



# Machine Learning

## Introduction

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# R

# Previously...

## R

### What is Intelligence?

What does it means to Think?

Thought processes and Reasoning      VS      Behavior



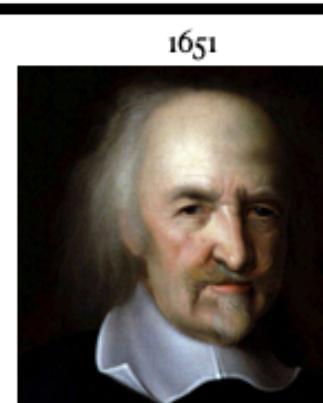
"I think, therefore I am,"  
René Descartes



Knowledge through senses  
Adam Smith



Dialectic method  
Hegel



Reason is nothing but 'reckoning'  
Thomas Hobbes



Critique of Pure Reason  
Kant



Russell's Paradox  
Bertrand Russell



#### Thinking Humanly

"The exciting new effort to make computers think . . . *machines with minds*, in the full and literal sense." (Haugeland, 1985)

"[The automation of] activities that we associate with human thinking, activities such as decision-making, problem solving, learning . . ." (Bellman, 1978)

#### Thinking Rationally

"The study of mental faculties through the use of computational models." (Charniak and McDermott, 1985)

"The study of the computations that make it possible to perceive, reason, and act." (Winston, 1992)

#### Acting Humanly

"The art of creating machines that perform functions that require intelligence when performed by people." (Kurzweil, 1990)

"The study of how to make computers do things at which, at the moment, people are better." (Rich and Knight, 1991)

#### Acting Rationally

"Computational Intelligence is the study of the design of intelligent agents." (Poole *et al.*, 1998)

"AI . . . is concerned with intelligent behavior in artifacts." (Nilsson, 1998)

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# Intelligence

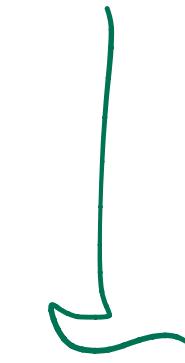
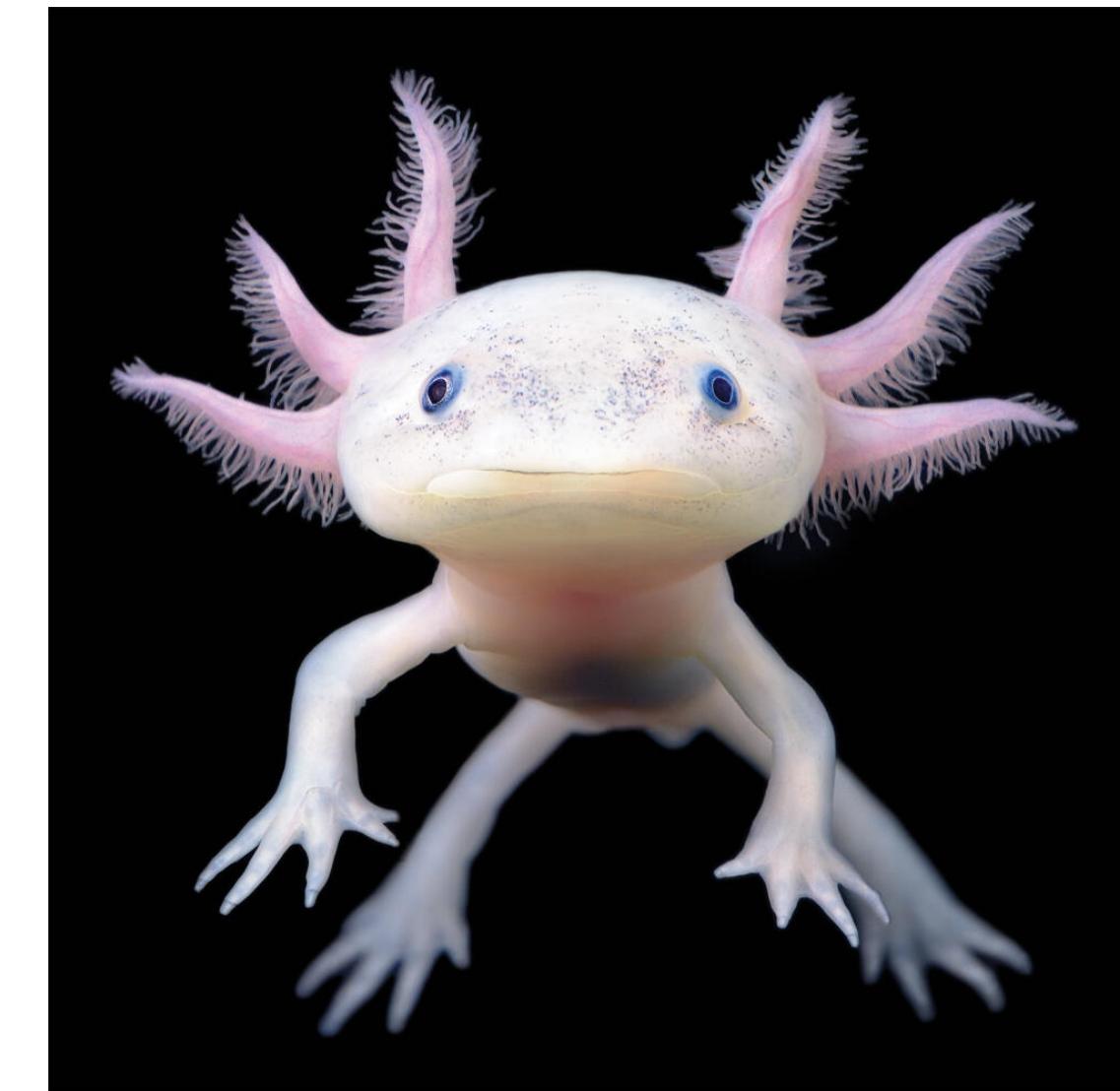
Assume...

- Intelligence is a **goal-directed adaptive behavior**

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# Adaptive Behavior



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## Learn Definition

- To gain **knowledge** or understanding of or skill in by study, instruction, or **experience**

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## Learn Definitions



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- **Kupfermann (1985):** Learning is the acquisition of knowledge about the world

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- **Shepherd (1988):** Learning is an adaptive change in behavior caused by experience

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## Learn Definitions



To gain knowledge or **understanding** of or skill in by study, instruction, or experience

- **Kupfermann (1985):** Learning is the acquisition of knowledge about the world
- **Shepherd (1988):** Learning is an adaptive change in behavior caused by experience
- **Einstein?:** *If you can't explain it simply, you don't understand it well enough*

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# Machine Learning

## What is it?

- **Arthur Samuel:** a field of study that gives computers the ability to learn without being explicitly programmed.
- **Tom M Mitchell:** a computer program is said to learn from experience E with respect to some class of tasks T and performance measure P if its performance at tasks in T, as measured by P, improves with experience E.
- **Ron Kohavi; Foster Provost (1988):** subfield of computer science that explores the study and construction of algorithms that can learn from and make predictions on data.

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# Machine Learning

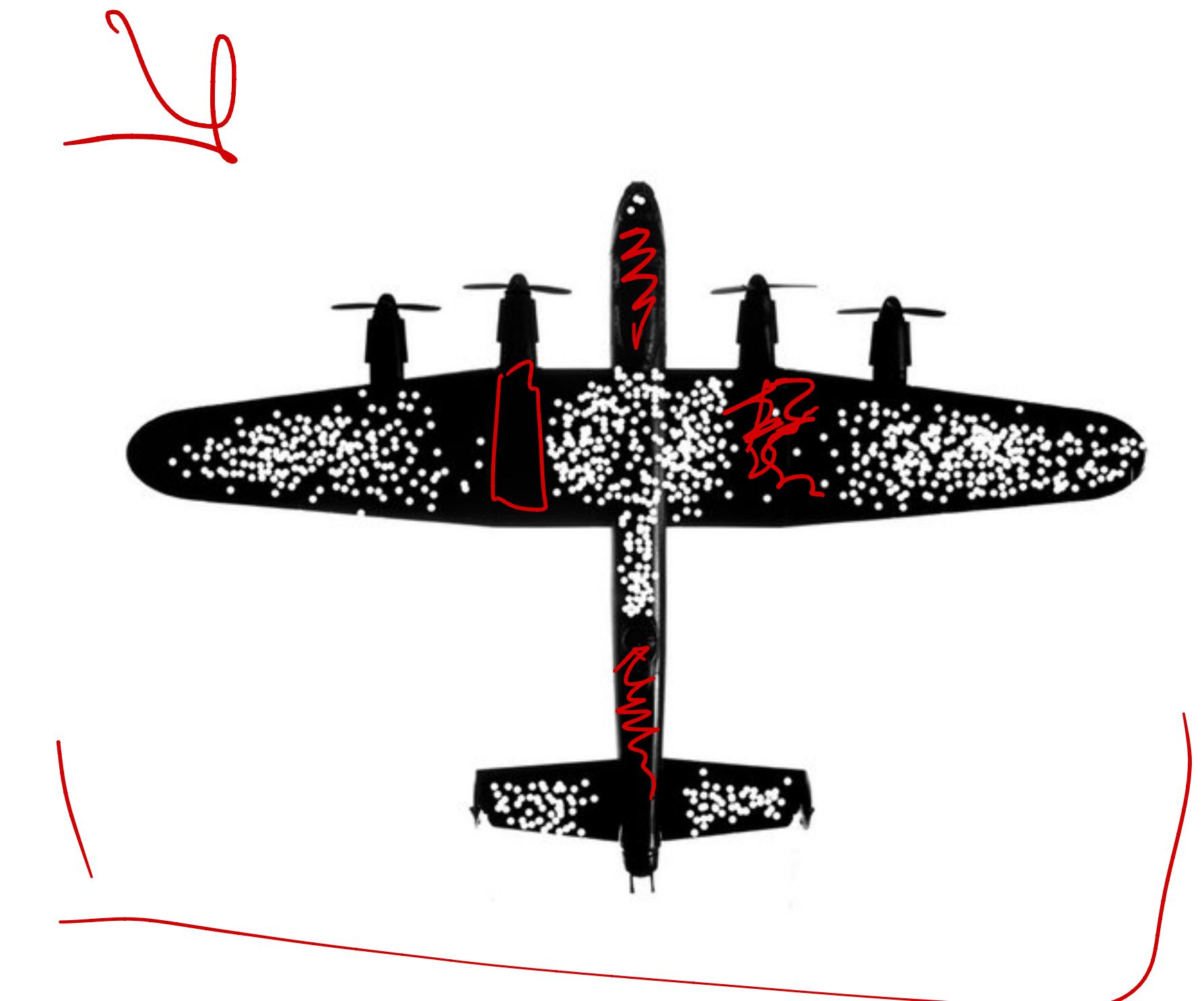
What is it?

- Search for patterns in data
- Predict

$$y = f(x)$$

Label ↗

↳ function  
+  
Learn



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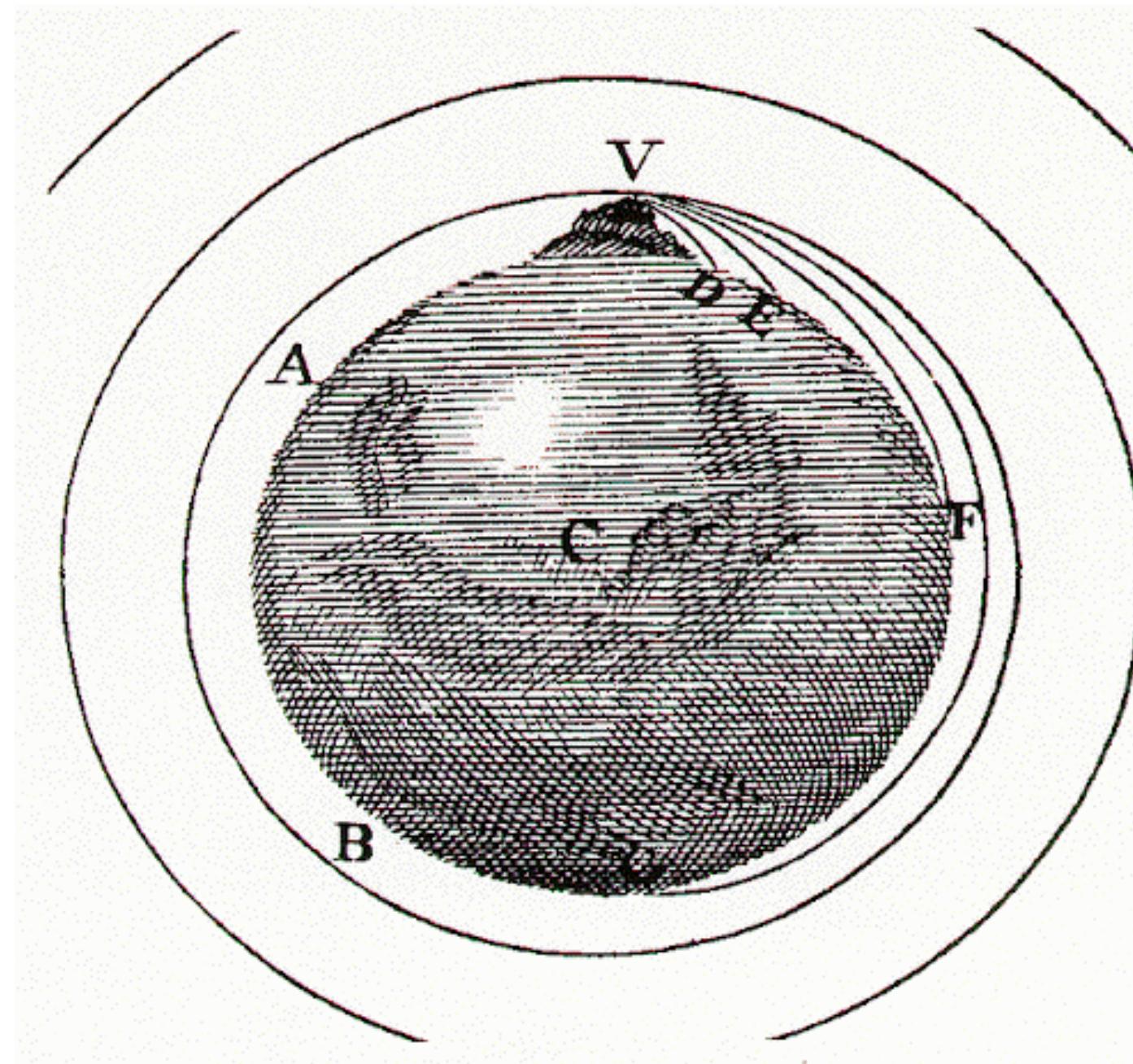
# ML vs Statistics

- Both: Summarize data
- Statistics: Data analysis - hypothesis testing
- ML: Efficient algorithms —> Fast, low data requirements, make predictions

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# Empirical Inference

- $\underbrace{(\text{Data, observations})}_{\text{Data}} \Rightarrow \underbrace{(\text{Rules, models})}_{\text{Models}}$

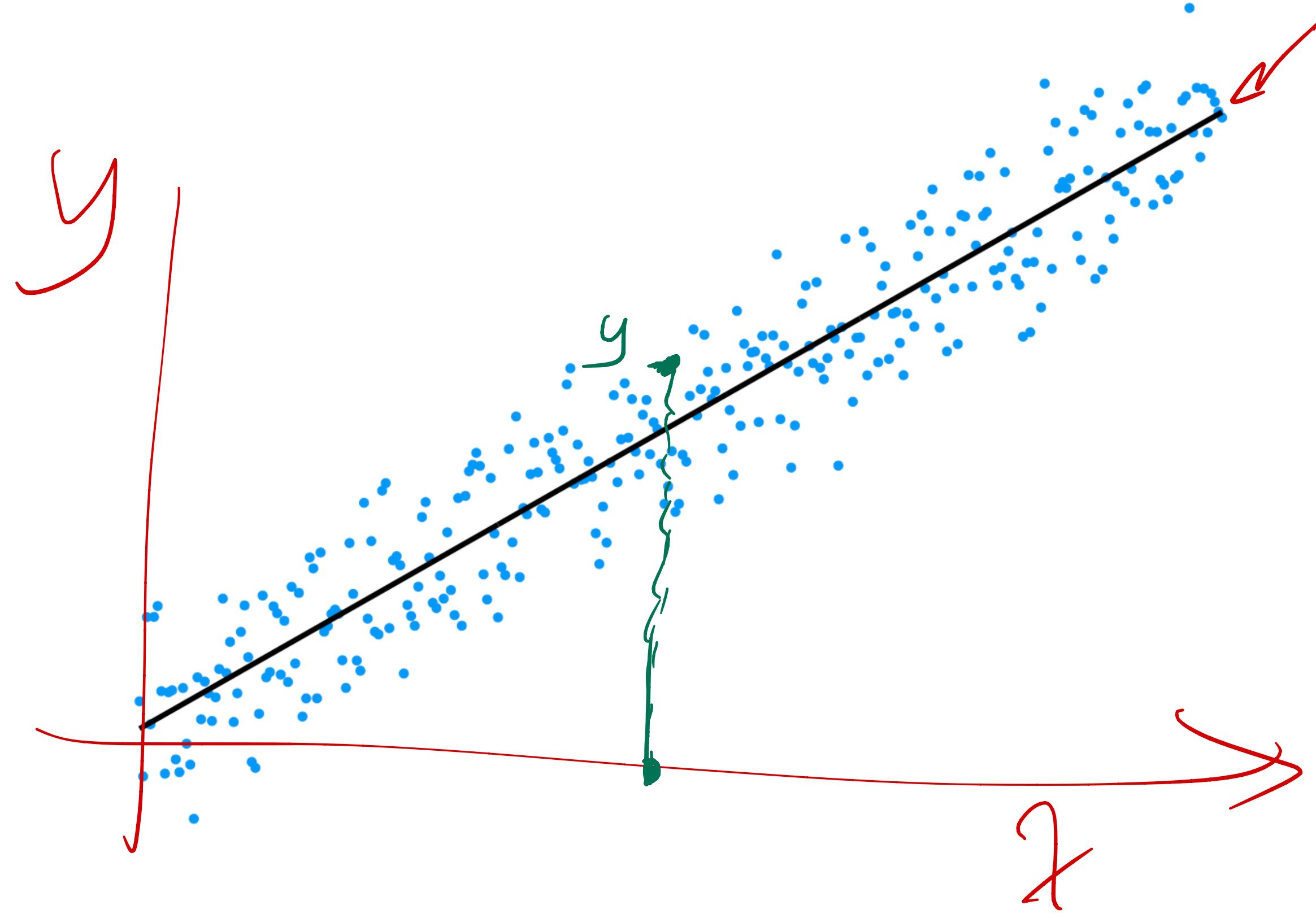


$$F = \frac{GM_1M_2}{r^2}$$

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# Empirical Inference

- (Data, observations) => (Rules, models)



$$Y = aX + b$$

Law

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## Generalization

- The rule/model should be used to predict new observations

- Example:  $\boxed{1, 2, 4, 7, \dots}$

- What is next?

- $1, 2, 4, 7, 11, \dots$

$$a_{n+1} = a_n + n$$

- $1, 2, 4, 7, 12, \dots$

$$a_{n+2} = a_{n+1} + a_n + 1$$

- $1, 2, 4, 7, 14, 28$

divisors of 28

- $1, 2, 4, 7, 1, 1, 5$

$$\pi = \cancel{3}1415$$

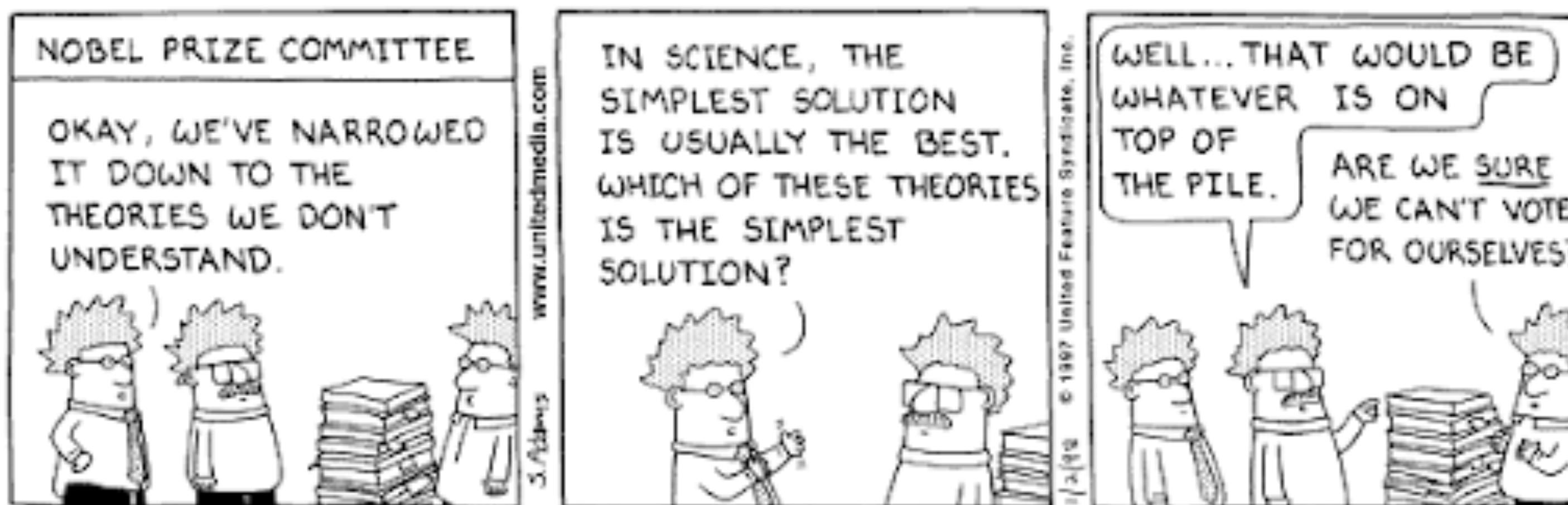
$$e = 2.71$$

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# Generalization

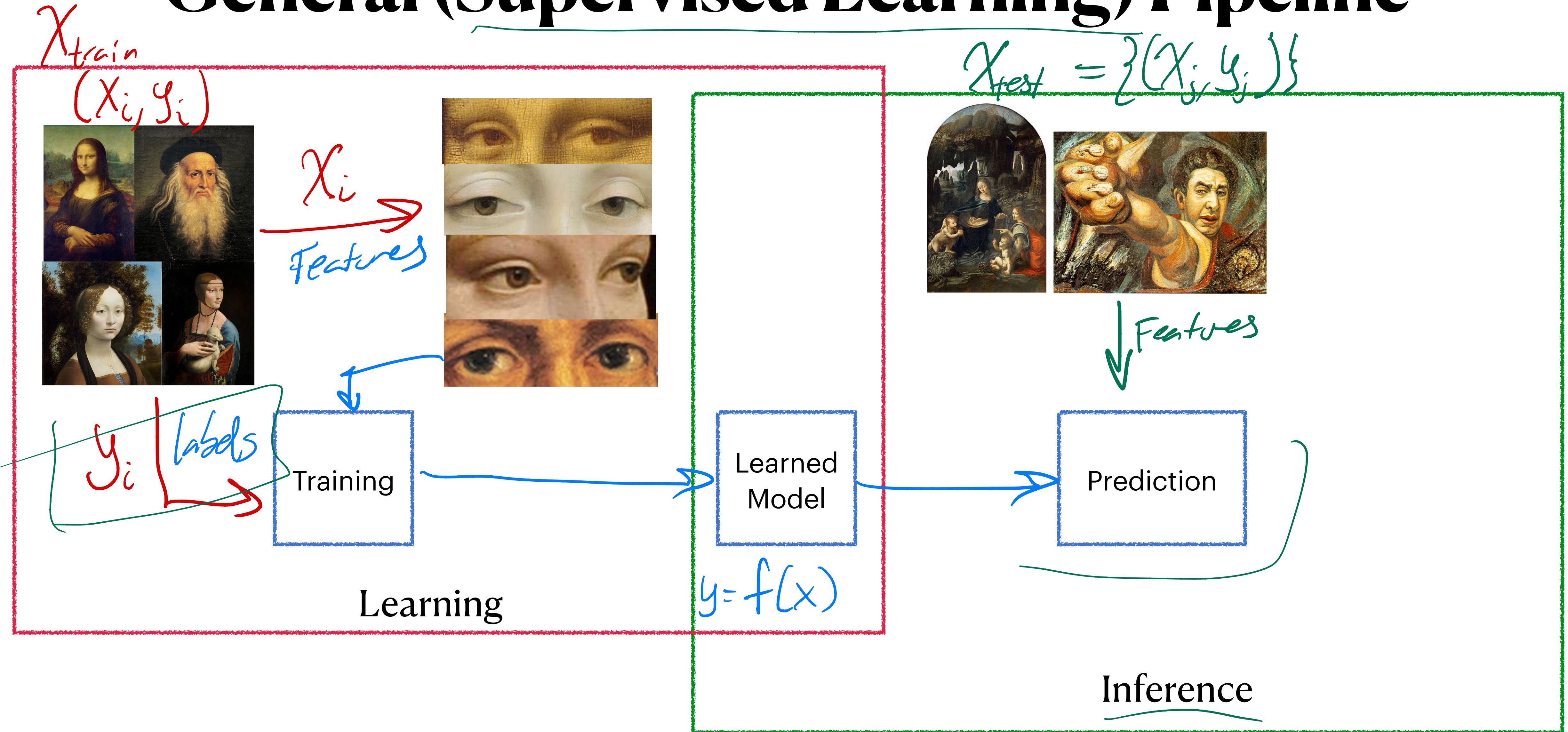
## Occam's razor

- Among competing hypotheses, the one with the fewest assumptions should be selected.



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# General (Supervised Learning) Pipeline

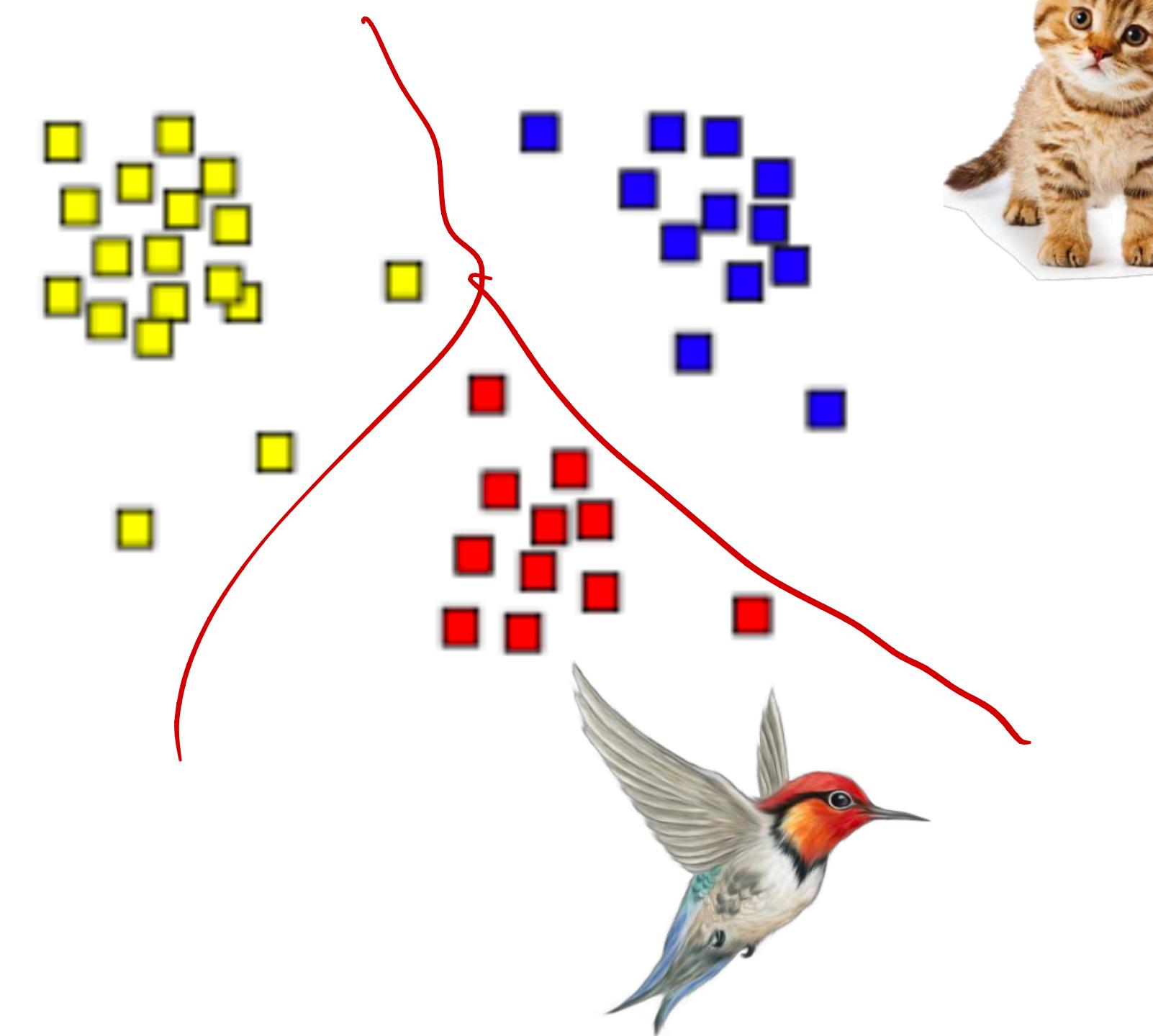


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# Unsupervised Learning

No labels

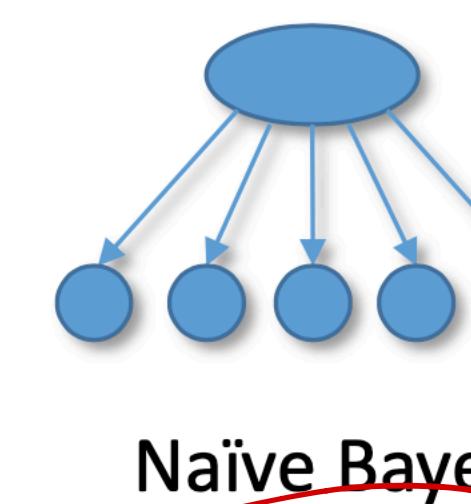
- clustering
- dimension  
Robotics



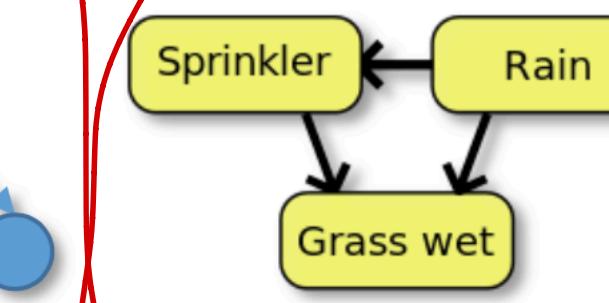
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# ML

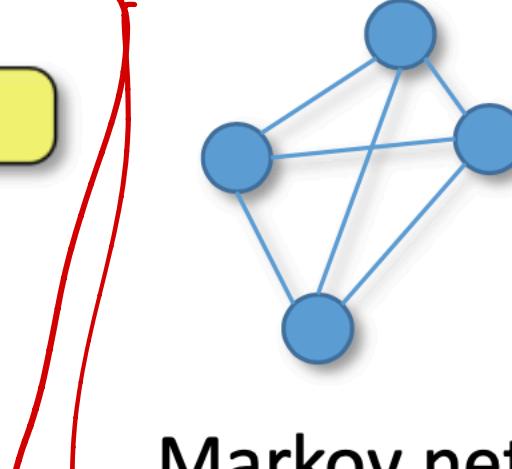
## Approaches



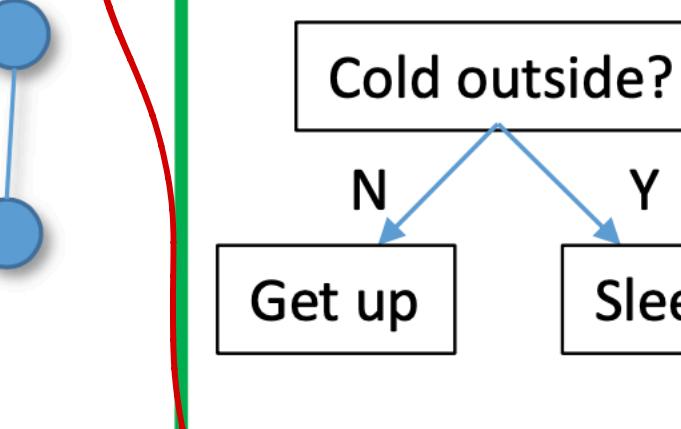
Naïve Bayes



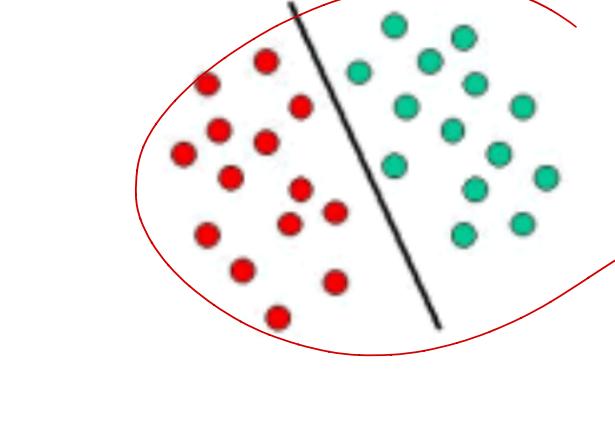
Bayes net



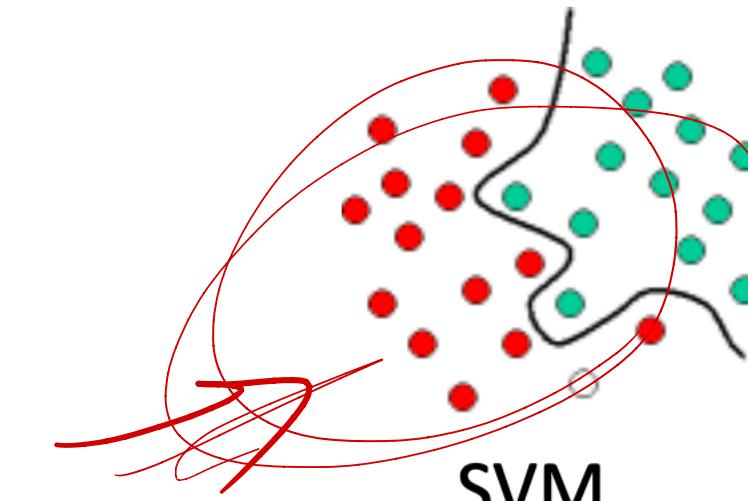
Markov network



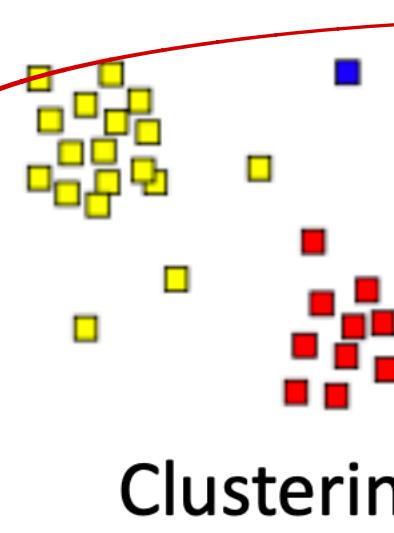
Decision tree



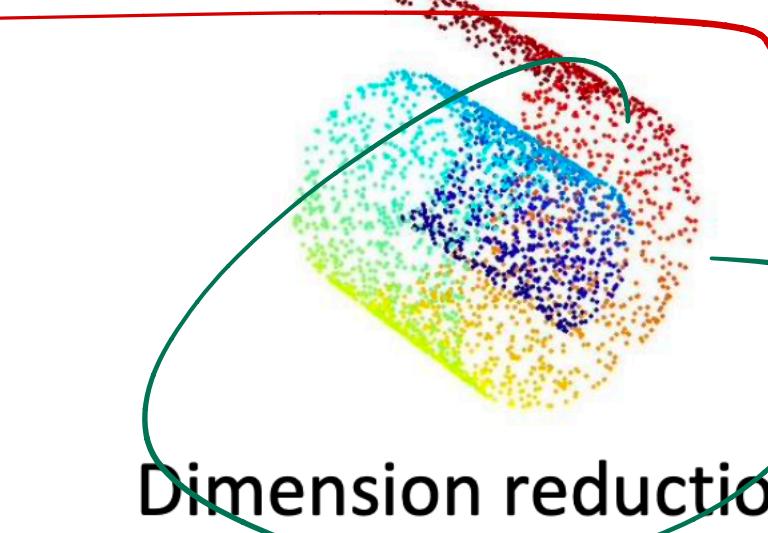
Linear classifier



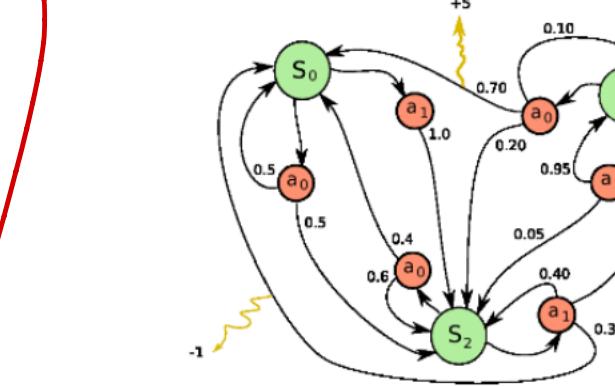
SVM



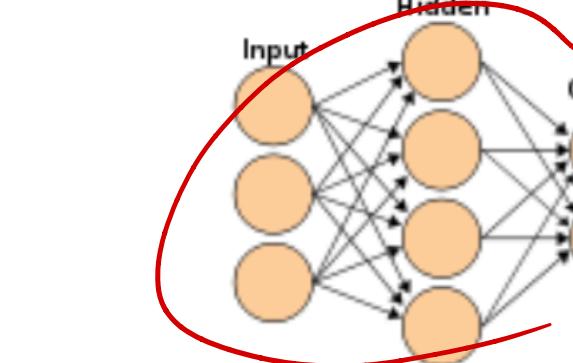
Clustering



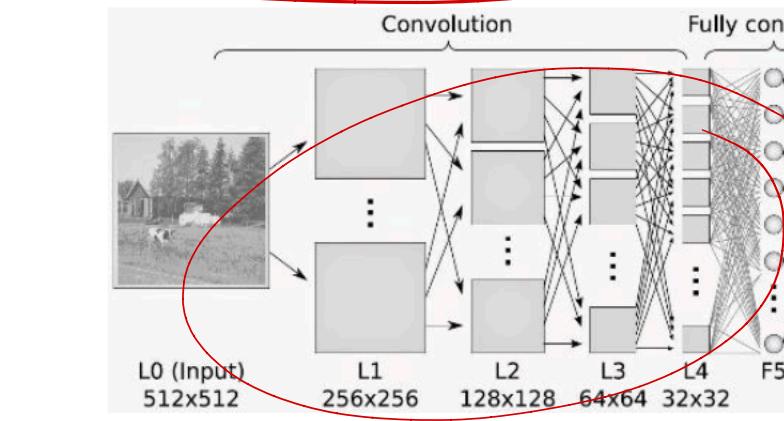
Dimension reduction



Reinforcement learning



Artificial neural network



Deep learning / CNN

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# Classification

- Set of training examples:
- Learn:
- Evaluate performance:

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## Naive Bayes Algorithm

- Recall Bayes Rule:

- Likelihood:

$$\bullet \quad L(x) = \frac{P(y = true \mid x)}{P(y = false \mid x)}$$

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# Naive Bayes Algorithm

- Estimate  $P(y)$  from the training examples
  - $P(y = \text{true}) =$
  - $P(y = \text{false}) =$

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# Naive Bayes Algorithm

- From the training examples
  1. Define  $\phi$ , a function that maps each input  $x$  into a features vector

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## Naive Bayes Algorithm

- From the training examples
  1. Define  $\phi$ , a function that maps each input  $x$  into a features vector
  2. Estimate the probability distribution of each feature  $\phi_j$

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## Naive Bayes Algorithm

- From the training examples
  1. Define  $\phi$ , a function that maps each input  $x$  into a features vector
  2. Estimate the probability distribution of each feature  $\phi_j$
  3. Estimate  $P(x | y)$  using the naive Bayes assumption

$$p(x|y = \text{true}) = \prod_{j=1}^l p(\phi_j(x)|y = \text{true})$$

$$p(x|y = \text{false}) = \prod_{j=1}^l p(\phi_j(x)|y = \text{false})$$

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