Applied Machine Learning

Introduction

Siamak Ravanbakhsh

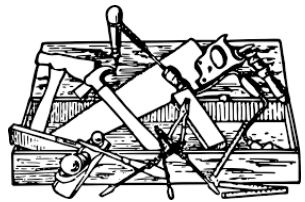
COMP 551 **(winter 2020)**

Objectives

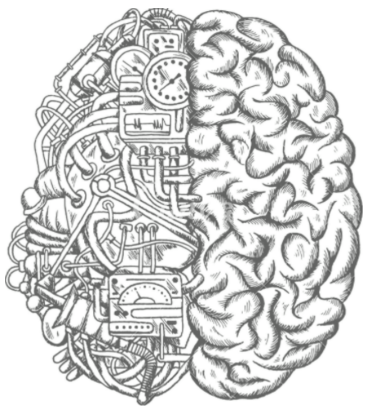
- understanding the scope of machine learning
  - relation to other areas
- understanding types of machine learning

What is Machine Learning?

ML is the set of "algorithms and statistical models that computer systems use to perform a specific task without using explicit instructions"



an inadequate history of ML



- 1950: Turing test
- 1956: checker player that learned as it played (Arthur Samuel)
  - coined the term Machine Learning
- 1958: first artificial neural networks called Perceptron (Frank Rosenblatt),
  - 1959: ADELINE (Widrow and Hoff)
- 1963: support vector machines (Vapnick & Ya)
- 1969: Minskey and Pappert show the limitations of single-layer neural networks
- 1970-80s rule-based and symbolic AI dominates (two AI winters)
- 1980's Bayesian networks (Judea Pearl)
- 1986 Backpropagation rediscovered (Rumelhart, Hinton & Williams)
- 1991 Kernel trick for SVM
- 2012 AlexNet wins Imagenet by a large margin
- 2012 - now deep learning explosion...
- next? AI winter? AGI?

Placing ML: overlapping fields

- **Artificial Intelligence**: its a broader domain (includes search, planning, multiagent systems, robotics, etc.)

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- **Data science**: an umbrella term for the above mostly used in industry when the output is knowledge/information to be used for decision making

Placing ML: main venues

top computer science conferences

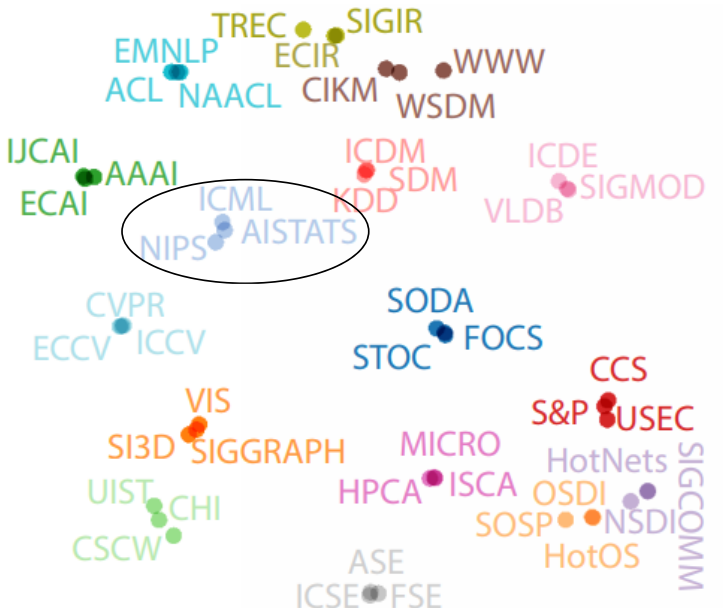
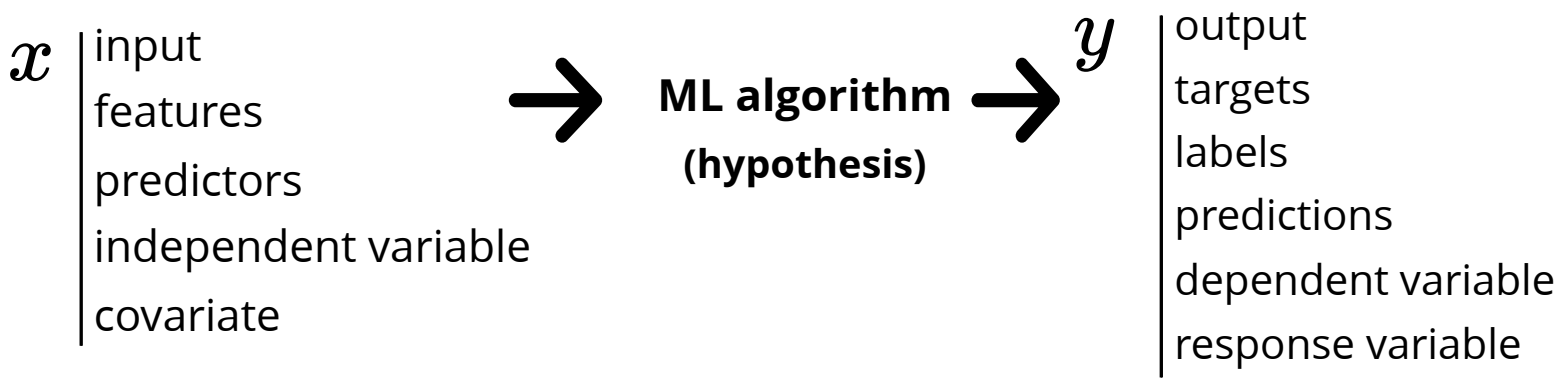


Figure from Dong et al.

Some terminology

x	input
	features
	predictors
	independent variable
	covariate

Some terminology



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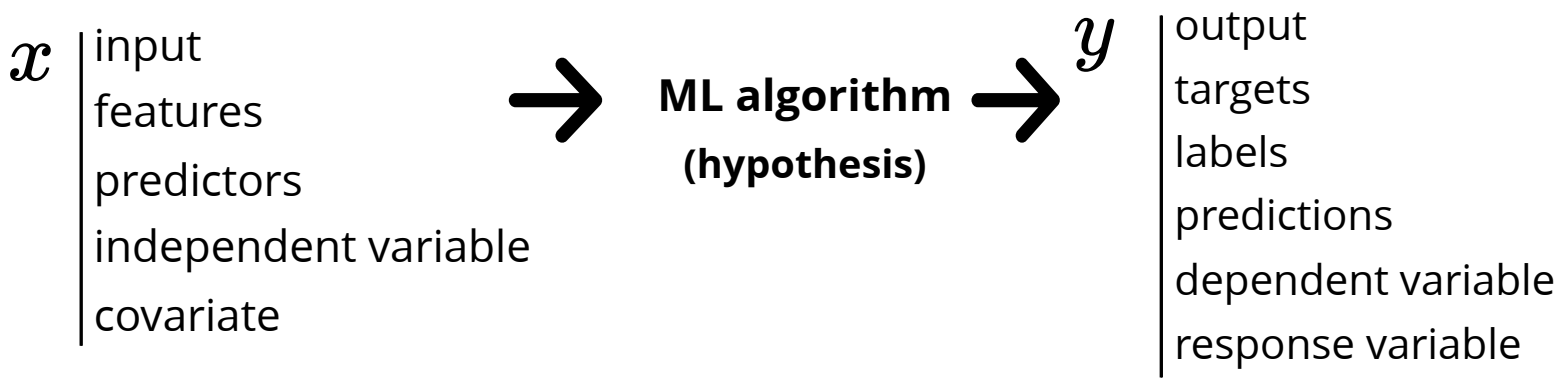

example

<tumorsize, texture, perimeter> = <18.2, 27.6, 117.5> → cancer = No

Some terminology

(labelled) **datasets**: consist of many training examples or **instances**

<tumorsize,	texture,	perimeter>	,	<cancer,	size change>
<18.2,	27.6,	117.5>	,	< No ,	+2 >
<17.9,	10.3,	122.8>	,	< No ,	-4 >
<20.2,	14.3,	111.2>	,	< Yes ,	+3 >
		⋮			
<15.5,	15.2,	135.5>	,	< No ,	0 >

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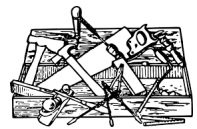
<tumorsize,	texture,	perimeter>	,	<cancer,	size change>		
<18.2,	27.6,	117.5>	,	< No ,	+2	>	$x^{(1)}$	
<17.9,	10.3,	122.8>	,	< No ,	-4	>	$x^{(2)}$	one instance
<20.2,	14.3,	111.2>	,	< Yes ,	+3	>	$x^{(3)}$	
		⋮					⋮	
<15.5,	15.2,	135.5>	,	< No ,	0	>	$x^{(N)}$	

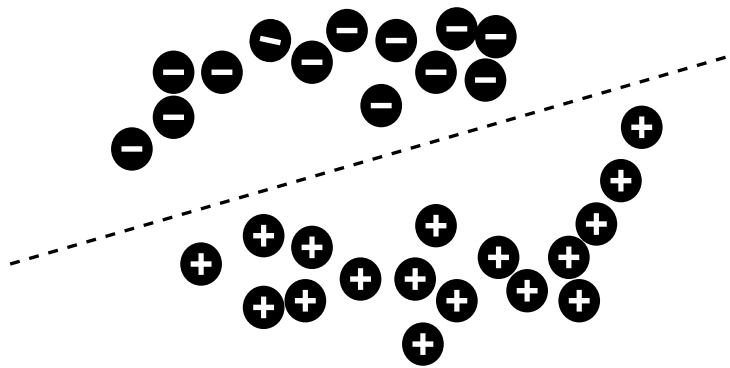
"Types" of Machine Learning



Supervised Learning: we have labeled data

- classification
- regression
- structured prediction

most of this course!



Supervised Learning

Regression: continuous output

<tumorsize, texture, perimeter>	,	<size change>
<18.2, 27.6, 117.5>	,	< +2 >
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<15.5, 15.2, 135.5>	,	< 0 >

target

Supervised Learning

Classification: categorical/discrete output

<tumorsize, texture, perimeter>	,	<cancer>
<18.2, 27.6, 117.5>	,	< No >
<17.9, 10.3, 122.8>	,	< No >
<20.2, 14.3, 111.2>	,	< Yes >
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Regression: continuous output

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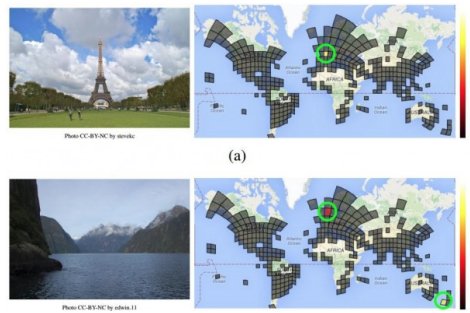
target

Supervised Learning: Example

MIT Technology Review

Topics+ Tl



(a)

Intelligent Machines

Google Unveils Neural Network with "Superhuman" Ability to Determine the Location of Almost Any Image

Guessing the location of a randomly chosen Street View image is hard, even for well-traveled humans. But Google's latest artificial-intelligence machine manages it with relative ease.

by Emerging Technology from the arXiv February 24, 2016

Supervised Learning: Example

Machine Translation:
data consists of input-output sentence pairs (x,y)

DeepL schools other online translators with clever machine learning

Devin Coldewey, Frederic Lardinois / 1:57 pm EDT • August 29, 2017 Comment

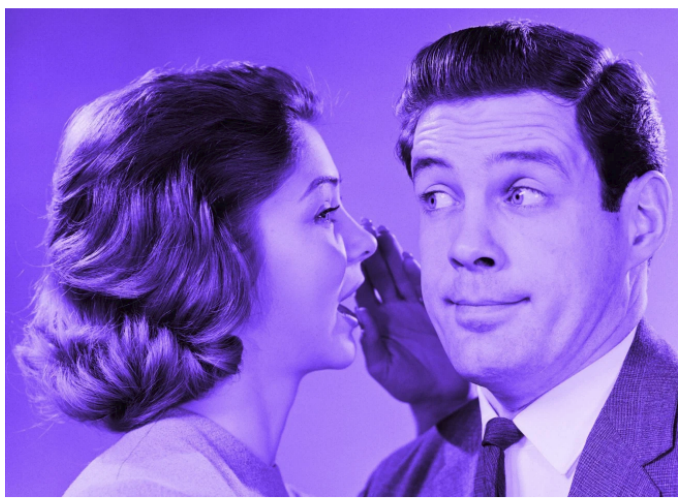


Image Credits: H. Armstrong Roberts/Getty Images

Supervised Learning: Example

Image captioning

input: image

output: text

The man at bat readies to swing at the pitch while the umpire looks on.

A large bus sitting next to a very tall building.

image: COCO dataset

Supervised Learning: Example

Object detection

input: image

output: a set of bounding box coordinates

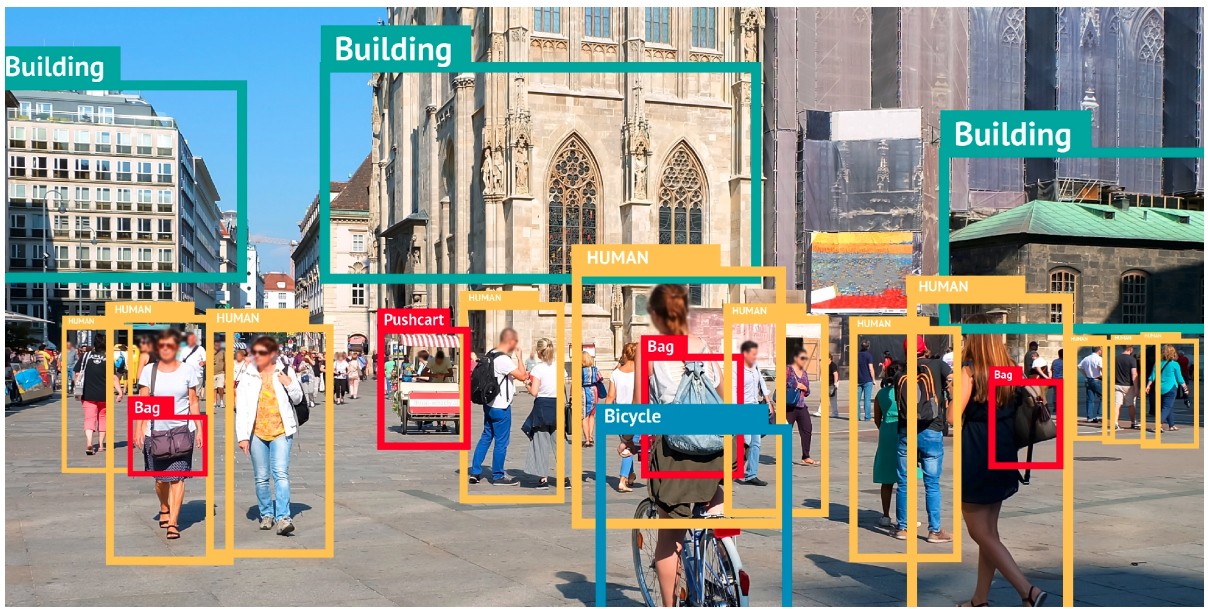
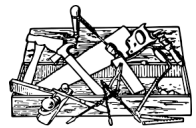


image: https://bitmovin.com/object-detection/

"Types" of Machine Learning



Unsupervised Learning: only unlabeled data

- clustering
- dimensionality reduction
- density estimation / generative modeling
- anomaly detection
- discovering latent factors and structures

helps explore and understand the data

closer to data mining

we have much more unlabeled data

more open challeges

Unsupervised Learning: Example

clustering

similar to classification but labels/classes should be inferred and are not given to the algorithm

<tumorsize,	texture,	perimeter>
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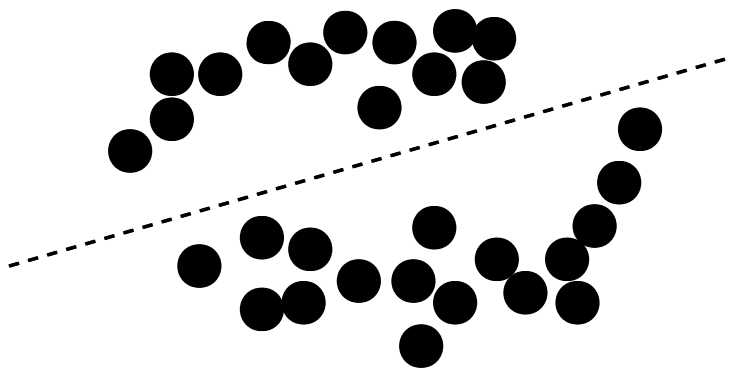


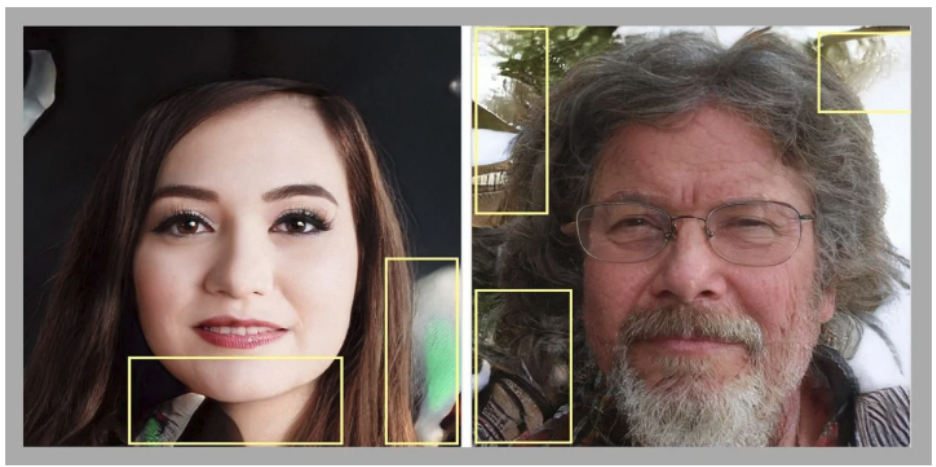
Unsupervised Learning: Example

Generative modeling (density estimation):
learn the data distribution $p(x)$

PARIS MARTINEAU BUSINESS 12.20.2019 08:21 PM

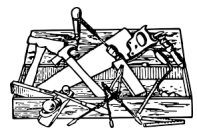
Facebook Removes Accounts With AI-Generated Profile Photos

Researchers said it appears to be the first use of artificial intelligence to support an inauthentic social media campaign.



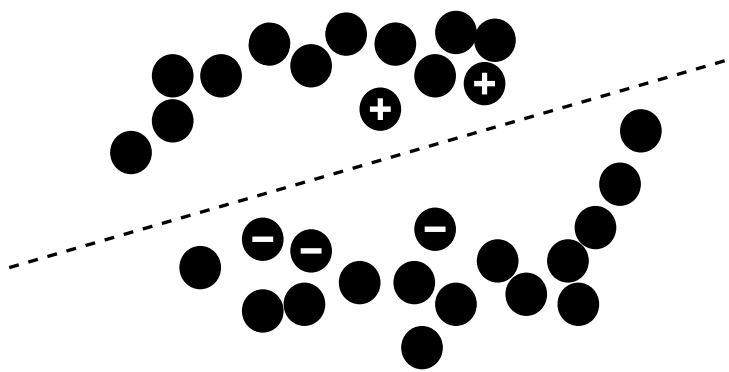
Profile pictures for Facebook accounts "Mary Keen" and "Jacobs Guillermo," admins on groups associated with The BL highlighted by Graphika. COURTESY OF GRAPHIKA

"Types" of Machine Learning



Semisupervised learning: a few labeled examples

- we can include structured problems such as
  - matrix completion (a few entries are observed)
  - link prediction



Netflix Awards $1 Million Prize and Starts a New Contest

BY STEVE LOHR SEPTEMBER 21, 2009 10:15 AM

Jason Kempin/Getty Images Netflix prize winners, f
Chabbert, Martin Piotte, Michael Jahrer, Andreas To

Update | 1:45 p.m. *Adding details announced close finish to the contest.*

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RYAN SINGEL SECURITY 03.12.10 02:48 PM

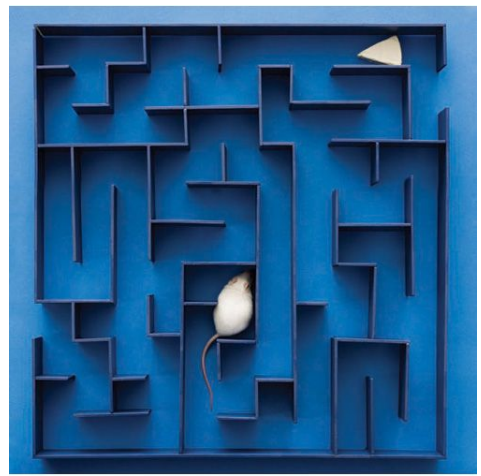
NetFlix Cancels Recommendation Contest After Privacy Lawsuit

Netflix is canceling its second $1 million Netflix Prize to settle a legal challenge that it breached customer privacy as part of the first contest's race for a better movie-recommendation engine.

"Types" of Machine Learning

Reinforcement Learning:

- weak supervision through the reward signal
- sequential decision making
- biologically motivated

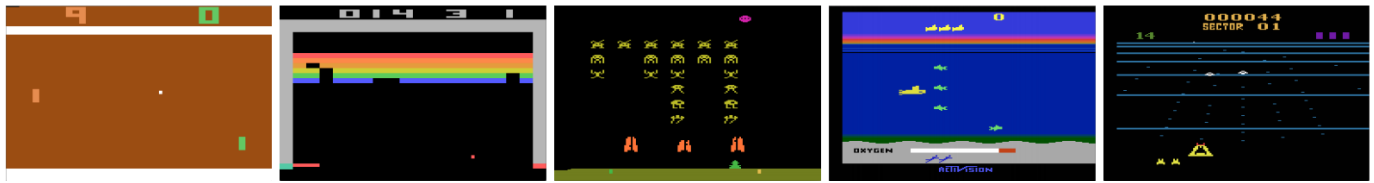


also related:

imitation learning: learning from demonstrations

- behavior cloning (is supervised learning!)
- inverse reinforcement learning (learning the reward function)

Reinforcement Learning: Example



Human Level Control Through Deep Reinforcement Learning

Abstract

The theory of reinforcement learning provides a normative account deeply rooted in psychological and neuroscientific perspectives on animal behaviour, of how agents may optimize their control of an environment. To use reinforcement learning successfully in situations approaching real-world complexity, however, agents are confronted with a difficult task: they must derive efficient representations of the environment from high-dimensional sensory inputs, and use these to generalize

Reinforcement Learning: Example

Google's AlphaGo Defeats Chinese Go Master in Win for A.I.

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By **PAUL MOZUR** MAY 23, 2017

Ke Jie, the world's top Go player, reacting during his match on Tuesday against AlphaGo, artificial intelligence software developed by a Google affiliate. China Stringer Network, via Reuters

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Summary

- **Supervised Learning:** we have labeled data
  - classification
  - regression
  - structured prediction
- **Unsupervised Learning:** only unlabeled data
  - clustering
  - dimensionality reduction
  - density estimation / generative modeling
  - anomaly detection
  - discovering latent factors and structures
- **Semisupervised learning:** a few labeled examples
- **Reinforcement Learning:** reward signal