**LECTURER: Nghia Duong-Trung** 

## **ARTIFICIAL INTELLIGENCE**

### **TOPIC OUTLINE**

History of Artificial Intelligence	1
Early Systems in Artificial Intelligence	2
Neuroscience and Cognitive Science	3
Modern Artificial Intelligence Systems	4
Applications of Artificial Intelligence	5

### MODERN ARTIFICIAL INTELLIGENCE SYSTEMS



## On completion of this unit, you will have...

- ... understanding of basic components within
   Natural Language Processing (NLP) components.
- ... knowledge of main steps within Computer Vision
   (CV) applications.
- ... a basic idea of common features extracted from images.



- 1. Describe the concept of stemming within NLP using your own words with the example of change.
- 2. Explain the relation between AI, ML and Computer Vision.
- 3. Can you list some of the typical features used within Computer Vision?

### NATURAL LANGUAGE PROCESSING (NLP)





identification of words in spoken language



# Language understanding

extraction of meaning from words and sentences, and reading comprehension



Language generation

ability to express information





Speech recognition

Virtual assistants

Language understanding

Language generation

Spam classification

Sentiment analysis

**Summaries** 

Chatbots

**NLP COMPONENTS** 

Semantics

**Pen** = writing tool

Word stemming

{Change, changing, changer, changed} = **chang** 

Word stemming

Part-of-speech-recognition

I like reading books

Pronoun Verb Verb Noun

Word stemming

Part-of-speech-recognition

Named entities recognition (NER)

Jane Doe is studying at IU since 2020.

Person Org Date

### **NLP COMPONENTS**

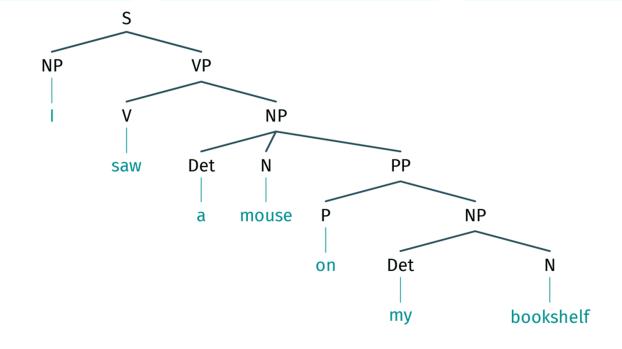
**Semantics** 

Word stemming

Part-of-speech-recognition

Named entities recognition (NER)

Parsing



Word stemming

Part-of-speech-recognition

Named entities recognition (NER)

Parsing

Sentiment analysis

The product is good.



The product can be improved.



The product is useless.



Word stemming

Part-of-speech-recognition

Named entities recognition (NER)

Parsing

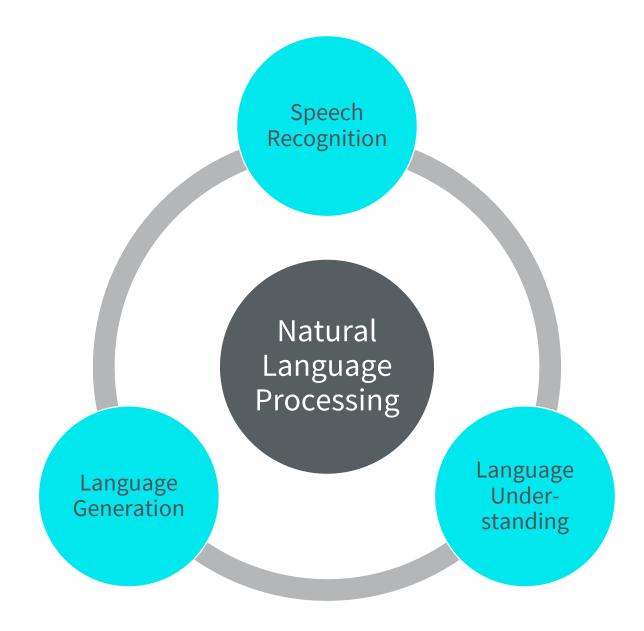
Sentiment analysis

Question and answering

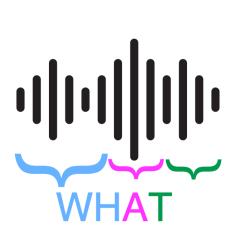
**Question:** Who was the first president of the United States?

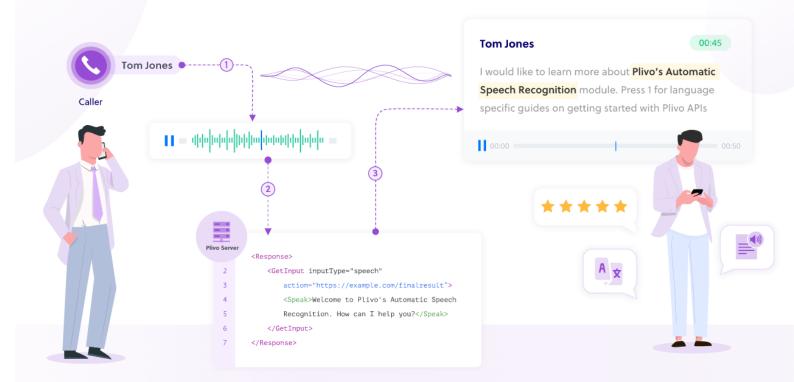
**Answer:** George Washington

### **SUBDOMAINS OF NLP**



- Speech Recognition is a subfield of computational linguistics dealing with the recognition and translation of spoken language into text by computers, a process known as "speech to text" in some cases.
- The phrase "speech recognition" refers to the process of converting spoken words into text in general; however, subfields such as voice recognition and speaker identification specialize in identifying both the spoken content and the speaker's identity.





There are two types of speech recognition systems, speaker-dependent, and speaker-independent

- Speaker-dependent systems are designed in such a way that training is required.
   This works by having a speaker read text into the system or a succession of discrete vocabulary. The algorithm will then analyze the vocal recordings and link them to the text collection.
- Speaker independent systems are speech recognition systems that do not rely on vocal training.

There are two types of models used in speech recognition systems:

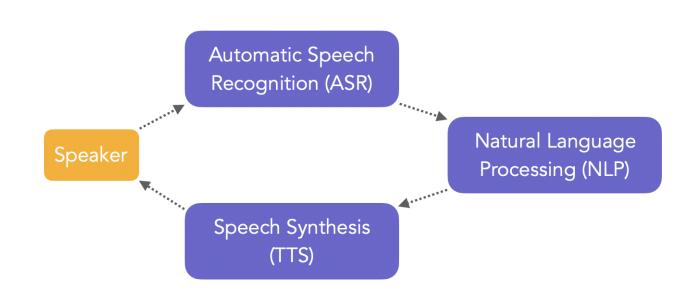
- Acoustic Model: A file containing statistical representations of each of the various sounds that make up a word is known as an acoustic model. A phoneme is a label given to each of these statistical representations. There are approximately 40 distinct sounds in the English language that is suitable for speech recognition, resulting in 40 separate phonemes.
- Language Model: To discriminate between words that sound similar, sounds are matched with word sequences. We presume our audio sample is grammatically and semantically sound, even if it is not grammatically perfect or has skipped words. As a result, incorporating a language model into decoding can enhance ASR accuracy.

Steps involved in the process of speech recognition:

- Analog-to-Digital Conversion: In most cases, speech is recorded and available in analog format. To convert analog voice to digital utilizing sampling and quantization techniques, standard sampling techniques or devices are available. A one-dimensional vector of voice samples, each of which is an integer, is typically used to represent digital speech.
- **Speech Pre-processing**: Background noise and long periods of quiet are common in a recorded conversation. Identification and removal of silent frames, as well as signal processing techniques to reduce/eliminate noise, are all part of speech pre-processing. Following pre-processing, the speech is divided into 20-second frames for subsequent feature extraction stages.
- **Feature Extraction**: It is the conversion of speech frames into a feature vector that specifies which phoneme or syllable is being spoken.
- Word Selection: The sequence of phonemes/features is translated into the spoken word using a language model/probability model.

### SPEECH RECOGNITION & NATURAL LANGUAGE PROCESSING (NLP)

- The combination of linguistics and machine learning (ML) is known as Natural Language Processing (NLP). To produce actionable results, NLP seeks to understand human-human and human-computer interactions in the form of language (voice or text). NLP is an ML application in which machines "learn" to understand the natural language from millions of example datasets.
- Speech recognition is concerned with turning speech input into text, whereas NLP is concerned with "understanding" language in order to feed subsequent activities
- Applications
  - Automatic subtitle with speech recognition
  - Mobile telephony, including mobile email
  - People with disabilities
  - Home automation
  - Virtual assistant
  - Dictation



- Natural language generation (NLG) is a software process that produces natural language output.
- NLG is characterized as "the subfield of artificial intelligence and computational linguistics that is concerned with the construction of computer systems than can produce understandable texts in English or other human languages from some underlying non-linguistic representation of information".
  - Common applications of NLG methods include the production of various reports, for example weather and patient reports; image captions; and chatbots.

# 6 steps to natural language generation

Content analysis	Data understanding	Document structuring	Sentence aggregation	Grammar structuring	6 Language presentation
Filtering data and identifying the important content	Interpreting data, identifying patterns and putting it into context	Creating a document plan and narrative structure	Combining relevant sentences to summarize the topic	Applying grammatical rules to create natural-sounding text	Formatting and presenting the final output

Natural language generation is being used in an array of ways. Some of the many uses include the following:

- generating the responses of chatbots and voice assistants such as Google's Alexa and Apple's Siri;
- converting financial reports and other types of business data into easily understood content for employees and customers;
  - automating lead nurturing email, messaging and chat responses;
  - personalizing responses to customer emails and messages;
  - generating and personalizing scripts used by customer service representatives;
  - aggregating and summarizing news reports;
  - reporting on the status of internet of things devices; and
  - creating product descriptions for e-commerce webpages and customer messaging.

# How NLP, NLU and NLG are related Natural language processing The overall term for how computers understand, interpret and use human language Natural language understanding

NLG relies on machine learning algorithms and other approaches to create machine-generated text in response to user inputs. Some of the methodologies used include the following:

- Recurrent neural network
- Long short-term memory
- Transformer
  - Generative pre-trained transformer (GPT)
  - Bidirectional encoder representations from transformers (BERT)
  - XLNet

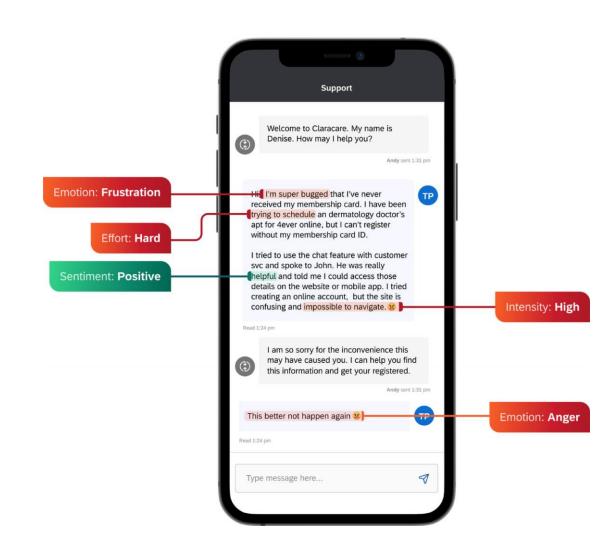
- Natural-language understanding (NLU) or natural-language interpretation (NLI) is a subtopic of natural-language processing in artificial intelligence that deals with machine reading comprehension. Natural-language understanding is considered an AI-hard problem.
  - what human language means, rather than simply what individual words say.
  - comprehend and respond accurately to the sentiments expressed in natural language text.
  - Natural Language Understanding deconstructs human speech using trained algorithms until it forms
    a structured ontology, or a set of concepts and categories that have established relationships with
    one another.
  - computational linguistics

### Extract information from the data source

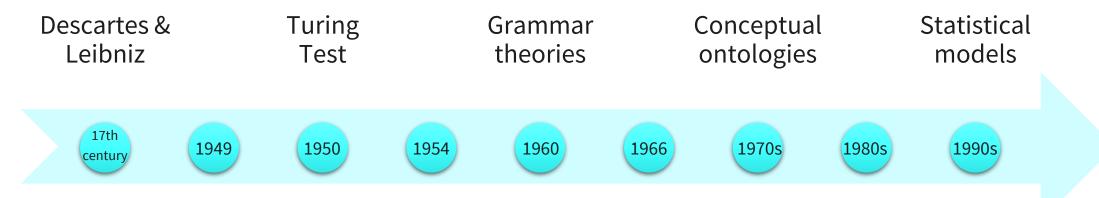
- "ICE ticket Berlin to München" 15.11, 9am"
  - ICE: type of train in Germany
  - Ticket: intent to buy a train ticket
  - Berlin, München: locations
  - 15.11, 9am: date and time

NLU deconstructs human speech/text using trained algorithms until it forms a structured ontology, or a set of concepts and categories that have established relationships with one another.

- Intent recognition, entity recognition



### **HISTORICAL BACKGROUND**



Weaver's memorandum

Georgetown experiment

ELIZA

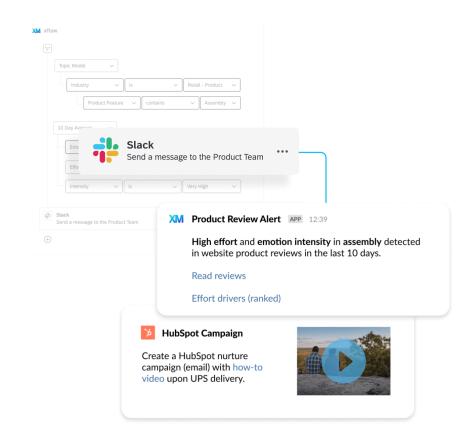
Symbolic models

https://aylien.com/blog/a-review-of-the-recenthistory-of-natural-language-processing



Image Source: Custom Depiction

### TYPICAL AREAS OF APPLICATION FOR NLP

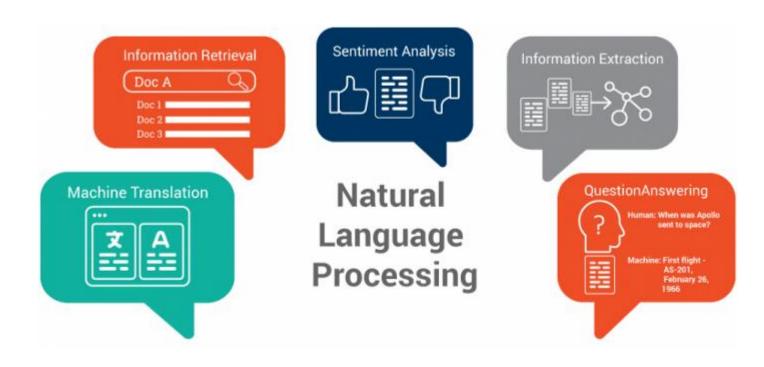




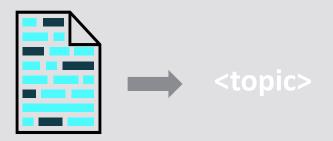
### **STATISTICS**

Worldwide revenue from the natural language processing (NLP) market is forecast to increase rapidly in the next few years. The NLP market is predicted be almost 14 times larger in 2025 than it was in 2017, increasing from around three billion U.S. dollars in 2017 to **over 43 billion in 2025**. Natural language processing (NLP) is a branch of artificial intelligence (AI) that helps computers understand, interpret and manipulate human language. (https://www.statista.com/statistics/607891/worldwide-natural-language-processing-market-

revenues/#:~:text=The%20NLP%20market%20is%20predicted,interpret%20and%20manipulate%20human%20language.



### **TOPIC IDENTIFICATION**

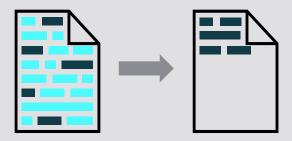


"One quick way to summarize the milestones in AI history is to list the Turing Award winners: Marvin Minsky (1969) and John McCarthy (1971) for defining the foundations of the field based on representation and reasoning; Allen Newell and Herbert Simon (1975) for symbolic models of problem solving and human cognition" (Russell & Norvig, 2021, p. 35).



Text Source: Russell, S., & Norvig, P. (2021), p. 35. Image Source: Custom Depiction.

### **TEXT SUMMARIZATION**

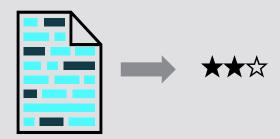


"Mr and Mrs Dursley, of number four, Privet Drive, were proud to say that they were perfectly normal, thank you very much. They were the last people you'd expect to be involved in anything strange or mysterious, because they just didn't hold with such nonsense" (Rowling, 1998, p. 1).



Mr and Mrs Dursley were perfectly normal. They were the last people to be involved in anything mysterious.

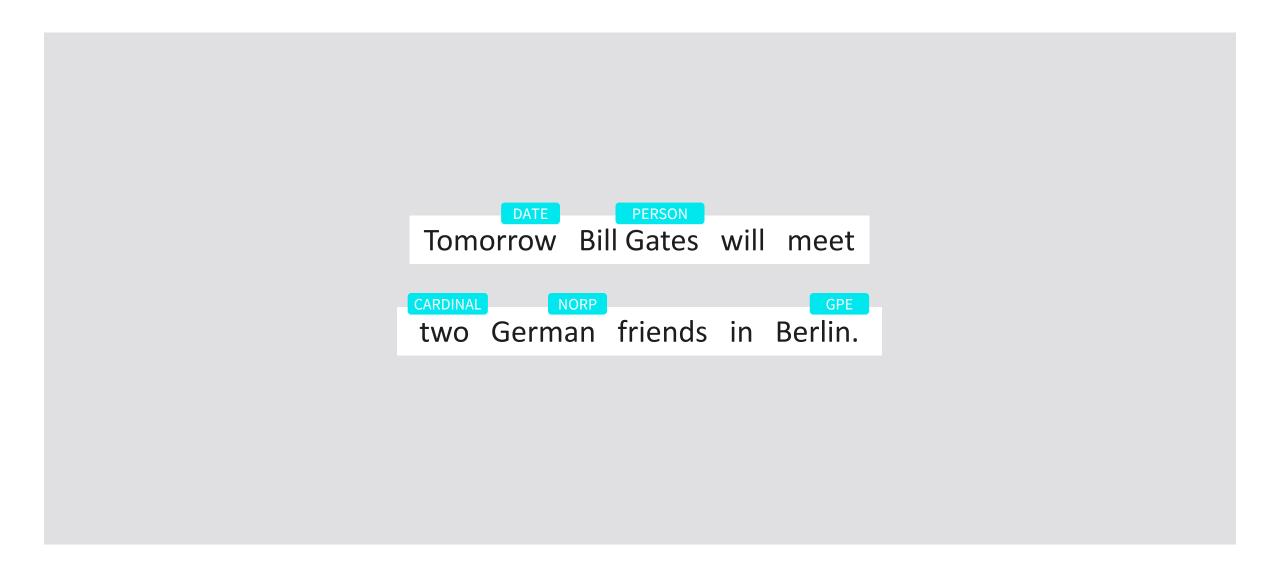
#### **SENTIMENT ANALYSIS**



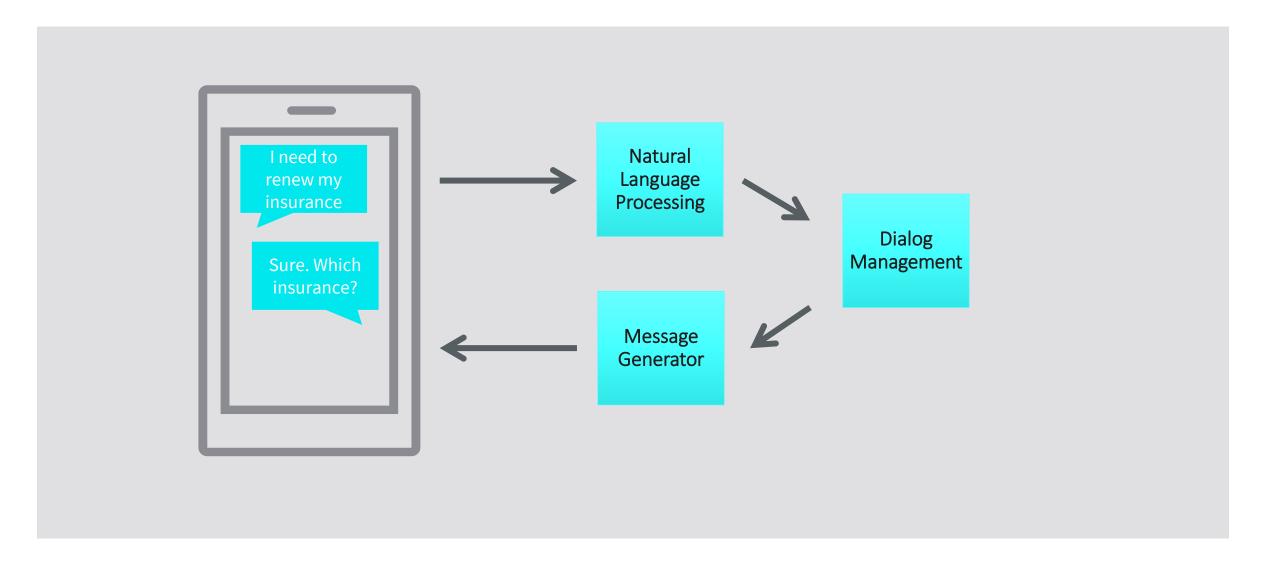
"Exactly what I thought it would be. Perfect early Father's Day gift for my husband – we had so much fun playing ping pong at a resort recently I just had to get it for him."



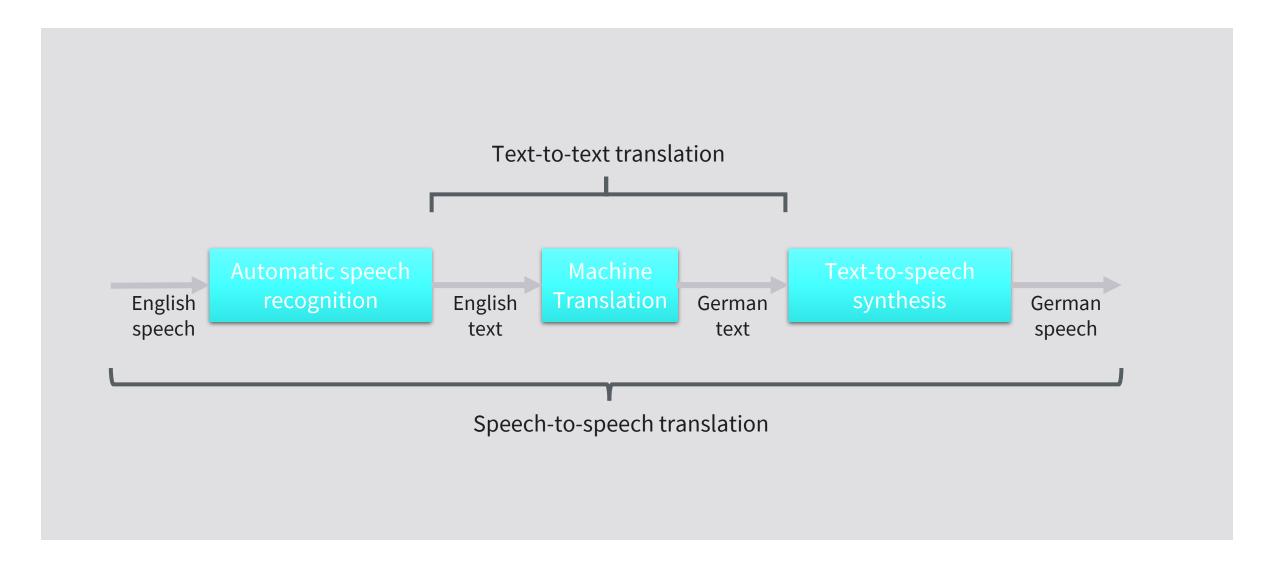
### NAMED ENTITY RECOGNITION



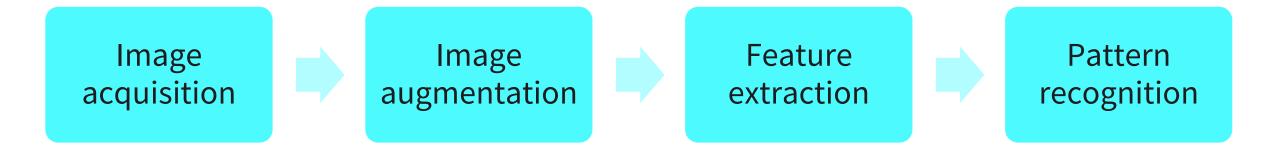
### **CHATBOTS**



### **MACHINE TRANSLATION**



# Computer Vision = understand the content of images and videos



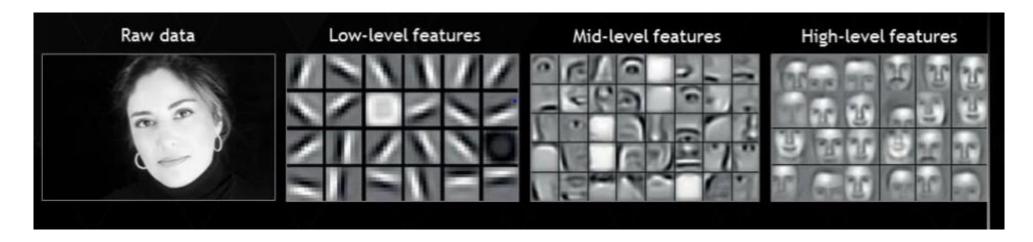
- Camera vision
- Pinhole cameras
- Film cameras
- Digital cameras

- Image sharpening
- Contrastenhancement

Extract higherlevel image features Infermeaningfulimage content

#### **IMAGE FEATURES**

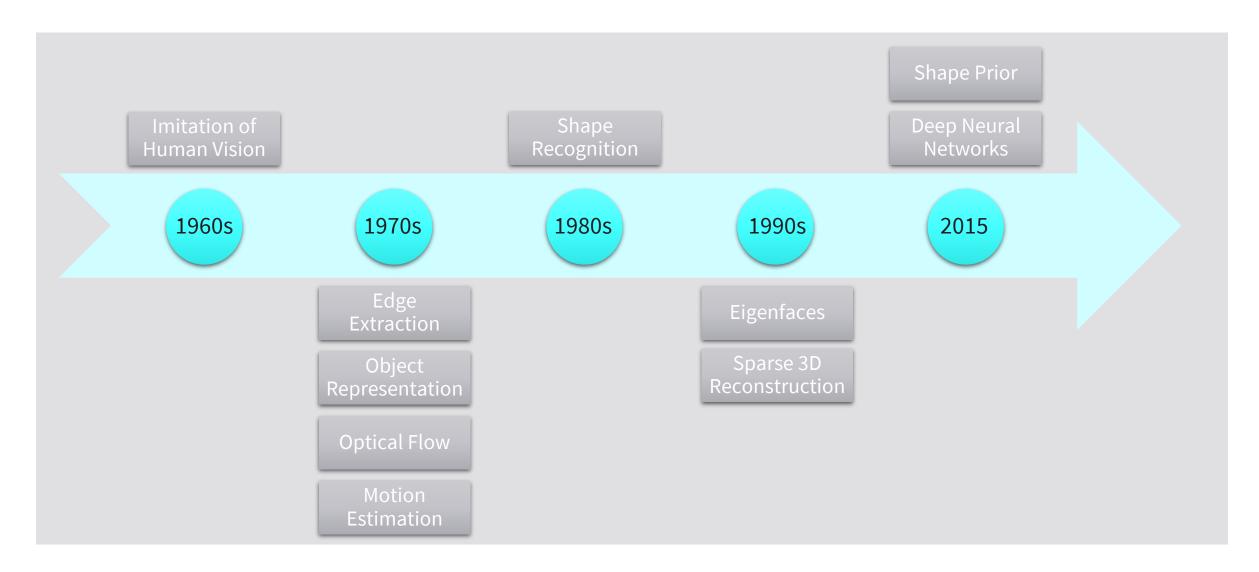
Extract salient features from images such as edges, colors, blobs and ridges:



### **COMPUTER VISION DEMO**

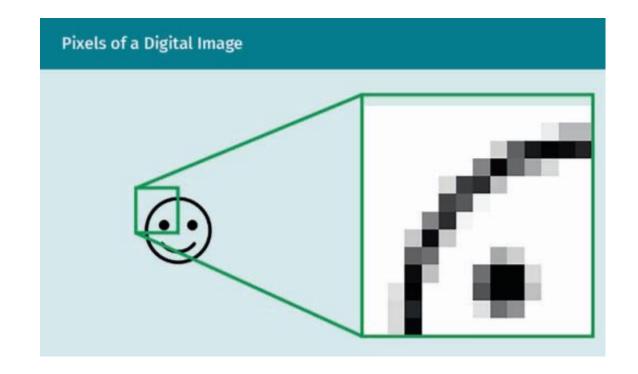


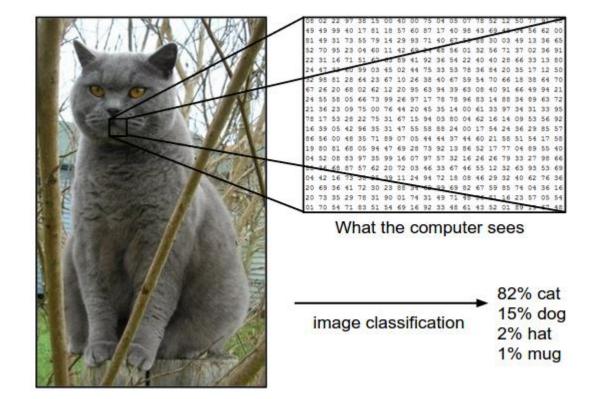
### **HISTORICAL DEVELOPMENTS**



### **HUMAN VS MACHINE**

Term: pixel
 color representations
 monochrome
 real, high, true, deep color





#### **TYPICAL COMPUTER VISION TASKS**

### Computer vision tasks

### Recognition

Detect and classify different objects or persons in images

# Motion analysis

Detect and track motion of objects, persons or the camera

# Image restoration

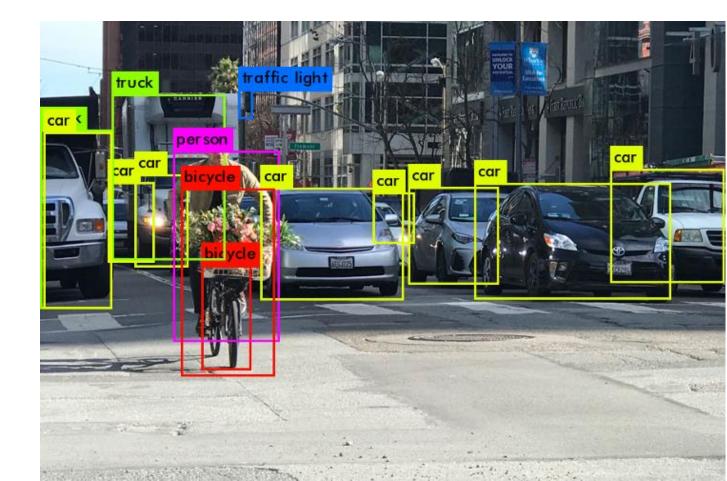
Filter and remove noise from images and videos

# Geometry reconstruction

Estimation of a 3D model of a real-world object or scene

### 10 INTERESTING COMPUTER VISION TASKS: OBJECT DETECTION

- identify and locate an object of particular interest from still image or video data.

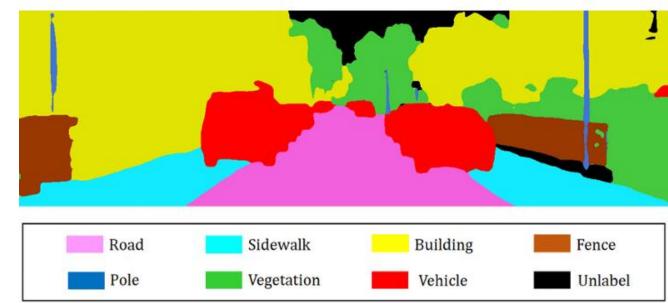


#### 10 INTERESTING COMPUTER VISION TASKS: IMAGE SEGMENTATION

- The process of partitioning an image into multiple segments. the aim is to transform the image in a representation that is easier to analyze, there are many flavors of image

segmentation





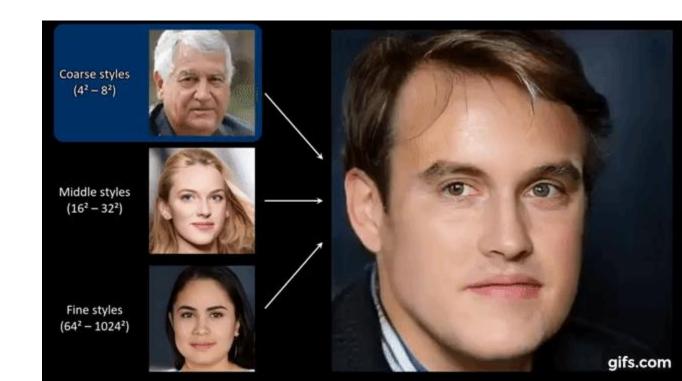
### 10 INTERESTING COMPUTER VISION TASKS: POSE ESTIMATION

Estimating the position and the orientation of the joints of a human body. this has a wide range of applications in AR/VR (augmented and virtual reality)

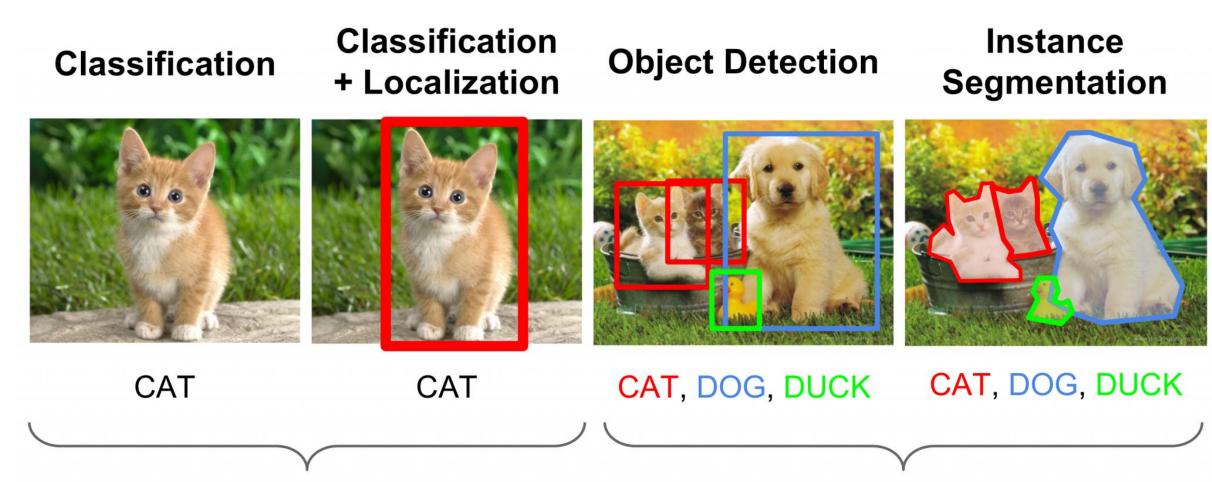


#### 10 INTERESTING COMPUTER VISION TASKS: IMAGE GENERATION

Generating new images from an existing data set, this task has been made popular by the use Generative adversarial networks (GANs) to generate seemingly real images or videos which were actually fake.



Identifying classes of objects within the image

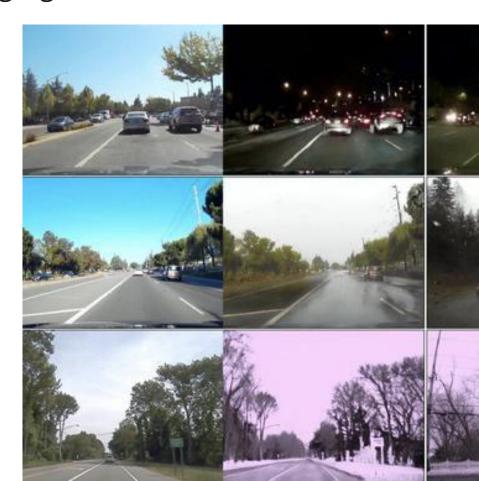


Single object

Multiple objects

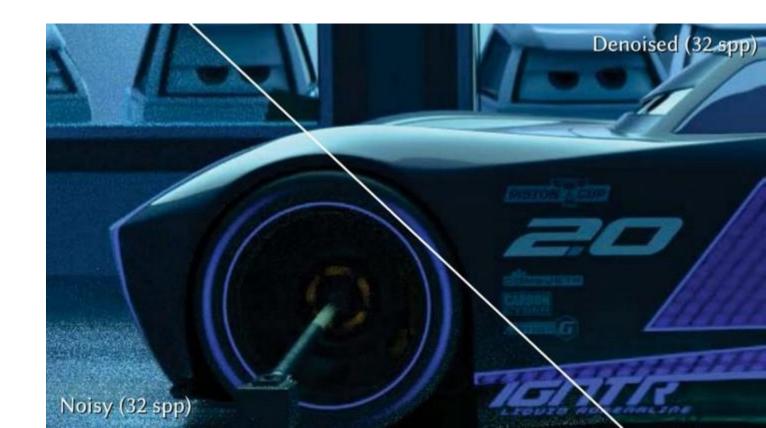
#### 10 INTERESTING COMPUTER VISION TASKS: DOMAIN ADAPTATION

maintaining a predictive model performance even if the data distribution the model was trained on has changed. domain adaptation uses labeled data in one or more data distribution and tries to solve a new task with data belonging to a different distribution



### 10 INTERESTING COMPUTER VISION TASKS: DE-NOISING

predicting the original image by removing noise from a noise-contaminated image



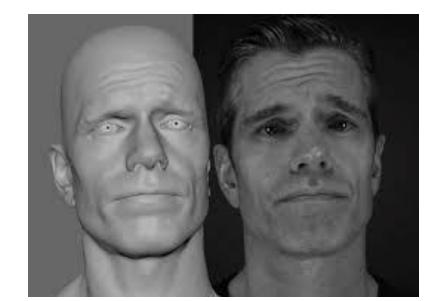
#### 10 INTERESTING COMPUTER VISION TASKS: SUPER-RESOLUTION

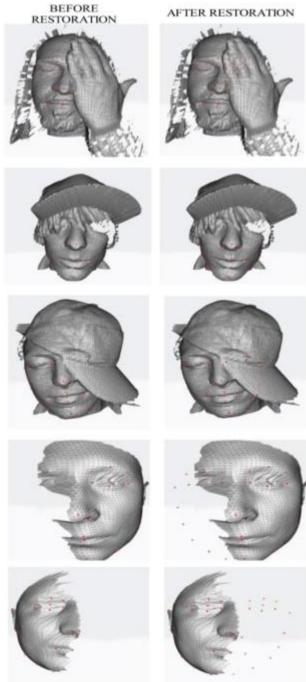
taking an input from a low resolution image and up-scaling it to a higher resolution



### 10 INTERESTING COMPUTER VISION TASKS: 3D FACE ANIMATION

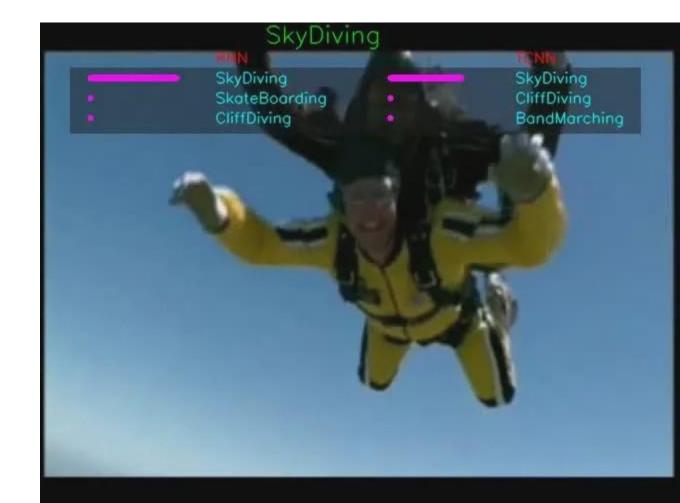
### reconstructing an animated 3D human face from a still image





#### 10 INTERESTING COMPUTER VISION TASKS: ACTION CLASSIFICATION IN VIDEO

predicting the movement of an object from observations made in a video stream



### **IMAGES AS FUNCTIONS**

A function that can map a two-dimensional coordinate (x,y) to a specific color value.

On the x-axis we begin on the left with a value of 0 and continue to the right until the maximum width of an image is reached.

On the y-axis, we begin with 0 at the top and reach the height of an image at the bottom.

f(42,100)

## Filter:

Function which receives an image as an input, applies modifications, and returns the filtered image as an output.

# Example: 2D image convolution

Convolution of an image I with a kernel k with a size of n and a center coordinate  $\alpha$ :

$$I \cdot (x, y) = \sum_{i=1}^{n} \sum_{i=j}^{n} I(x - 1 + a, y - j + a) k(i, j)$$

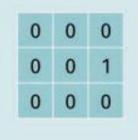
I · (x, y): value of the resulting image I · at position (x, y), I is the original image.

### **FILTERS**

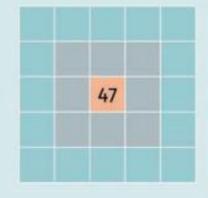
### **2D Image Convolution**

Step 1

63	80	90	62
41	26	86	41
27	42	47	93
44	88	41	24
71	58	80	4
	41 27 44	41 26 27 42 44 88	63 80 90 41 26 86 27 42 47 44 88 41 71 58 80

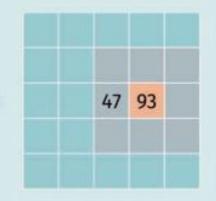


Х



Step 2

82	63	80	90	62
73	41	26	86	41
89	27	42	47	93
55	44	88	41	24
86	71	58	80	4



Final result

63	80	90	62	0
41	26	86	41	0
27	42	47	93	0
44	88	41	24	0
71	58	80	4	0

### **PADDING TECHNIQUES**

# 82 63 80 90 62 82 63 80 90 62 82 63 80 90 62 73 41 26 86 41 73 41 26 86 41 73 41 26 86 41 89 27 42 47 93 89 27 42 47 93 89 27 42 47 93 55 44 88 41 24 0 55 44 88 41 24 24 55 44 88 41 24 41

86 71 58 80

Replication padding

80

4

86 71 58 80

Reflection padding

41 24 41

**Different Padding Techniques** 

86 71 58 80

Constant padding

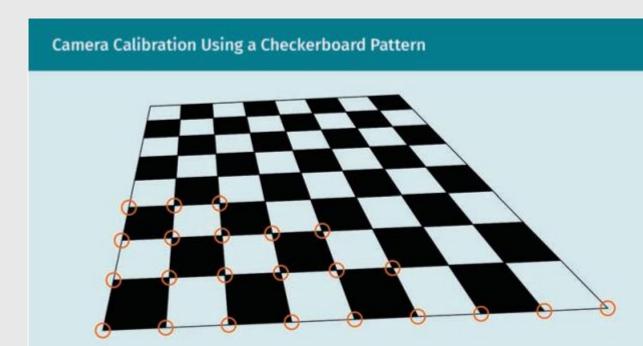
0

0

### **CAMERA CALIBRATION**

The calibration makes it possible to extract distortion from the images





## **Feature Detection**

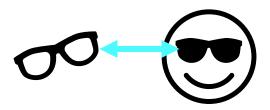


- Blobs (blue)
- Edges (red)
- Corners (yellow)

# Feature Description Feature Matching

Semantic information

- Common algorithms:
  - BRIEF
  - ORB
  - SIFT
  - SURF



 Identification of similar features in different images

https://medium.com/data-breach/introduction-to-surf-speeded-up-robust-features-c7396d6e7c4e https://docs.opencv.org/3.4/df/dd2/tutorial\_py\_surf\_intro.html

### **IMPORTANT CHARACTERISTICS OF FEATURE DETECTION ALGORITHMS**



### **SEMANTIC SEGMENTATION**

Original image

Segmentation map

Segmentation overlay

Background

Chair

Coffee table



### You now have...

- ... understanding of basic components within
   Natural Language Processing (NLP) components.
- ... knowledge of main steps within Computer Vision (CV) applications.
- ... a basic idea of common features extracted from images.

### SESSION 5

# **TRANSFER TASK**

### **TRANSFER TASK**

Brainstorm about factors of images that make Computer Vision difficult. While doing so, reflect on Human Vision throughout daily life and what information the human brain needs to translate.

## TRANSFER TASK PRESENTATION OF THE RESULTS

Please present your results.

The results will be discussed in plenary.





# 1. Speech-to-Text

- a) has no relation to natural language processing.
- b) is one of many sub-fields of natural language processing.
- c) is what natural language processing is all about.
- d) has no practical application.



# 2. Artificial General Intelligence

- a) is a natural extension of weak artificial intelligence.
- b) is provably impossible.
- c) will require a considerably more variegated set of abilities than what weak artificial intelligence approaches can currently provide.
- d) is guaranteed to come about in the near future.



# 3. Computer vision refers to

- a) image acquisition by computers.
- b) computer graphics.
- c) the semantic understanding of visual scenes.
- d) filter operations such as smoothing or de-noising.

### **LIST OF SOURCES**

Fomodan, E.R. & Caleanu, C. (2018). Bag of Features vs Deep Neural Networks for Face Recognition. 2018 International Symposium on Electronics and Telecommunications (ISETC). DOI: 10.1109/ISETC	2018.8583846.

