

# Special Topic: Machine Learning (ML) and Artificial Intelligence (AI)

ENGR 103: Engineering Computation and Algorithmic Thinking

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# Today's Lecture

- What is AI, ML, Deep Learning?
- A Typical ML Workflow
- Data Collection and Preparation
- Different Models

**How many of you have used *AI today*?**

# Applications

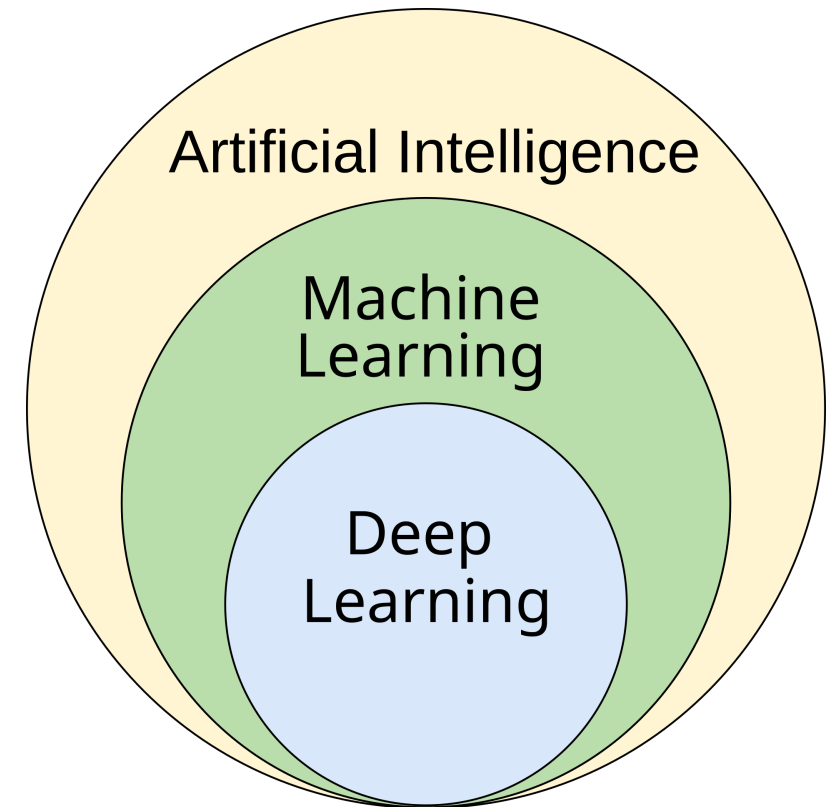
- Agriculture: optimal plant growth
- Banking: fraud detection
- Cybersecurity: attack detection and handling
- Civil: structural health monitoring
- Mechanical: predictive maintenance in machines
- Healthcare: medical diagnosis of patients
- Logistics: routing and optimization

# AI, ML, and Deep Learning

**Artificial Intelligence (AI)** refers to the capability of computational systems to perform tasks typically associated with human *intelligence* (learning, reasoning, problem-solving, perception, and decision-making).

**Machine Learning (ML)** is a subset of AI that learns patterns from data.

**Deep learning** is a subset of machine learning that focuses on utilizing *neural networks*.



# Basic ML Workflow

1. Collect and prepare data (images, sensor readings, text, ...)
2. Train a model (feeding examples)
3. Make predictions (on new data)
4. Improve over time (learning from mistakes)

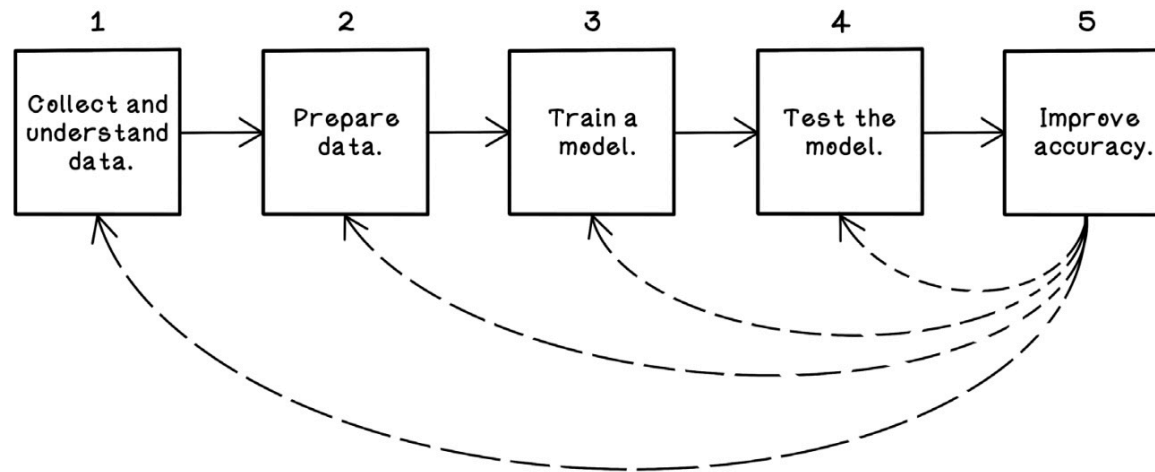


Image from R. Hurbans, Grokking Artificial Intelligence Algorithms.

# Data Collection and Processing

Data scientists often spend **60-80% of their time on data-related tasks**: collecting, cleaning, transforming, and exploring data.

Raw data is **messy**: missing values, outliers, inconsistent formats, or irrelevant features can tank a model's performance.

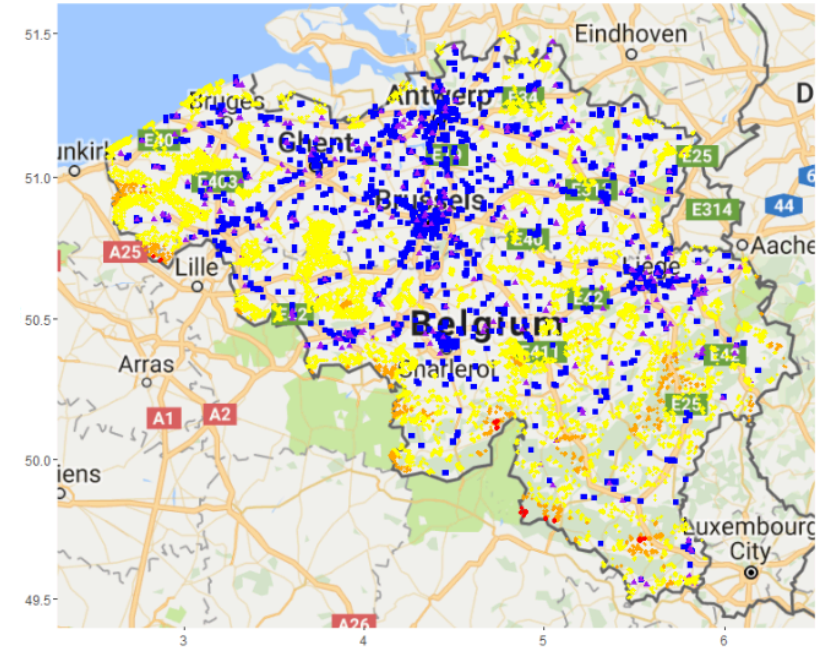
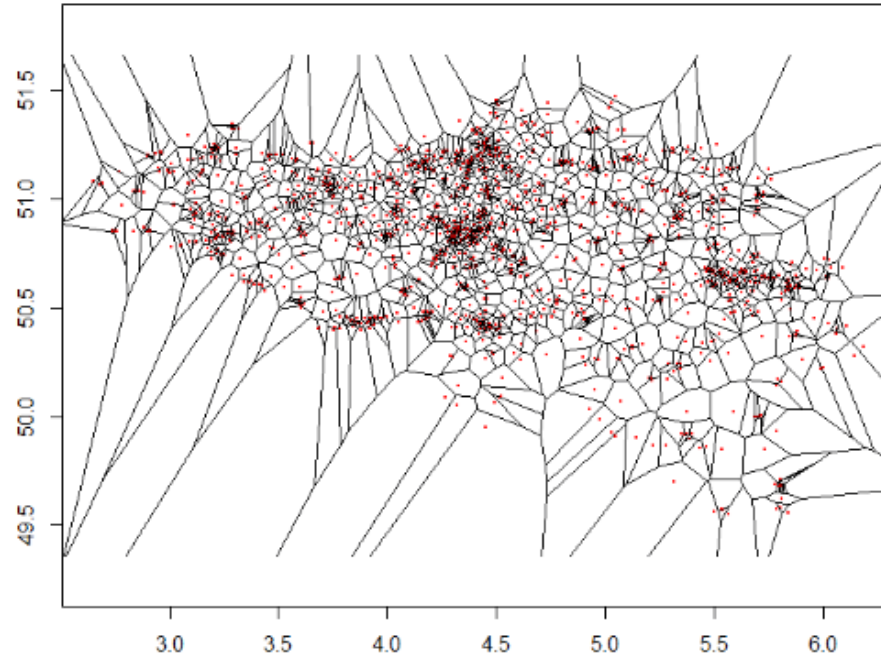
**Garbage in, garbage out.** A model's only as good as its data. Clean, well-prepared data ensures the model learns meaningful patterns, not noise.

Not all datasets are terrible.

When doing supervised learning, we want to have *train* and *test* data.

# Data Processing Example

Voronoi tessellation  
to compute  
distances between  
~700k people and  
~ 1.2k repair shops





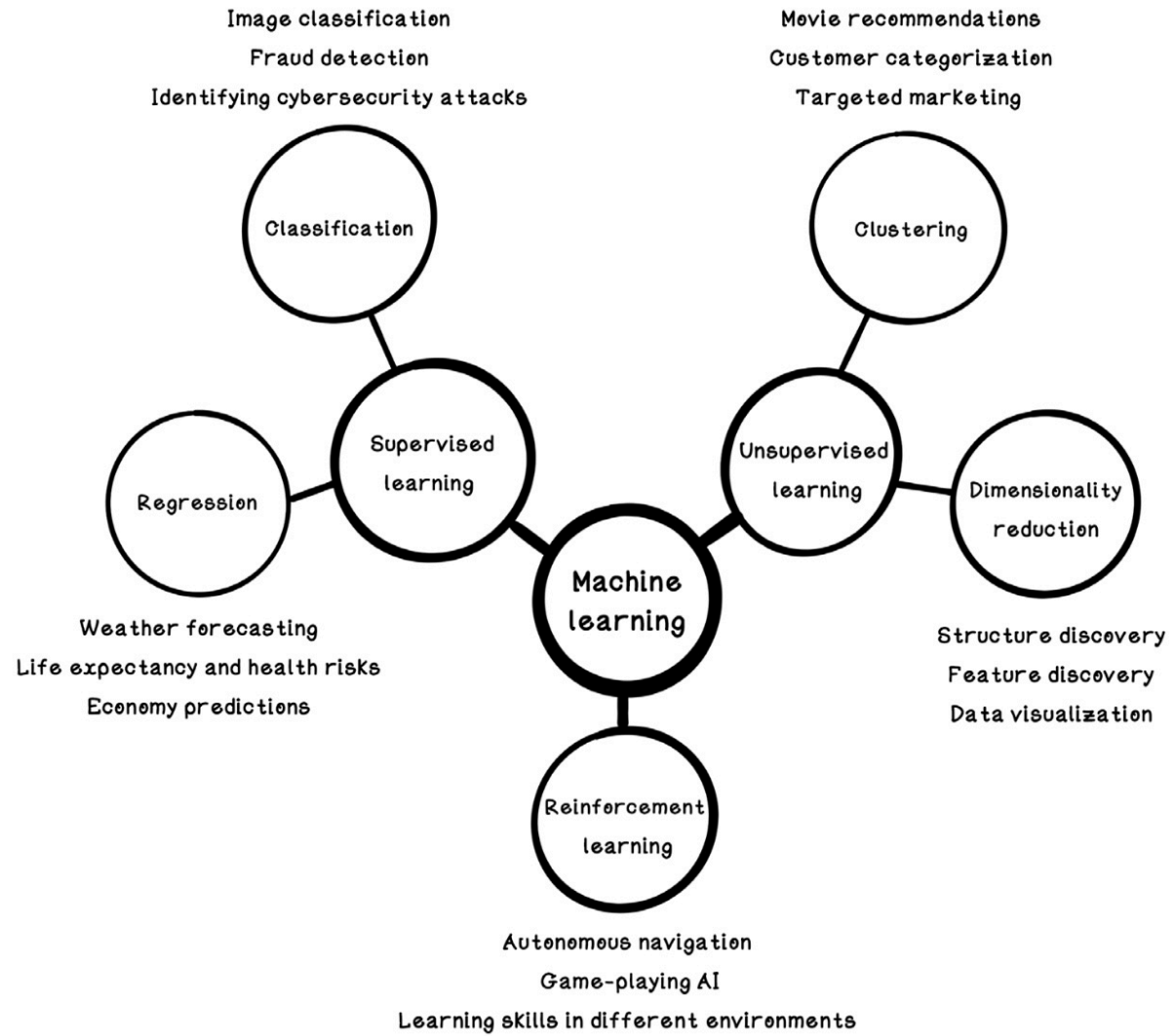
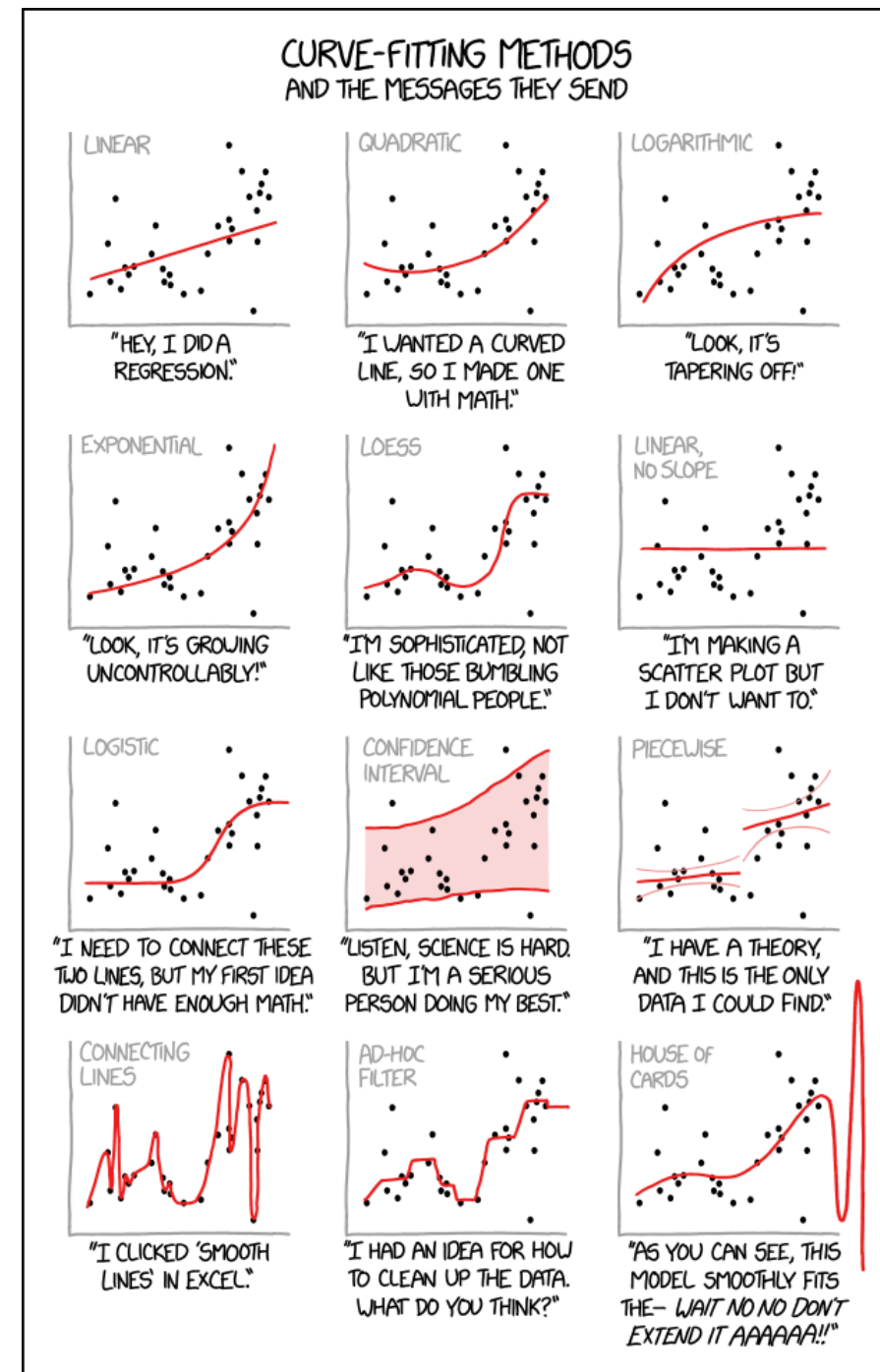


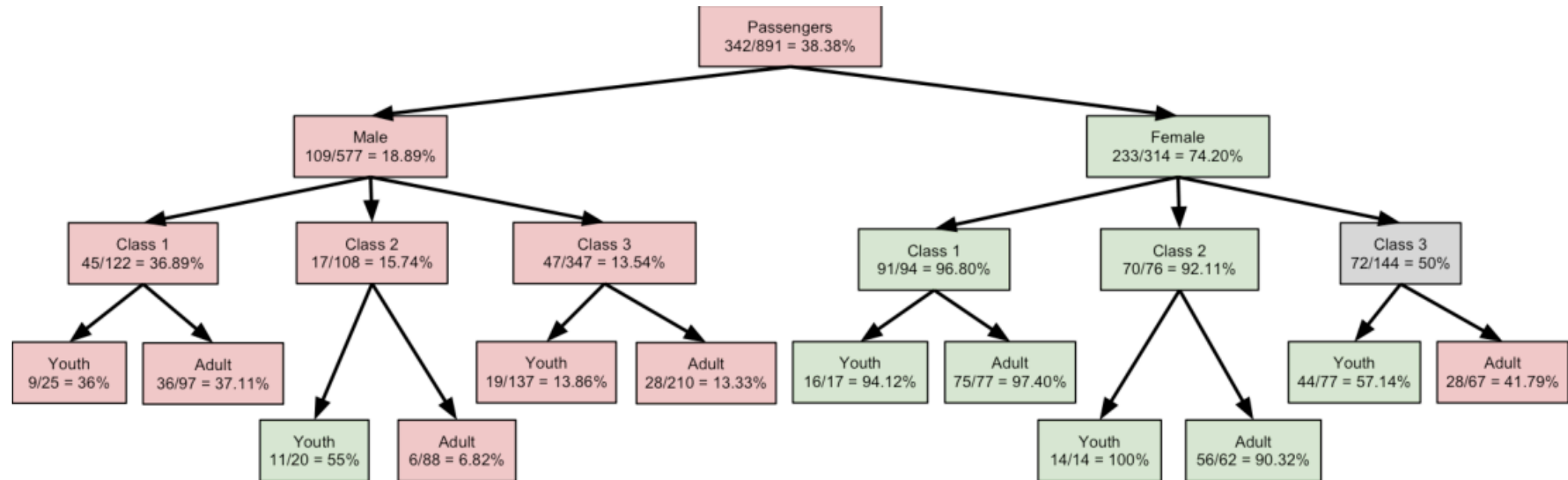
Image from R. Hurbans, Grokking Artificial Intelligence Algorithms.

# Regression

<https://xkcd.com/2048/>



# Classification



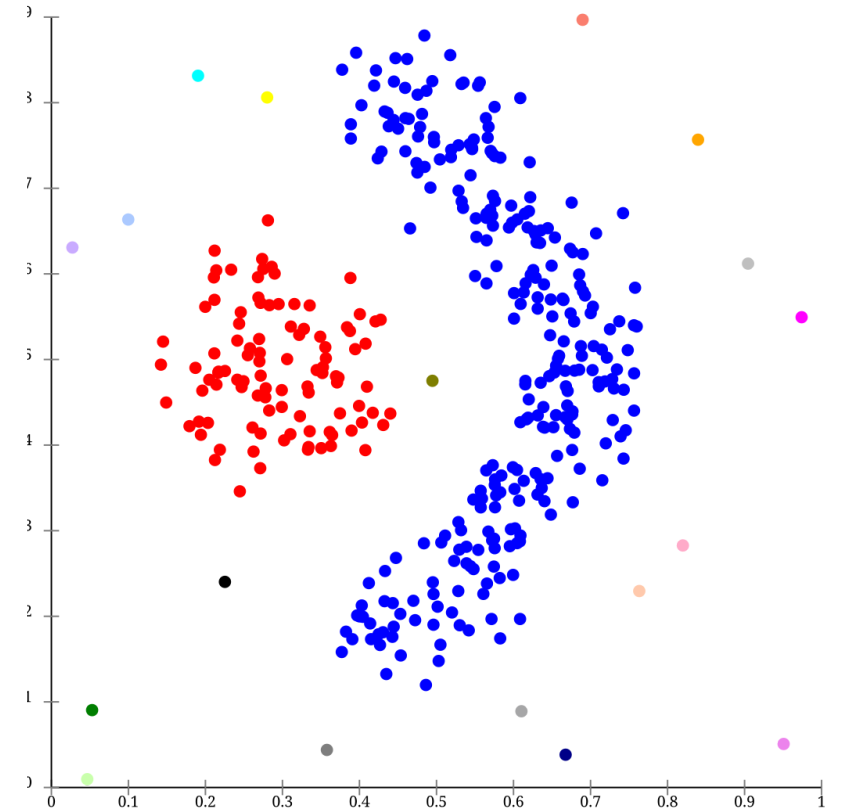
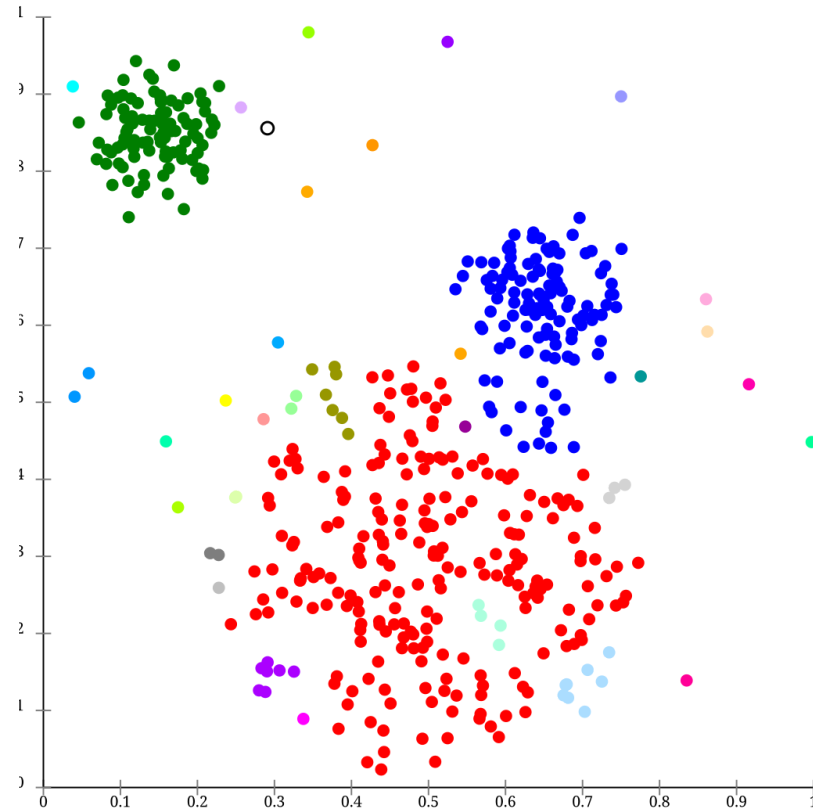
Can we predict Titanic survivors? **Green** mostly survived. **Red** mostly died.

CS229 Titanic – Machine Learning From Disaster

# Clustering

Centroid vs  
Density

Source

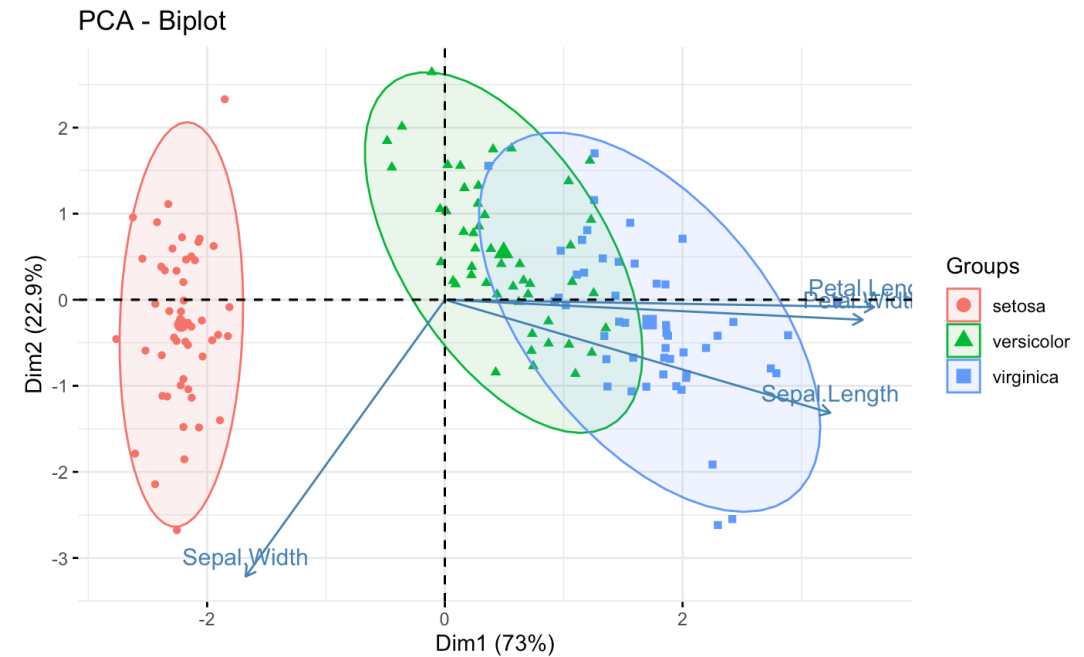
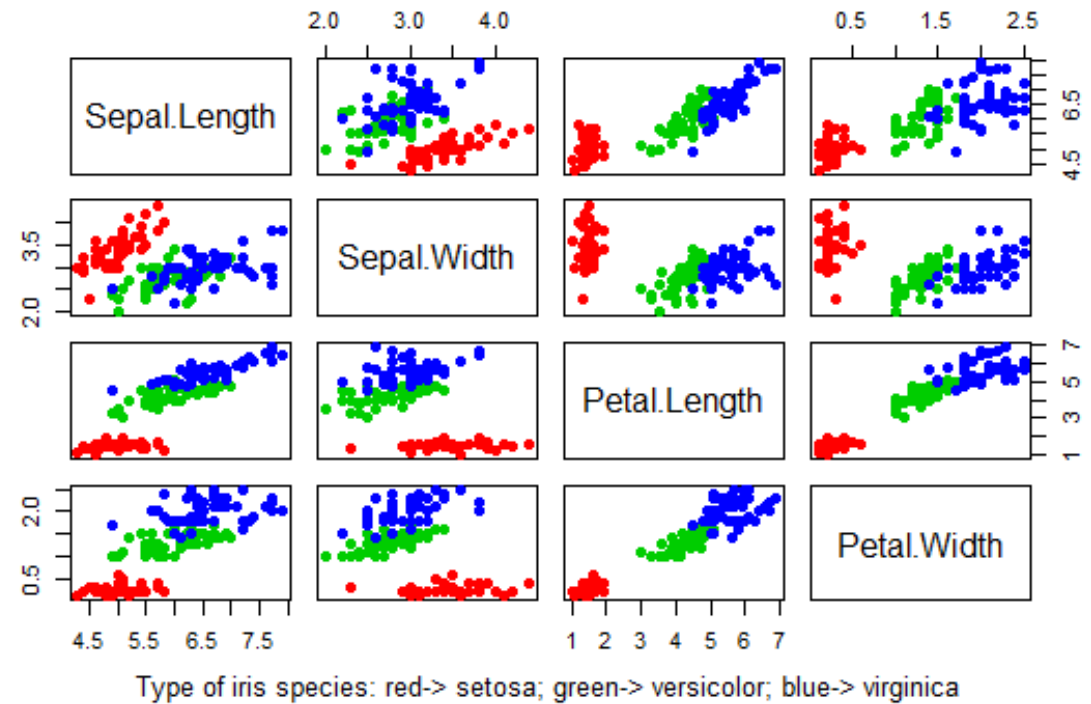


# Dimensionality Reduction

Scatter Plots

Principal Component Analysis (PCA)

*Iris Dataset*

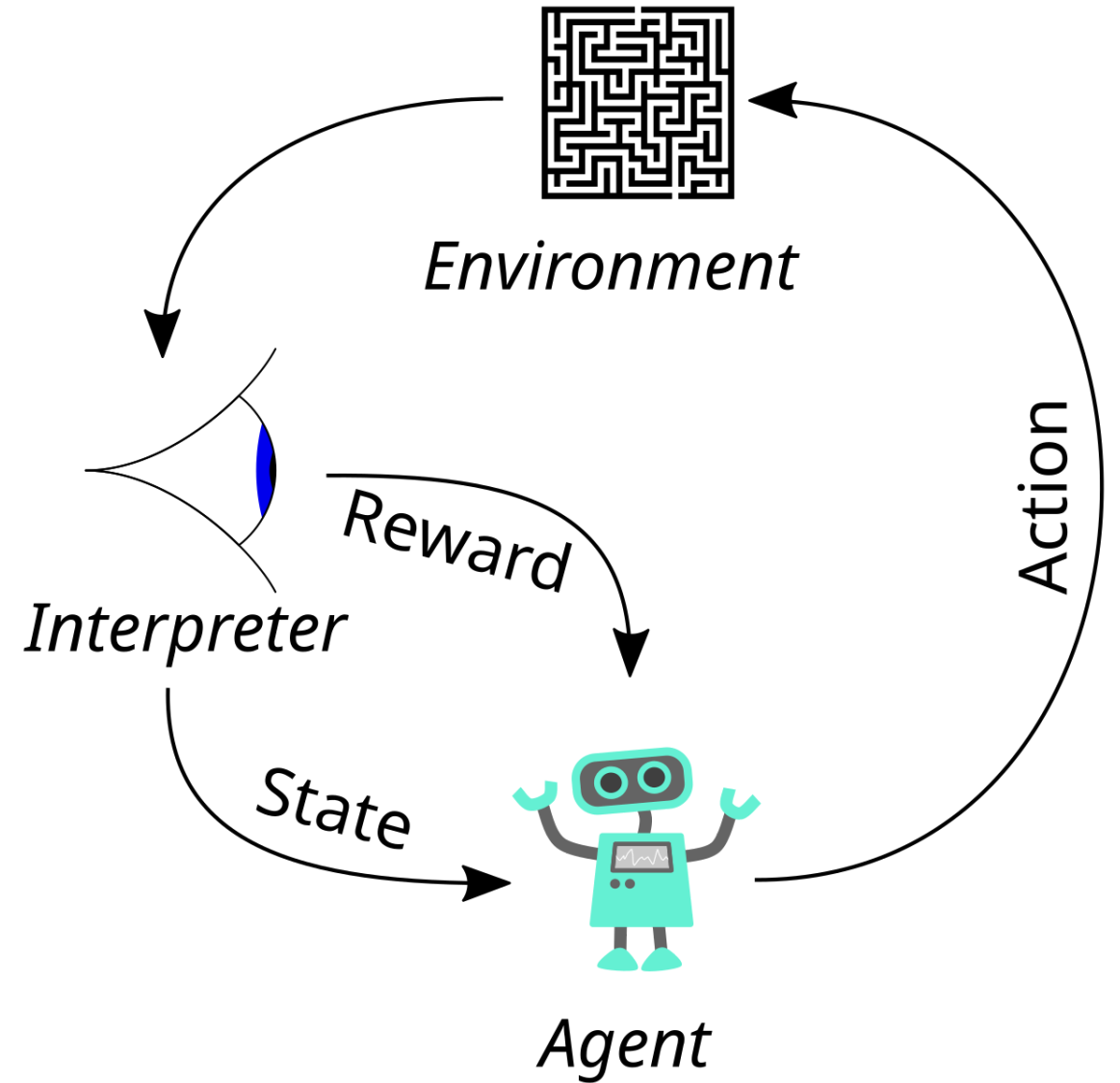


# Reinforcement Learning

Imagine trying to teach a pup how to sit by giving them a reward if they succeed.

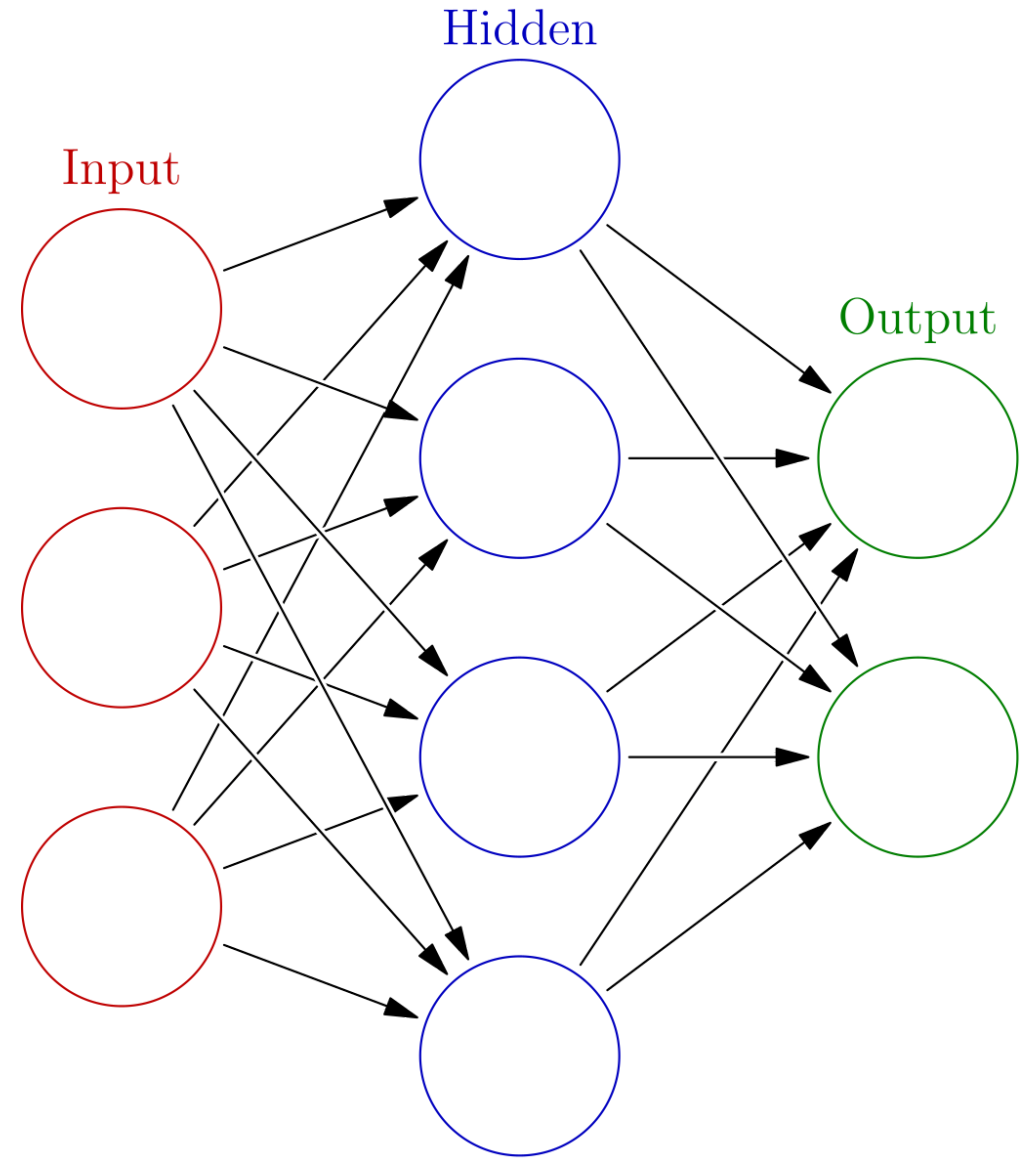
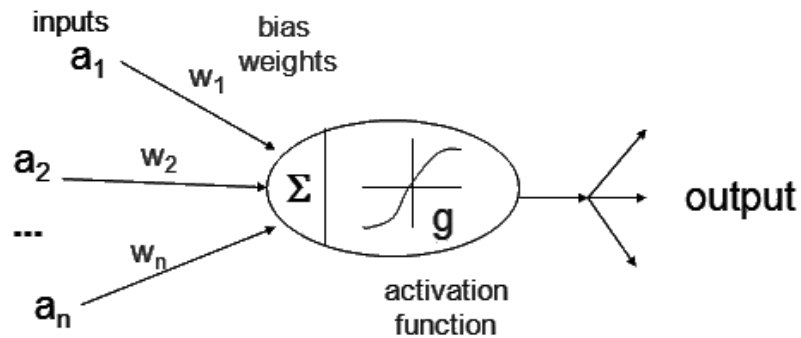
In this case, the model/agent improves when making successful attempts.

Source



# Neural Networks

Nodes (neurons) connected by edges (synapses). Each node receives information from connected nodes, processes it, then sends it to other nodes.



Source

# Evaluating Model Quality

It depends on the type of model:

- classification: **confusion matrix** or similar
- regression: some measure of the "error"
- reinforcement learning: some measure of the "reward"

All metrics have some trade-offs.

		Predicted condition	
		Positive (PP)	Negative (PN)
Actual condition	Total population = P + N		
	Positive (P)	True positive (TP)	False negative (FN)
	Negative (N)	False positive (FP)	True negative (TN)



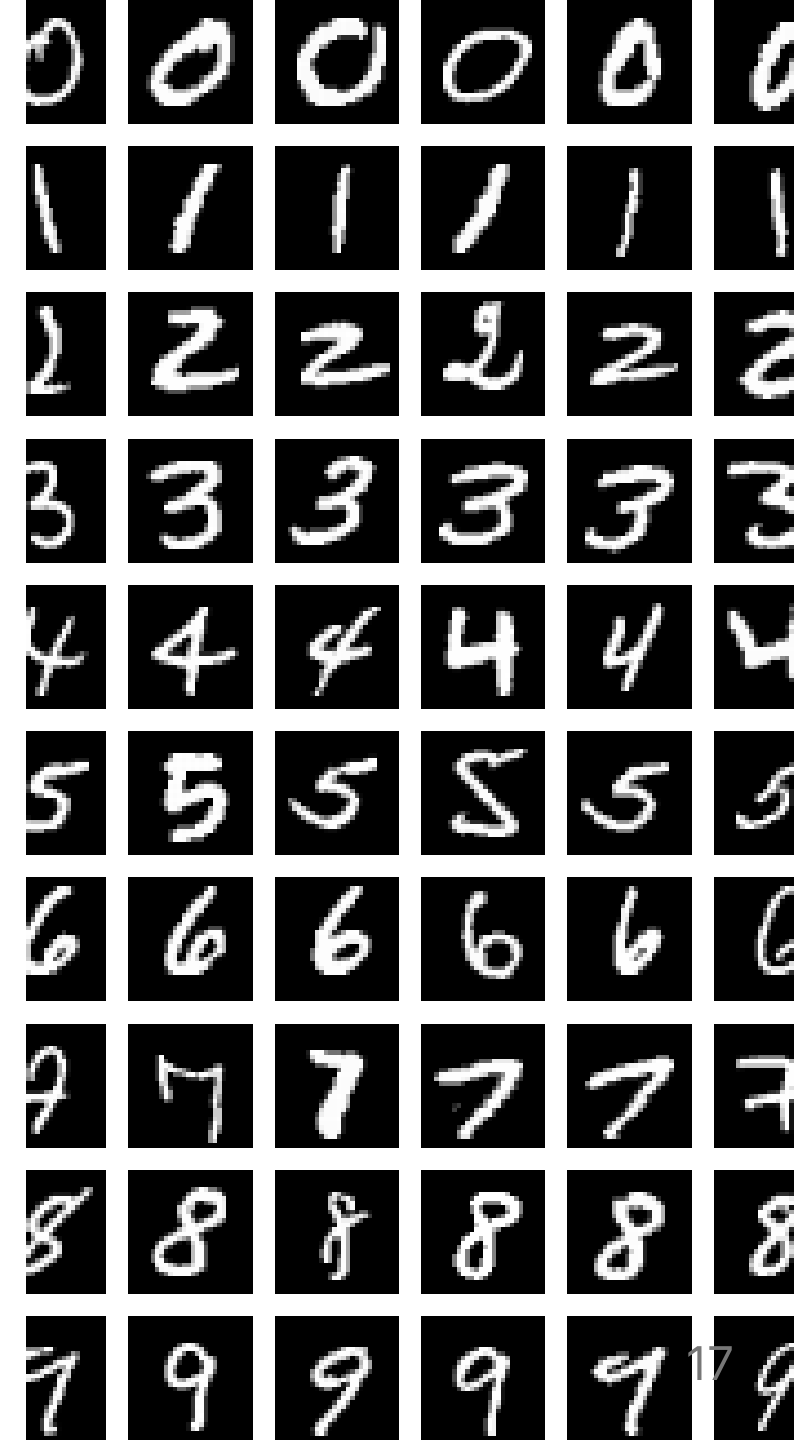
# Google Colab Exercise

Classifying handwritten digits: [MNIST Database](#)

Using neural networks (simple to complex).

Open <https://colab.research.google.com/> to get started.

Code is [available on GitHub](#) as well.



# Worth Checking

- [AlphaGo Documentary](#)
- [HuggingFace LeRobot](#)
- Conversational AI and Search: Grok, ChatGPT, Gemini, LeChat, Copilot, Perplexity
- [Teachable Machine](#)
- [TensorFlow Playground](#)
- [Kaggle's Intro to Machine Learning](#)
- [AI Dungeon](#)
- [Quick, Draw!](#)