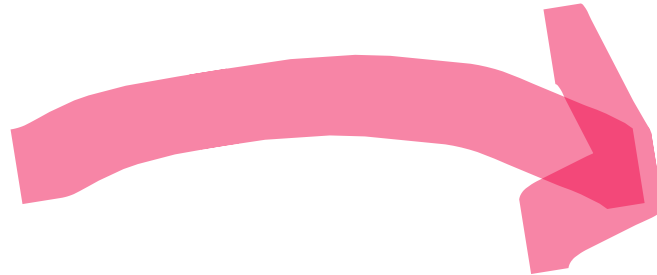


# An introduction to pattern recognition and clustering

**Lecture materials available on GitHub:**

[https://github.com/mluerig/jena\\_clustering\\_lecture](https://github.com/mluerig/jena_clustering_lecture)

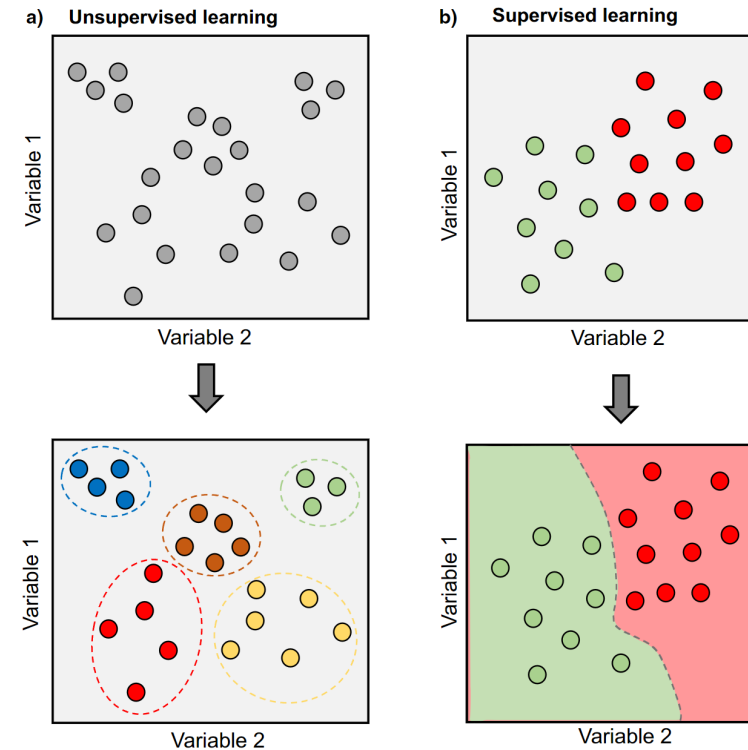
(scan this to  
experience an  
example of pattern  
recognition!)



# An introduction to pattern recognition and clustering



Part 1: Computer vision



Part 2: Clustering

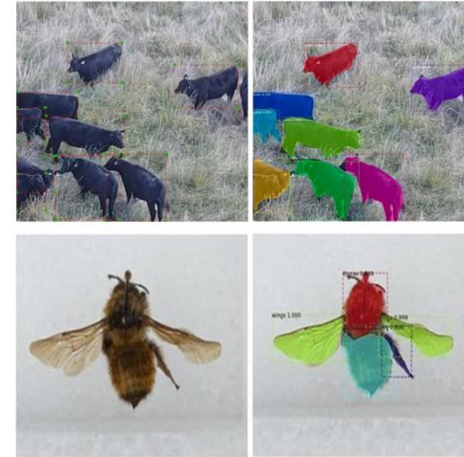
# Introduction: Computer vision

## Automated image analysis

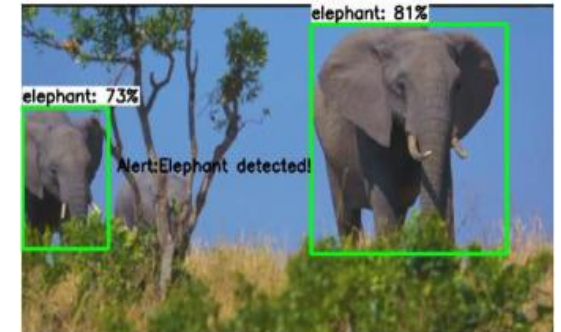
External phenotypes and morphology, unlike internal phenotypes or DNA, are physical properties that can be captured and analyzed using digital imaging.

**Computer vision**, is an effective toolbox to extract meaningful features from images, and thereby facilitating, for example, the automated measurement of traits.

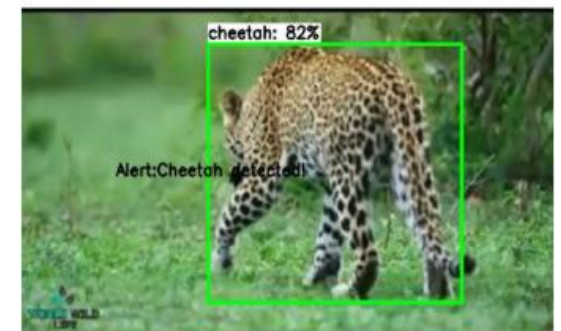
Especially if we are interested in measuring morphology in many specimens, modern, AI-based CV is indispensable



(a)



(b)

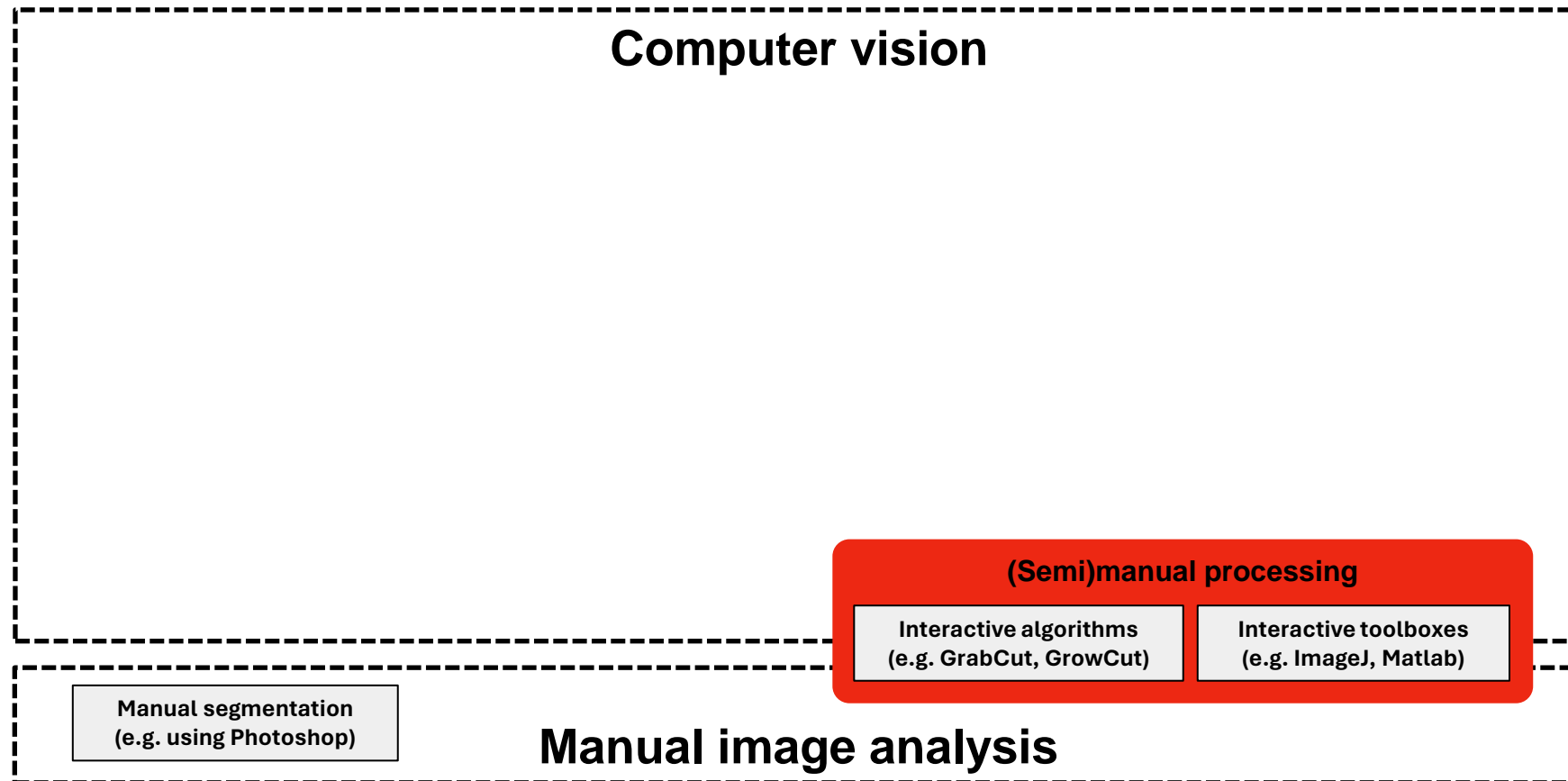


(c)

# Introduction: Computer vision

## Computer vision

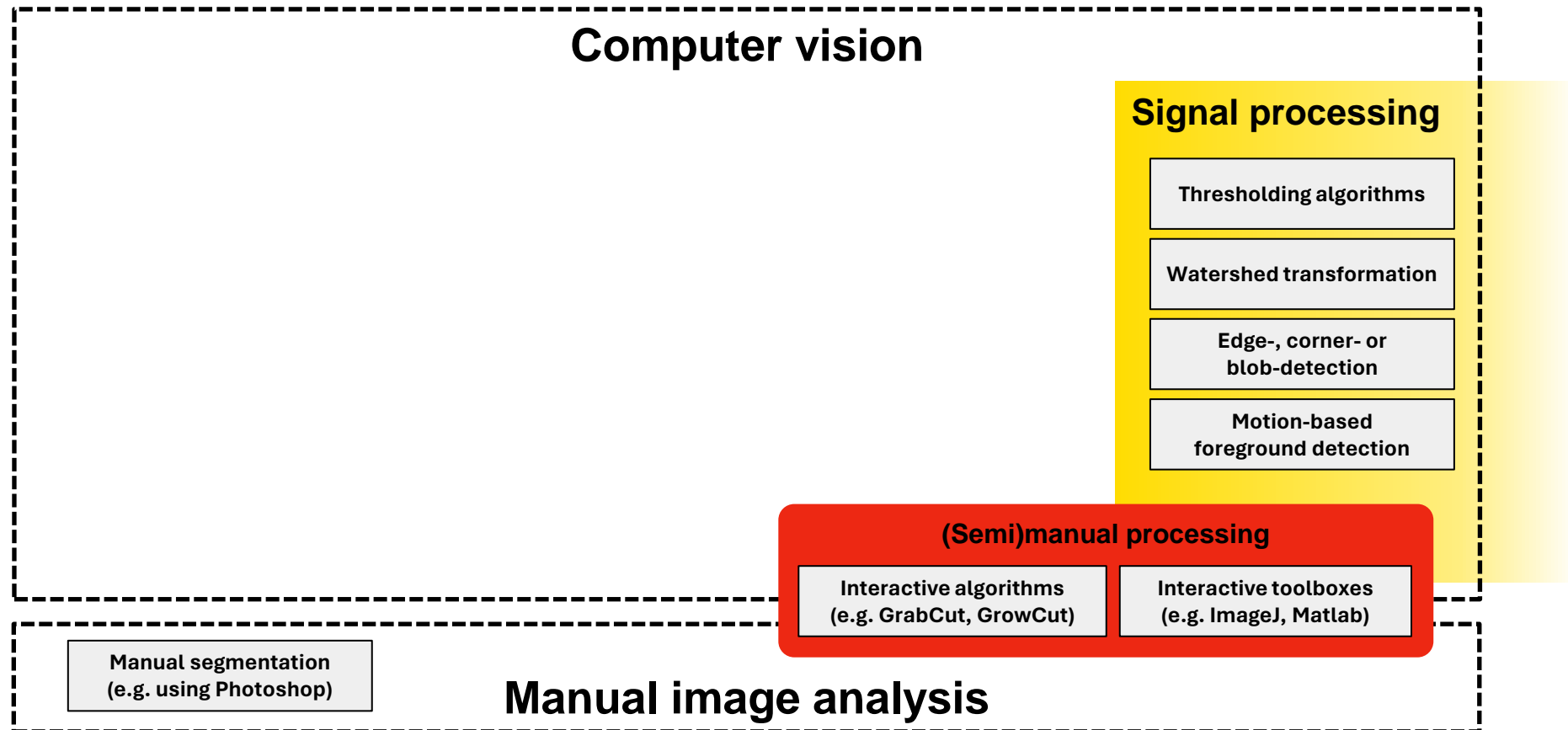
A field at the intersection of machine learning and signal processing



# Introduction: Computer vision

## Computer vision

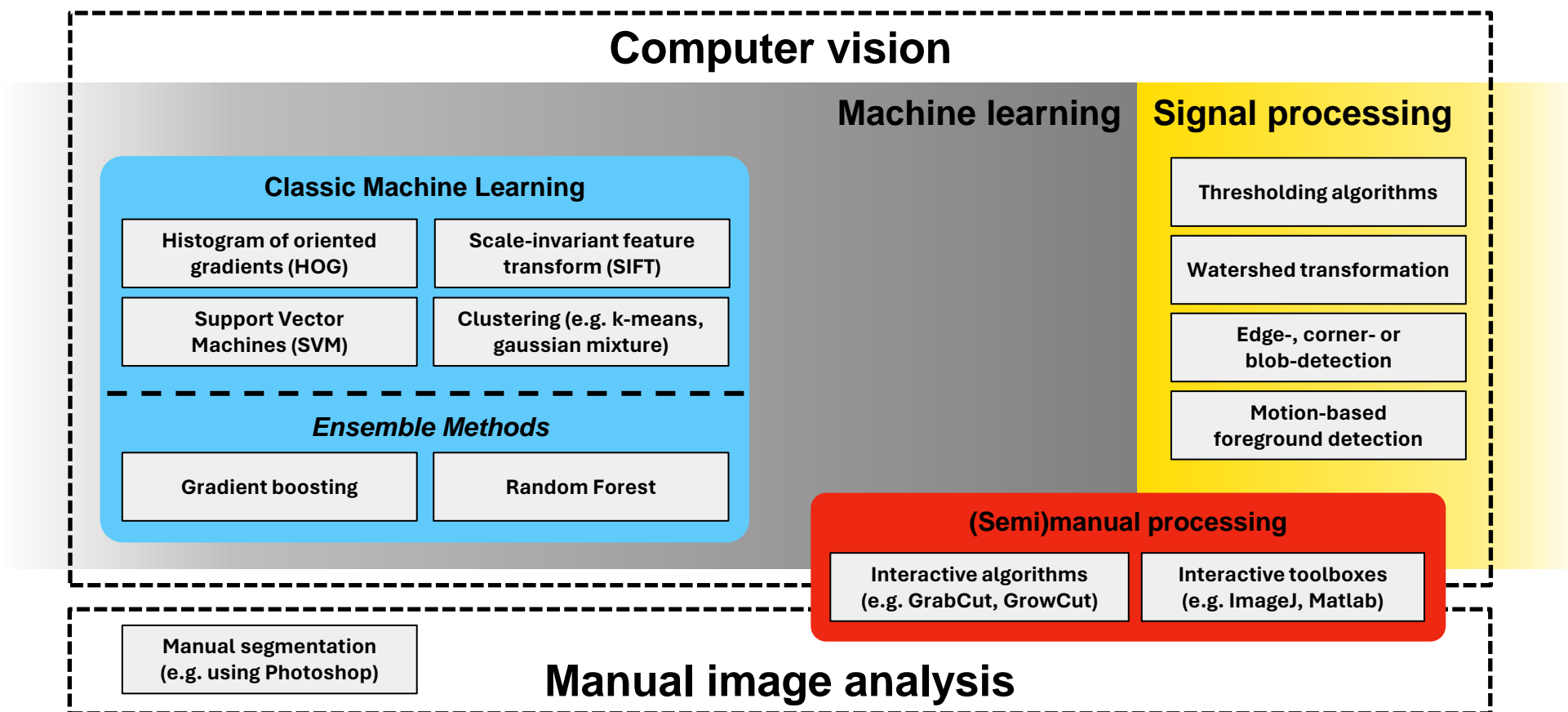
A field at the intersection of machine learning and signal processing



# Introduction: Computer vision

## Computer vision

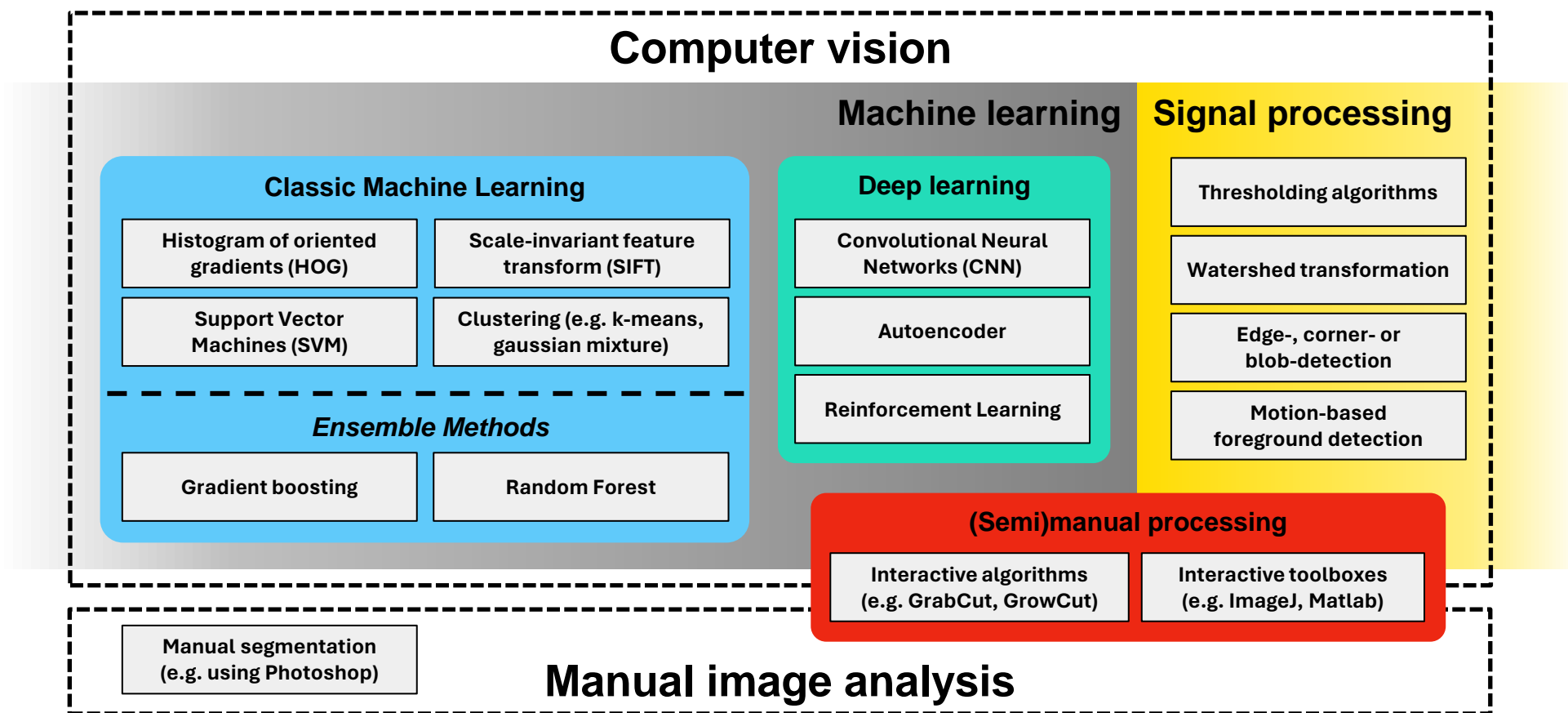
A field at the intersection of machine learning and signal processing



# Introduction: Computer vision

## Computer vision

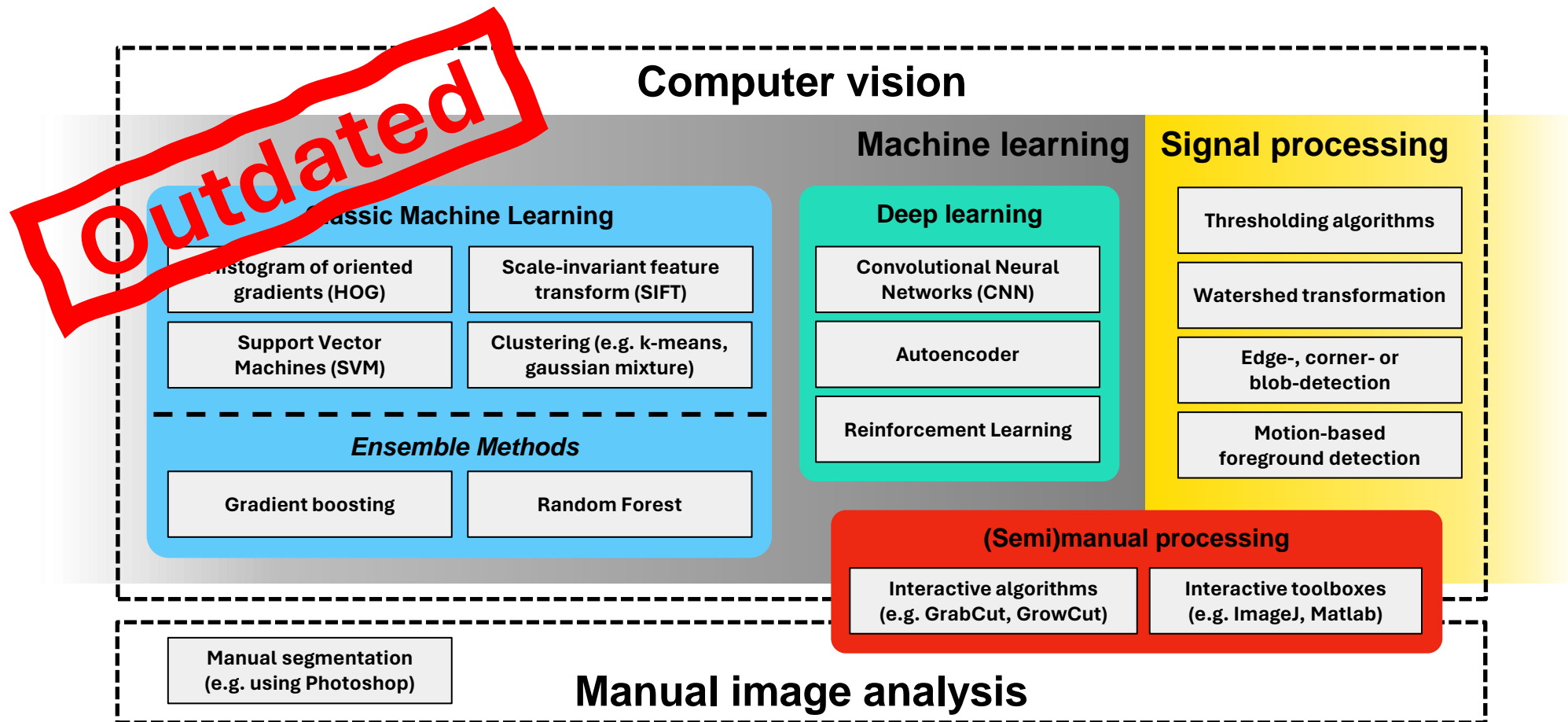
A field at the intersection of machine learning and signal processing



# Introduction: Computer vision

## Computer vision

A field at the intersection of machine learning and signal processing

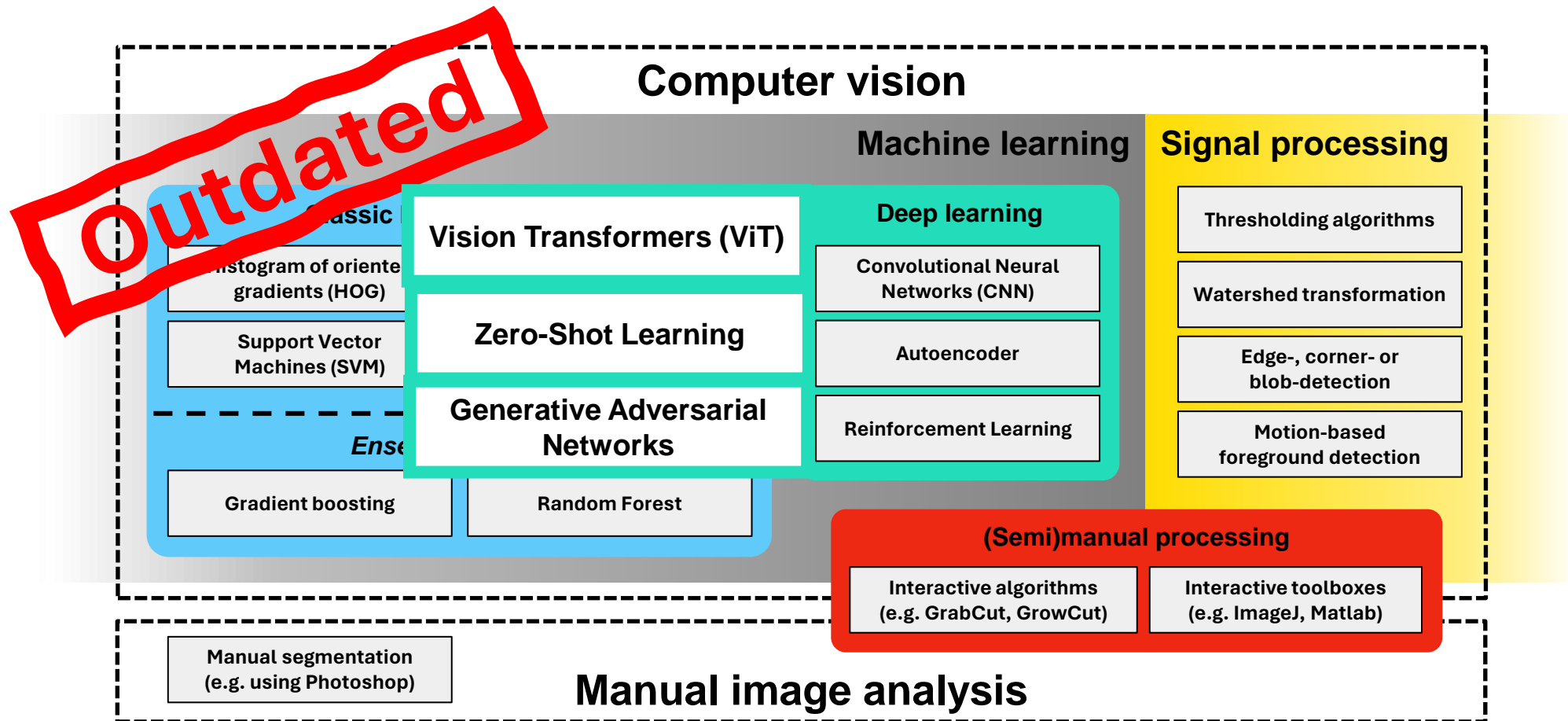




# Introduction: Computer vision

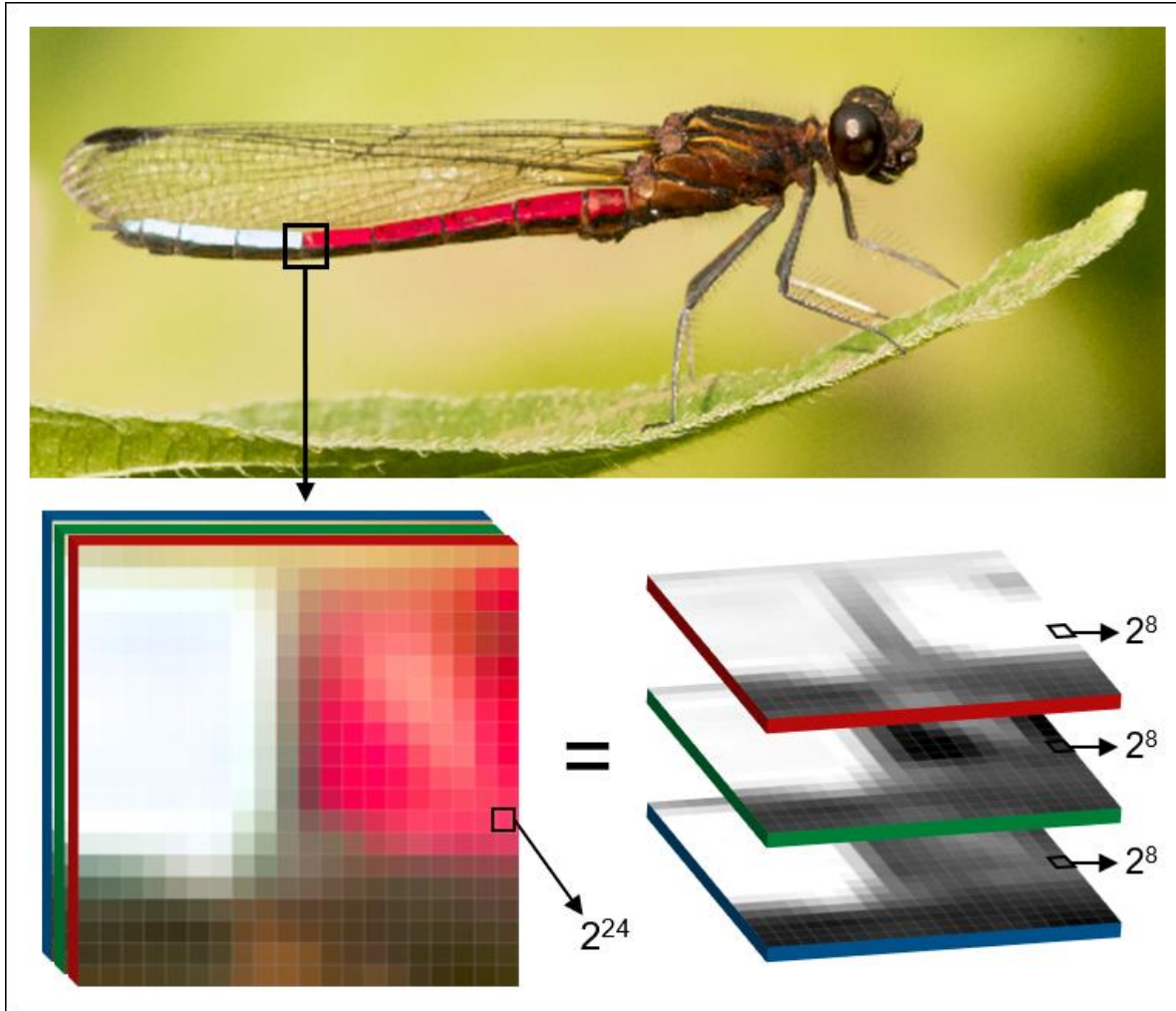
## Computer vision

A field at the intersection of machine learning and signal processing



# Introduction: Computer vision

## Digital images



Lürig et al. 2021

red



green



blue



# Computer vision

## CV tasks

- Classification
- Object detection
- Segmentation (semantic + instance)

(Not shown)

- Key points / pose estimation
- Object tracking

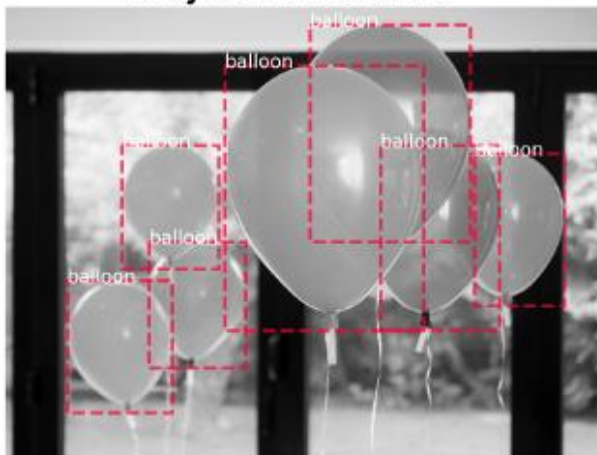
Classification



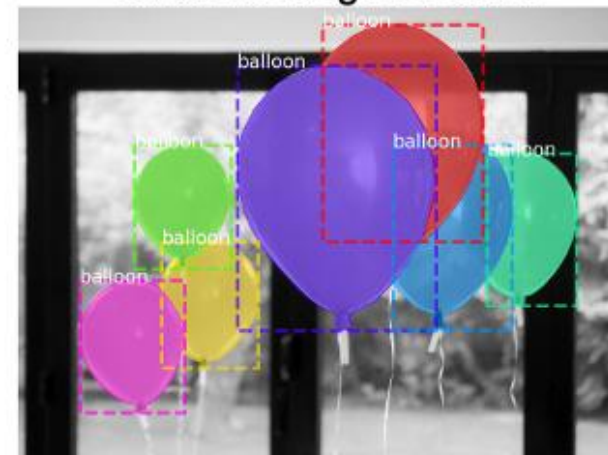
Semantic Segmentation



Object Detection



Instance Segmentation



# Introduction: Computer vision

## Signal-processing-based segmentation



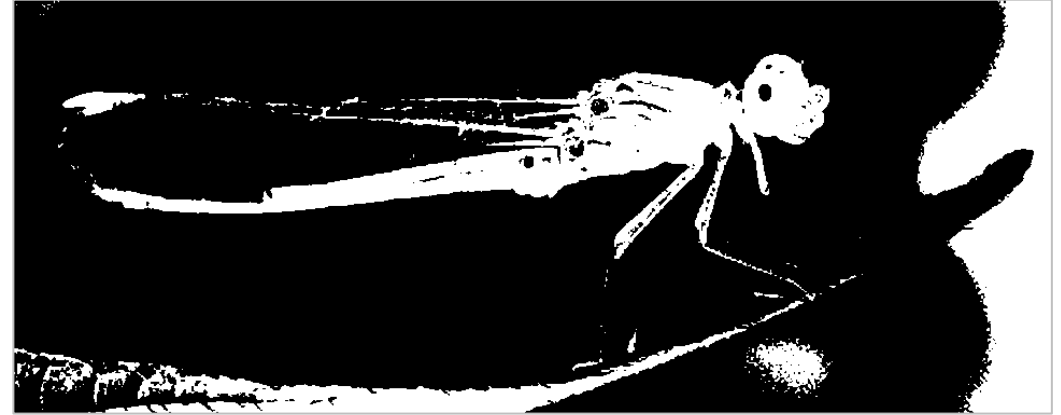
Grayscale

# Introduction: Computer vision

## Signal-processing-based segmentation



Grayscale



Binary threshold (val=127)

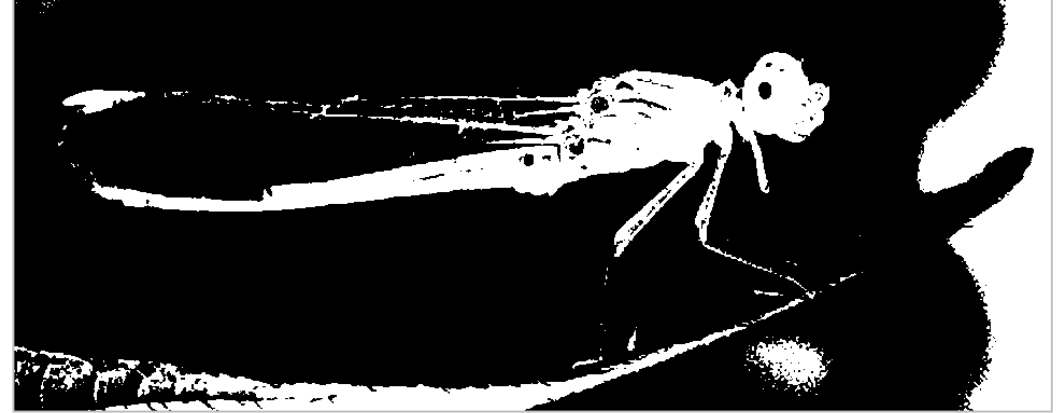


# Introduction: Computer vision

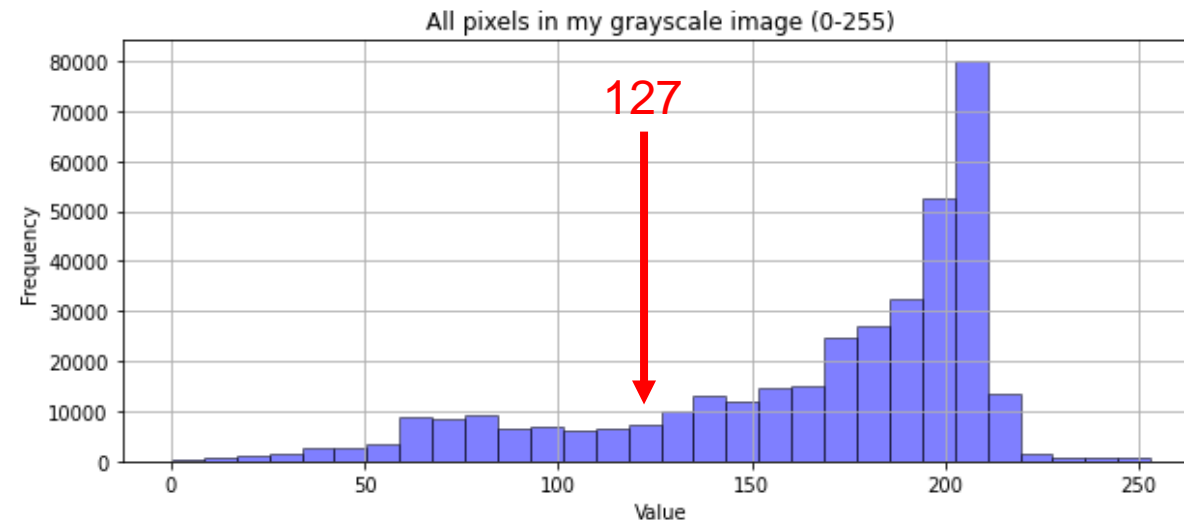
## Signal-processing-based segmentation



Grayscale



Binary threshold (val=127)

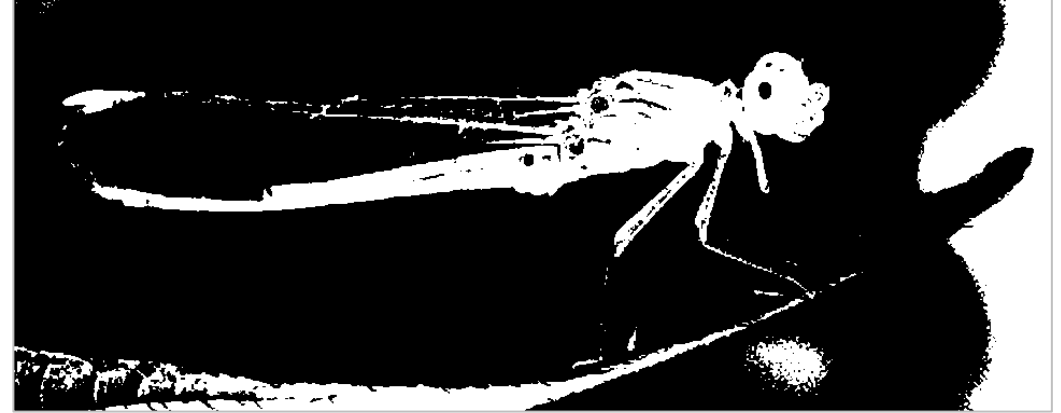


# Introduction: Computer vision

## Signal-processing-based segmentation



Grayscale



Binary threshold (val=127)



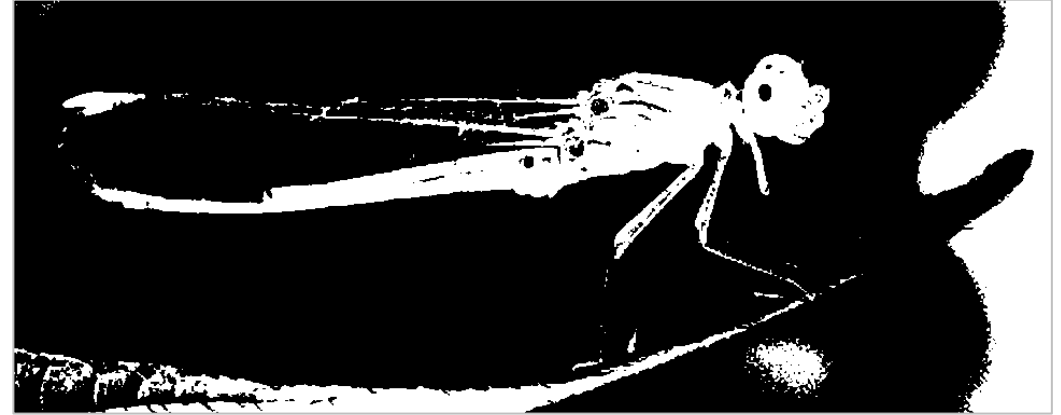
Gaussian blur + Otsu threshold

# Introduction: Computer vision

## Signal-processing-based segmentation



Grayscale



Binary threshold (val=127)



Gaussian blur + Otsu threshold



Final mask

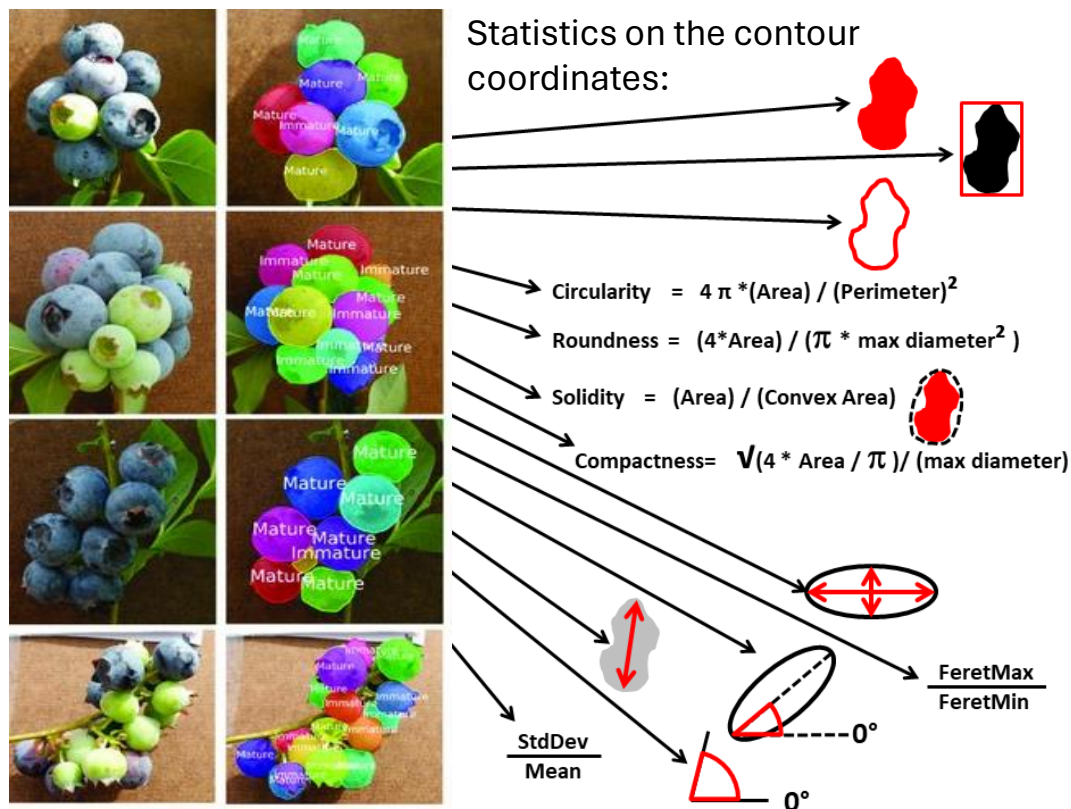


# Introduction: Computer vision

## Trait extraction

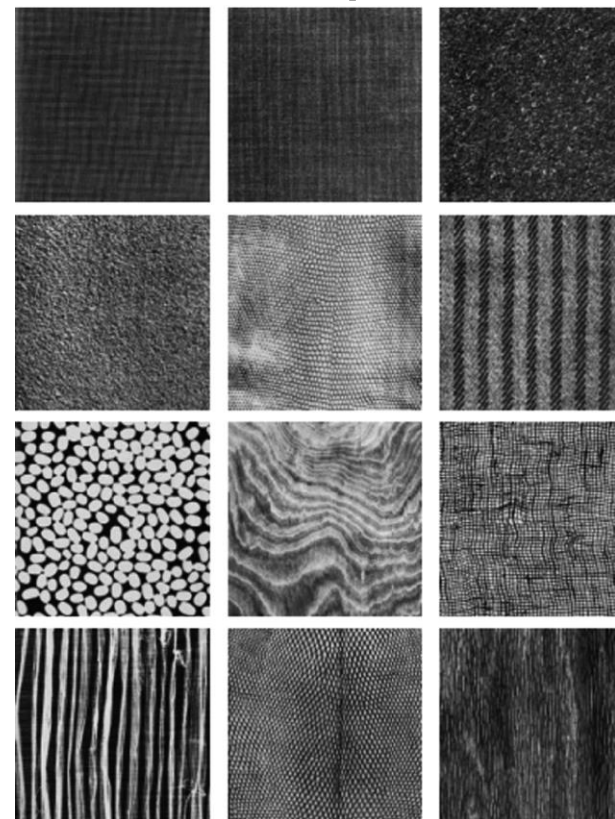
### Shape features

use contour pixels of mask



### Texture features

use all pixels within mask



Statistics on the spatial distribution of pixels in each channel (RGB / HSV):

- Minimum, maximum, Mean, Median
- Skewness, Kurtosis, Variance
- Uniformity, Clusteredness
- Entropy
- Gray-Level matrices
- Higher level neighborhood interactions
- ...

# Practical: Computer vision

< Practical part - open 01\_computer\_vision.ipynb >

# Introduction: Clustering

## Finding our way through big data

Biological datasets can be vastly diverse and complex, which requires not just examination, but also organization and understanding.

**Machine learning** offers a powerful approach to make sense of such complex data. Specifically, unsupervised methods such as clustering algorithms aim to identify inherent groupings in the data based on similarities among data points.

This process is somewhat analogous to the way biologists classify organisms into taxa based on shared characteristics.

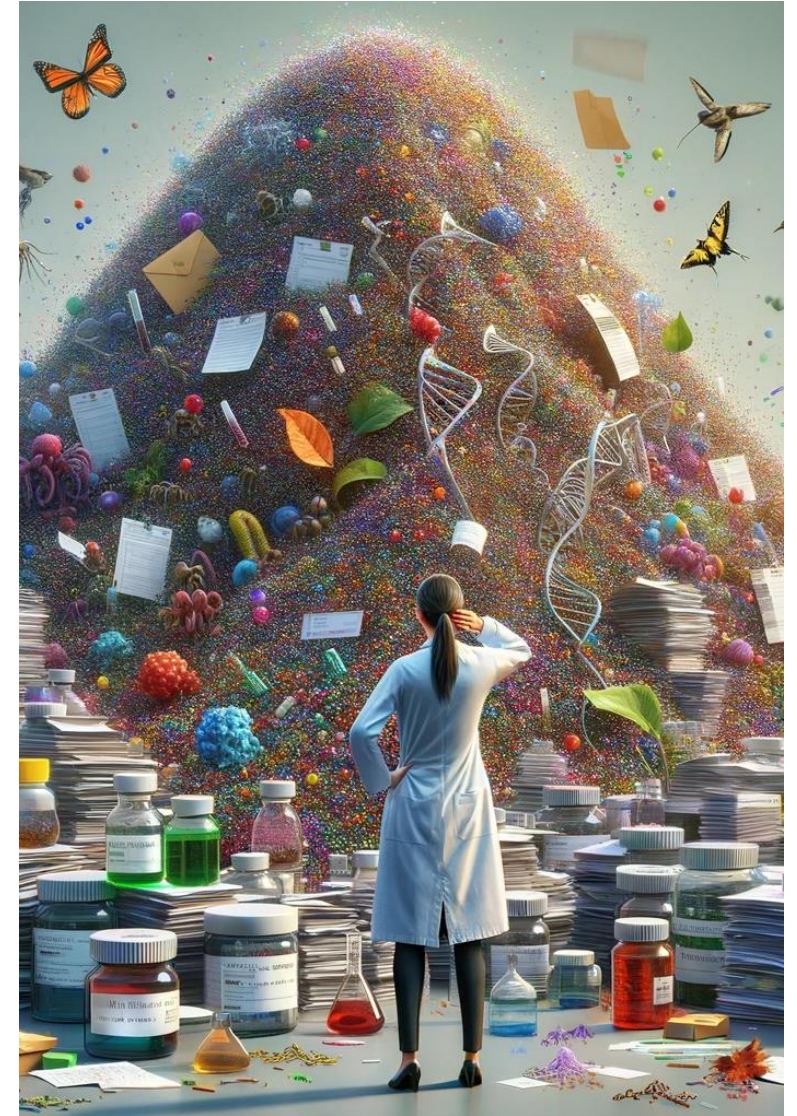


Figure: DALL-E

# Introduction: Clustering

## Unsupervised vs. supervised machine learning

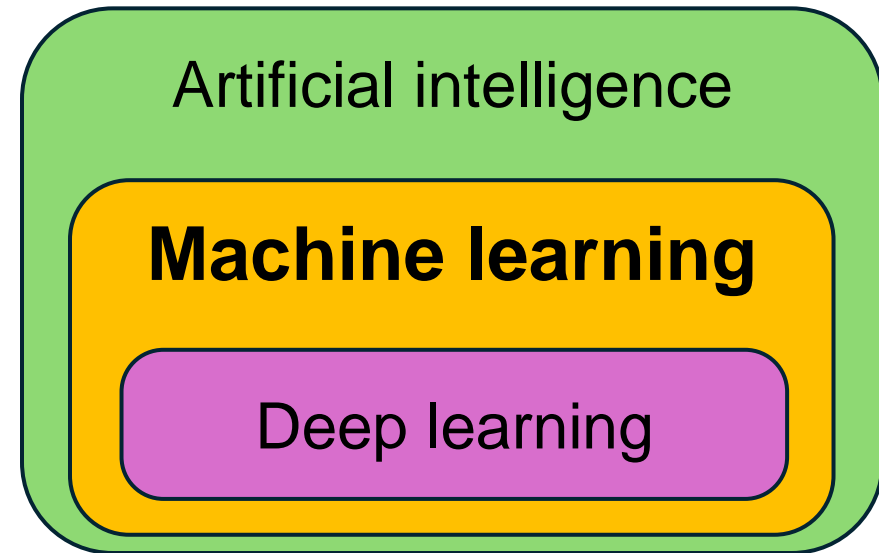
**Machine learning** – the study of algorithms that can learn from observational data, which involves training (on labeled/unlabeled data) and making predictions or decisions based on input data.

### Subset of

Artificial intelligence (broader in scope, may combine different model-types and architectures to simulate “intelligence”)

### Superset of

Deep learning (artificial neural networks, often with many layers [=deep])





# Introduction: Clustering

## Unsupervised vs. supervised machine learning

**Machine learning** – the study of algorithms that can learn from observational data, which involves training (on labeled/unlabeled data) and making predictions or decisions based on input data.

### Purpose

- Prediction of a class or a value
- Pattern recognition and feature extraction

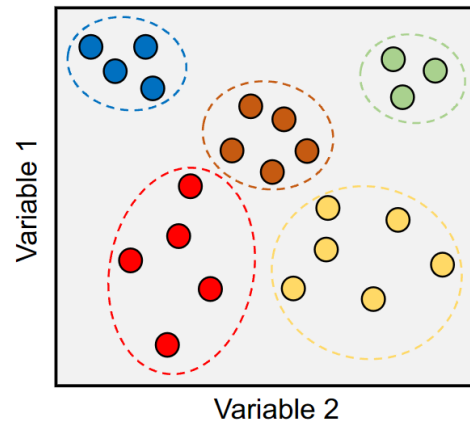
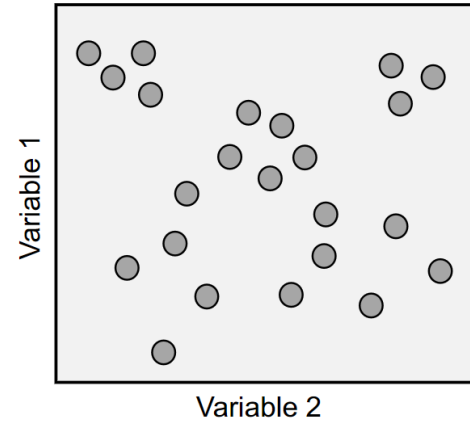
### Useful when:

- Human expertise is limited or does not exist (taxonomists)
- Huge amounts of data are available (~2 billion museum specimens)
- Models need to be customized (to a taxonomic group)

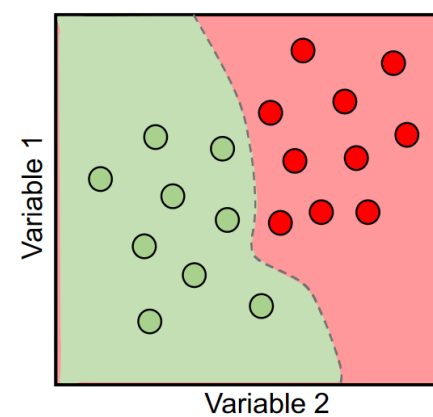
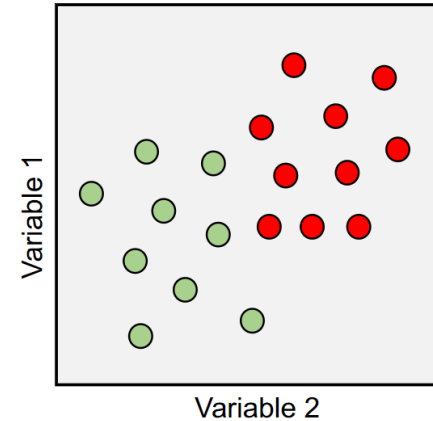
# Introduction: Clustering

## Unsupervised vs. supervised machine learning

a) Unsupervised learning



b) Supervised learning



# Introduction: Clustering

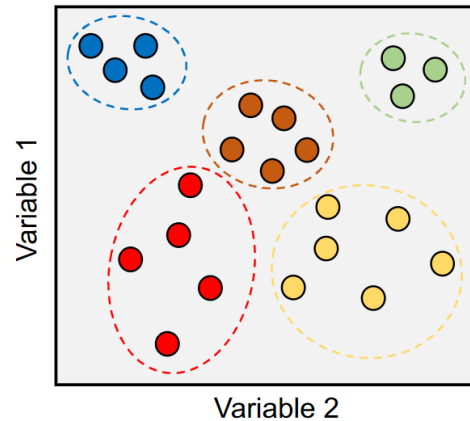
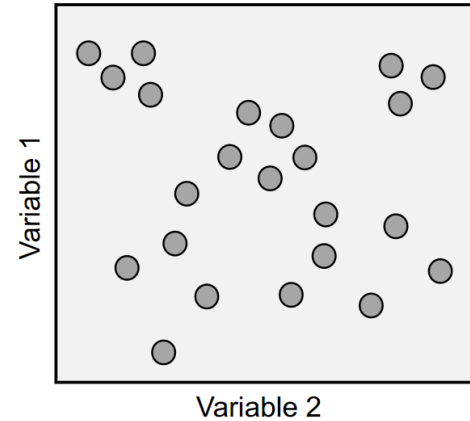
## Unsupervised vs. supervised machine learning

Traits  
(morphological,  
functional,  
chemical)

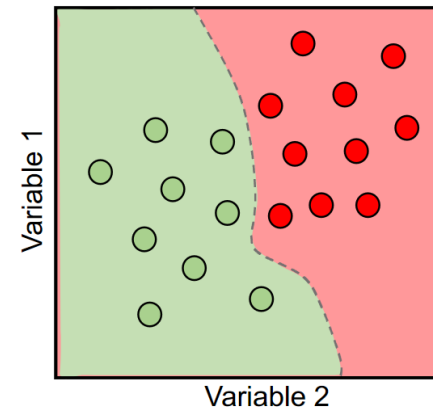
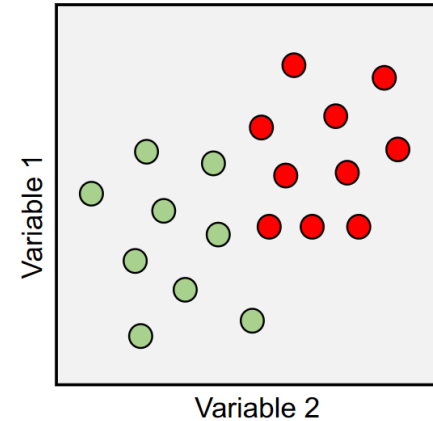


Separation  
into functional  
groups

a) Unsupervised learning



b) Supervised learning



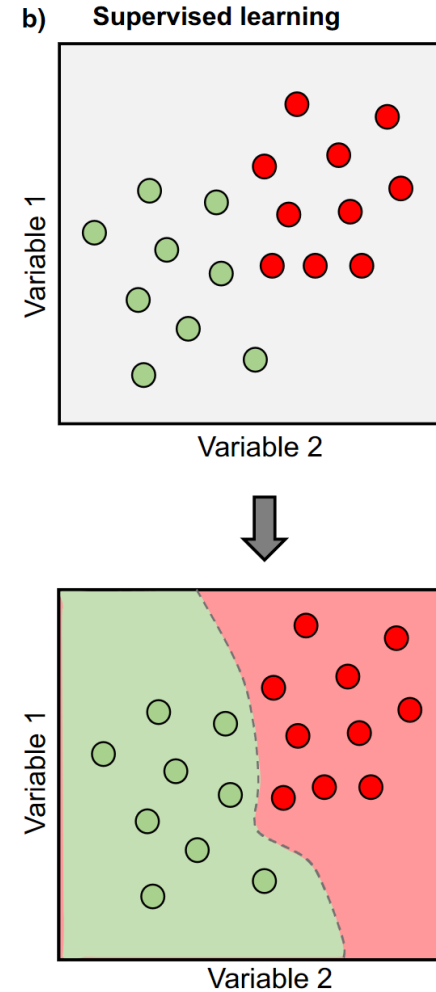
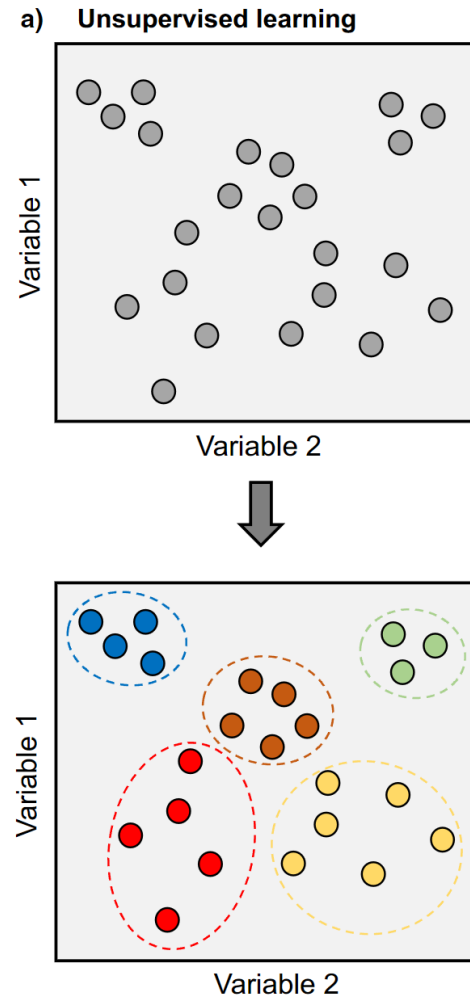
# Introduction: Clustering

## Unsupervised vs. supervised machine learning

Traits  
(morphological,  
functional,  
chemical)



Separation  
into functional  
groups



Traits + labels  
(e.g., species)



Species  
identification



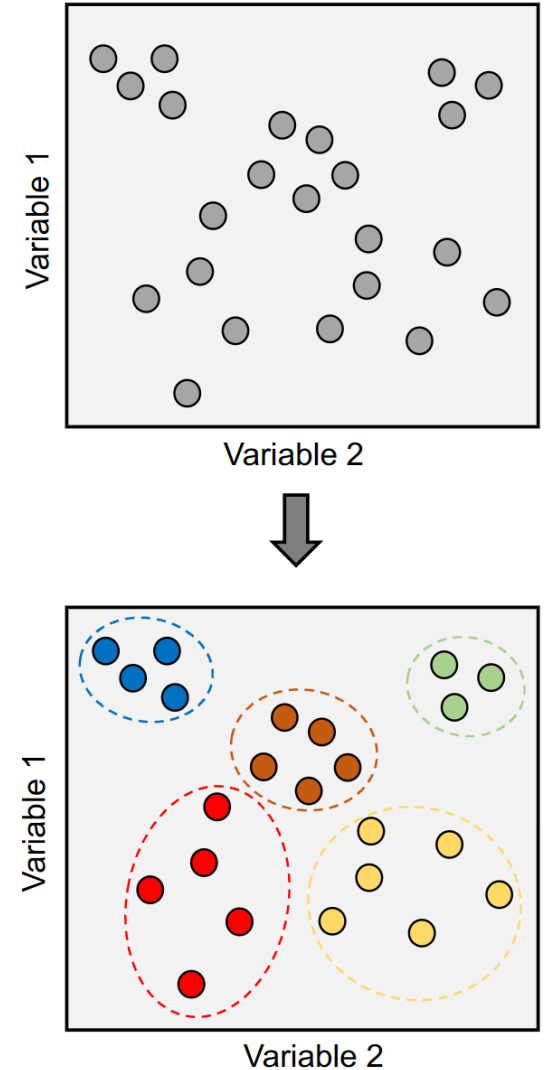
# Introduction: Clustering

## Clustering

**Definition** - type of unsupervised learning that involves grouping a set of observations in a way that members of a group (the *cluster*) are more similar to each other than to those in other groups.

**Purpose** - discovery of inherent groupings within in datasets

**Common algorithm:** k-means



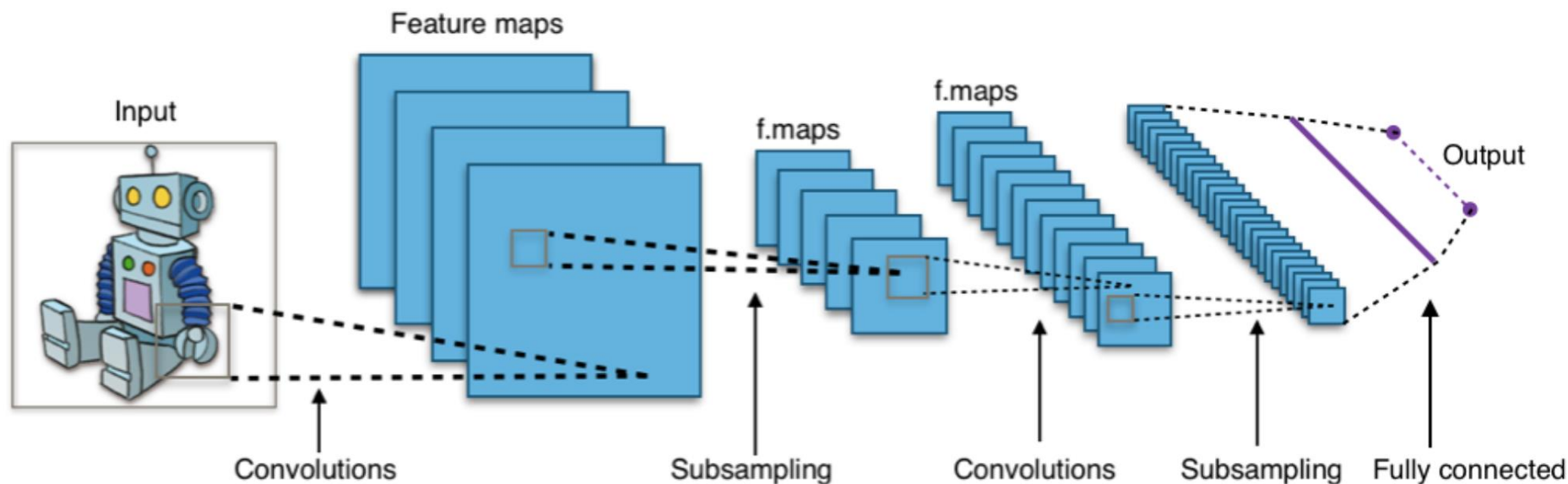
# Practical: Clustering

< Practical part - open 02\_clustering.ipynb >

# Computer vision

## Deep learning

Family of machine learning models based on neural networks – e.g., convolutional neural networks

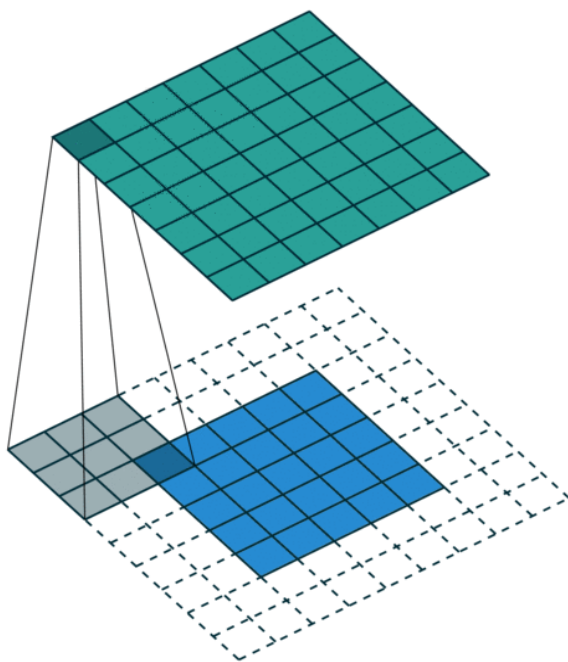


**Condensation of information**

# Computer vision

## Deep learning

Convolutions are mathematical operations by which information contained in images is condensed through multiple layers of abstraction.



Kernel iterates over image array

1 <sub>x1</sub>	1 <sub>x0</sub>	1 <sub>x1</sub>	0	0
0 <sub>x0</sub>	1 <sub>x1</sub>	1 <sub>x0</sub>	1	0
0 <sub>x1</sub>	0 <sub>x0</sub>	1 <sub>x1</sub>	1	1
0	0	1	1	0
0	1	1	0	0

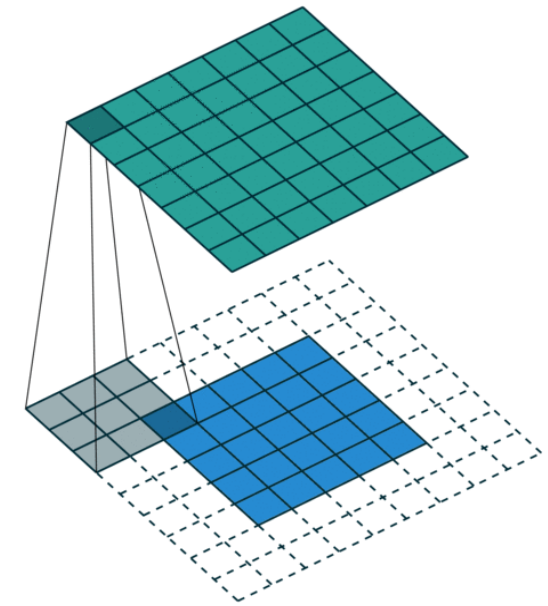
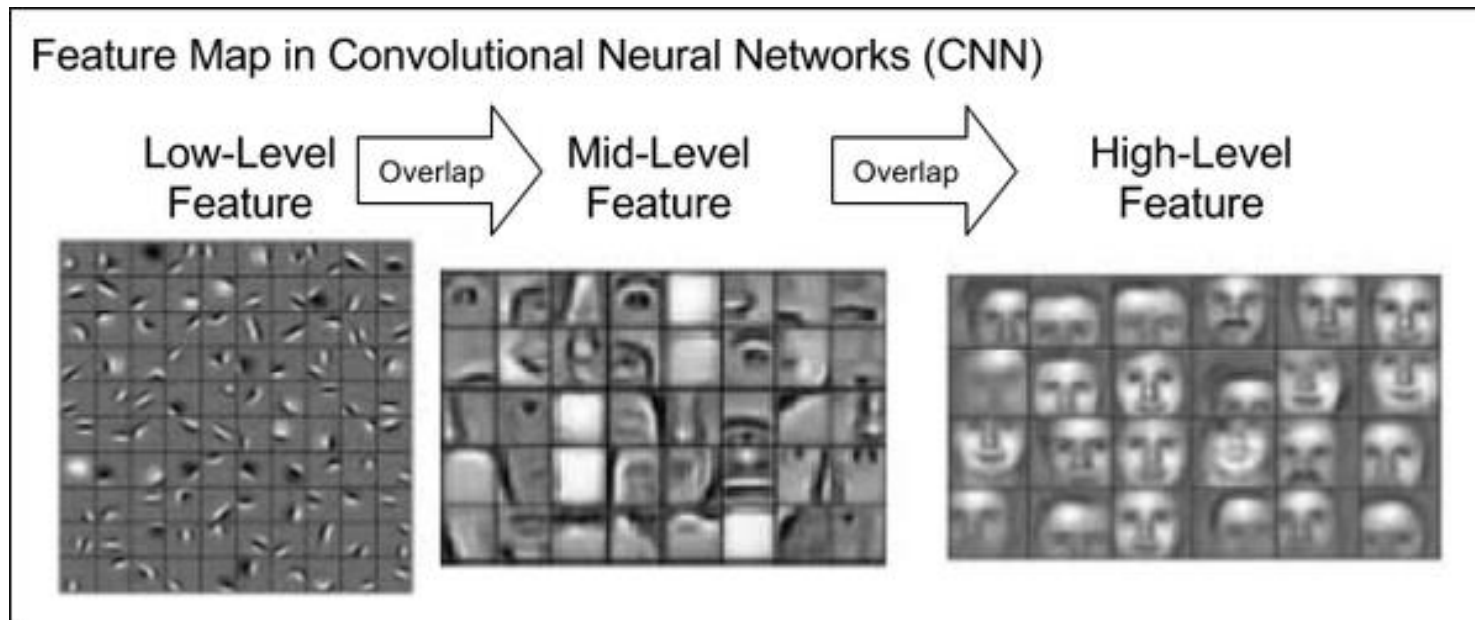
Image

4		

Convolved  
Feature

## Deep learning

Convolutions are mathematical operations by which information contained in images is condensed through multiple layers of abstraction.



Model compares image sections with learned feature / feature space for segmentation

# Computer vision

## Deep learning

Convolutions are mathematical operations by which information contained in images is condensed through multiple layers of abstraction.

