

Machine Learning for Satellite Imaging

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IEEE
HYDERABAD SECTION



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Introductory Comments

- This presentation is for educational purposes only.
- Information provided is intended to initiate discussions
- This presentation makes considerable use of public sources of information. Every attempt is made to cite the sources.
- Any mention of a company, public-sector organization, individual, or technology is neither an endorsement nor criticism of the stated party or technology

A photograph of two farmers, a man and a woman, wearing wide-brimmed straw hats and gloves, harvesting large heads of green lettuce in a field. They are both smiling and looking at each other. The background is a blurred field of more lettuce plants.

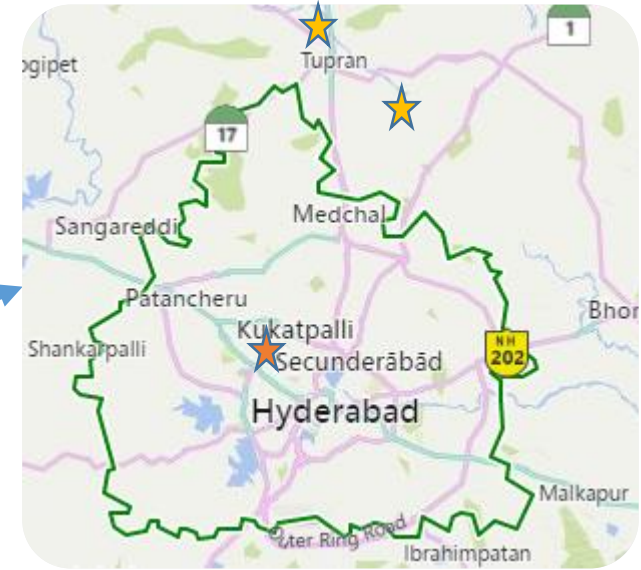
Our Purpose

To enrich the lives of those who produce and those who consume, ensuring progress for generations to come.



SUSTAINABLE DEVELOPMENT GOALS





Corporate



MCRC



TRC





UAV



Satellite
Imaging



IoT

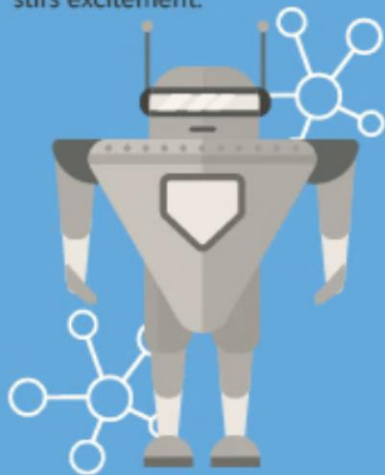


Proximal
Imaging

Performance $\propto f(G, E, M)$

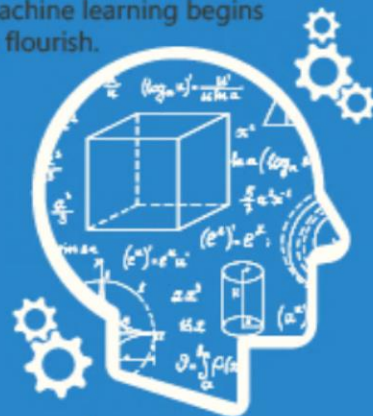
ARTIFICIAL INTELLIGENCE

Early artificial intelligence stirs excitement.



MACHINE LEARNING

Machine learning begins to flourish.



DEEP LEARNING

Deep learning breakthroughs drive AI boom.



1950's 1960's 1970's 1980's 1990's 2000's 2010's

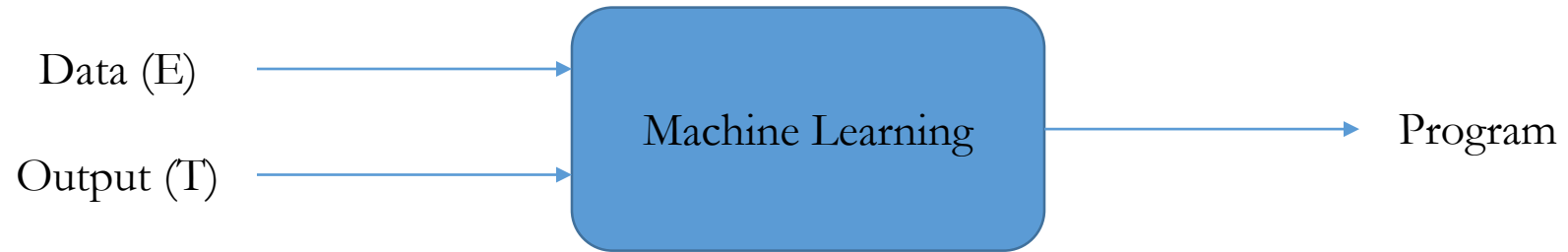
Since an early flush of optimism in the 1950's, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.

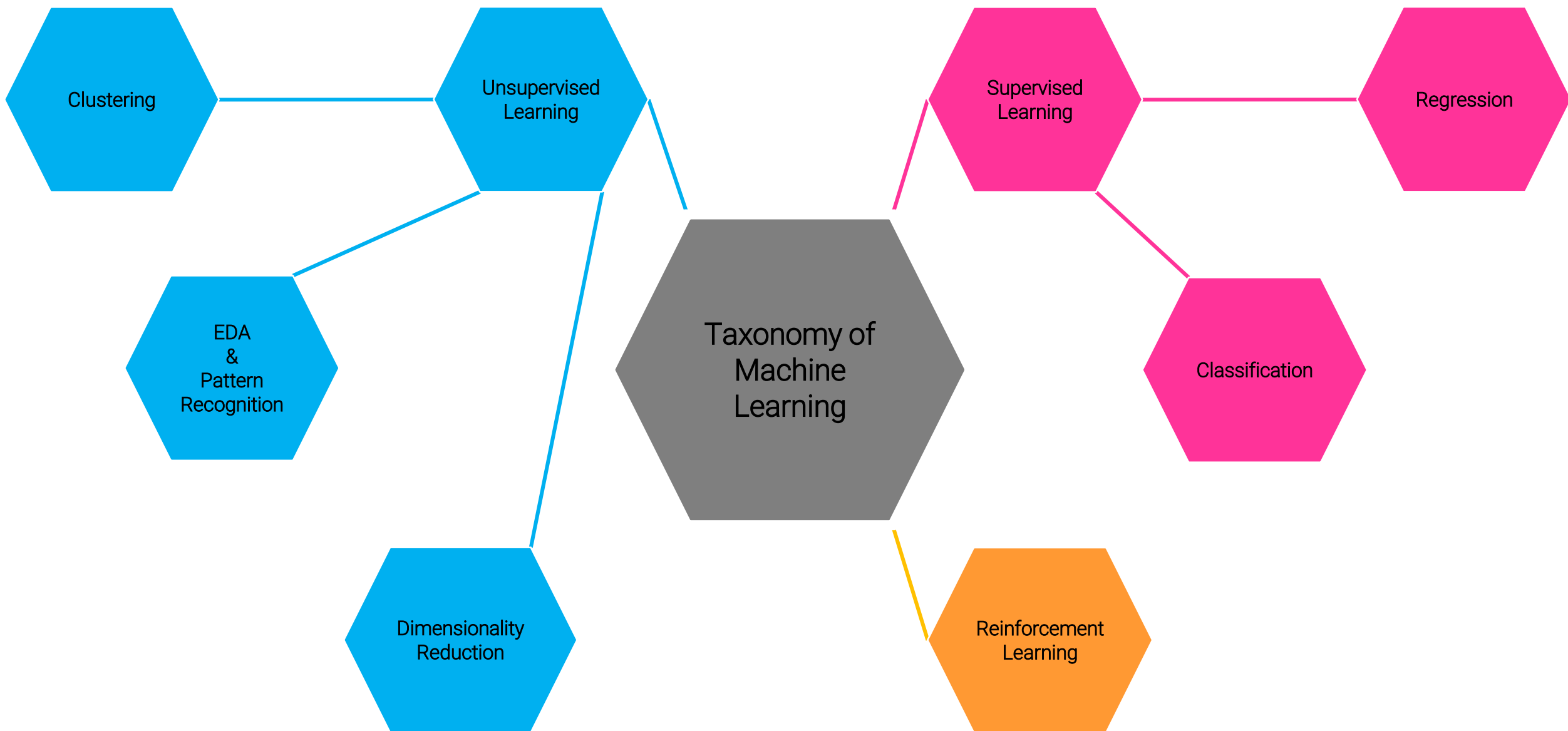
What is Machine Learning?

Definition

A computer program is said to learn from experience \mathbf{E} with respect to some class of tasks \mathbf{T} and performance measure \mathbf{P} , if its performance at tasks T , improves with experience E

-Mitchell 1997





Prerequisites for this course

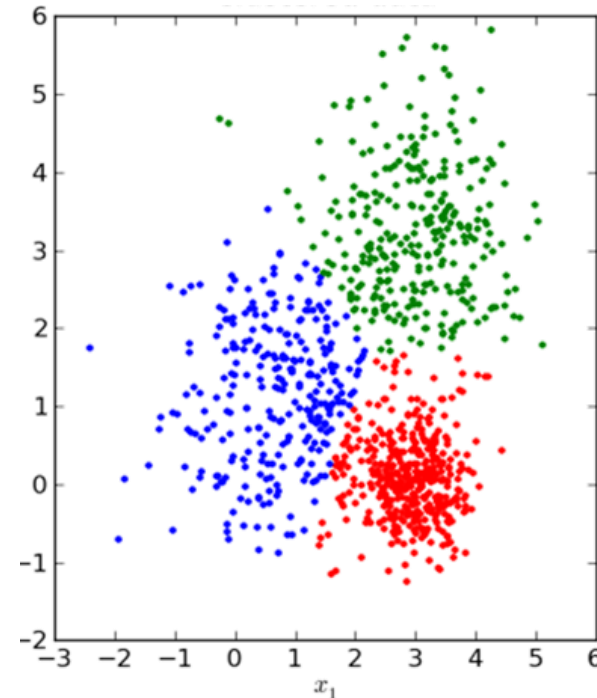
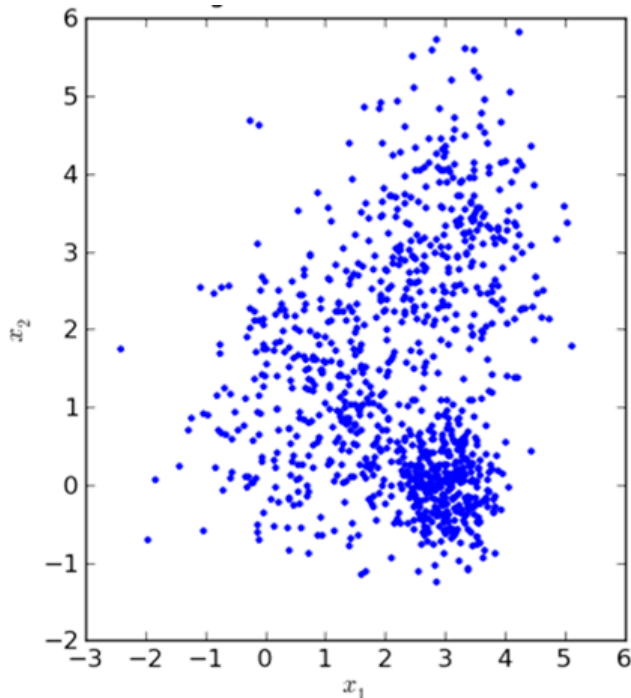
- Linear Algebra
- Basic of image processing
- Python programming
- Google Account

Dataset

<http://tiny.cc/91oxhz>

Unsupervised Learning

- Dataset (**E**): $X = (x_1, x_2, \dots, x_n)$
- Task (**T**): To find the interesting structures in the data (No labels)



K-means

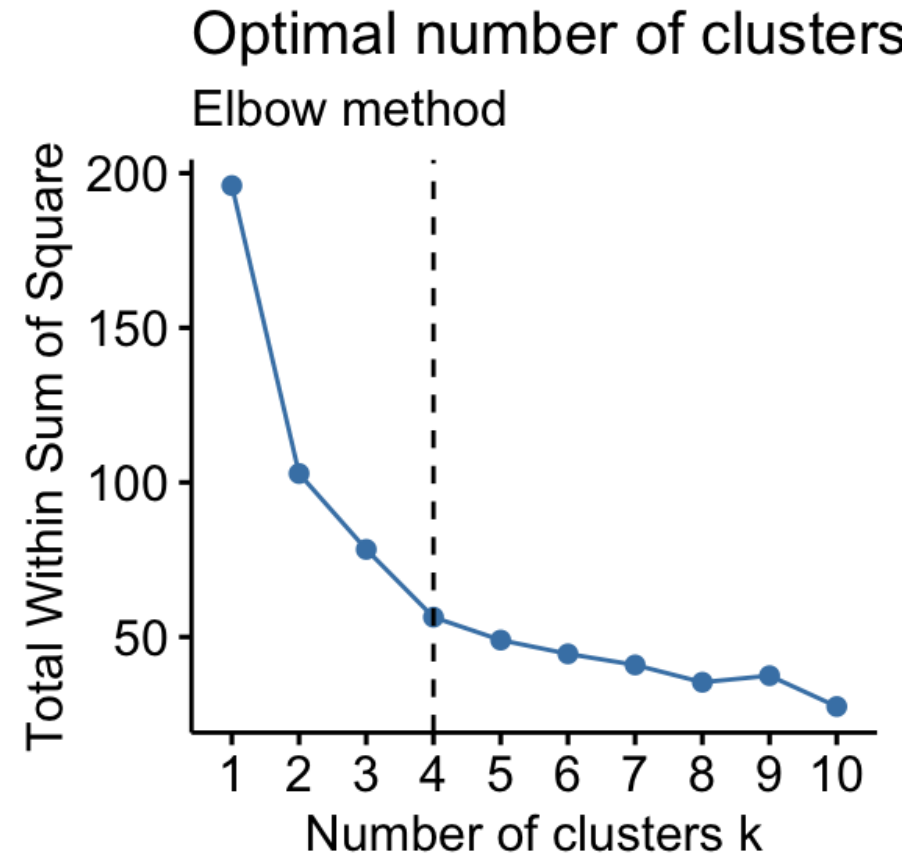
- Given a set of points: $X: (x_1, x_2, \dots, x_n)$
 - Objective function to minimize: $\min_{\{l_c\}_{c=1}^k} \sum_{c=1}^k \sum_{x \in l_c} \|x - \mu_c\|_2^2$, where $\mu_c = \frac{1}{|l_c|} \sum_{x \in l_c} x$
1. Choose the number of clusters: k
 2. Assign X to the k clusters
 3. M: Calculate the mean of each cluster and assign cluster centers to it
 4. E: Reassign X to the clusters based on Euclidean distances

B. Kulis, *et al.* "Revisiting k -means: New Algorithms via Bayesian Nonparametrics", ICML, 2012.

$$l^*(i) = \operatorname{argmin}_c \|x_i - \mu_c\|_2^2$$

<https://icml.cc/2012/papers/291.pdf>

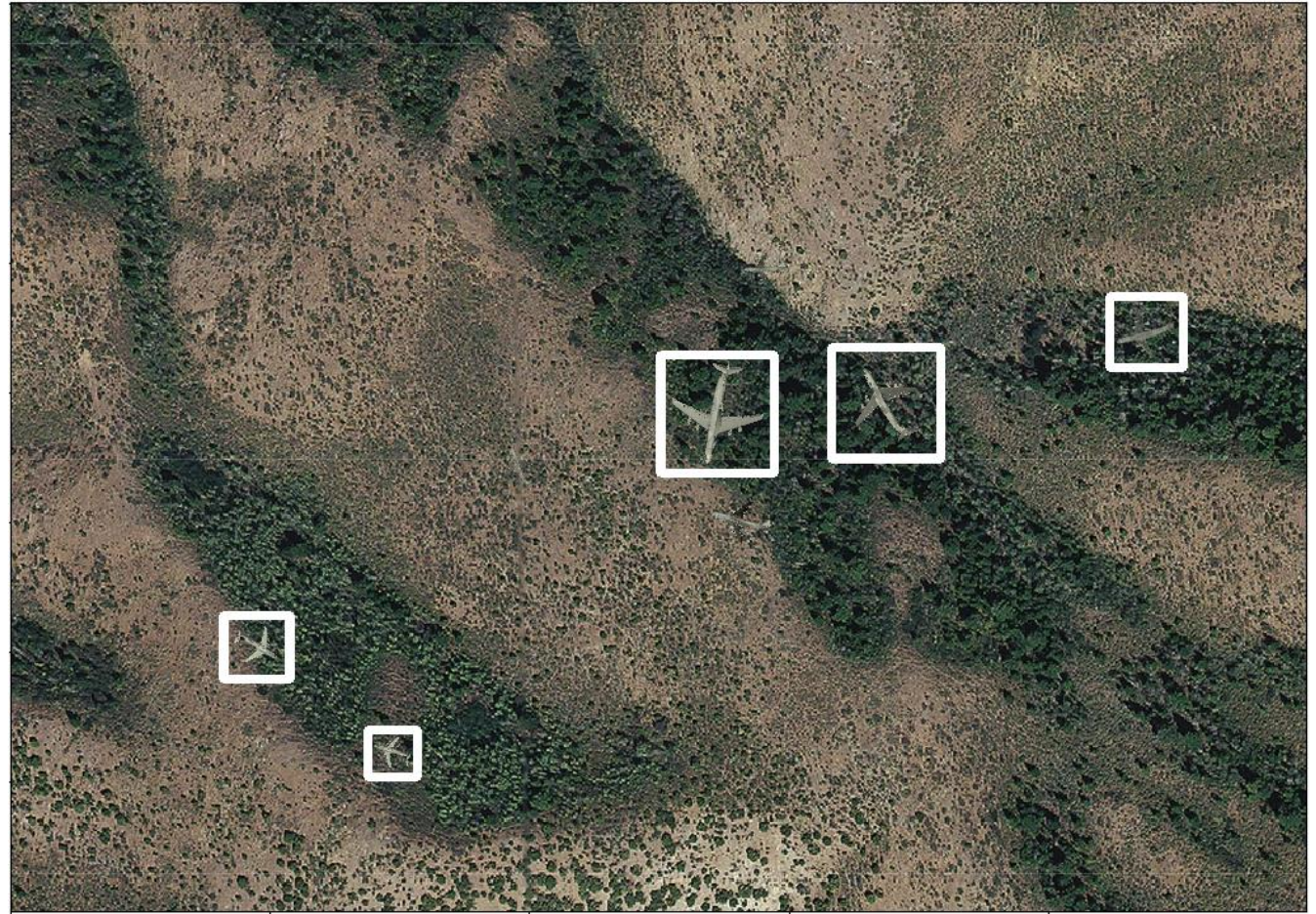
Optimal Cluster Number



Supervised Learning

Example Object Identification

- X : Pixel values
- y : Bounding box

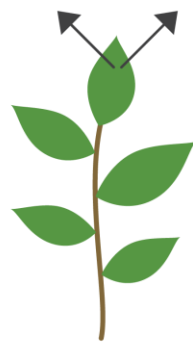


Feature Engineering

- Example: Normalized difference vegetation index (NDVI)

HEALTHY
VEGETATION REFLECTANCE

50% NIR 8% RED



NDVI = 0.72

STRESSED
VEGETATION REFLECTANCE

40% NIR 30% RED



NDVI = 0.14

$$\text{NDVI} = \frac{\text{NIR} - \text{RED}}{\text{NIR} + \text{RED}}$$

Feature Engineering

- Example: Normalized difference Water index (NDWI)

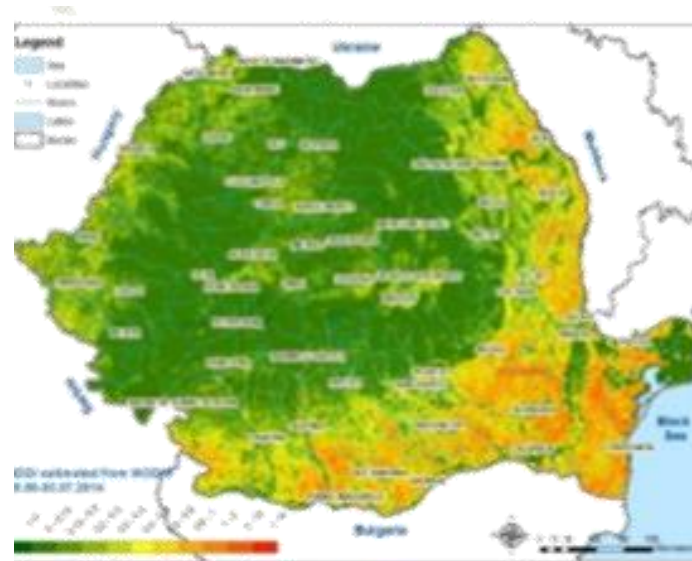
$$NDWI = \frac{B - NIR}{B + NIR}$$



Feature Engineering

- Example: Normalized Difference Drought Index (NDDI)

$$NDDI = \frac{NDVI - NDWI}{NDVI + NDWI}$$

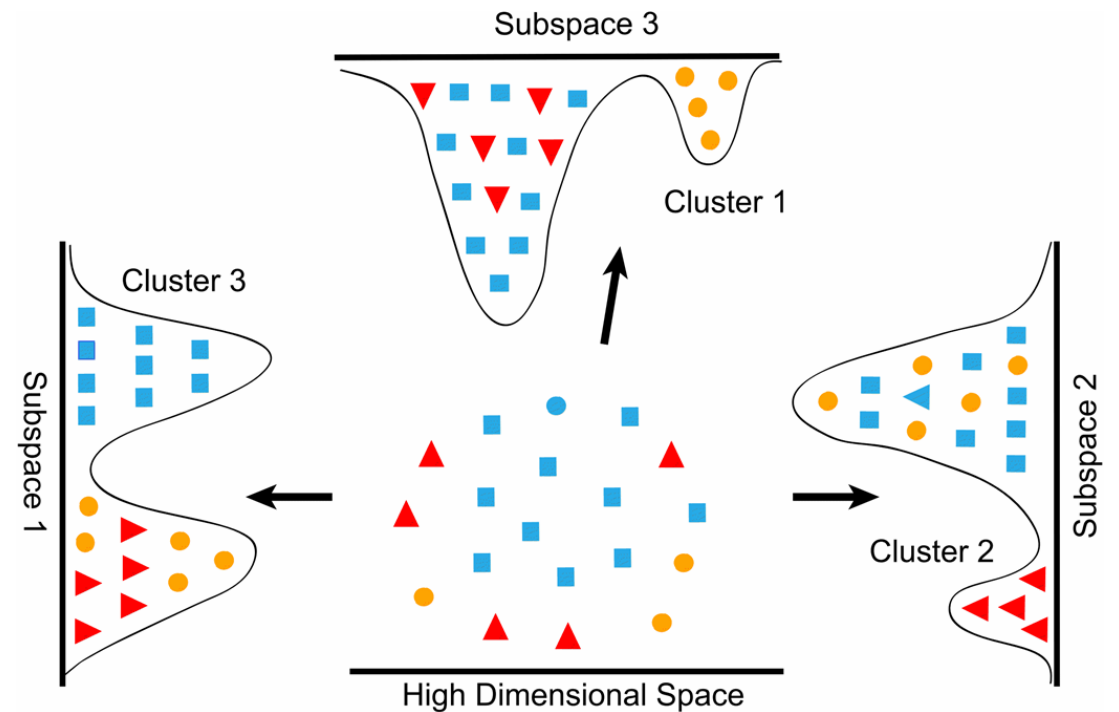


Dimensionality Reduction

- Reducing the dimensions of the feature space is called

Dimensionality Reduction

- Feature Elimination
- Feature Extraction



Steps for PCA

- Find the sample mean

$$\boldsymbol{\mu} = \frac{1}{N} \sum_i \mathbf{x}_i$$

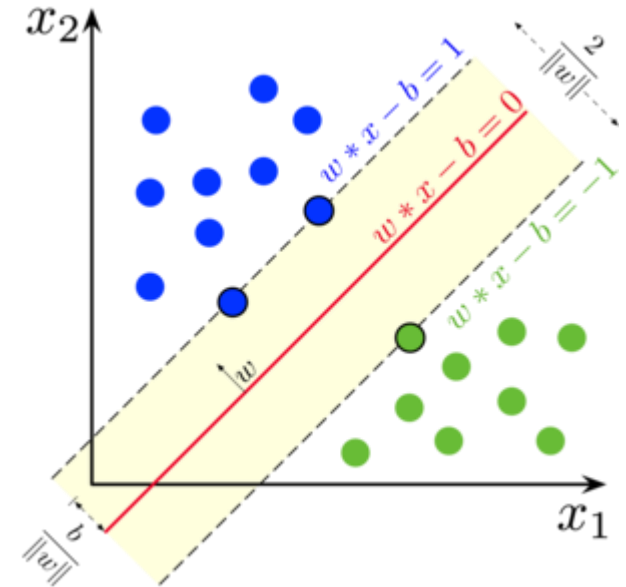
- Compute the Covariance $C = \frac{1}{N} X X^T = \frac{1}{N} \sum_i (\mathbf{x}_i - \boldsymbol{\mu})(\mathbf{x}_i - \boldsymbol{\mu})^T$

- Find the Eigen values and vectors for C $VC = \Lambda C$

- Arrange them and transform $Y = V^T X$

Supervised Learning

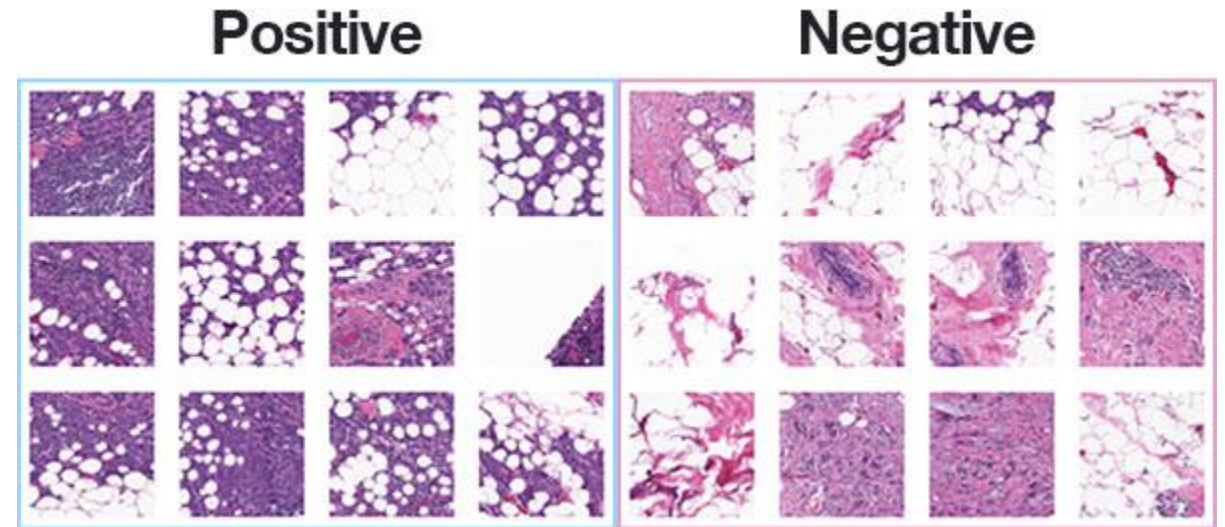
- X : input (E) $\in \mathbb{R}^M \times \mathbb{R}^N$
- y : Output (T) $\in \mathbb{R}^M$
- Regression: when Y is a continuous variable
- Classification: when Y is discrete



Supervised Learning

Example Image Classification

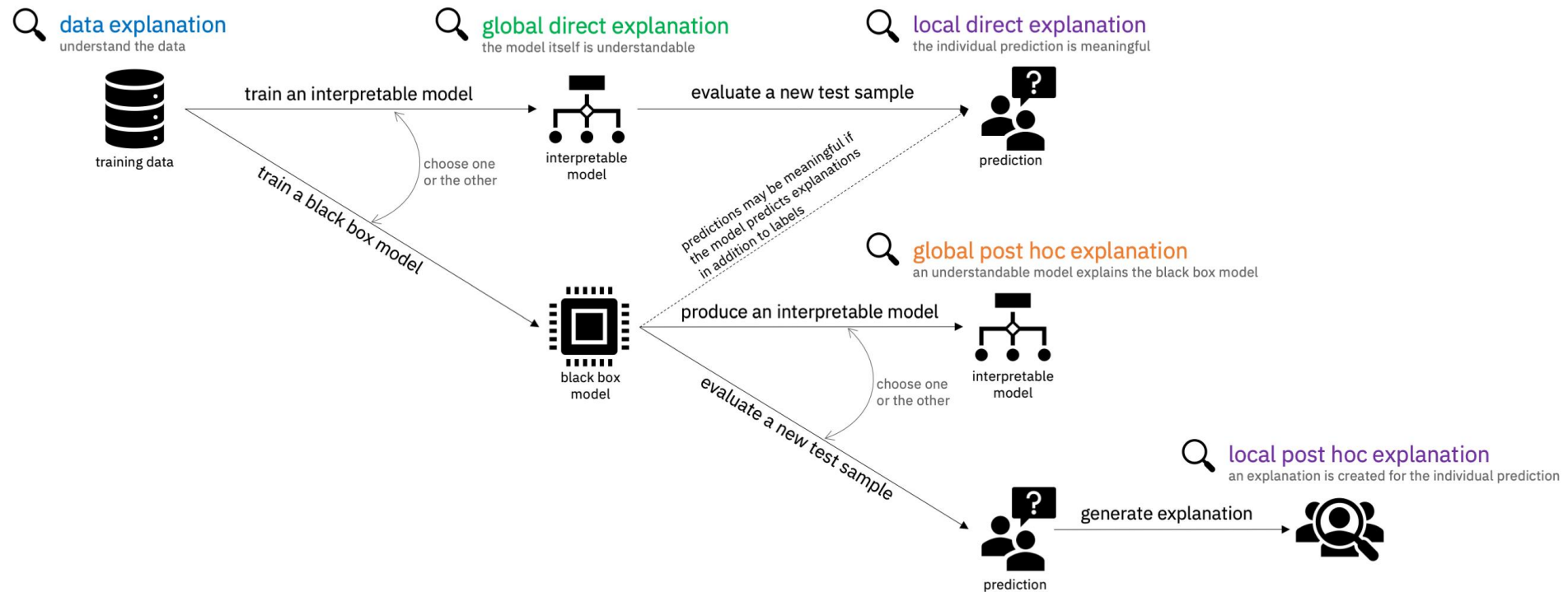
- X : Pixel values
- y : Class that the images belong to





Human-Centered AI

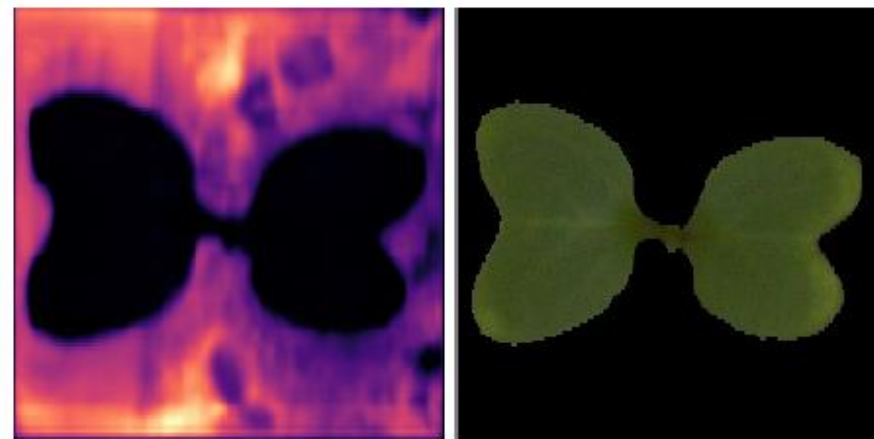
Explainable Model and Trust



Feature 1



Post-hoc

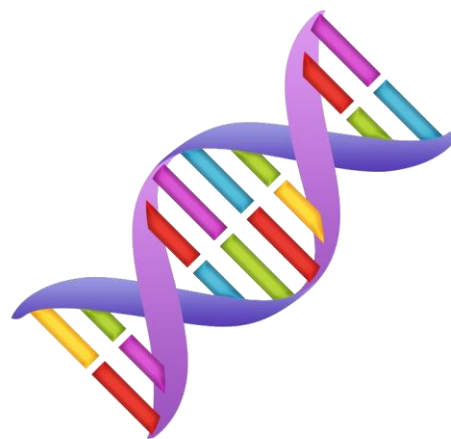


(a) Checker board pattern

(b) Output activation as a mask on the



Technology



Science



Collaboration



Thank You!



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