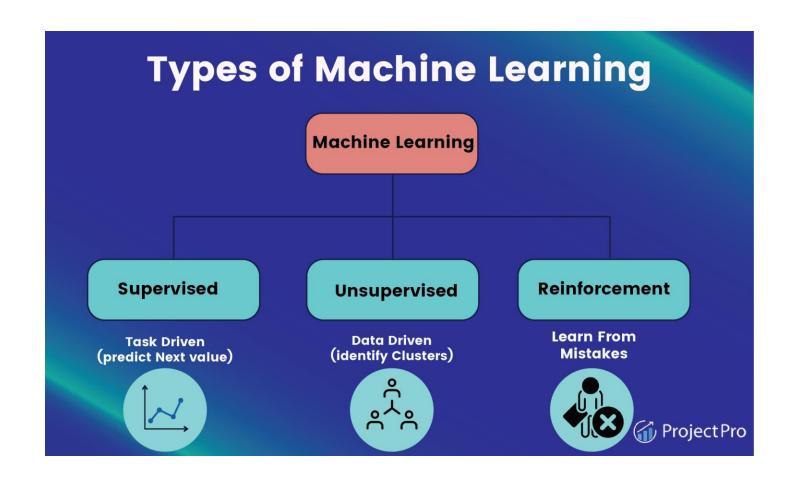
#### MACHINE LEARNING IN POLITICAL SCIENCE

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Assistant Professor
Dept of Government and Public Administration
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February 12, 2025

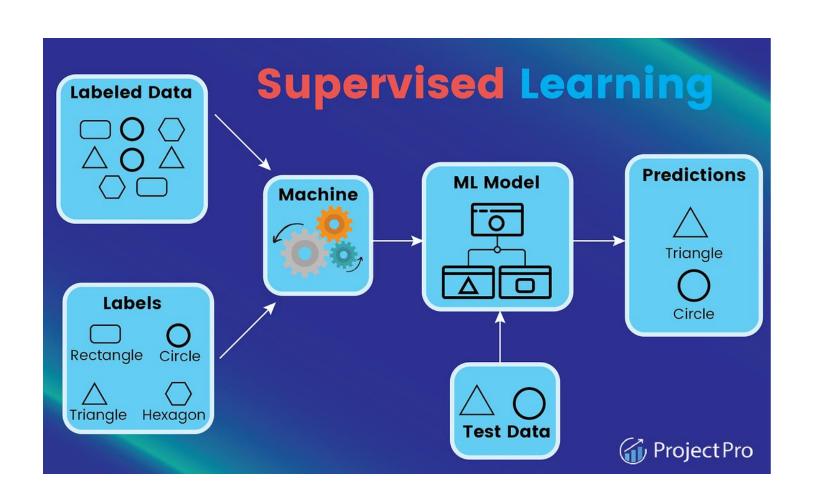


### MACHINE LEARNING





#### I. SUPERVISED LEARNING ALGORITHMS





#### 1.1 CLASSIFICATION

- Predicts the label of a class
- Predict the dataset's categories
- Example: "Yes" or "No"
- Commonly Used Algorithms:
- Decision Tree Algorithm
- Logistic Regression
- Random Forest Algorithm
- Support Vector Machine Algorithm

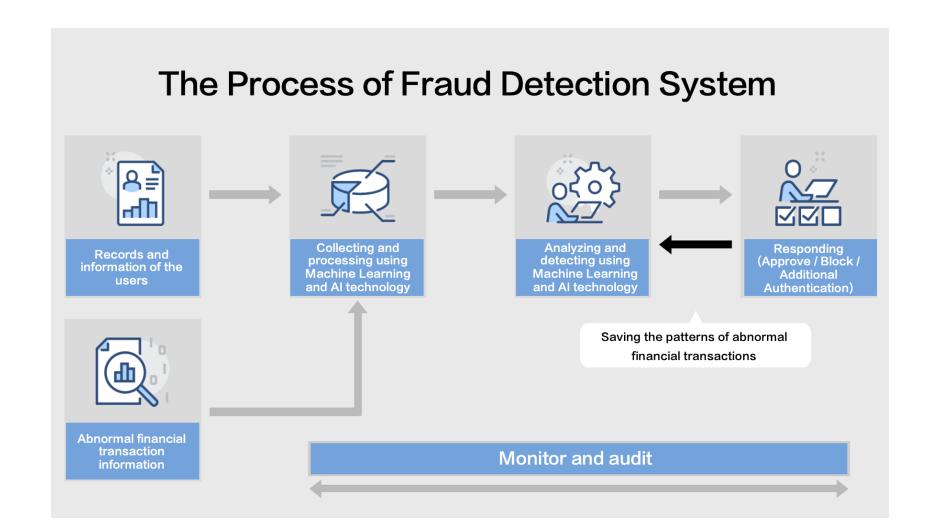


#### 1.2 REGRESSION

- Predicts the numerical label/continuous variables
- Example: weather prediction
- Commonly used algorithms:
- Decision Tree Algorithm
- Lasso Regression
- Multivariate Regression Algorithm
- Simple Linear Regression Algorithm



#### APPLICATION I: FRAUD DETECTION





#### APPLICATION II: IMAGE SEGMENTATION

Classification

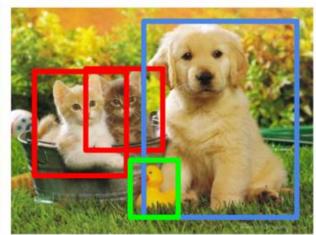
Classification + Localization

**Object Detection** 

Instance Segmentation









CAT

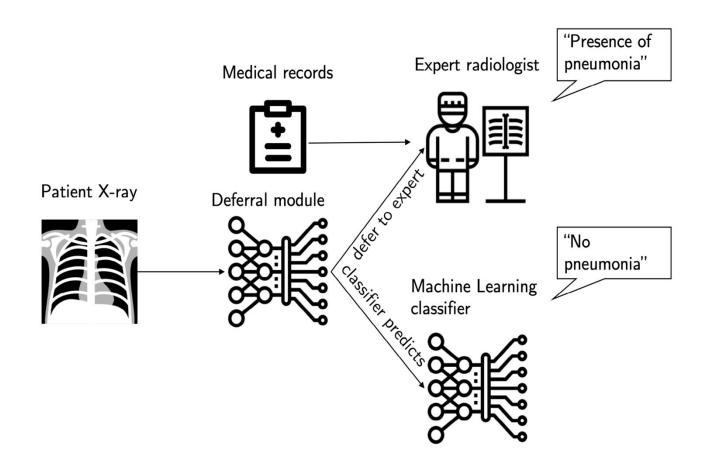
CAT

CAT, DOG, DUCK

CAT, DOG, DUCK

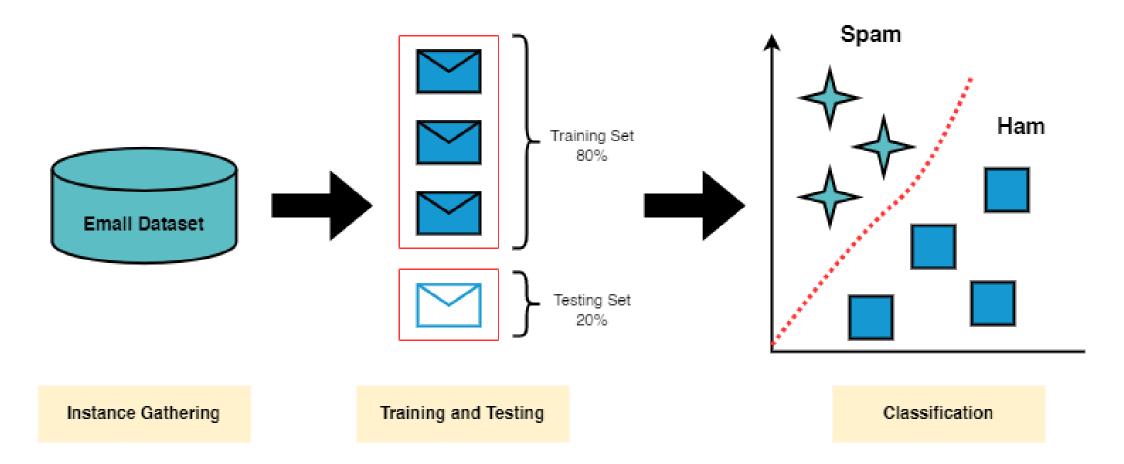
Single object Multiple objects

#### APPLICATION III: MEDICAL DIAGNOSIS



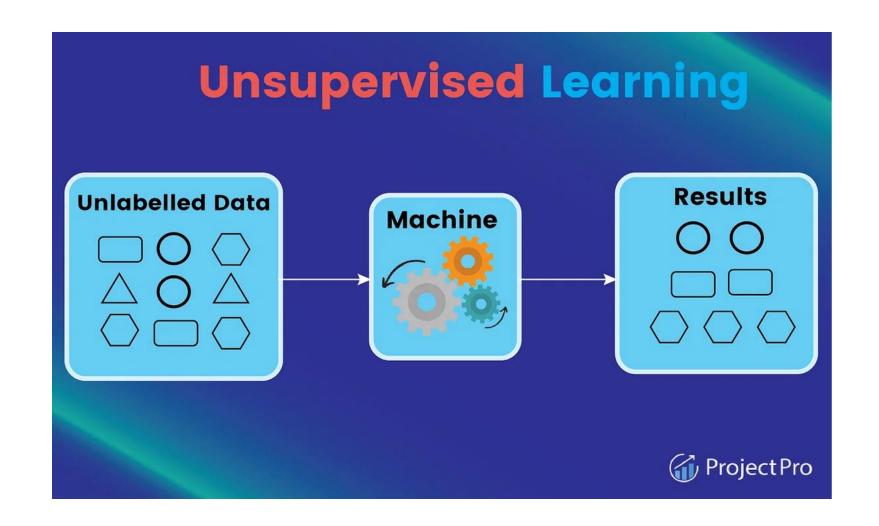


#### APPLICATION IV: SPAM DETECTION





#### 2. UNSUPERVISED LEARNING ALGORITHMS





#### 2.1 ASSOCIATION

- Finds relations between variables in a large dataset
- Goal: discover and map data dependent on the other to produce maximum profit
- Example: web usage mining
- Commonly used algorithms:
- Apriori algorithm
- Eclat
- FP-growth algorithm

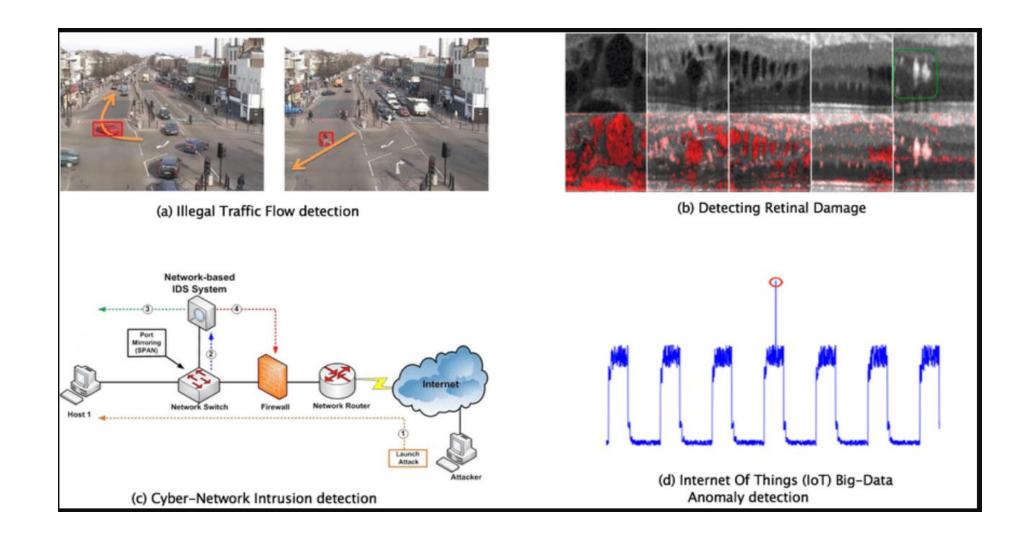


#### 2.2 CLUSTERING

- A method of grouping each set of similar objects into a cluster
- Goal: discover inherent groups from the dataset
- Example: retail marketing
- Commonly used algorithms:
- K-Means Clustering Algorithm
- DBSCAN Algorithm
- Independent Component Analysis
- Mean-Shift Algorithm
- Principal Component Analysis

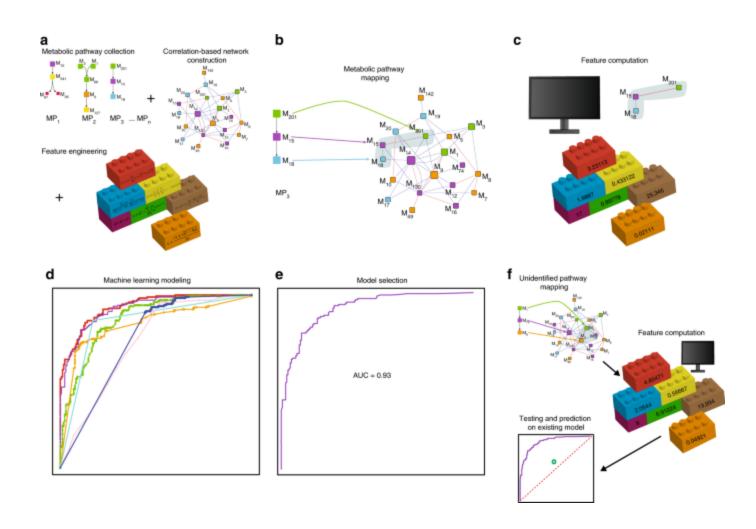


#### APPLICATION I: ANOMALY DETECTION



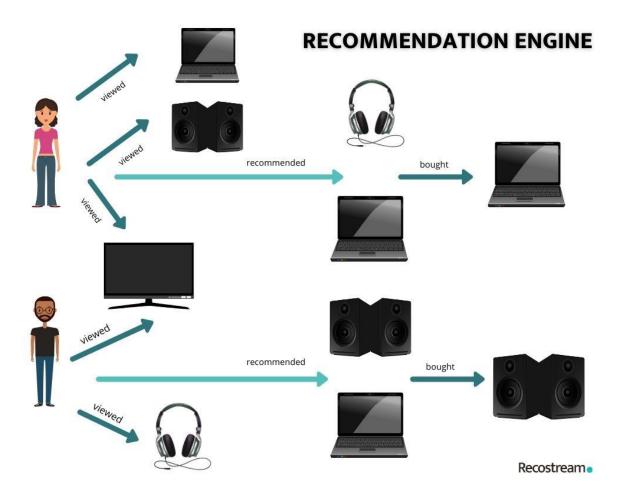


### APPLICATION II: NETWORK ANALYSIS



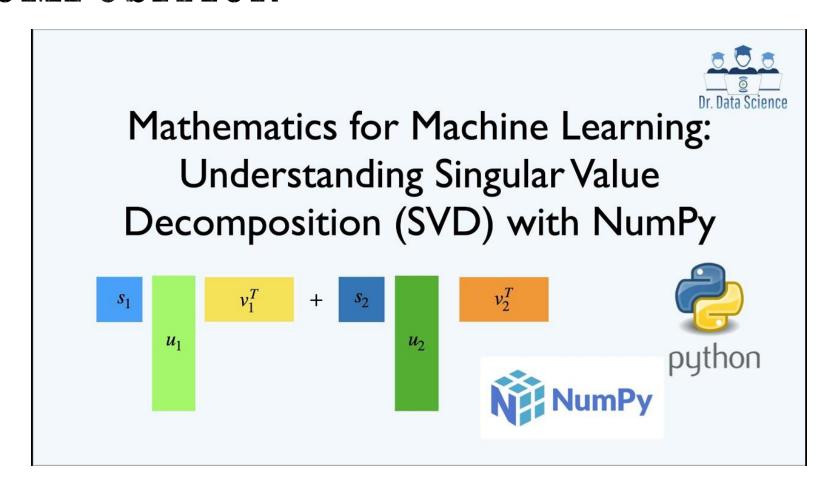


# APPLICATION III: RECOMMENDATION SYSTEMS





## APPLICATION IV: SINGULAR-VALUE DECOMPOSITION

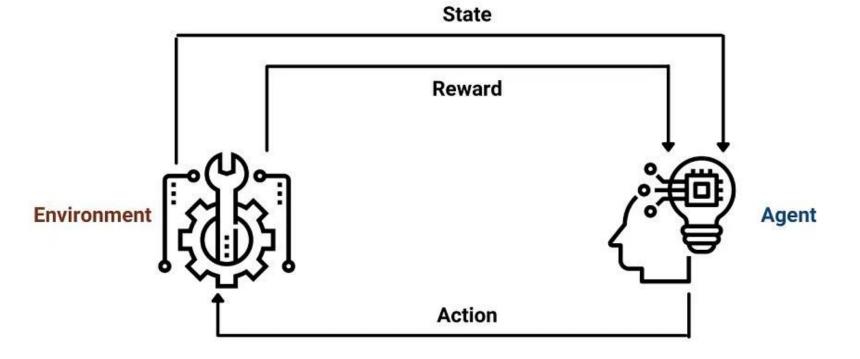




# 3. REINFORCEMENT LEARNING ALGORITHMS

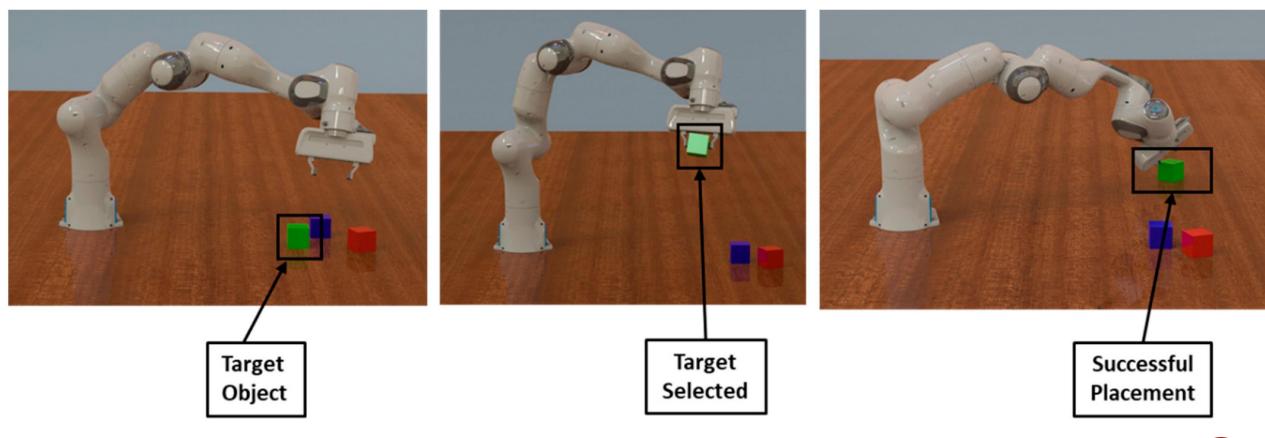
#### **Reinforcement Learning**





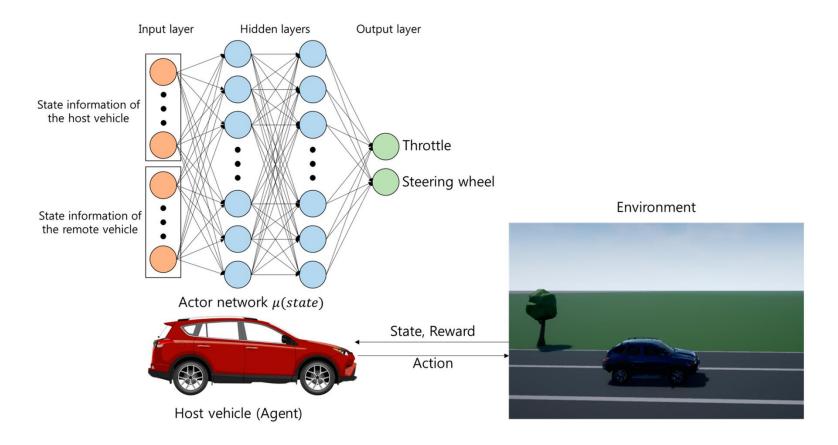


### APPLICATION I: ROBOTICS



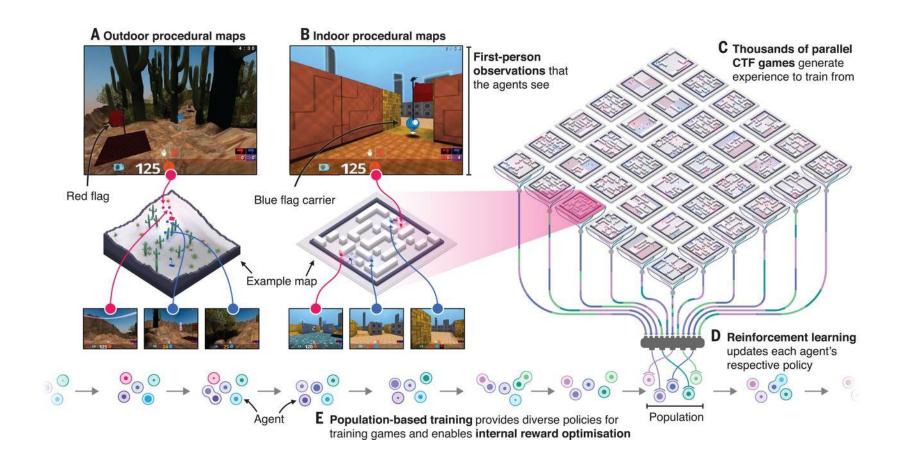


# APPLICATION II: SELF-DRIVING AUTOMOBILES





### APPLICATION III: VIDEO GAMES





#### APPLICATION IV: CHATGPT

Step 1

Collect demonstration data and train a supervised policy.

A prompt is sampled from our prompt dataset.

A labeler demonstrates the desired output behavior.

This data is used to fine-tune GPT-3.5 with supervised learning.



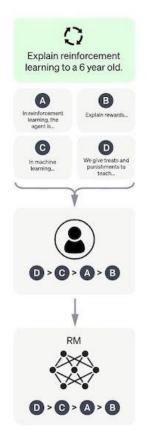
Step 2

Collect comparison data and train a reward model.

A prompt and several model outputs are sampled.

A labeler ranks the outputs from best to worst.

This data is used to train our reward model.



Step 3

Optimize a policy against the reward model using the PPO reinforcement learning algorithm.

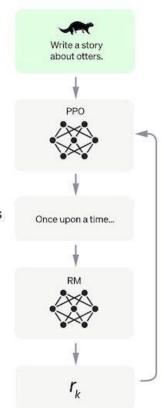
A new prompt is sampled from the dataset.

The PPO model is initialized from the supervised policy.

The policy generates an output.

The reward model calculates a reward for the output.

The reward is used to update the policy using PPO.





- Introduction to the Virtual Issue: Machine Learning in Political Science
- https://www.cambridge.org/core/services/aop-filemanager/file/5c348274e401b41903dae11b/PAN-VSI-Intro-0119-Machinelearning.pdf
- Machine Learning for Social Science: An Agnostic Approach
- <a href="https://www.annualreviews.org/doi/pdf/10.1146/annurev-polisci-053119-015921?download=true">https://www.annualreviews.org/doi/pdf/10.1146/annurev-polisci-053119-015921?download=true</a>



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