

University of Illinois at Urbana-Champaign
Dept. of Electrical and Computer Engineering

ECE 101: Exploring Digital Information Technologies for Non-Engineers Fall 2024

Exam 3 Review

Theme: Sense-Compute-Communicate-Actuate Loop

- **Drone delivery systems**
- **Self-driving cars (autonomous vehicles)**
- **Farm robots**
- **Digital assistants: Alexa, Siri, etc.**
- **Cleaning Robots: Roomba**
- **Smart Treadmill**
- **Smart water bottle**
- ...

Example: Drone Delivery System

- **Sense:**
 - Sensors for temperature, pressure, wind speed—environment conditions,
 - GPS—location
 - Camera, IMU
- **Compute:**
 - Process input—sensed data coming in as input,
 - Provide output—decision on what to do
- **Communicate:**
 - Send updates to and get instructions from control station
 - Communicate with other drones (if part of a fleet)
- **Actuate:**
 - Control rotors to land safely or hover—speed up/down, change directions etc.
 - Control mechanism to deposit package safely.

Example: Autonomous Vehicles

- **Sense:**
 - Camera
 - LiDAR
 - IMU, GPS, Microphone
- **Compute:**
 - Process input—
 - Provide output—
- **Communicate:**
 - With the human in the car
 - With a centralized control system
- **Actuate:**
 - Control engine, speed up/down, turn wheels, apply brakes, turn signal
 -

Types of Sensors

Types of sensors

1. **Cameras:** Regular cameras, IR, thermal, radar, Lidar
2. **Microphones:** audible, ultrasound
3. **IMU: Inertial Measurement Unit**
(accelerometer, gyroscope, magnetometer)
4. **Wireless:** GPS, Wifi (WiGig 60GHz, THz), UWB
5. **Assorted:** pressure, humidity, proximity, temperature, chemical traces

Basic working principle

- **Pin-hole camera**
- **Microphone**
- **LIDAR**
- **GPS**

Computer Vision: Basic Image Processing Techniques

- **Edge detection**
- **Image segmentation**
- **Feature extraction/registration**
- **Depth perception**

Computer Vision: Things to remember

- Active vs. passive sensors ... why do we need both?
- Sensor fusion
- Occlusion
- Safety envelope

Speech and NLP: Issues

- Context
- Ambiguity
- Sarcasm, irony, euphemism, metaphors
- Accents, dialects, pidgin languages etc.

Computing steps to understand speech

1. **get rid of noise**: other voices (unauthorized), music, television, video games, pets, and so forth.
2. perform “**voice recognition**”: translate an audio signal into a sequence of words.
3. understand what the human is trying to communicate: **process their natural language** (English, for example).

Computational tools for understanding speech

1. Context— search for keywords
2. Share information between hierarchy of interacting, probabilistic models
3. MLE - Maximum Likelihood Estimate: Given an observation,
 - **choose the explanation** that is
 - **most likely to produce the observation.**
4. Bayes' Theorem:

$$\text{Probability (A AND B)} = \text{Probability (A)} \times \text{Probability (B | A)}$$

The chance of **A and B both happening** is equal to the **product of the chance of A happening and the chance of B happening if A has happened.**

A game of probability

Pat will roll either

- one (six-sided) die or
- two dice and add up the numbers.

Then Pat tells us the amount rolled. Can we guess whether Pat rolled one or two dice?

Probability Pat rolls 1 die = $1/2$ (one of two options they have)

Probability Pat rolls 2 dice = $1/2$ (one of two options they have)

What is the chance that Pat rolled a 4 with 1 die? $1/6$

What is the chance that Pat rolled a 4 (total) with 2 dice? $3/36$ or $1/12$

(Could have been $1+3$, $2+2$, or $3+1$)

probability (Pat rolled one die) X probability (got a 4 | Pat rolled one die) = $1/2 \times 1/6 = 1/12$

probability (Pat rolled one die) X probability (got a 4 | Pat rolled one die) = $1/2 \times 1/12 = 1/24$

Initial Probabilities are Important to Correct Choices

What if Pat tells us more about how they choose to roll 1 or 2 dice?

probability (Pat rolls one die) = $\frac{1}{4}$ and

probability (Pat rolls two dice) = $\frac{3}{4}$

In that case, our guess changes, as

$\frac{1}{4} \cdot$ probability (got a 4 | Pat rolled one die)

$$= \frac{1}{4} \times \frac{1}{6} = \frac{1}{24}$$

<

$\frac{3}{4} \cdot$ probability (got a 4 | Pat rolled two dice)

$$= \frac{3}{4} \times \frac{1}{12} = \frac{1}{16}$$

How is MLE used?

Voice recognition answers the question, “**Given an audio input, what sequence of words was spoken?**”

A solution is generated by finding the sequence of words that is most likely to have generated the audio input.

Natural language processing answers the question, “**Given a sequence of words, what did the speaker want to communicate?**”

A solution is generated by finding the meaning that is most likely to have generated the sequence of words.

BERT: Bidirectional Encoder Representations from Transformers

Guessing Words Easier with Words on Both Sides

“I took my ... [more words].”

What’s the next word?

In 1953, journalists* realized that **the words AFTER the missing would help** in guessing.

“I took my ... for a walk.”

What’s the next word?

Dog, perhaps?

Could be lots of words, but dog may be a good MLE choice.

“Cat,” “snake,” “Ferrari,” maybe not so good.

BERT: Bidirectional Encoder Representations from Transformers

BERT (**B**idirectional **E**ncoder **R**epresentations from **T**ransformers)

- revolutionized how machines understand language
- **read text in both directions at once**, which helps it understand the context better (before BERT, many models read text in one direction i.e. left to right)
- first trained on a large amount of text (**pre-training**) to learn general language patterns
- **fine-tuned** to the specific task, like answering questions or sentiment analysis (versatility)

Large Language Models (LLMs)

GPT (Generative Pre-trained Transformer), developed by OpenAI.

1. Generative Model: While BERT is great at understanding and processing text, GPT is designed to generate text. This means it can write essays, create dialogue, and even compose poetry.

2. Transformer Architecture: Both BERT and GPT use a type of neural network called a Transformer, a neural network that learns context and thus meaning by tracking relationships in sequential data like the words in this sentence. GPT uses this architecture to predict the next word in a sentence, making it very good at generating coherent and contextually relevant text.

3. Pre-training on a Massive Scale: GPT models are pre-trained on a vast amount of text from the internet. This extensive training helps them generate more natural and diverse text.

Applications of NLP (Generic or Specific)

- **Text Classification:** Categorizing text into predefined categories.
 - Did the reviewer like a movie?
 - What sentiment did they express?
- **Interpretation:**
 - Is anything in a patient's electronic medical records relevant to the patient's current symptoms?
- **Named Entity Recognition (NER)**
Identifying and classifying entities like names, dates, and locations in text.
- **Question Answering:** Finding answers to questions within a text
- **Text Generation:** Creating new text based on a given prompt.
- **Dialogue Systems:** Powering chatbots and virtual assistants.
- **Creative Writing:** Generating stories, poems, and other creative content.

AR, VR, MR and ER

- **AR - Augmented reality**, designed to **add digital elements over real-world views** with limited interaction.
- **VR - Virtual reality** immersive experiences helping to **isolate users from the real world**, usually via a headset device and headphones designed for such activities.
- **MR - Mixed reality** combining AR and VR elements so that **digital objects can interact with the real world**, means businesses can design elements anchored within a real environment.
- **ER -Extended reality (XR)**— covering all types of technologies that **enhance our senses**, including the three types previously mentioned.

AR/VR Applications

- **Ones discussed in class**
- **New ones**

AR/VR Helpers

- **Motion Capture**

- With markers and marker less

- Applications

- **Haptics**

- Whats is it?

- Applications

Autonomous Driving: Major Technologies

- computer (robot) vision,
- sensor fusion, and
- machine learning

Autonomous Driving: Issues

- Unlike many machine learning applications, we have **relatively little** of the most important types of **data** for training.
 - It's **neither easy nor cheap** to stage a potential accident to make sure that autonomous drivers “learn” to avoid them.
 - An **autonomous vehicle must be able to respond** to rare events **safely**
- **Failed** image recognition
 - ML Models Can be Brittle
 - ML models can fall prey to adversarial attacks leading to

Autonomous Driving: Data Scarcity Solution

- Sophisticated **simulations that can generate sensor data** for a range of physically realistic situations in order to train the ML models needed to drive safely. (Computer games for computers)
- Collect more data (millions of Teslas on the road)

Autonomous Driving: Actuation

- **Examples** of how actuation needs of a self driving car lend themselves to great application of computation
 - 3 point turn
 - Driving on different surfaces
 - Calculation of stopping distance

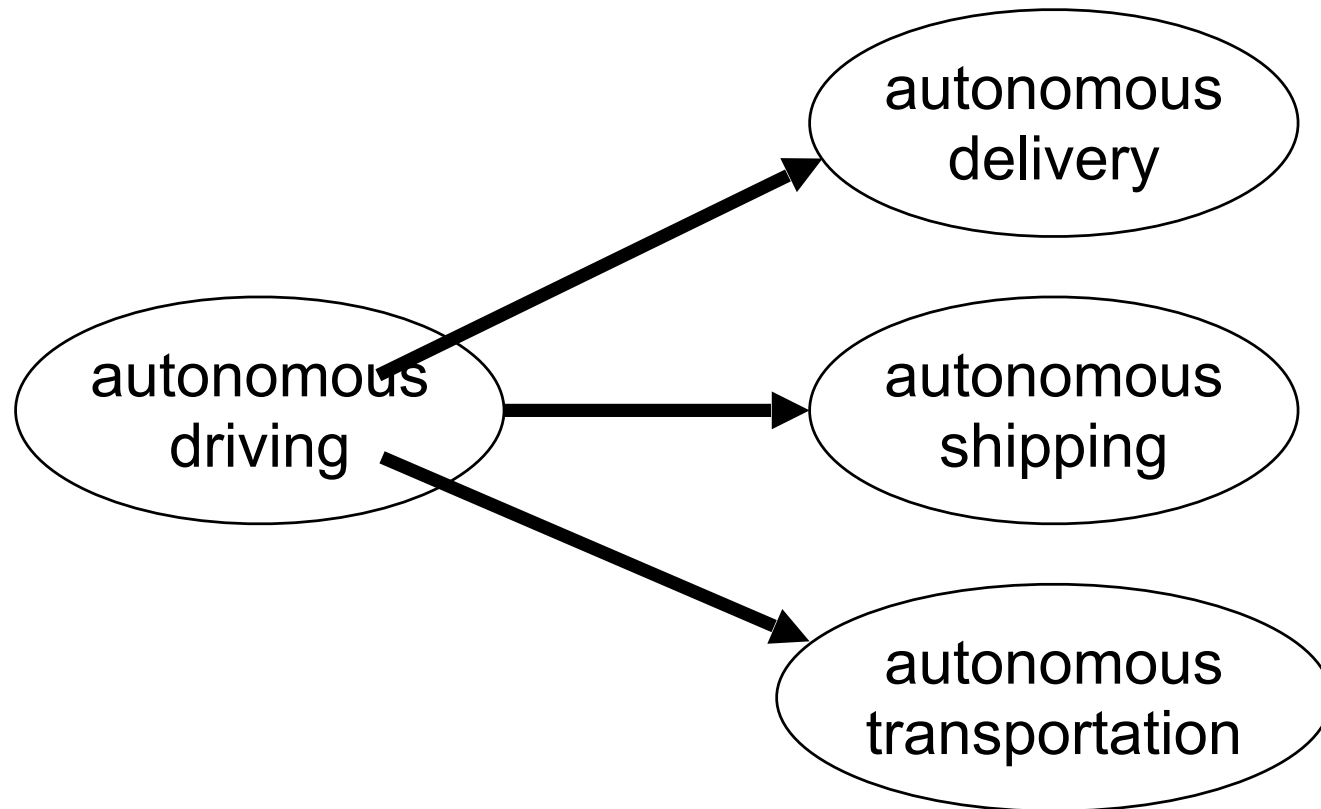
Autonomous Driving: Safety Concerns

**What's more important,
pedestrian's life or passenger safety?**

If a car has to decide between
hitting a pedestrian and
endangering the vehicle (and thus the passengers),
what should it do?

Many questions that need to be answered.

Autonomous Driving: Applications, Advantages and Disadvantages



File Systems

- **Cloud service:** Desired properties: Availability, Reliability, Consistency
- Undoing operations: Undo log, snapshots, version control
- Why does order of operations matter?
- Why might inconsistency occur in a shared file? How is it resolved? (Server version that serializes operations)
- Why would we tolerate inconsistencies? For Speed

File Systems

- **Eventual consistency**
- **Push and pull models**

Last Class Participation Assignment

- **Complete ICES survey and confirm on Canvas.**
 - Abrita
 - Sattwik
- **Your feedback matters! I would like to make the course more useful to future students.**