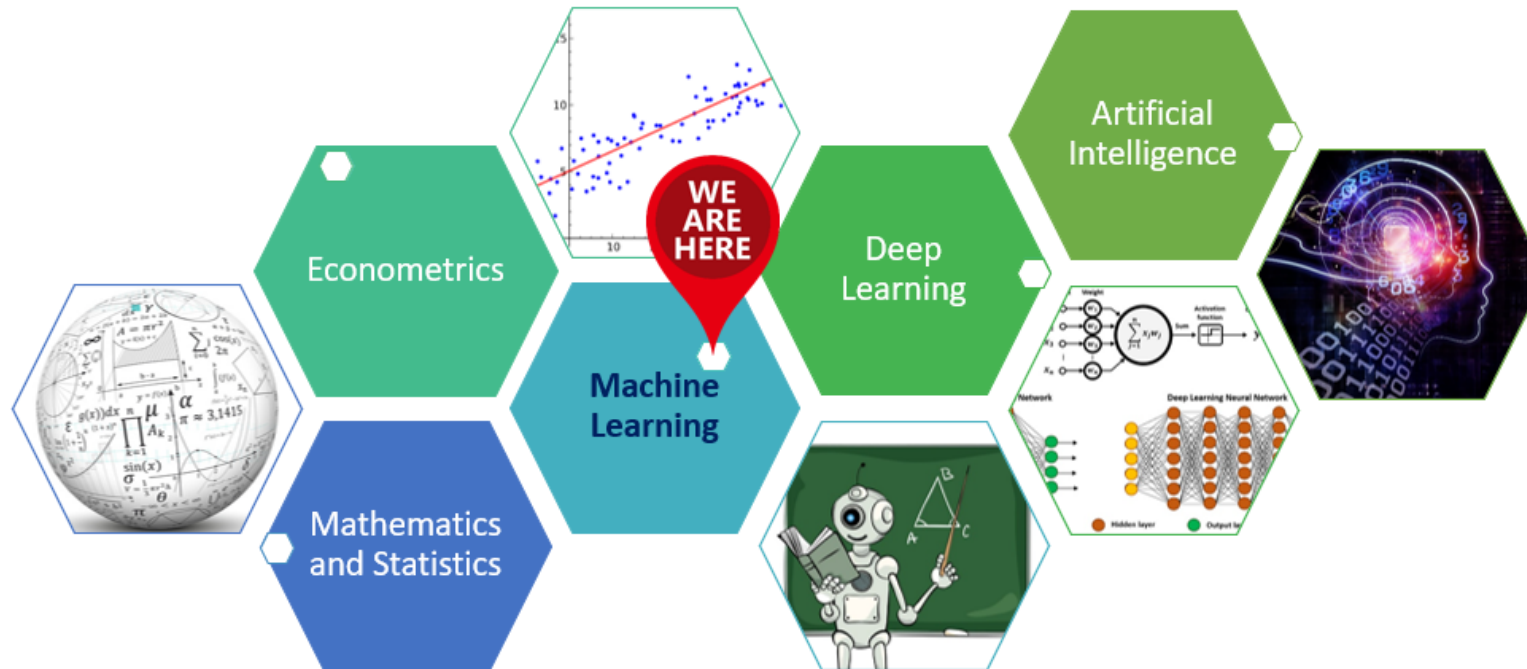
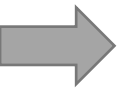


# Module 1- Part I

## Welcome to the magic world of Machine Learning





# Big picture: Econometrics vs Machine Learning



What are we trying to do as a researcher?



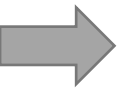
Solve real world problems, right?



Is there a theory?

What is the relationship between

- Sales and advertisement / R&D expenditure / seasonality / industry / ... ?
- Quantity demanded and price / income / technology / price of competitors / ... ?
- Wage and education/ age/ gender/ experience/ ...?

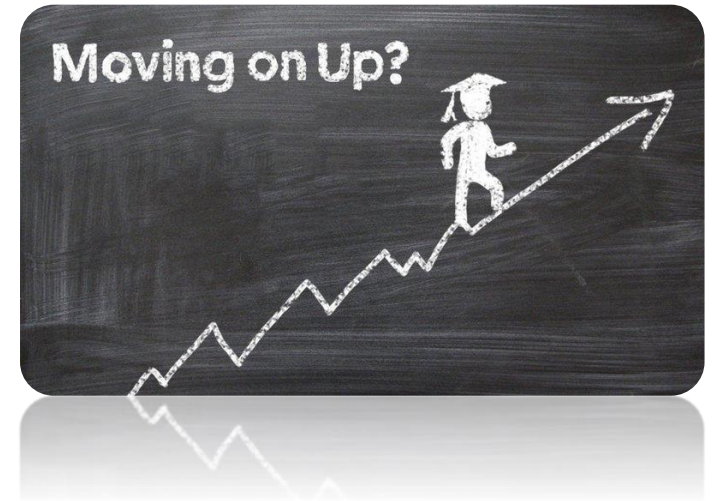


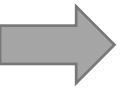
# A simple example

- Quantifying wage components! (is there a theory?)
- What are the drivers:
  - Education, age, experience, IQ, ...
  - Ethnicity, race, gender, ...
  - Industry, location, working hours, ...
- Let's build a model (**assuming** a linear functional form!)

$$wage = \beta_0 + \beta_1 educ + \beta_2 age + \beta_3 exper + \beta_4 IQ + \dots + \beta_k hours + u$$

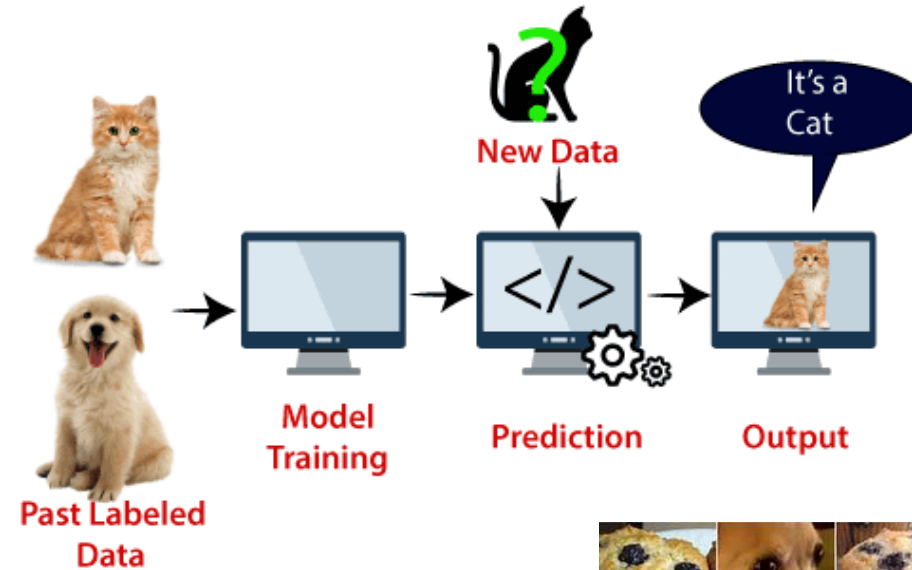
- Can you **interpret** this model? Do you care about the interpretability?
- Can you make **predictions** using your model?
- Can you make this functional form more flexible? What are the caveats?





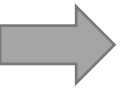
# A different example

- Cat vs dog classification problem (image classification)



- Do you really care about **interpretability** of the model here?
- What about accuracy of your **predictions**?





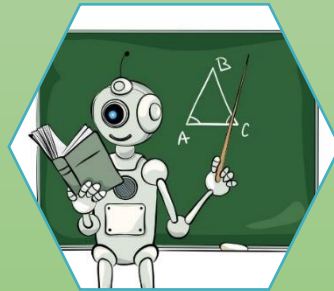
# Artificial intelligence vs Machine learning vs Deep learning

Artificial intelligence: Any technique which enables machines to mimic human behavior



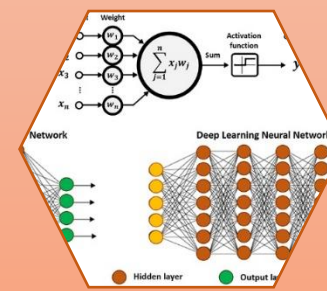
1950's

Machine Learning: Subset of AI that enables computers to learn from data. the model is trained with a set of algorithms

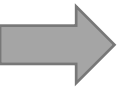


1980's

Deep Learning: Subset of ML that extract patterns from data using neural networks.



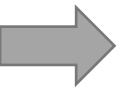
2010's



# Statistical learning vs machine learning

|                          | Statistical Learning                             | Machine Learning / Deep Learning                    |
|--------------------------|--|---|
| Focus                    | Hypothesis testing & interpretability            | Predictive accuracy and extracting complex patterns |
| Driver                   | Math, theory, hypothesis                         | Fitting data  |
| <b>Data size</b>         | Any reasonable set                               | Big data  |
| <b>Data type</b>         | Structured                                       | Structured, unstructured, semi-structured           |
| Dimensions / scalability | Mostly <b>low</b> dimensional data               | <b>High</b> dimensional data                        |
| Strength                 | Understand <b>causal</b> relationship & behavior | Prediction (forecasting and nowcasting)             |
| <b>Interpretability</b>  | <b>High</b>                                      | <b>Medium to Low</b>                                |



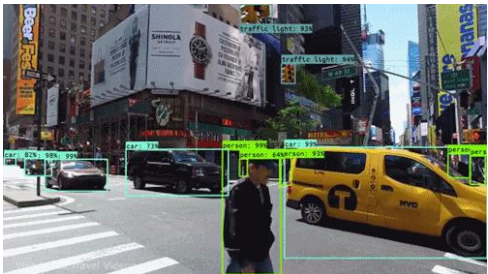


# Limitations of Econometrics/Structured ML

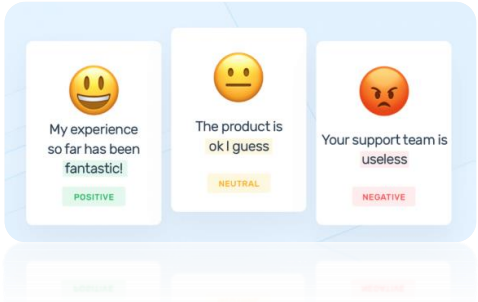
Econometrics/structured ML can only handle structured data (tabular data)!

Structured Data

|    | A          | B       | C                | D        |
|----|------------|---------|------------------|----------|
| 1  | Date       | Account | Transaction Type | Amount   |
| 2  | 2017-01-12 | 123     | Credit           | 6089.78  |
| 3  | 2017-01-12 | 123     | Fee              | 9.99     |
| 4  | 2017-01-12 | 456     | Debit            | 1997     |
| 5  | 2017-01-12 | 123     | Debit            | 20996.12 |
| 6  | 2017-01-13 | 123     | Debit            | 17       |
| 7  | 2017-01-13 | 123     | Debit            | 914.36   |
| 8  | 2017-01-14 | 789     | Credit           | 11314    |
| 9  | 2017-01-14 | 789     | Fee              | 9.99     |
| 10 | 2017-01-14 | 456     | Debit            | 15247.89 |
| 11 | 2017-01-14 | 123     | Debit            | 671.28   |
| 12 | 2017-01-15 | 456     | Credit           | 5072.1   |
| 13 | 2017-01-15 | 456     | Fee              | 9.99     |
| 14 | 2017-01-16 | 456     | Debit            | 5109.07  |
| 15 | 2017-01-19 | 123     | Credit           | 482.01   |



Unstructured Data  
(everything else!!)



# ➔ A more complex example

Stock price prediction \$\$\$

- What are the classical drivers:
  - Company's fundamentals (balance sheet, income statement, cash flow statement)
  - Competitors (comparing multiples)
  - Technical analysis!
  - Seasonality (holidays, months, days, ...)



What else?

- Market sentiment (news, tweets, blogger opinions, conference calls, ...)
- Satellite images from parking lots!

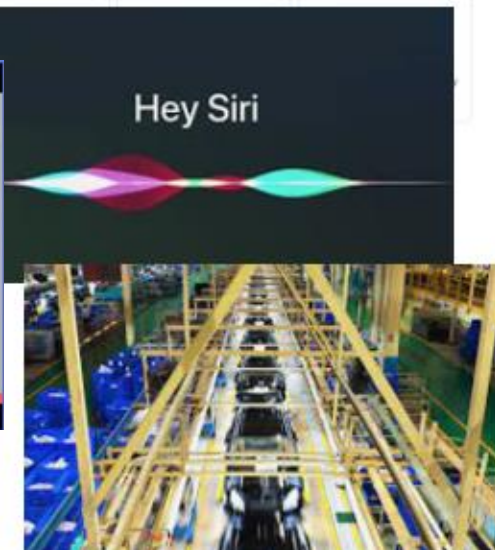


# Why should I learn it?

- It's a bid deal, it is **everywhere!**
- Better career opportunities
- Hedge against next recession



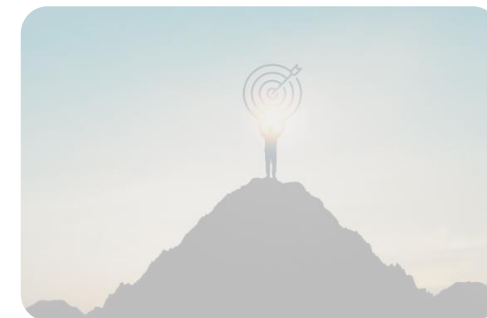
OpenAI





# Class Modules

- Module 1- Introduction to Deep Learning
- Module 2- Setting up Machine Learning Environment
- Module 3- Linear Regression (Econometrics approach)
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# Module 1- Part II

## What is Machine Learning



**Artificial intelligence:** Any technique which enables machines to mimic human behavior



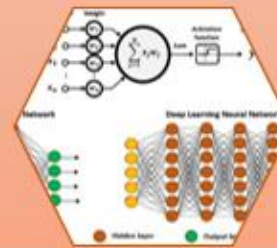
1950's

**Machine Learning:** Subset of AI that enables computers to learn from data. the model is trained with a set of algorithms

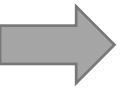


1980's

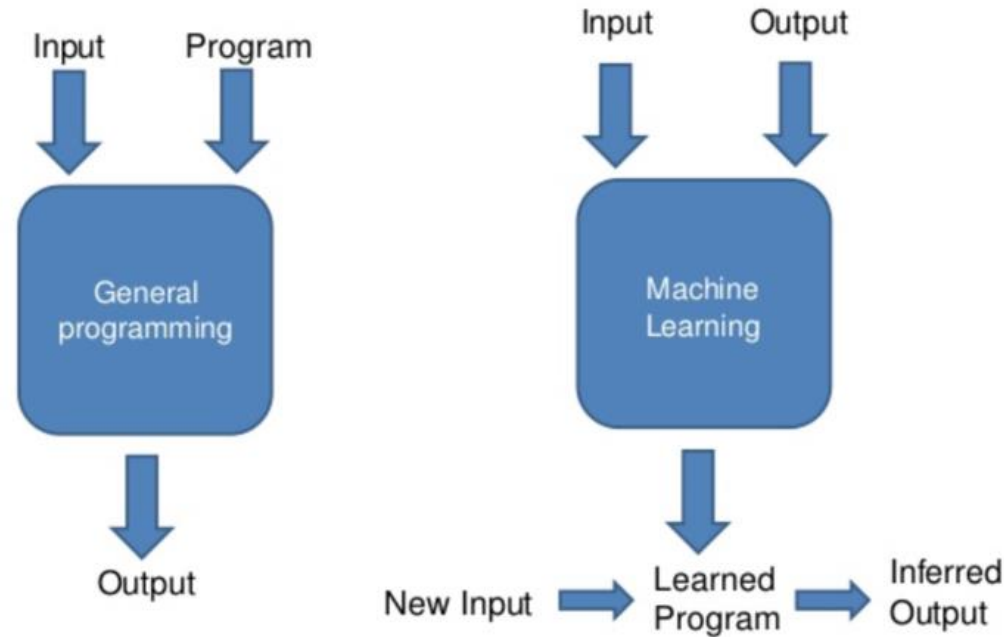
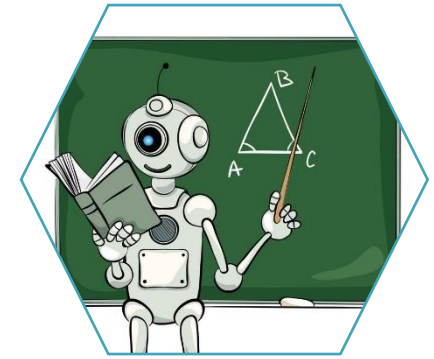
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2010's



# General programming vs Machine learning

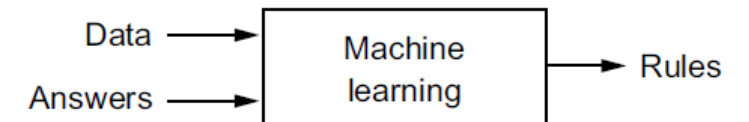
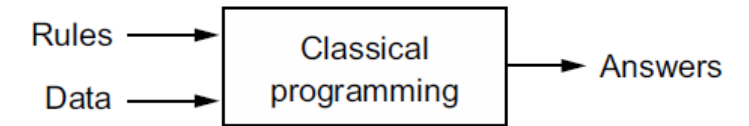


- Machine Learning: Involves automated detection of meaningful patterns in data and apply the pattern

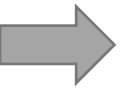


# ➔ What is Machine Learning?

- A machine learning system is **trained** (with algorithms) rather than explicitly **programmed**.
- Machine Learning is a subset of AI that enables computers to **learn** from data.
- ML involves automated detection of meaningful **patterns** in data and apply the pattern to make **predictions** on **unseen data**!
- This is done by **minimizing** the loss on the training data.
- The goal is to **maximize** the performance on the unseen data.







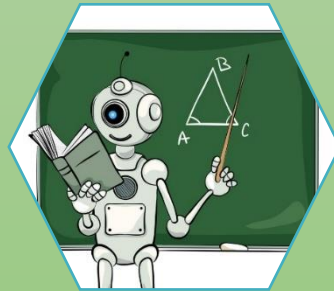
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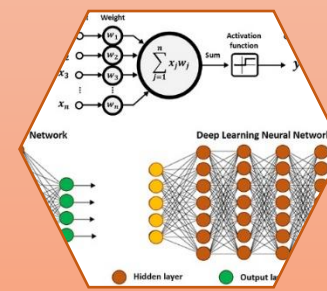
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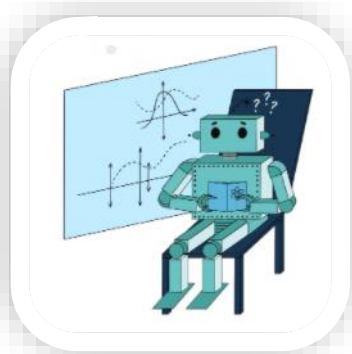
1980's

Deep Learning: Subset of ML that extract patterns from data using neural networks.



2010's

# Types of Machine Learning



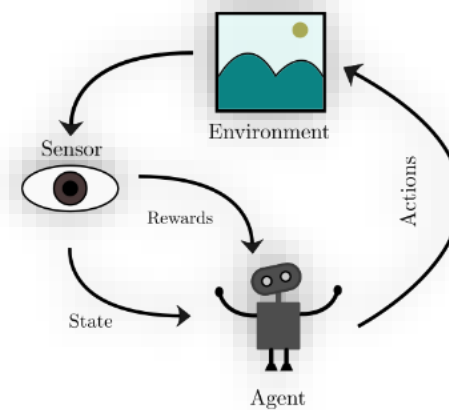
Unsupervised  
Learning

Machine  
Learning

Supervised  
Learning

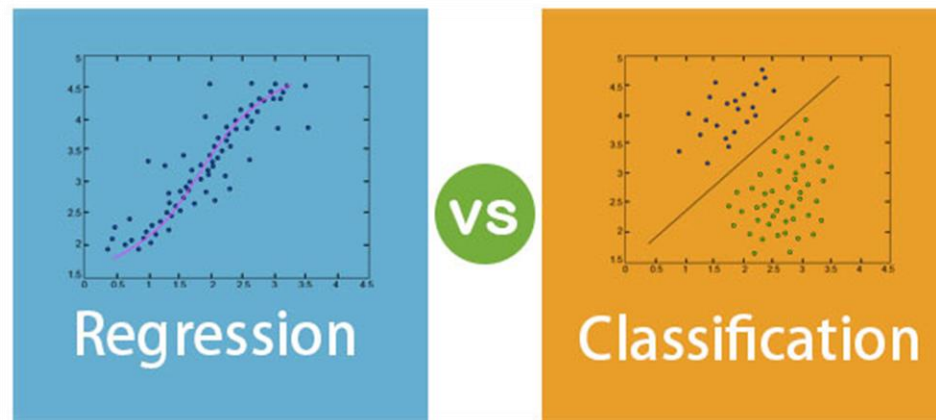


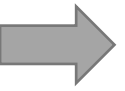
Reinforcement  
Learning



# ➔ Supervised Learning

- Supervised learning is a type of machine learning where the algorithm is **trained on labeled data**.
- The data is labeled, meaning that the data has been **tagged with the correct output**.
- The model is then able to learn **the relationship** between the input data and the corresponding output labels and can make predictions on new data.
- **Regression:**
  - Predicting housing price
  - Predicting stock market returns
- **Classification:**
  - Generating buy, sell, hold signals.
  - Predicting credit default rate.





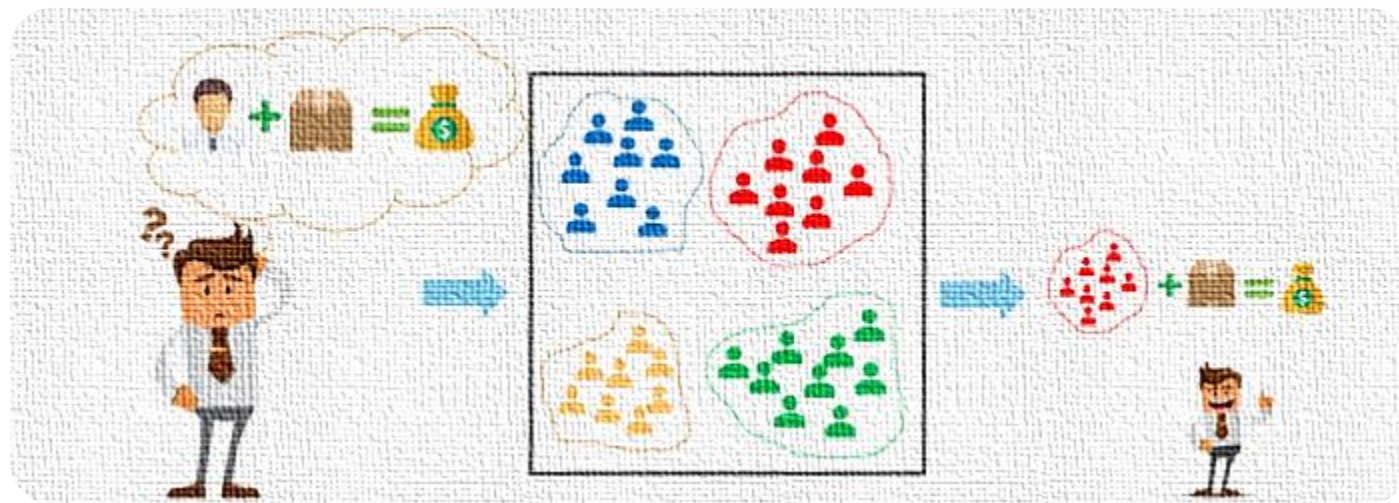
# Class exercise

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- Can you think of an example of Regression? Classification?

# ➔ Unsupervised Learning

- Unsupervised learning is a type of machine learning where the algorithm is **not given any labeled** training data.
- The goal is to discover the **underlying patterns** and find groups of samples that behave similarly. **Find something interesting!**
- **Clustering**: group similar data points together
  - ✓ Mall customer segmentation
  - ✓ Client profiling and asset allocation

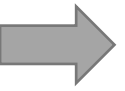




# → Unsupervised Learning

- Unsupervised learning is a type of machine learning where the algorithm is **not given any labeled** training data.
- **Anomaly detection:** Find anomalies (unusual data points)
  - ✓ Fraud detection
- **Dimensionality Reduction:** compress data in lower dimension
  - ✓ Identify the most predictive factors underlying asset price models.





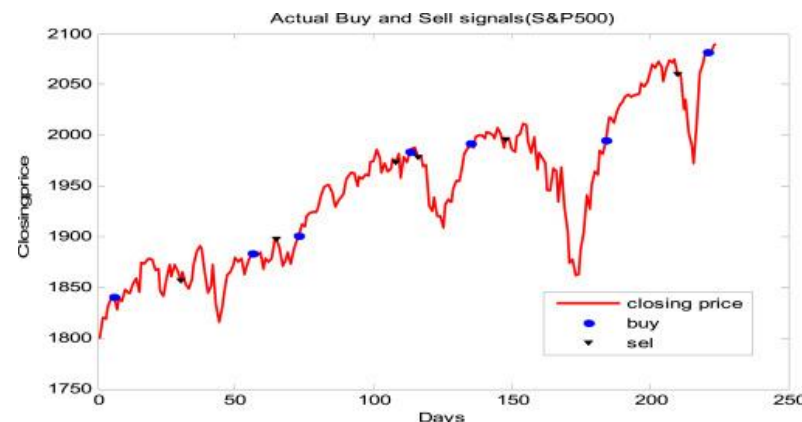
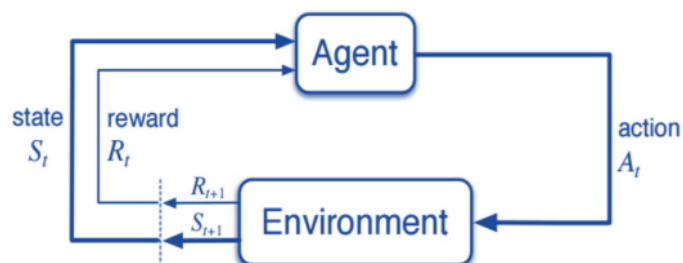
# Class exercise

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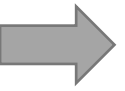
- Can you think of an example of Clustering? Anomaly detection?

# ➔ Reinforcement Learning

- **Reinforcement learning** is a type of machine learning where an **agent** learns to interact with its **environment** in order to maximize a **reward**.
- The agent receives rewards for performing actions that lead to successful outcomes and learns to repeat successful actions and avoid unsuccessful ones. (**Explore** and **Exploit**)



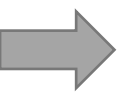
- Example: a virtual trader (**agent**) who follows certain trading rules (**actions**) in a specific market (**environment**) to maximize its profits (**reward**).



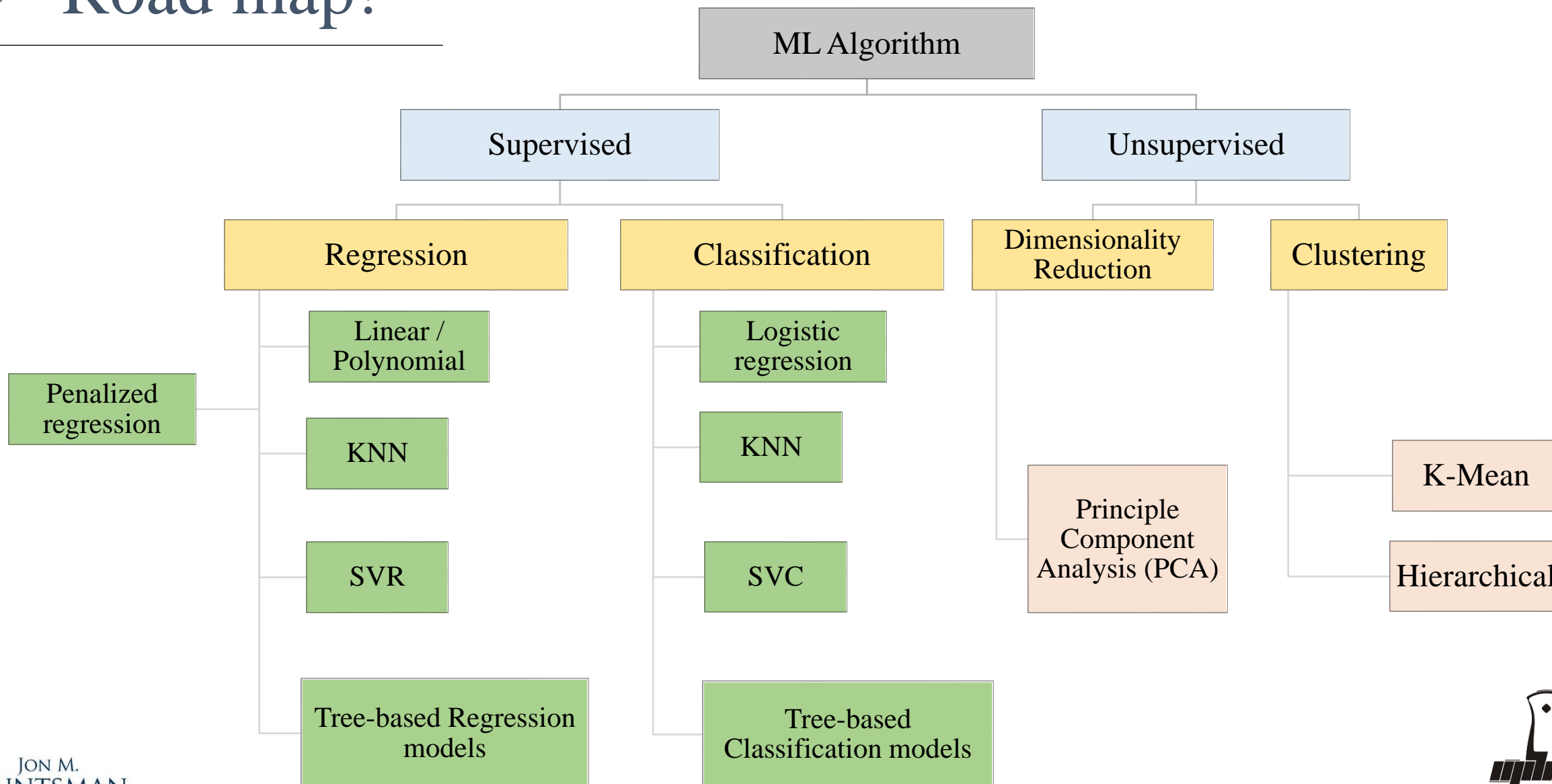
# Class exercise

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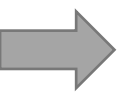
- Can you think of an example of reinforcement learning?



# Road map!



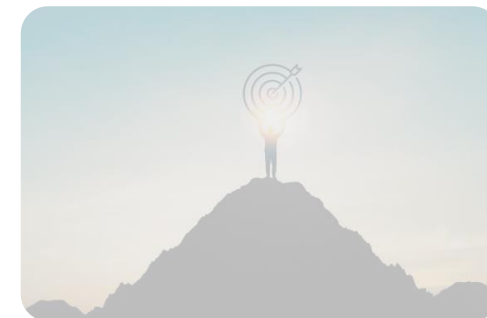


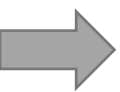


# What's on GitHub

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# Having said that...

- **Warning:** A ML algorithm will always find a pattern, even if there is none.

