 Turing provided some very persuasive arguments that a system passing the Turing test <i>is intelligent</i>. We can only really say it <i>behaves like a human</i> Nothing guarantees that it thinks like a human
• Weak Turing type tests:	

• The Turing test does not provide much traction on the question of how to actually build an intelligent system.

7

CSC384, University of Toronto

overlooks

See Luis von Ahn, Manuel Blum, Nicholas Hopper, and John Langford.

CAPTCHA: Using Hard AI Problems for Security. In Eurocrypt.

Type the two words:

inquiry

Human Intelligence

- Recently some claims have been made of Al systems that can pass the Turing Test.
- However, these systems operate on subterfuge, and were able to convince a rather naïve jury that they were human like.
- The main technique used is obfuscation...rather than answering questions the system changed the topic!
- This is not what Turing described in his Turing Test CSC384, University of Toronto

Human Intelligence

- But more importantly, we know very little about how the human brain performs its higher level processes. Hence, this point of view provides very little information from which a scientific understanding of these processes can be built.
- Nevertheless, Neuroscience has been very influential in some areas of AI. For example, in robotic sensing, vision processing, etc.
- Humans might not be best comparison?
 - Don't always make the best decisions
- Computer intelligence can aid in our decision making
 CSC384, University of Toronto

```
Human Intelligence
```

- In general there are various reasons why trying to mimic humans might **not** be the best approach to AI:
 - Computers and Humans have a very different architecture with quite different abilities.
 - Numerical computations
 - Visual and sensory processing
 - Massively and slow parallel vs. fast serial

	Computer	Human Brain	
Computational Units	8 CPUs, 10 ¹⁰ gates	10 ¹¹ neurons	
Storage Units	10 ¹⁰ bits RAM	10 ¹¹ neurons	
	10 ¹³ bits disk	10 ¹⁴ synapses	
Cycle time	10 ⁻⁹ sec	10 ⁻³ sec	
Bandwidth	10 ¹⁰ bits/sec	10 ¹⁴ bits/sec	
Memory updates/sec	10 ¹⁰	10 ¹⁴	

CSC384, University of Toronto

9

11

10

Rationality

- The alternative approach relies on the notion of **rationality**.
- Typically this is a precise formal notion of what it means to do the right thing in any particular circumstance. Provides
 - A precise mechanism for analyzing and understanding the properties of this ideal behavior we are trying to achieve.
 - A precise benchmark against which we can measure the behavior the systems we build.

Rationality

- Formal characterizations of rationality have come from diverse areas like logic (laws of thought) and economics (utility theory—how best to act under uncertainty, game theory how self-interested agents interact).
- There is no universal agreement about which notion of rationality is best, but since these notions are precise we can study them and give exact characterizations of their properties, good and bad.

Four AI Definitions by Russell + Norvig

Not necessarily like humans

• We'll focus on acting rationally

Like humans

Systems that think like

- this has implications for thinking/reasoning

	 	 	3	

• Al tries to understand and model intelligence as a computational process.

Computational Intelligence

- Thus we try to construct systems whose computation achieves or approximates the desired notion of rationality.
- Hence AI is part of Computer Science.
 - Other areas interested in the study of intelligence lie in other areas or study, e.g., cognitive science which focuses on human intelligence. Such areas are very related, but their central focus tends to be different.

CSC384, University of Toronto

Subareas of Al

- Perception: vision, speech understanding, etc.
- Machine Learning, Neural networks
- Robotics
- Natural language processing
- Reasoning and decision making our focus
 - Knowledge representation
 - Reasoning (logical, probabilistic)
 - **Decision making** (search, planning, decision theory)

humans Systems that think rationally

F			
Act	Systems that act like humans	Systems that act rationally Our focus	
	1		
	Cognitive Science		

CSC384, University of Toronto

Subareas of Al	Subareas of Al
• Many of the popular recent applications of AI in industry have been based on Machine Learning, e.g., voice recognition systems on	 Nor will we discuss Computer Vision nor Natural Language to any significant extent.
your cell phone.	 All of these areas have developed a number of specialized theories and methods specific to the problems they study.
• We will not say much in this course about machine learning, although the last part of the course will introduce Bayes Nets a form of probabilistic graphical model.	 The topics we will study here are fundamental techniques used in various AI systems, and often appear in advanced research in many other sub-areas of AI.
Probabilistic graphical models are fundamental in machine learning.	 In short, what we cover here is not sufficient for a deep understanding of AI, but it is a good start.
SC384, University of Toronto 17	CSC384, University of Toronto
Further Courses in AI	CSC384, University of Toronto What We Cover in CSC384
Further Courses in AI Perception: vision, speech understanding, etc.	What We Cover in CSC384
Further Courses in Al Perception: vision, speech understanding, etc. – CSC487H1 "Computational Vision"	•Search (Chapter 3, 5, 6)
Further Courses in Al Perception: vision, speech understanding, etc. - CSC487H1 "Computational Vision" - CSC420H1 "Introduction to Image Understanding"	What We Cover in CSC384
Further Courses in Al Perception: vision, speech understanding, etc. - CSC487H1 "Computational Vision" - CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks - CSC321H "Introduction to Neural Networks and Machine Learning" - CSC411H "Machine Learning and Data Mining"	What We Cover in CSC384 • Search (Chapter 3, 5, 6) – Uninformed Search (3.4)
Further Courses in Al Perception: vision, speech understanding, etc. - CSC487H1 "Computational Vision" - CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks - CSC321H "Introduction to Neural Networks and Machine Learning" - CSC411H "Machine Learning and Data Mining" - CSC412H1 "Uncertainty and Learning in Artificial Intelligence"	What We Cover in CSC384 •Search (Chapter 3, 5, 6) – Uninformed Search (3.4) – Heuristic Search (3.5, 3.6)
Further Courses in Al Perception: vision, speech understanding, etc. - CSC487H1 "Computational Vision" - CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks - CSC321H "Introduction to Neural Networks and Machine Learning" - CSC411H "Machine Learning and Data Mining" - CSC412H1 "Uncertainty and Learning in Artificial Intelligence" Robotics - Engineering courses	What We Cover in CSC384 •Search (Chapter 3, 5, 6) – Uninformed Search (3.4) – Heuristic Search (3.5, 3.6) – Game Tree Search (5)
 Further Courses in Al Perception: vision, speech understanding, etc. CSC487H1 "Computational Vision" CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks CSC411H "Introduction to Neural Networks and Machine Learning" CSC411H "Machine Learning and Data Mining" CSC412H1 "Uncertainty and Learning in Artificial Intelligence" Robotics Engineering courses 	What We Cover in CSC384 •Search (Chapter 3, 5, 6) – Uninformed Search (3.4) – Heuristic Search (3.5, 3.6) – Game Tree Search (5) •Knowledge Representation (Chapter 8, 9)
 Further Courses in Al Perception: vision, speech understanding, etc. CSC487H1 "Computational Vision" CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks CSC411H "Introduction to Neural Networks and Machine Learning" CSC411H "Machine Learning and Data Mining" CSC412H1 "Uncertainty and Learning in Artificial Intelligence" Robotics Engineering courses 	What We Cover in CSC384 • Search (Chapter 3, 5, 6) – Uninformed Search (3.4) – Heuristic Search (3.5, 3.6) – Game Tree Search (5) • Knowledge Representation (Chapter 8, 9) – First order logic for more general knowledge (8)
 Further Courses in Al Perception: vision, speech understanding, etc. CSC487H1 "Computational Vision" CSC420H1 "Introduction to Image Understanding" Machine Learning, Neural networks CSC321H "Introduction to Neural Networks and Machine Learning" CSC411H "Machine Learning and Data Mining" CSC412H1 "Uncertainty and Learning in Artificial Intelligence" Robotics Engineering courses Natural language processing CSC401H1 "Natural Language Computing" 	What We Cover in CSC384 • Search (Chapter 3, 5, 6) – Uninformed Search (3.4) – Heuristic Search (3.5, 3.6) – Game Tree Search (5) • Knowledge Representation (Chapter 8, 9) – First order logic for more general knowledge (8)

CSC384, University of Toronto

CSC384, University of Toronto

What We Cover in CSC384		AI Successes
 Classical Planning (Chapter 10) Predicate representation of states Planning Algorithms Quantifying Uncertainty and Probabilistic Reasoning (Chapter 13, 14, 16) Uncertainties, Probabilities Probabilistic Reasoning, Bayesian Networks 		 Games: chess, checkers, poker, bridge, backgammon Search Physical skills: driving a car, flying a plane or helicopter, vacuuming Sensing, machine learning, planning, search, probabilistic reasoning Language: machine translation, speech recognition, character recognition, Knowledge representation, machine learning, probabilistic reasoning Vision: face recognition, face detection, digital photographic processing, motion tracking, Commerce and industry: page rank for searching, fraud detection, trading on financial markets Search, machine learning, probabilistic reasoning
CSC384, University of Toronto	21	CSC384, University of Toronto 22
Recent AI Successes		Degrees of Intelligence
 Darpa Grand Challenges Goal: build a fully autonomous car that can drive a 240 km course in the Mojave desert 2004: none went further than 12 km 2005: 5 finished 2007: Urban Challenge: 96 km urban course (former air force base) with obstacles, moving traffic, and traffic regulations: 6 finishers 2011: Google testing its autonomous car for over 150,000 km on real roads 2011: IBM Watson competing successfully against two Jeopardy grand-champions 		 Building an intelligent system as capable as humans remains an elusive goal. However, systems have been built which exhibit various specialized degrees of intelligence. Formalisms and algorithmic ideas have been identified as being useful in the construction of these "intelligent" systems. Together these formalisms and algorithms form the foundation of our attempt to understand intelligence as a computational process. In this course we will study some of these formalisms and see how they can be used to achieve various degrees of intelligence.
CSC384, University of Toronto	23	CSC384, University of Toronto 2

Readings

- 1.1: What is Al?
- 2: Intelligent Agents
- •Other interesting readings:
 - 1.2: Foundations
 - 1.3: History