

# INTRO TO MACHINE LEARNING

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CMSC 671

By the end of class today, you will be able to:

- Distinguish between types of machine learning algorithms
- Identify when you would want to use particular algorithm types

*Modified from slides by Dr. Chris Callison-Burch, Dr. Cassandra Kent, & Dr. Cynthia Matuszek*

## Naive Bayes is naive because it assumes...

(A) all features are conditioned on the same label

0%

(B) the probabilities of features don't need to be conditioned on their labels

32%

✓ (C) all features are independent effects of their label

68%

What do you think of when you hear "machine learning"?

outcome input imitate applied knn classifiers intended outcomes  
 neural getting prediction algorithms explicitly improving  
 models consumes accuracy also use branch detection learns want  
 gaining things task ai learn machines intelligence  
 algorithm humans learning data based actions past  
 focuses patterns machine computer end perform make way  
 given regression "ai" complete  
 iterate humanity allows artificial supervised cnn (ai) behavior give sgd  
 networks gradually think new agi unsupervised focused knowledge  
 predicts man-made future automation etc linear experience learner predict

# COURSE UPDATES


- Course Evaluations are available until December 12 at 11:59pm
- Milestone 2 is due tonight (11:59pm)
  - Take feedback into consideration!!
- Milestone 3 is due December 7<sup>th</sup> (next Thursday)
  - Draft of paper, official conference template
  - Get preliminary results

# EXTRA CREDIT REPORT

- Goal: use research to back up your stance on something from the news about AI
- 0.5-2 points on final grade
  - 0.5 if you give a solid effort but don't satisfy other requirements
  - 2 points if you satisfy all the requirements

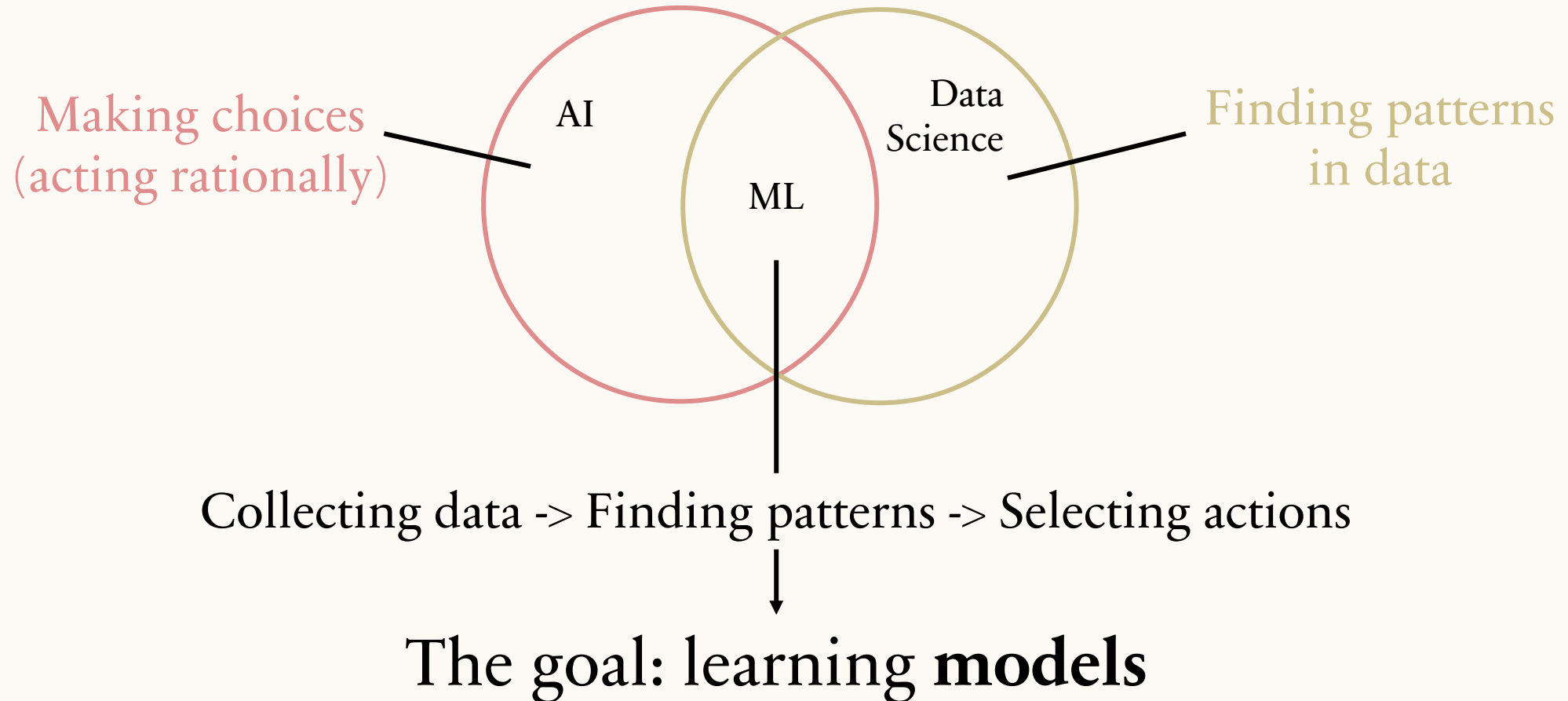
<https://laramartin.net/Principles-of-AI//homeworks/extra-credit-report.html>

11/28/2023 – Intro to Machine Learning



# **MACHINE LEARNING**

# WHAT IS MACHINE LEARNING?

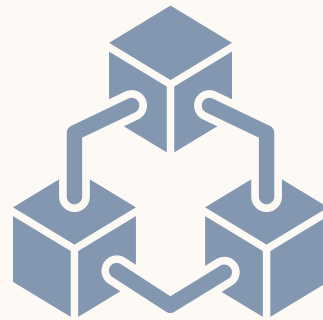


# WHAT IS A MODEL?

- Model: simplification of some phenomenon
  - Learning patterns from evidence
- Typically model something we **don't know** or **have trouble specifying**
  - How the environment can change
  - Actual states given only (partial) observations
  - How my sensors work
  - How my actions work
  - ...



# HOW DO WE LEARN MODELS?



Take past experiences  
(lots of data; corpus)

Find patterns  
(the ML algorithm)

Use on new experiences  
(save & test the model)

# TYPES OF MODELS

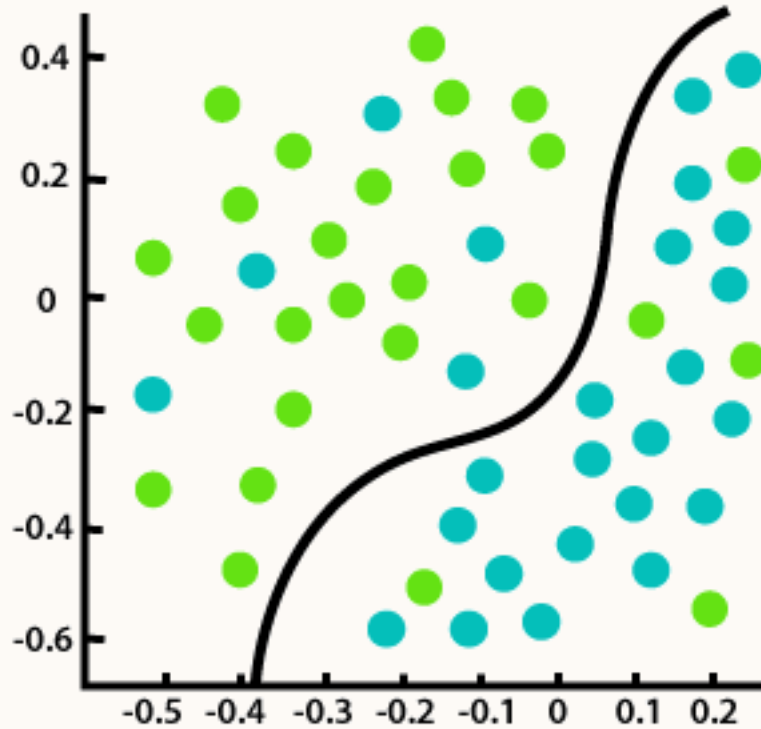
## Classification

- Model outputs comes from a finite set of values
- Discrete result
- *Examples:*
  - What type of animal is this a picture of?
  - Predicting the weather (sunny, cloudy, or rainy?)
  - Ranking: Is this result *better* than this result?

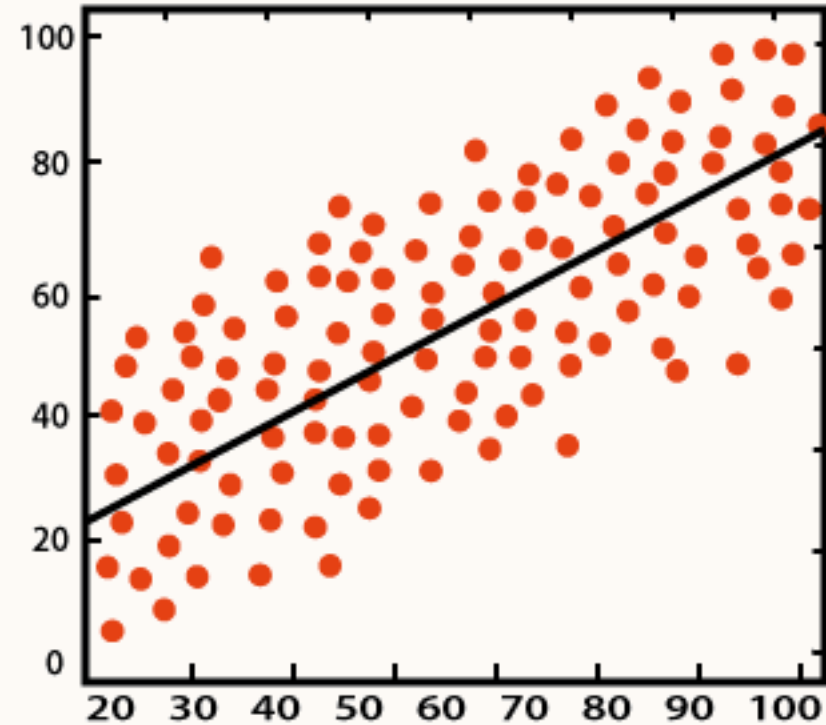
## Regression

- Model outputs are continuous values
- Continuous result
- *Examples:*
  - How far will I move if I drive my motors at this speed for 1 second?
  - Predicting the weather (temperature)
  - Ranking: *how good* is this result?

# TYPES OF MODELS



Classification



Regression

# WHAT ARE SOME OTHER EXAMPLES OF THESE?

## Classification

- Spam vs not spam
- Is a food poisonous
- Picture identification/labeling

## Regression

- House prices
- Finding relationship between two features
- Weather forecasting

# TYPES OF LEARNING

## Supervised learning

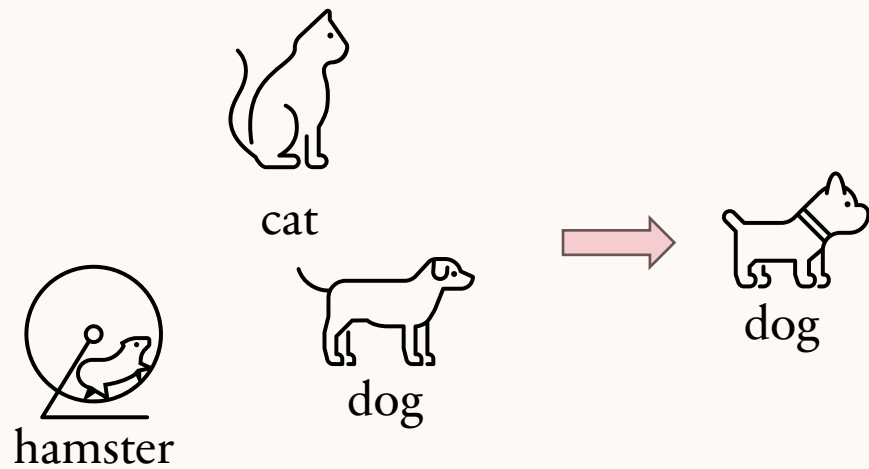
- Data has feedback (labels)
- Data consists of input-output pairs
- Learn mapping from input to output
- Examples:
  - Dataset classification
  - How likely is it that this person will get into a car accident?

## Unsupervised learning

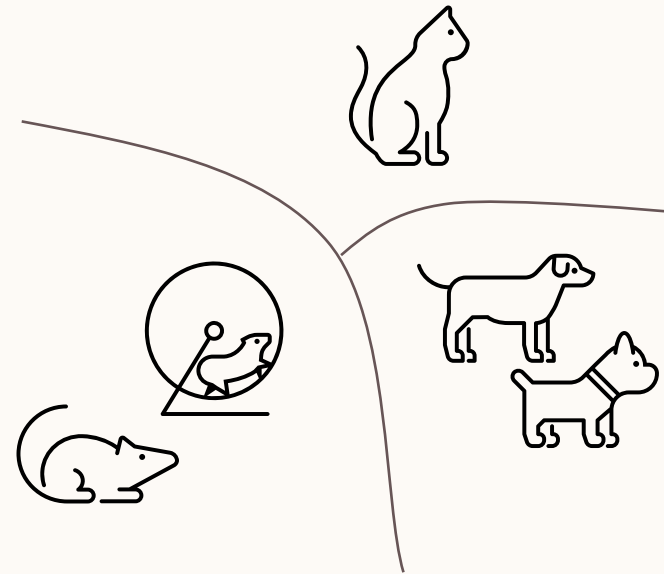
- No explicit feedback in data
- Learn patterns directly from data
- *Examples:*
  - Clustering
  - Do these people fall under multiple groups?

# TYPES OF LEARNING

## Supervised learning



## Unsupervised learning



# WHAT ARE SOME OTHER EXAMPLES OF THESE?

## Supervised learning

- Spam vs not-spam (if labeled)
- Disease detection
- Loan approval -- prediction

## Unsupervised learning

- Text clustering (topic modeling)
- Fraud detection (via outliers)
- Similarities across domain (e.g., clusters of grades)
- Suggesting similar images

# TYPES OF ALGORITHMS

*Supervised Learning*

*Unsupervised Learning*

*Discrete*  
*Continuous*

classification or categorization	clustering
regression	dimensionality reduction



# TYPES OF LEARNING

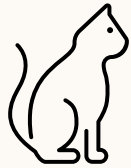
## Offline learning

- We have data in advance
- Predictions do not change over time
- Data is independent and identically distributed
- *Examples:*
  - Database learning tasks
  - Image captioning
  - Machine translation

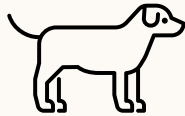
## Online learning

- Model receives input, produces output, and *then* is told what the correct output was
- Assume predictions may change over time
- Data is related over time
- *Examples:*
  - Long-term applications
  - Personalization
  - Field robotics

# HOW DO WE LEARN MODELS? (2)



cat



dog



hamster

[position of lines]  
[loops and dots]  
[any other information  
relevant to our problem]

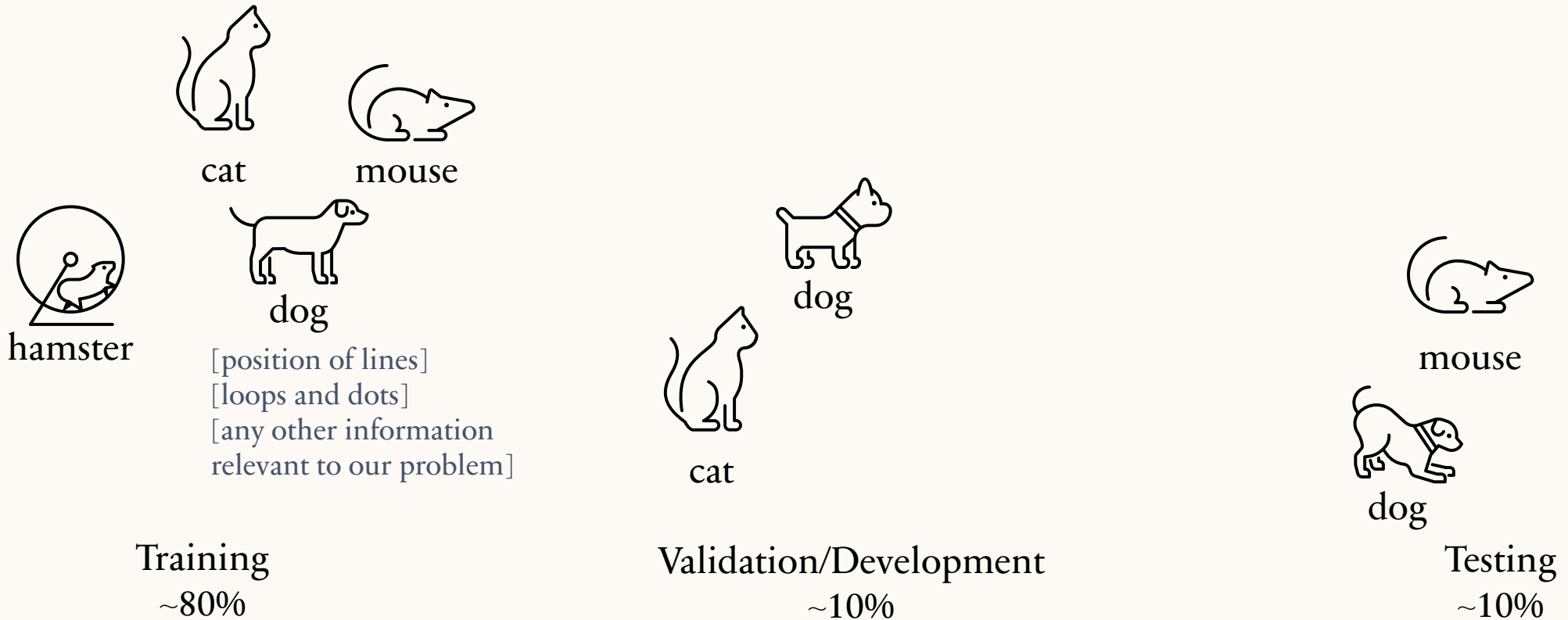
$P(\text{hamster} | [\text{line in this position}], \dots)$

$P(\text{dog} | [\text{line in this other position}], \dots)$

Have data with  
features extracted  
(and possibly labels)

Learn associations  
between features  
and labels

# DIVIDING UP DATA FOR TRAINING



Why would we do this? To prevent overfitting!