

# INTRODUCTION TO ARTIFICIAL INTELLIGENCE

**ECE 47300, Spring 2023** 

CRN: 17248 (001)

**BHEE 170** 

**Instructor: David I. Inouye** 

Website: <a href="https://www.davidinouye.com/course/ece47300-spring-2023/">https://www.davidinouye.com/course/ece47300-spring-2023/</a>



#### WHAT IS ARTIFICIAL INTELLIGENCE?

- Merriam-Webster Dictionary
  - "a branch of computer science dealing with the simulation of intelligent behavior in computers"
- Oxford Dictionary
  - "the theory and development of computer systems able to perform tasks that normally require human intelligence, such as visual perception, speech recognition, decision-making, and translation between languages."



#### **POSSIBLY A.I. IS A MOVING TARGET**

- Are these A.I.?
  - Chess solvers
  - TurboTax
  - Chatbot
- What about these?
  - Speech recognition
  - Face recognition
  - Personalized recommendations
  - Self-driving cars
- "Al is anything that humans can do that computers cannot **yet** do."



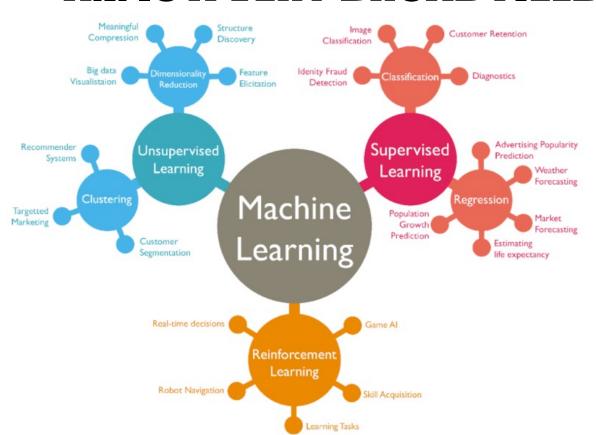
#### A MORE PRAGMATIC DEFINITION

"Al is that which appears in academic conferences on Al" \*

<sup>\*</sup> From slides by Prof. Zico Kolter at CMU: <a href="http://www.cs.cmu.edu/~./15780/slides/intro.pdf">http://www.cs.cmu.edu/~./15780/slides/intro.pdf</a>)



A.I. IS A VERY BROAD FIELD



Artificial Intelligence

Machine
Learning

Deep
Learning

#### Major applications

- Computer vision
- Natural language processing
- Robotics

Original credit unknown, retrieved from <a href="https://www.sharper.ai/taxonomy-ai/">https://www.sharper.ai/taxonomy-ai/</a>



### THIS COURSE WILL ONLY COVER A SMALL SET OF TOPICS

- 1. Introduction to artificial intelligence
- 2. Machine learning basics
- 3. Deep learning basics
- 4. Dimensionality reduction
- 5. Deep generative models
- 6. Natural language processing
- 7. Markov decision processes
- 8. Special topics



#### **OTHER RELATED CLASSES**

- ECE 57000: Artificial Intelligence by Prof. Inouye (cannot also take this course)
- ECE 60146: Deep Learning (spring semester) by Prof. Kak and Prof. Bouman
- ECE 50024: Machine Learning I (spring semester) by Prof. Qi Quo
- ECE 69500: Optimization in deep learning by Prof. Abolfazl Hashemi



#### **AGENDA**

- 1. Announcements
  - Emergency preparedness
- 2. Syllabus
- 3. Logistics
- 4. (time-permitting) Defining AI



#### **EMERGENCY PREPAREDNESS**

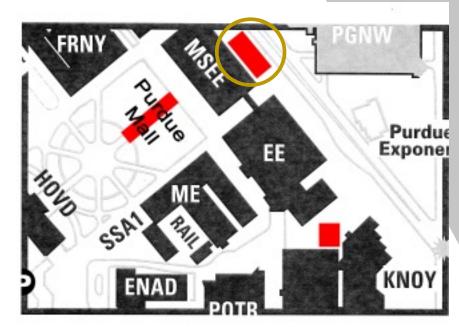
As we begin this semester, I want to take a few minutes and discuss emergency preparedness. Purdue University is a very safe campus and there is a low probability that a serious incident will occur here at Purdue. However, just as we receive a "safety briefing" each time we get on an aircraft, we want to emphasize our emergency procedures for evacuation and shelter-in-place incidents. Our preparedness will be critical IF an unexpected event occurs!



#### **EMERGENCY PREPAREDNESS**

Emergency preparedness is your personal responsibility. Purdue University is actively preparing for natural disasters or human-caused incidents with the ultimate goal of maintaining a safe and secure campus. Let's review the following procedure

- For any emergency text or call 911.
- There are more than 300 Emergency Telephones (aka blue lights) throughout campus that connect directly to the Purdue Police Department (PUPD). If you feel threatened or need help, push the button and you will be connected right away.
- If we hear a fire alarm, we will immediately evacuate the building and proceed to the grassy area north of the MSEE building. Do not use the elevator.





#### **EMERGENCY PREPAREDNESS**

• If we are notified of a Shelter in Place requirement for a **tornado** warning, active shooter, or hazardous waste, we will stop classroom or research activities and shelter in **this classroom**.



#### **SYLLABUS!**

All significant updates to the syllabus or course schedule will be posted on **Piazza**.

See **course website** for syllabus and schedule.



#### **LOGISTICS**

- Brightspace
  - <u>First time</u>: Click on Gradescope and Piazza links in Brightspace to connect Purdue account
- The course website will have all other necessary links and content



#### **NOW BACK TO AI...**



#### A MORE PRAGMATIC DEFINITION

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#### NOT "MOVIE" A.I.



C-3PO and R2D2 in Star Wars.



The rogue A.I. HAL9000 from the movie 2001: A Space Odyssey.



Skynet from Terminator.



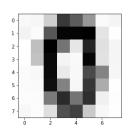
#### COMPUTERS DON'T "THINK" LIKE WE DO

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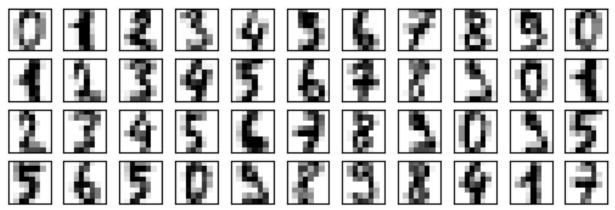
A matrix of numbers as the computer "sees". Do you know what this matrix represents?



### COMPUTERS DON'T "THINK" LIKE WE DO



The same matrix of numbers displayed as an image.



Other examples from this dataset.

Do you know what the numbers represent now?



### COMPUTERS DON'T "THINK" LIKE WE DO



 $+ .007 \, \times$ 

x
"panda"
57.7% confidence

 $\operatorname{sign}(
abla_{m{x}}J(m{ heta},m{x},y))$  "nematode"

8.2% confidence



 $x + \epsilon sign(\nabla_x J(\theta, x, y))$ "gibbon"

99.3 % confidence



Real gibbon...

<sup>\*</sup> From Explaining and Harnessing Adversarial Examples by Goodfellow et al.



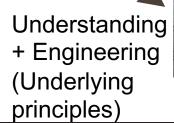
### NOR IS HUMAN IMITATION NECESSARILY THE GOAL

Consider flight









**Imitation** 







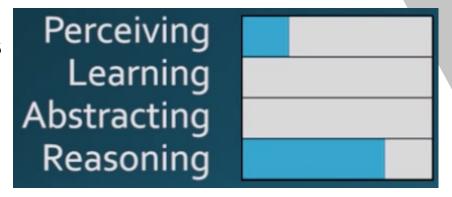
#### DARPA'S PERSPECTIVE ON AI

- First wave Handcrafted Knowledge
- Second wave Statistical Learning
- Third wave (future) Contextual Adaptation



# 1<sup>ST</sup> WAVE: HANDCRAFTED KNOWLEDGE

- Examples
  - Auto-scheduling systems
  - Chess
  - TurboTax
  - Simple medical diagnosis
- Good at
  - Logical reasoning from rules
- Bad at
  - Perceiving
  - Learning



Excellent DARPA video (16 min) on AI (content above based on this): <a href="https://www.youtube.com/watch?v=-001G3tSYpU">https://www.youtube.com/watch?v=-001G3tSYpU</a>



#### **ONE KEY PROBLEM:**

#### (FORMALLY) STATING WHY IS HARD BUT LABELING IS EASY

What is this a photo of?



https://unsplash.com/search/photos/cute-cat

- Why is this an image of a cat?
- Can you state a rule for all cats?



#### 2<sup>ND</sup> WAVE: STATISTICAL LEARNING

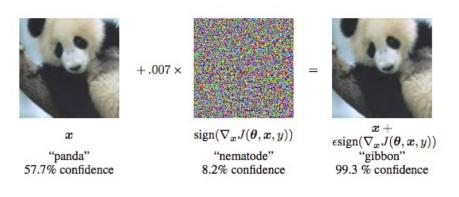
- Examples
  - Voice/face recognition
  - Personalized recommendations
- Good at
  - Perceiving
  - Learning
- Bad at
  - Logical reasoning
  - Robustness





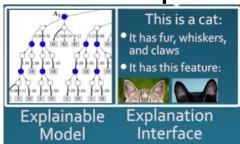
#### **KEY PROBLEMS WITH 2ND WAVE**

Lack of robustness / fragile systems





Lack of explanations



I understand why
I understand why not
I know when you'll succeed
I know when you'll fail
I know when to trust you

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# 3<sup>RD</sup> WAVE (FUTURE): CONTEXTUAL ADAPTATION

- Combination of previous two waves
- System will construct <u>explanatory models</u>
  - Causation
  - · Some abstraction
  - Explainable
- Examples
  - Incorporate handwriting knowledge for recognizing new characters
  - ? (Maybe you can work on this)
- Good at
  - Perceiving
  - Learning
  - Reasoning
- Slightly better at
  - Abstracting





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### MODELING, INFERENCE, AND LEARNING PARADIGM OF AI

- **1. Modeling** How do you model the real-world problem?
  - How should you represent the input to the model? (e.g.,
  - How to represent the **model functions**? (e.g., explicit, implicit, deep)
- **2. Inference** Given a model, how do you perform various tasks or answer questions based on the model?
  - How do you categorize photos?
  - How do you generate novel fake images?
  - How do you determine the next action to take to reach a goal?
- 3. Learning How do you train/adapt the models given experience or data?
  - Which objective function are you optimizing?
  - What are the learning constraints?
  - Which algorithm do you use to find the model parameters?



### QUESTIONSP