

**STAT 37710 / CAAM 37710 / CMSC 35400**  
**Machine Learning**

**Introduction**

Cong Ma

# What is machine learning?

- Wiki's definition of machine learning (ML):
  - Machine learning (ML) is a field of inquiry devoted to understanding and building methods that 'learn', that is, methods that **leverage data** to improve performance on some set of **tasks** ---adapted from Tom Mitchell

# (Almost) everybody knows ML

- Learn the mean of a normal distribution
- Data: iid data from normal distribution
- Task and performance: estimation and mean squared error
- Learning methods: sample mean

# Examples of machine learning

- **spam filter**

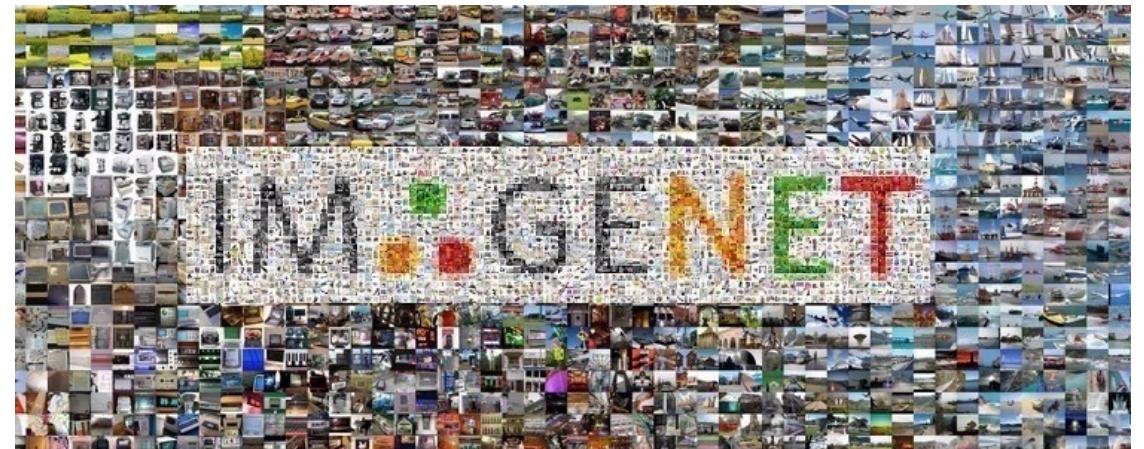
- Data: emails and labels
- Task: label emails to either spam or non-spam
- Performance: accuracy in labelling emails



# Examples of machine learning

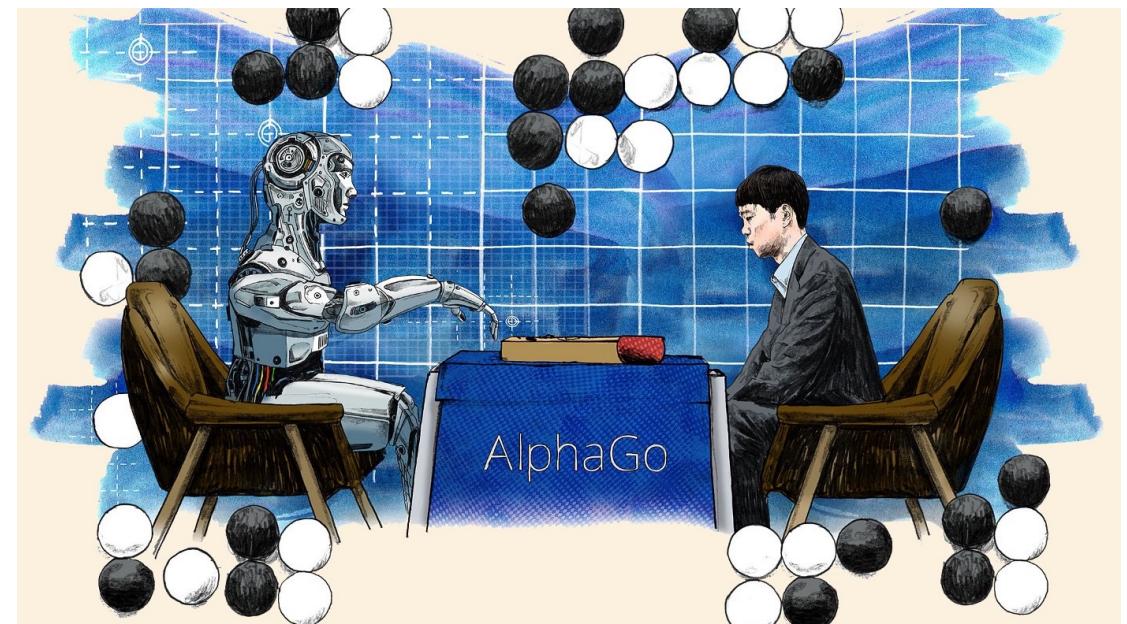
- **image classification**

- Data: images together with labels (ImageNet)
- Task: label images to categories (e.g., cat, dog)
- Performance: accuracy in labelling images



# Examples of machine learning

- Playing Go
  - Data: history of game playing
  - Task: playing Go well
  - Performance: winning rate against world champion



# Machine learning is ubiquitous...

**A large-scale crowd-sourced analysis of abuse against women journalists and politicians on Twitter**

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From Satellite Imagery to Disaster Insights

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**Wildlife Poaching Prediction with Data and Human Knowledge \***

We present a novel intelligent tutoring system which builds upon well-established

**A Scalable, Flexible Augmentation of the Student Education Process**

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Adithya Ramanathan  
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**Abstract**

# Helping Bees and Beekeepers with AI

Honeybee Identification with Machine Learning Using Augmented Microscopy

Peter He (Department of Computing, Imperial College London) · Alexis Gkantrigas (Department of Molecular Biology, University College London) · Gerard Glowacki (Imperial College London)

## Towards a Sustainable Food Supply Chain Powered By Artificial Intelligence

Volodymyr Kuleshov, Marlan Seymour, Danny Nemer, Nathan Fenner, Matthew Schwartz  
Afresh Technologies and Stanford University

Introduction

## Improving Traffic Safety in Jakarta Through video Analysis

João Caldeira, Alex Fout, Aniket Kesari, Raesetje Sefala, Katy Dupre, Joe Walsh

University of Chicago, Colorado State University

- **Problem:** ~2,000 people die in traffic accidents in Jakarta, Indonesia
- The city of Jakarta invests in traffic cameras to capture data on traffic behavior, but this data does not scale with an increasing population.

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## Next Hit Predictor - Self-exciting Risk Modeling for Predicting Next Locations of Serial Crimes

Yunyi Li  
The University of Iowa  
yunyi-li@uiowa.edu

Tong Wang  
The University of Iowa

# ML is interdisciplinary

statistics

information theory

## Machine Learning

optimization

algorithms

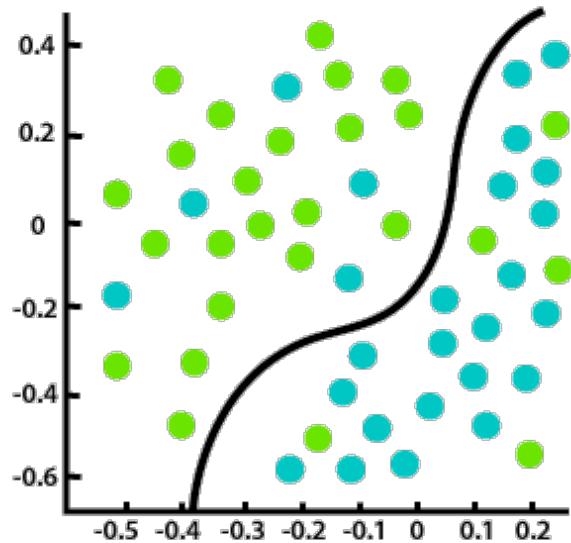
# Machine learning tasks

- *Depending on the feedbacks, machine learning can be decomposed into*
  - *Supervised Learning*
    - Regression
    - Classification
  - *Unsupervised Learning*
    - Clustering
    - Dimension reduction Anomaly detection, ...
- *Many other specialized tasks*

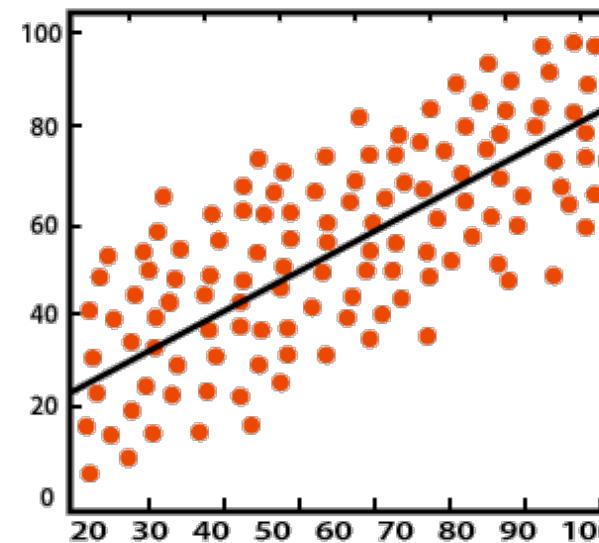
# Supervised learning

$$f : X \longrightarrow Y$$

# Supervised learning



Classification



Regression

# Image classification



# Regression

- **Goal:** Predict **real valued** labels (possibly vectors)
- Examples:

X

Flight route

Real estate objects

Patient & drug ....

Y

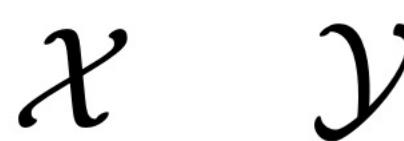
Delay (minutes)

Price

Treatment effectiveness ...

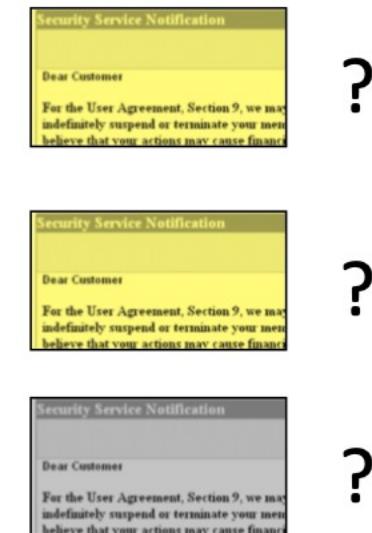
# Basic supervised learning pipeline

## Training Data



$$f : \mathcal{X} \rightarrow \mathcal{Y}$$

## Test Data



χ

# Example: Classifying documents

- **Input:**

- Documents with labels, but how to represent documents

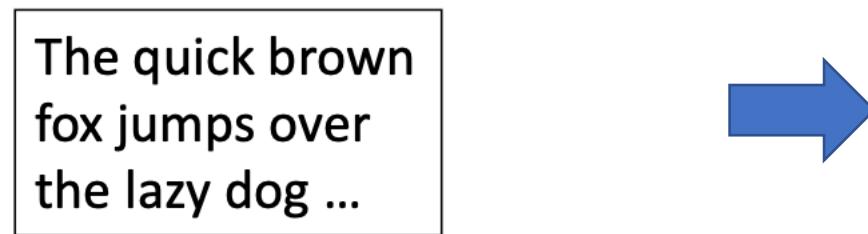
- **Goal:**

- Learn a good classifier



# Representing data

- Learning methods expect standardized representation of data
  - (e.g., Points in vector spaces, nodes in a graph, similarity matrices ...)



- Concrete choice of representation (“features”) is crucial for successful learning
- This class (typically): **feature vectors** in  $\mathbb{R}^d$

# Example: Bag-of-words

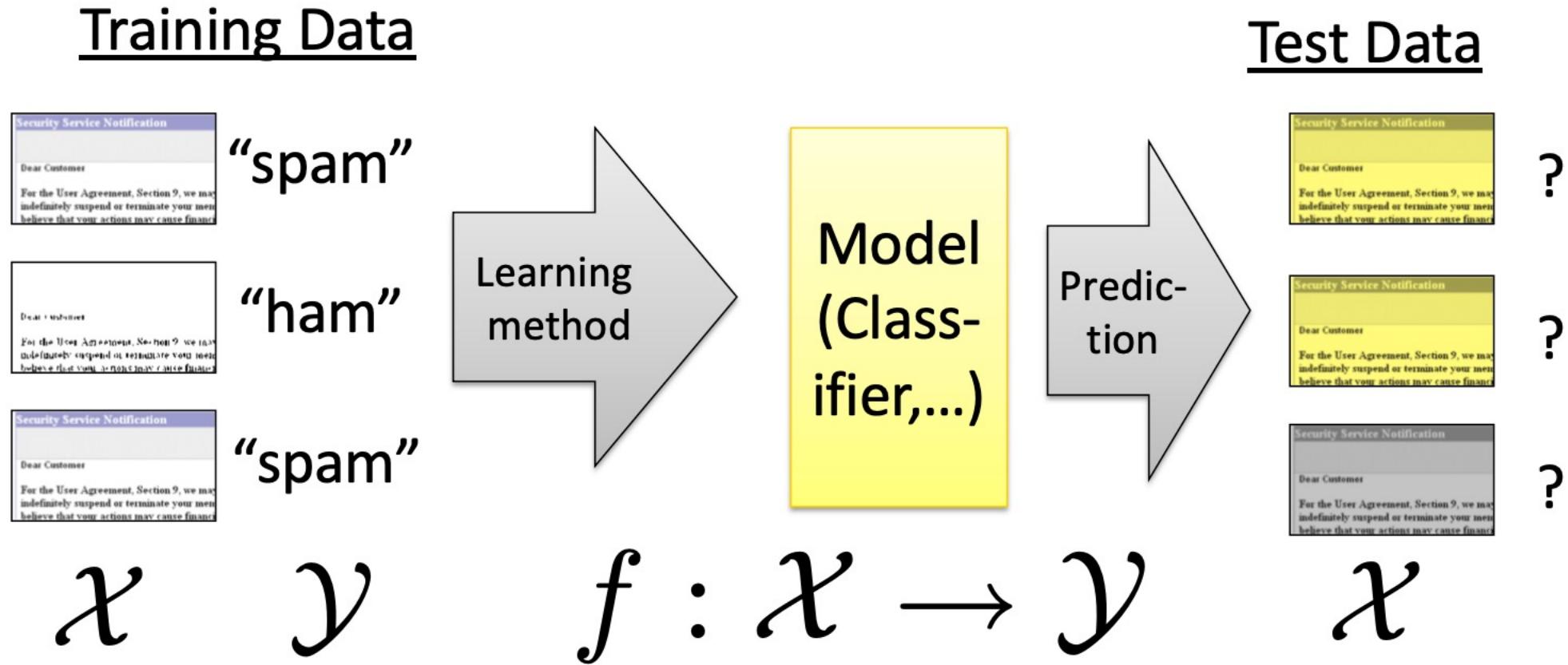
- Suppose language contains at most  $d=100000$  words
- Represent each document as a vector  $\mathbf{x}$  in  $\mathbb{R}^d$
- $i$ -th component  $x_i$  counts occurrence of  $i$ -th word

Word	Index
a	1
abandon	2
ability	3
...	
is	578
...	
test	2512
...	
this	2809
....	

# Bag-of-words: Improvements

- Some words more “important” than others
  - Remove “stopwords” (the, a, is, ...)
  - Stemming (learning, learner, learns -> learn)
  - Discount frequent words (tf-idf)
- Bag-of-words ignores order
  - Consider pairs (n-grams) of consecutive words
- Does not differentiate between similar and dissimilar words (ignores semantics)
  - Word embeddings (e.g., word2vec, GloVe)

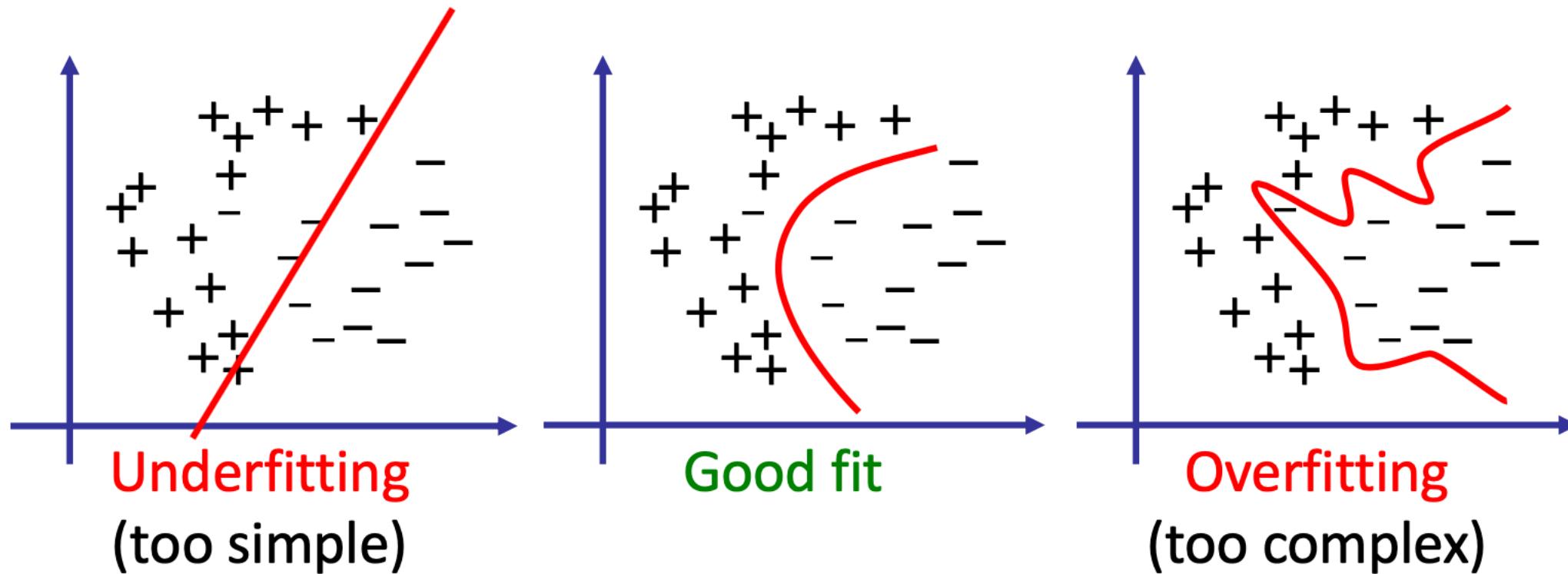
# Basic supervised learning pipeline



# Model class

- Linear
- decision tree
- random forests
- graphical models
- deep neural networks

# Model selection and validation



# Other models of learning

- **Unsupervised learning**
  - Learning without labels
- **Semi-supervised learning**
  - Learning from both labeled and unlabeled data
- **Transfer learning**
  - Learn on one domain and test on another
- **Active learning**
  - Acquiring most informative data for learning
- **Online learning**
  - Learning from examples as they arrive over time
- **Reinforcement learning**
  - Learning by interacting with an unknown environment

# Summary

- Where we are

- Basic forms of learning:
  - Supervised learning and other modes of machine learning
- Key challenge in ML
  - Trading goodness of fit and model complexity
- Representation of data is of key importance

- What's next

- Formally state machine learning problems
- Estimation theory and bias-variance tradeoff