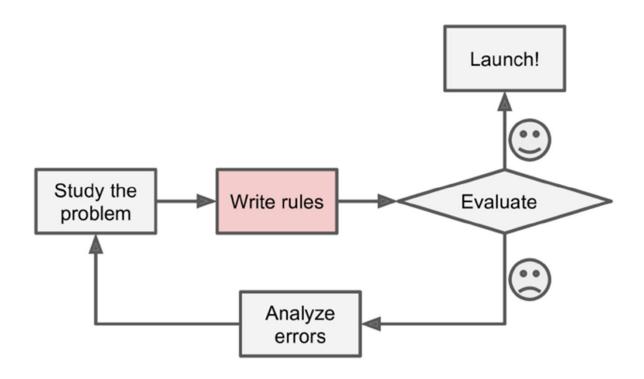
DS-GA 3001.007 Introduction to Machine Learning

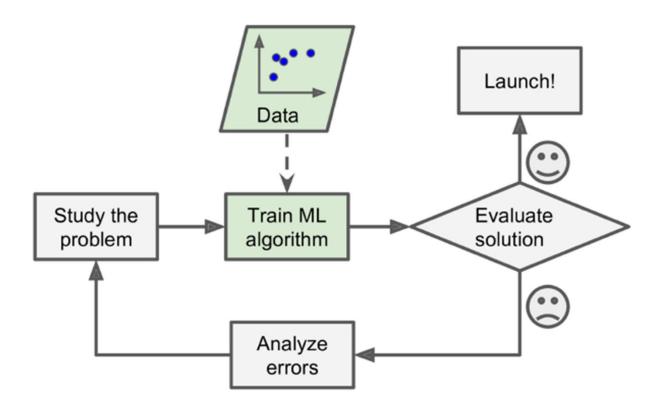
Lecture 1

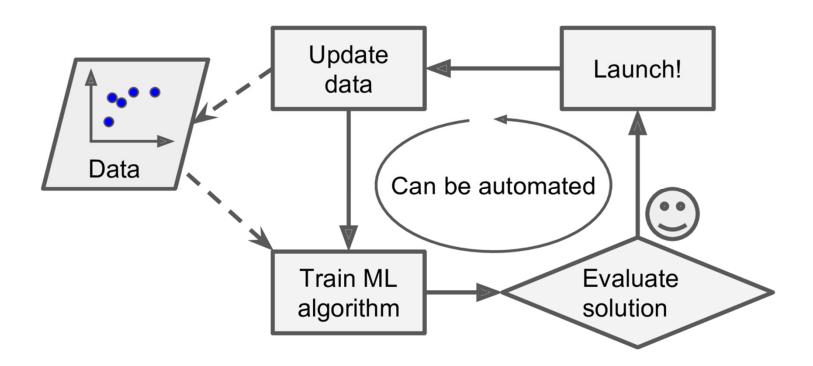
Agenda

- Overview
 - ▶ What is ML?
 - ▶ Who uses ML?
 - ► Why study ML?
- Lesson
 - ▶ What are the different types?
 - ▶ What are the components of an application?
 - ▶ How to use ML for data science?
- Demo
 - ► Churn analysis

A computer program learns from experience E with respect to some task T and some performance measure P, if its performance on task T, as measured by P, improves with experience E.



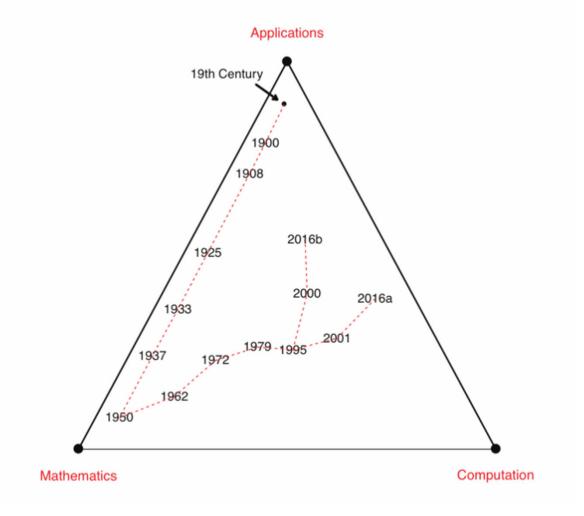




Who uses machine learning?

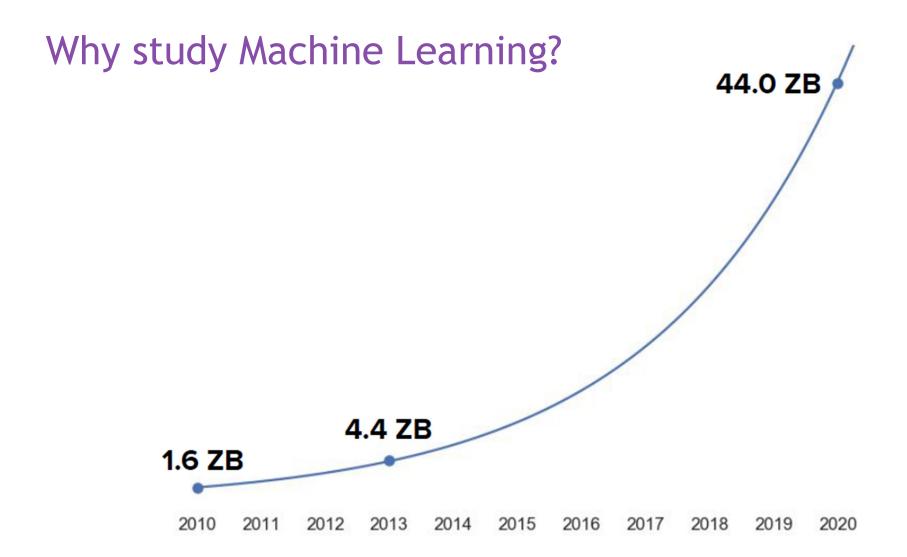
- ► Natural Language Processing
 - ► Text classification
 - ► Part of speech recognition
- ► Computer Vision
 - ► Character recognition
 - ► Image retrieval
- ► Speech Recognition
 - ► Source separation
 - ► Speaker Identification

Who uses Machine Learning?



Why study Machine Learning?

- ► Adaptive Systems
 - ▶ Deploy without hard-coded routines
- ► Scalable Systems
 - Determine patterns in large and complex sets of observations



DS-GA 3001.007 Introduction to Machine Learning

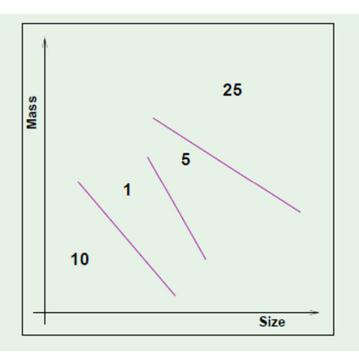
- Interdisciplinary course for students in the sciences, engineering and humanities
- ► Comparable to DS-GA 1003
- ► Goals
 - ▶ Prepare
 - **►** Empower
 - **►** Enable

Types of Machine Learning

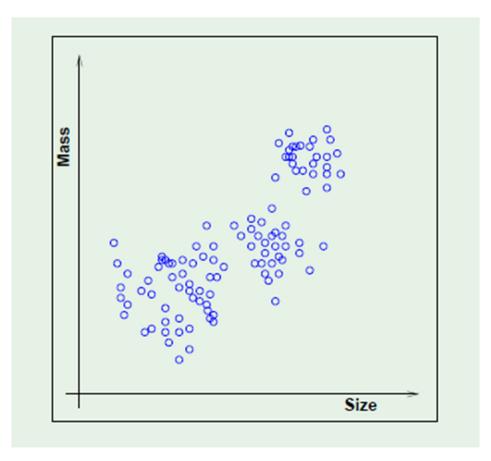
- Supervised Learning
 - ► Classification
 - ► Regression
- Unsupervised Learning
 - Clustering
 - ▶ Dimension Reduction
- ► Reinforcement Learning

Supervised Machine Learning

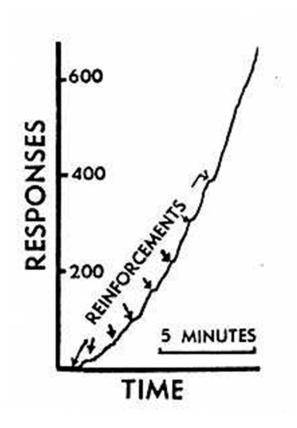




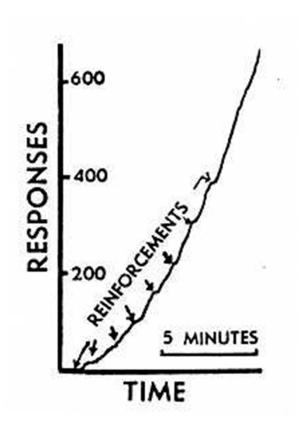
Unsupervised Machine Learning



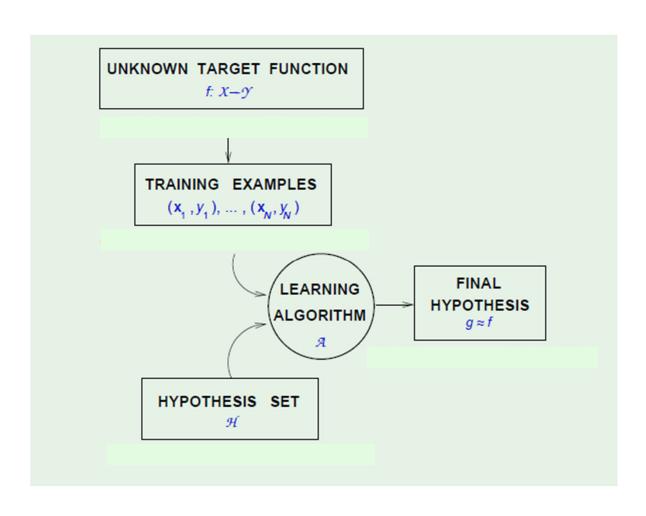
Reinforcement Learning



No Free Lunch



- ► For the spam detection, can you propose examples of...
 - Supervised Learning
 - Unsupervised Learning
 - ► Reinforcement Learning



- ► Sampling data
 - ► Train Sample
 - ► Validation Sample
 - ► Testing Sample

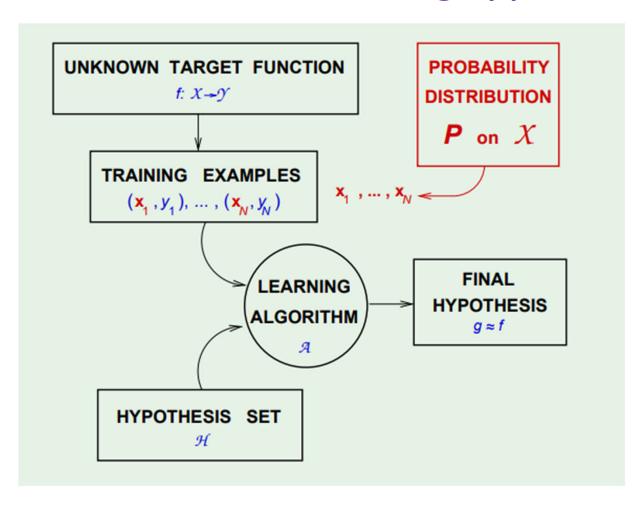
- Sampling data
 - ► Train Sample
 - ► Validation Sample
 - ► Testing Sample
- ► Input/Output
 - ▶ Features
 - ▶ Labels

- Sampling data
 - ► Train Sample
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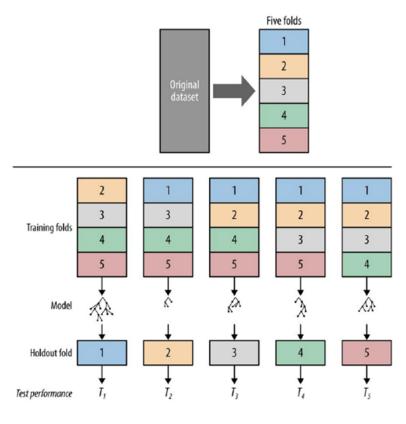
- ▶ Fitting a model
 - ▶ Hypotheses
 - ▶ Hyperparameters

- Sampling data
 - ► Train Sample
 - ► Validation Sample
 - ► Testing Sample
- ► Input/Output
 - Features
 - ▶ Labels

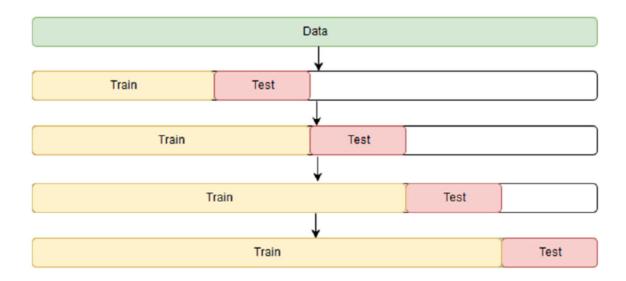
- Fitting a model
 - Hypotheses
 - ▶ Hyperparameters
- Error Analysis
 - ► Loss functions
 - Accuracy Metrics



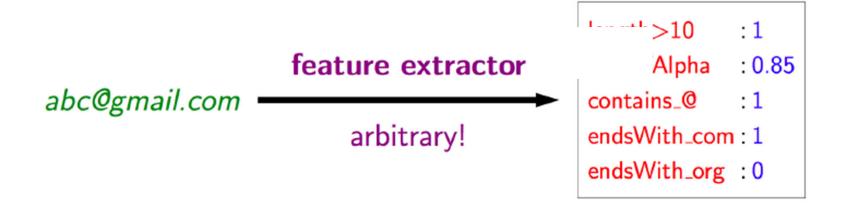
Train, Validate, Test



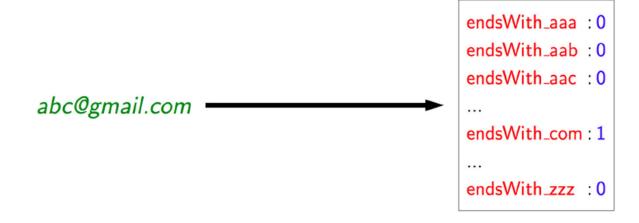
Train, Validate, Test

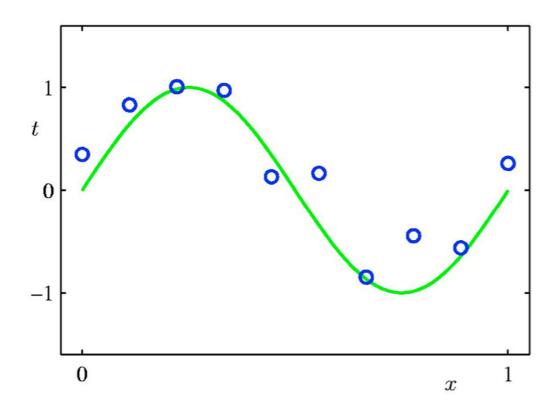


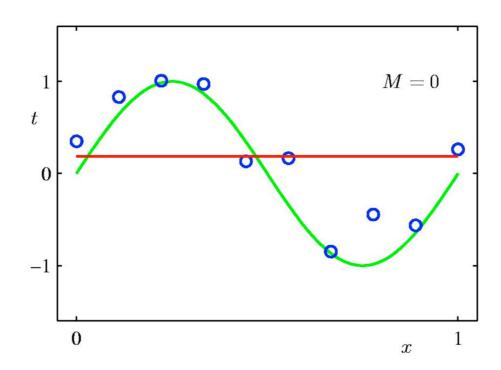
Features and Labels

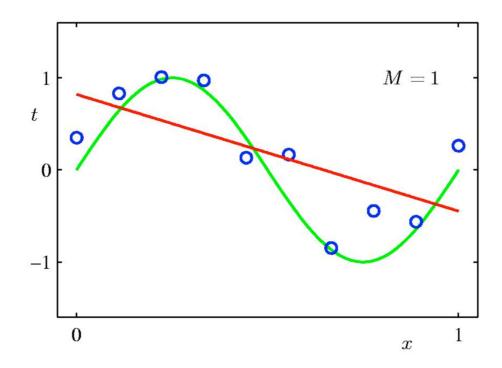


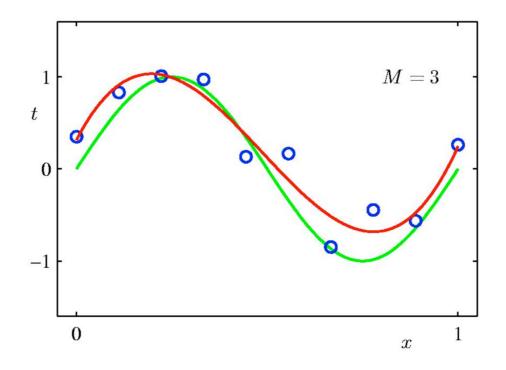
Features and Labels

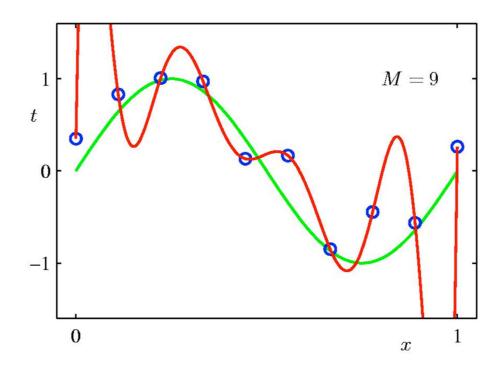


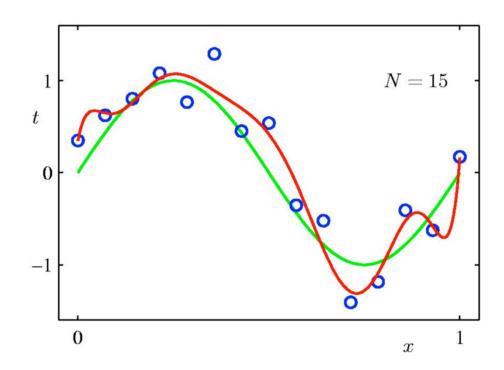


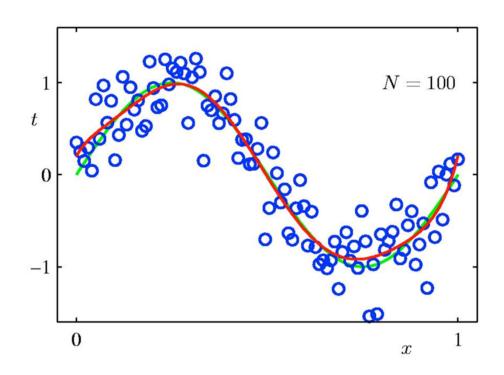






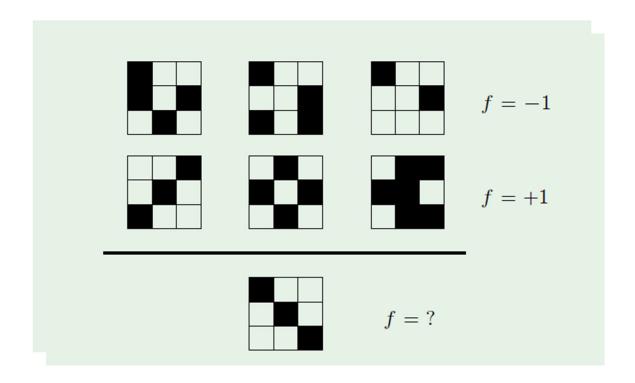






Loss Function and Metrics

		True condition		
	Total population	Condition positive	Condition negative	
Predicted	Predicted condition positive	True positive	False positive, Type I error	
condition	Predicted condition negative	False negative, Type II error	True negative	



Example	x_1	x_2	x_3	x_4	y
1	0	0	1	0	0
2	0	1	0	0	0
3	0	0	1	1	1
4	1	0	0	1	1
5	0	1	1	0	0
6	1	1	0	0	0
7	0	1	0	1	0

Example	x_1	x_2	x_3	x_4	y
1	0	0	1	0	0
2	0	1	0	0	0
3	0	0	1	1	1
4	1	0	0	1	1
5	0	1	1	0	0
6	1	1	0	0	0
7	0	1	0	1	0

x_1	x_2	x_3	x_4	y
0	0	0	0	?
0	0	0	1	?
0	0	1	0	0
0	0	1	1	1
0	1	0	0	0
0	1	0	1	0
0	1	1	0	0
0	1	1	1	?
1	0	0	0	?
1	0	0	1	1
1	0	1	0	?
1	0	1	1	?
1	1	0	0	0
1	1	0	1	?
1	1	1	0	?
1	1	1	1	?

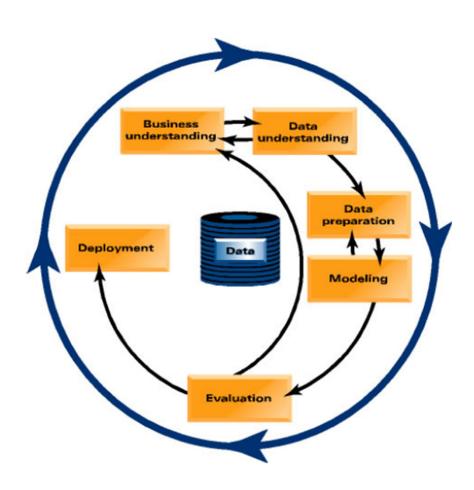
Rule	Counterexample
$\Rightarrow y$	1
$x_1 \Rightarrow y$	3
$x_2 \Rightarrow y$	2
$x_3 \Rightarrow y$	1
$x_4 \Rightarrow y$	7
$x_1 \wedge x_2 \Rightarrow y$	3
$x_1 \wedge x_3 \Rightarrow y$	3
$x_1 \wedge x_4 \Rightarrow y$	3
$x_2 \wedge x_3 \Rightarrow y$	3
$x_2 \wedge x_4 \Rightarrow y$	3
$x_3 \wedge x_4 \Rightarrow y$	4
$x_1 \wedge x_2 \wedge x_3 \Rightarrow y$	3
$x_1 \wedge x_2 \wedge x_4 \Rightarrow y$	3
$x_1 \wedge x_3 \wedge x_4 \Rightarrow y$	3
$x_2 \wedge x_3 \wedge x_4 \Rightarrow y$	3
$x_1 \wedge x_2 \wedge x_3 \wedge x_4 \Rightarrow y$	3

x_1	x_2	x_3	x_4	y
0	0	0	0	?
0	0	0	1	?
0	0	1	0	0
0	0	1	1	1
0	1	0	0	0
0	1	0	1	0
0	1	1	0	0
0	1	1	1	?
1	0	0	0	?
1	0	0	1	1
1	0	1	0	?
1	0	1	1	?
1	1	0	0	0
1	1	0	1	?
1	1	1	0	?
1	1	1	1	?

How to use machine learning for data science?

- ► Components of Data Science
 - ► Formulate a problem
 - ► Gather data
 - ► Explore data
 - ▶ Determine a model for prediction and inference
 - ► Evaluate findings

How to use machine learning for data science?



How to use machine learning for customer churn?

The Data

- 1. The historical data consist of 39,859 customers. The historical data contain 19,901 customers that churned (*i.e.* left the company) and 19,958 that did not churn (see the "churndep" variable).
- 3. Here are the data set's 11 possible predictor variables for churning behavior:

Position	Variable Name	Variable Description
1	revenue	Mean monthly revenue in dollars
2	outcalls	Mean number of outbound voice calls
3	incalls	Mean number of inbound voice calls
4	months	Months in Service
5	eqpdays	Number of days the customer has had his/her current equipment
6	webcap	Handset is web capable
7	marryyes	Married (1=Yes; 0=No)
8	travel	Has traveled to non-US country (1=Yes; 0=No)
9	pcown	Owns a personal computer (1=Yes; 0=No)
10	credited	Possesses a credit card (1=Yes; 0=No)
11	retcalls	Number of calls previously made to retention team

The dependent variable, Churndep, = 1 if the customer churned, = 0 otherwise.

Reminders

- Section
 - ► Section 008 Thursdays 2:25-3:15 pm (60th 5th Avenue, Room 1/15)
 - ► Access to https://iml-f19.jupyter.hpc.nyu.edu
- Syllabus
 - ▶ Please review the course policies about assignments and grading.
- Surveys
 - ▶ Please complete Survey 1