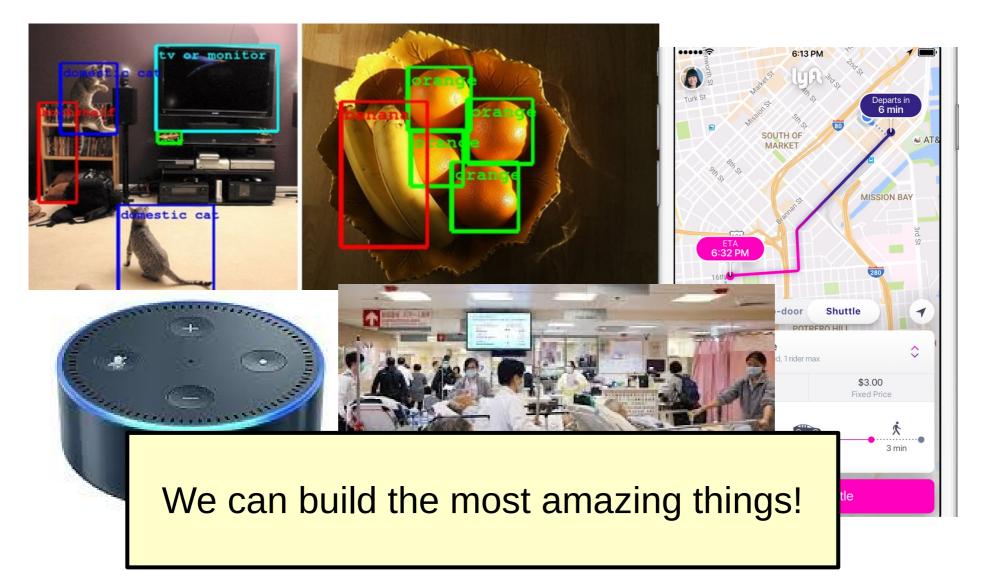
CS181: Introduction to Machine Learning

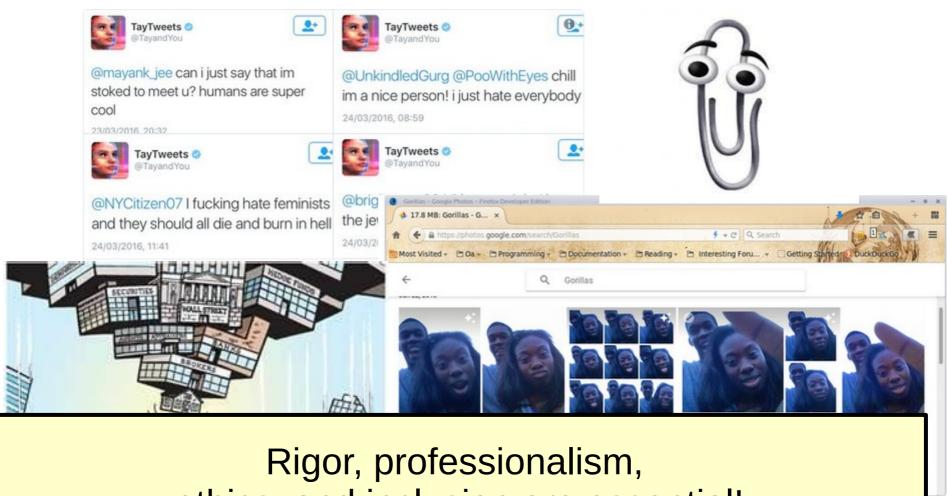
Finale Doshi-Velez











ethics, and inclusion are essential!

https://twitter.com/geraldmellor/status/712880710328139776/photo/1?ref src=twsrc%5Etfw&ref url=https%3A%2F%2Fwww.theverge.com%2F2016%2F3%2F24%2F11297050%2Ftay-microsoft-chatbot-racist https://twitter.com/jackvalcine/status/615331869266157568?ref_src=twsrc^tfw&ref_url=https%3A%2F%2Fwww.usatoday.com%2Fstory%2Ftech%2F2015%2F07%2F01%2Fgoogle-apologizes-after-photos-identify-blackpeople-as-gorillas%2F29567465%2F

How we will get there



- Make appropriate model choices
- Have sufficient understanding to learn and apply the new techniques
- Identify sources of error
- Evaluate carefully

How we will get there

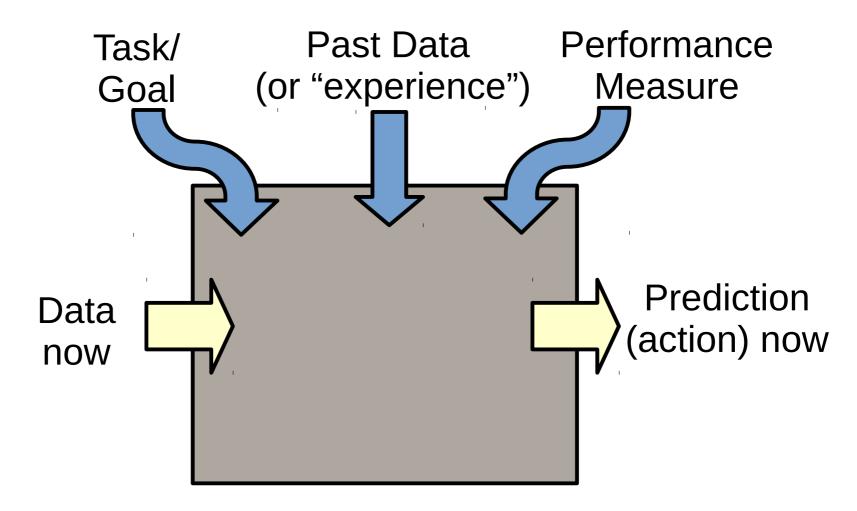


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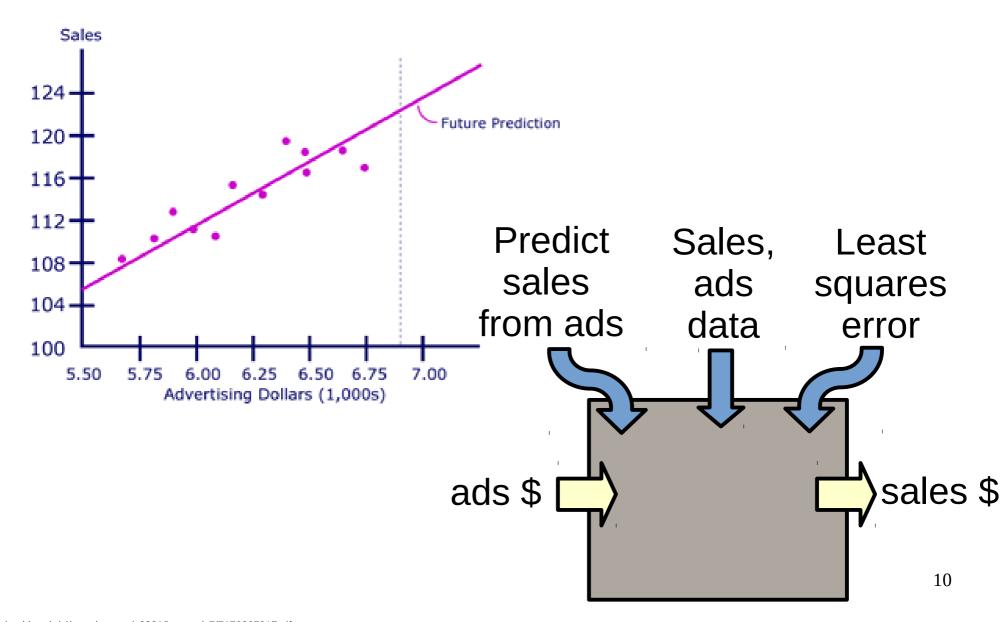


Let's get to it!

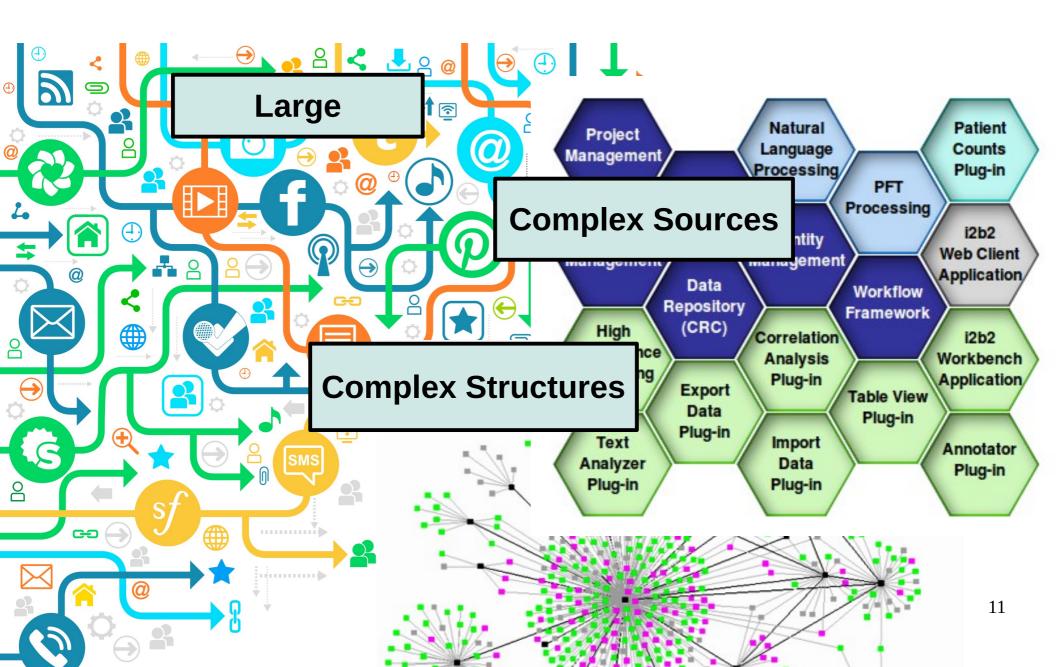
What is machine learning?



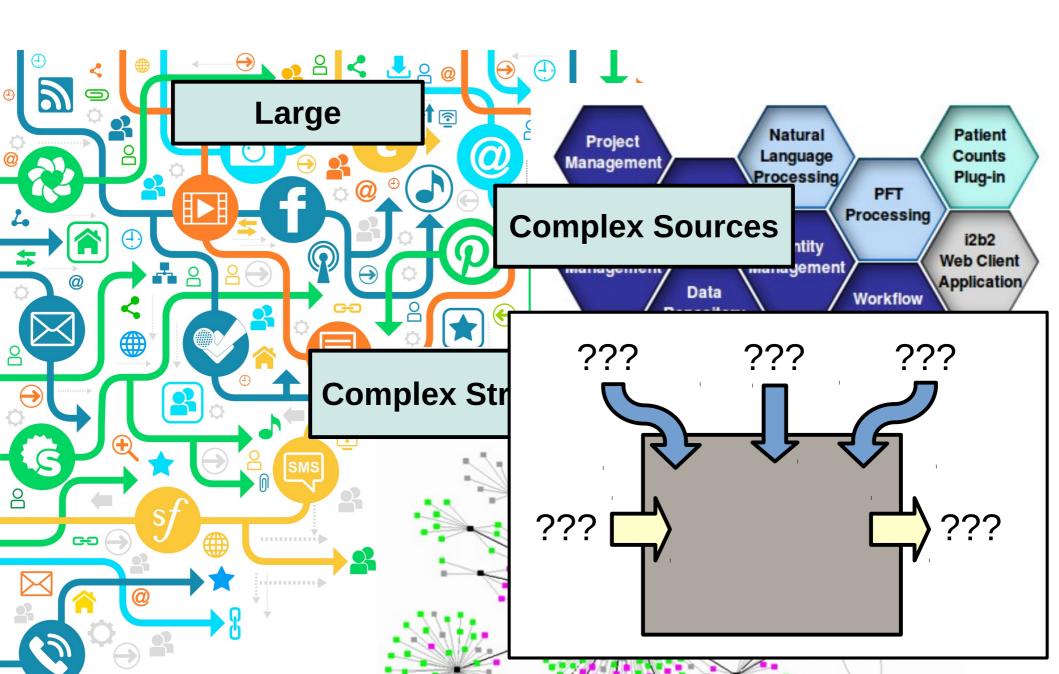
The starting point...



... where we are now



... where we are now



A Story: Treating Depression

Clinical Question: What meds to give what depression patients?



Clinical Question: What meds to give what depression patients?

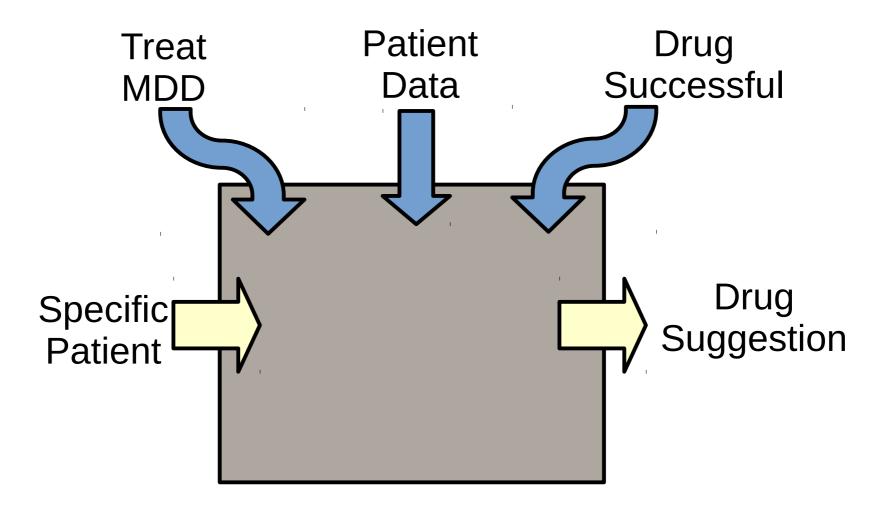


Clinical Question: What meds to give what depression patients?



Also: Can we explain who responds to what drug?

Formalizing...



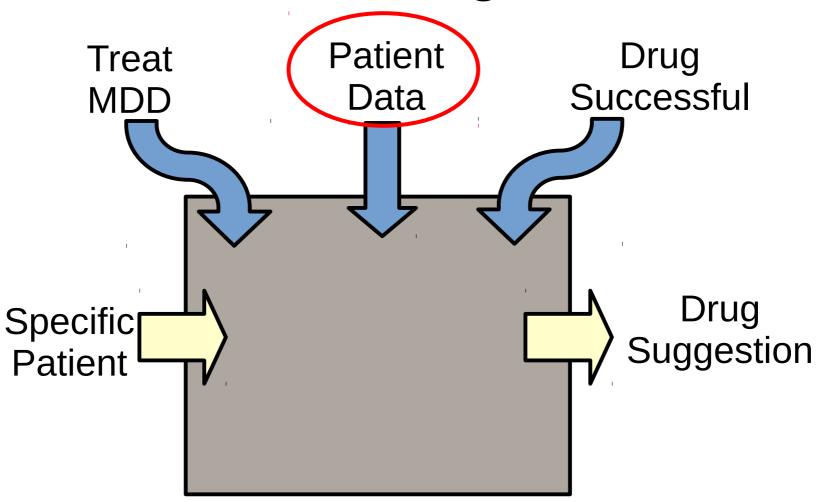
Current Approaches

Most current studies ask narrow questions and require specialized data. Examples:

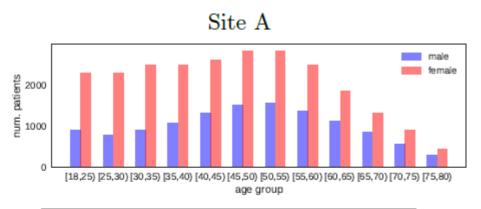
- Large-scale clinical trial (STAR*D, COMED, iSPOT-D) analyses to decide between certain features or pairs of drugs (e.g. Chekroud et al., Joyce et al., Lavretsky et al.)
- MRIs and biomarkers to determine subtype and treatment choice (e.g. Liston et al., Craighead et al., Breitenstein et al.)

Our goal: We want to be able to recommend all common drugs, for patients with any prior treatment history.

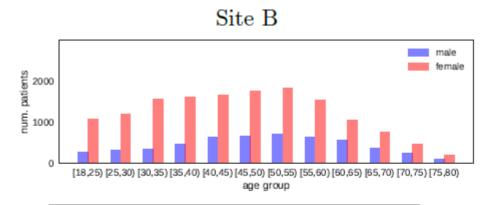
Formalizing...



Cohort Statistics



	female	male	total	frac.
Asian	603	222	825	0.022
Black	959	412	1371	0.037
Hispanic	1006	402	1408	0.038
Other	2153	885	3038	0.082
White	20158	10415	30575	0.822
total	24879	12336	37217	
frac.	0.668	0.331		



	female	male	total	frac.
Asian Black Hispanic Other White	233 1604 2264 1026 9662	56 419 657 386 3888	289 2023 2921 1413 13551	0.014 0.100 0.145 0.070 0.671
total frac.	14789 0.732	5406 0.268	20197	

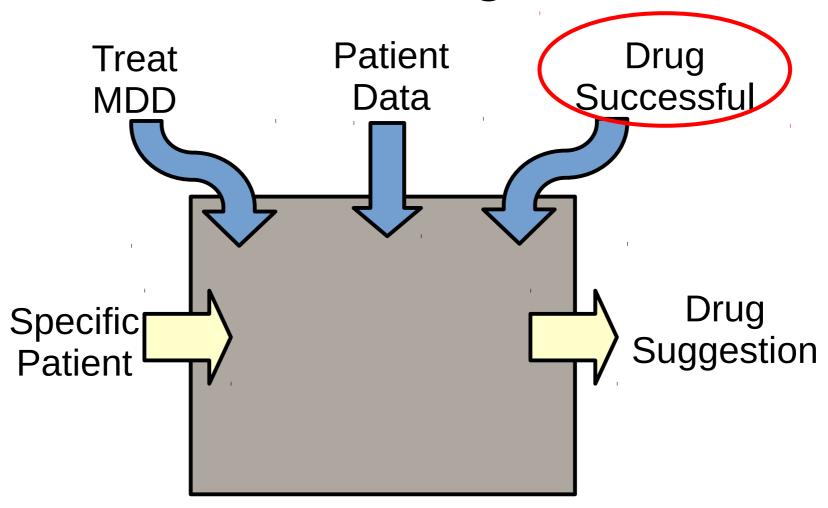
Featurization

- Of the 22,000 unique codes (ICDs, CPTs, RXNORM) in the EHR, retain 7,291 which occur in at least 1,000 distinct patients.
- Focus on 10 primary drugs that are prescribed in at least 1000 patients

nortriptyline amitriptyline bupropion fluoxetine sertraline paroxetine venlafaxine

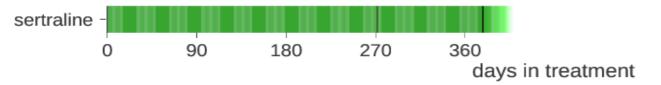
mirtazapine citalopram escitalopram

Formalizing...

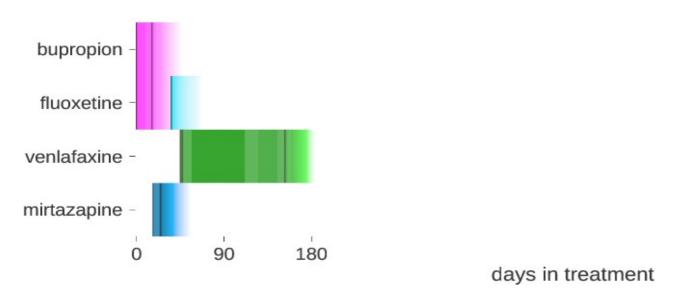


Performance Metric

"Instant Success" Patient

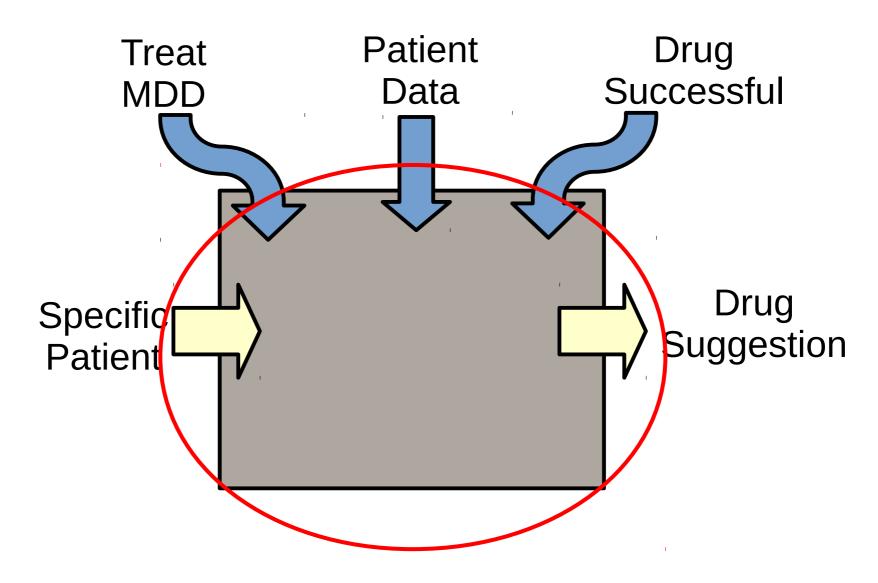


"Eventual Success" Patient



Success: require same primary over 90 days, with a visit frequency of at least every 13 months.

Formalizing...



Choices for Classifiers

In high dimensions, standard predictors (logistic regression, random forests, neural nets...) often have a tension between sparsity and interpretability.

- "Hip fracture" could code for "elderly"
- "Pregnant" could code for "female"

Thus, we choose to apply topic models for dimensionality reduction, and then predict based on the doc-topic probabilities.

Topic Models

Topics are distributions over words (codes)

$$\phi_k \sim Dir(\alpha_v)$$

Patients are distributions over topics

$$\pi_d \sim Dir(\alpha_p)$$
 $w_d^i \sim Dir(\pi_d \phi)$

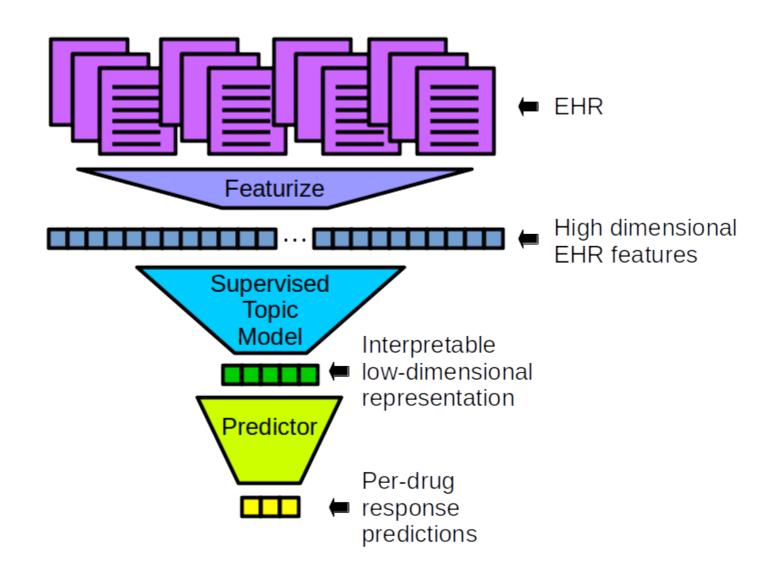
Example topic

```
1.0000
         29650:bipolar affective disorder, depres
0.9999
          2967:bipolar_affective_disorder,_unspec
0.9999
         29570:schizo-affective type schizophreni
0.9999
         29660: bipolar affective disorder, mixed,
0.9998
        c90870:electroconvulsive_therapy_(include
0.9998
        c00104:anesthesia for electroconvulsive t
0.9997
         29560:residual schizophrenia, unspecifie
0.9996
        p9427:other_electroshock_therapy
0.9993
        d00061:lithium
0.9993
         29653: bipolar affective disorder, depres
0.9985
         29651:bipolar_affective_disorder,_depres
0.9985
        d04825:aripiprazole
```

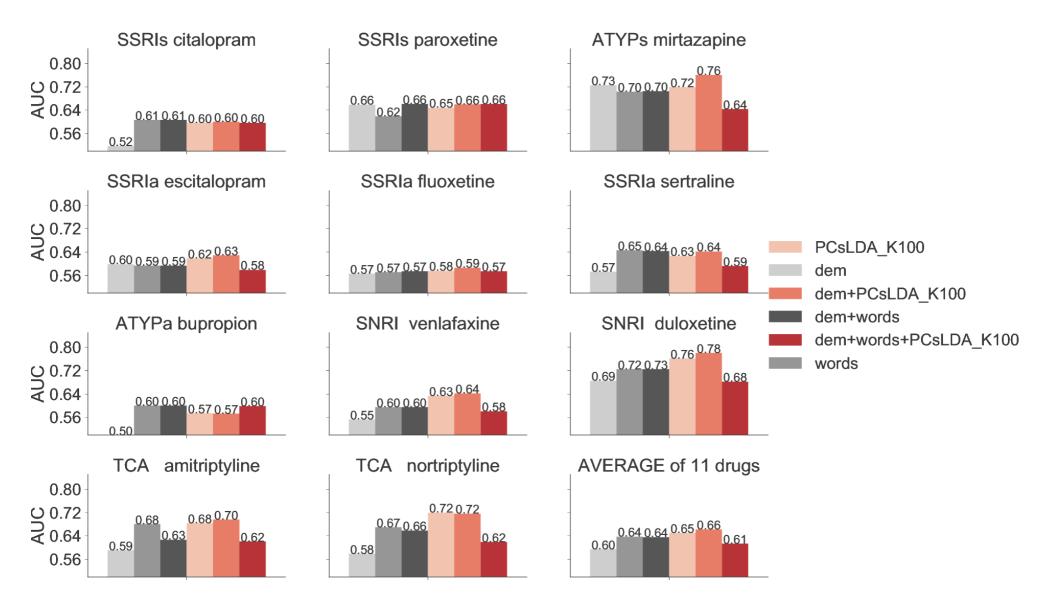


Patient has bipolar disorder

Pipeline



Drug-by-Drug Performance



Interpretation: Bupropion

BPsLDA +7.7

- 0.60 nortriptyline
- 0.27 nonspecific abnormal findings
- 0.21 other specified local infection
- 0.20 embrionic cyst of the fallopian tube
- 0.18 application of the intervertebrea...
- 0.16 other malignant neoplasm...
- 0.15 amoxicilllin/clarithromycin
- 0.15 need for prophylactic vaccine
- 0.15 observation or inpatient visit...

Gibbs -0.6

- 1.0000 bipolar, depressive
- 0.9999 bipolar, unspecified
- 0.9999 schizo-affective schizophrenia
- 0.9999 bipolar, mixed
- 0.9998 electroconvulsive therapy
- 0.9998 anesthesia for ECT
- 0.9997 residual schizophrenia
- 0.9996 other electroshock therapy
- 0.9993 lithium

PCLDA +3.8

- 0.99 migraine, unspecified, without...
- 0.99 other malaise and fatigue
- 0.99 common migraine...
- 0.99 sumatriptan
- 0.99 asa/butalbital/caffeine
- 0.99 zolmitriptan
- 0.99 migraine, unspecified, with..,
- 0.99 classical migraine, without...
- 0.99 classical migraine, with...

Interpretation: Bupropiom

BPsLDA -15.8

- 0.39 visual field defect, unspecified
- 0.39 citalopram
- 0.36 microdissection
- 0.35 need for prophylactic vaccine
- 0.31 pet imaging regional or wide area
- 0.29 visual discomfort
- 0.29 accident poison by heroin
- 0.29 personal history of alcohol
- 0.27 other specified intestinal disorder

Gibbs -2.7

- 0.9997 respiratory_failure
- 0.9995 cystic fibrosis...

PCI DA -26.4

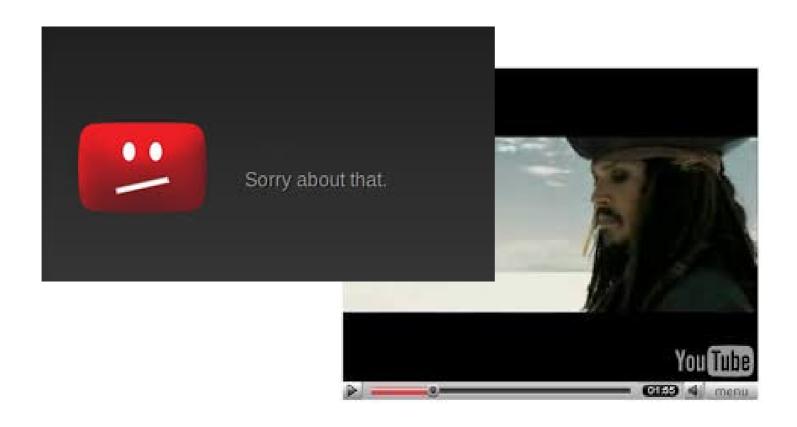
- 1.00 semen analysis, complete...
- 1.00 male infertility, unspecified
- 1.00 lipoprotein, direct measurement
- 0.99 sperm isolation, simple...
- 0.99 tissue culture for non-neoplasm...
- 0.99 conditions due to anomaly...
- 0.99 vasectomy, unilateral or...
- 0.99 arthrocentesis
- 0.99 scrotal varices

Currently working on how to present these results in a helpful way to practicing clinicians

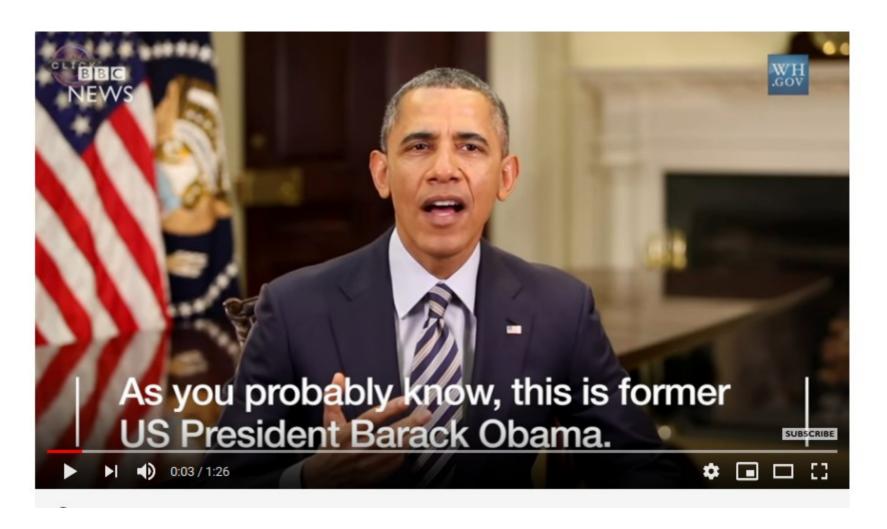
- 0.9964 continuous mech ventilation
- 0.9963 intubation endotracheal

Your Turn: Detecting Copyright Violations on Youtube...

(~300 hours of new video per minute)

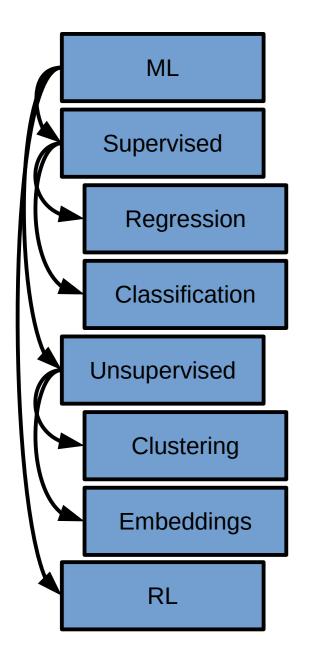


Your Turn: Detecting Fake Videos

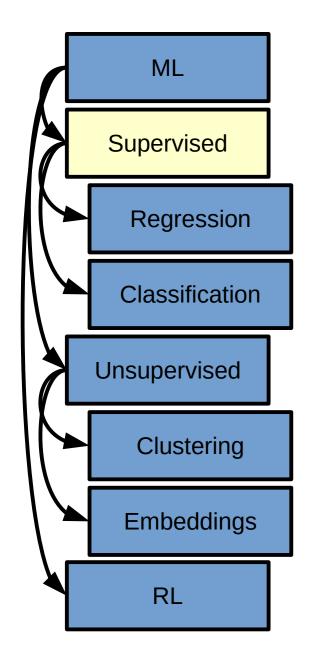


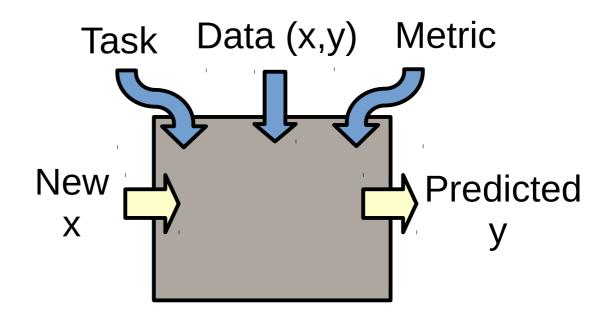
Taxonomy of Problems

Machine Learning Taxonomy

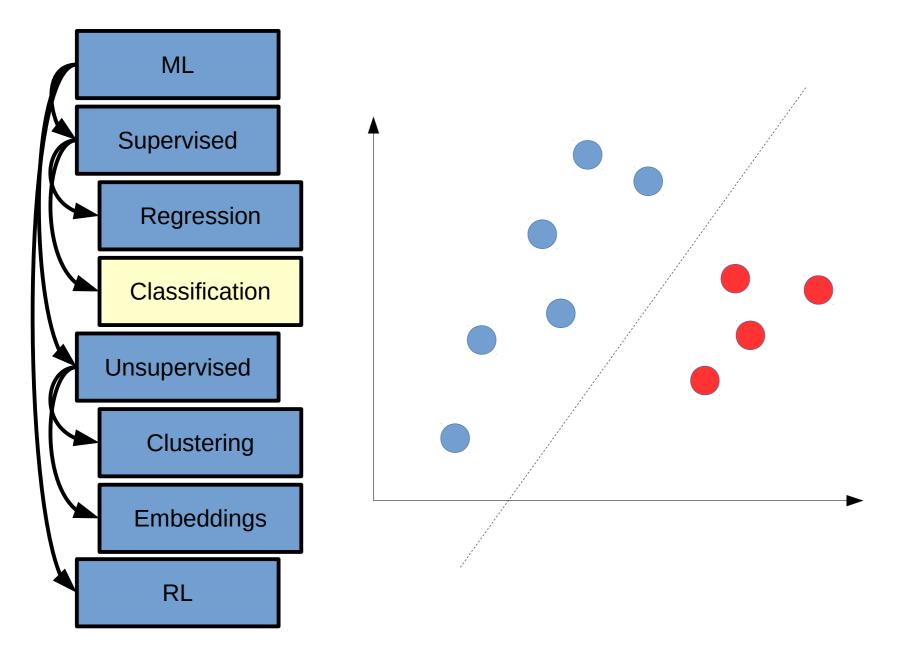


Machine Learning Taxonomy

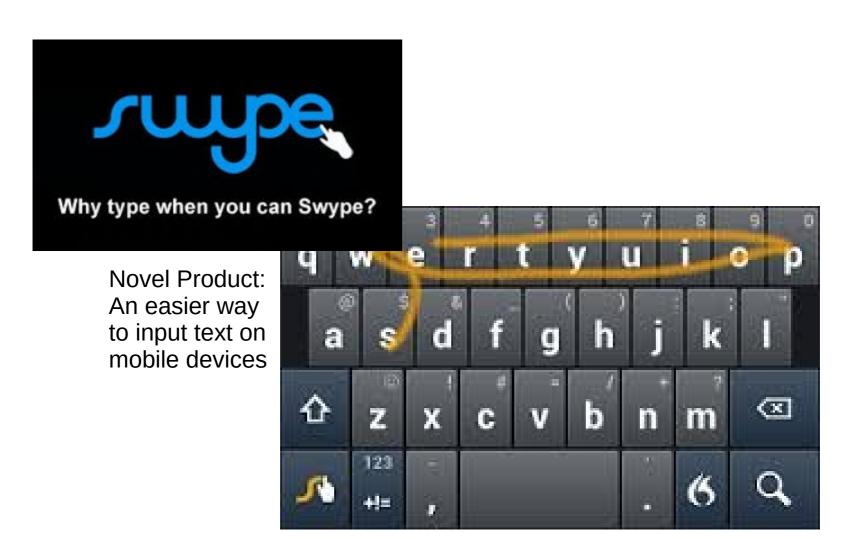




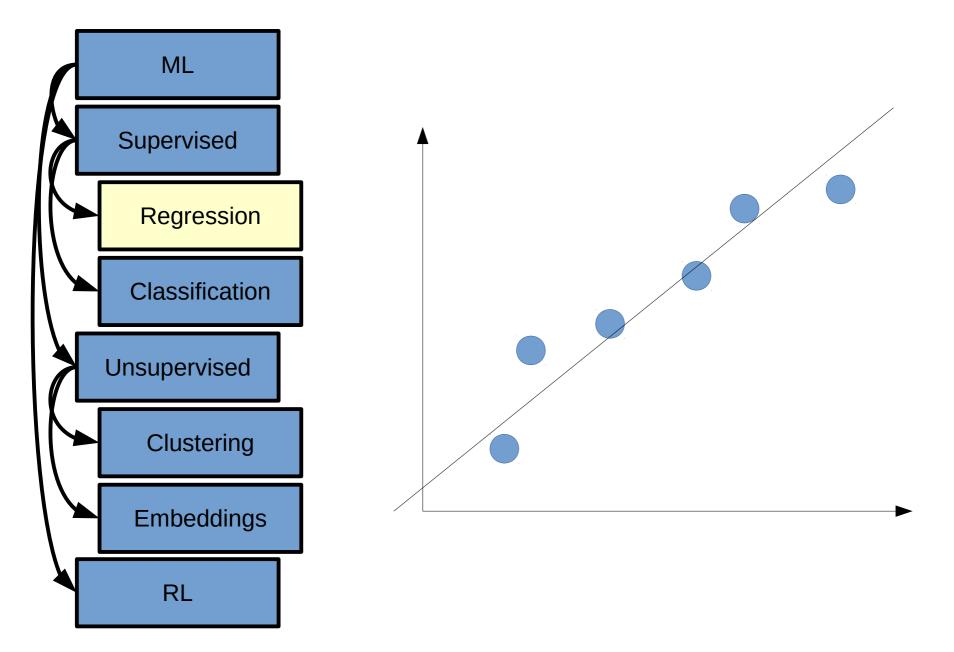
Terminology: Classification



Example: Swype



Terminology: Regression

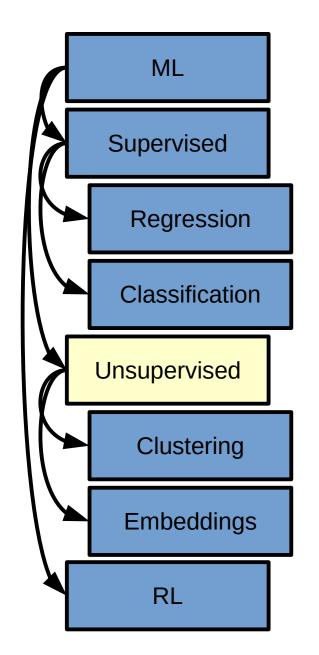


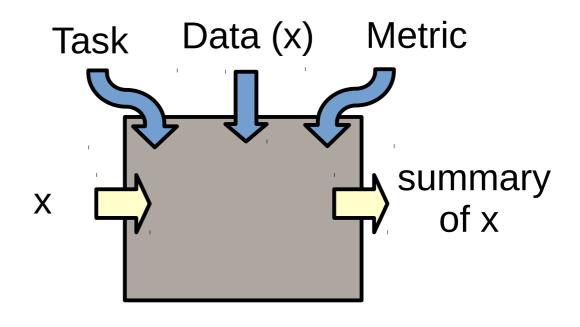
Example: Virtu



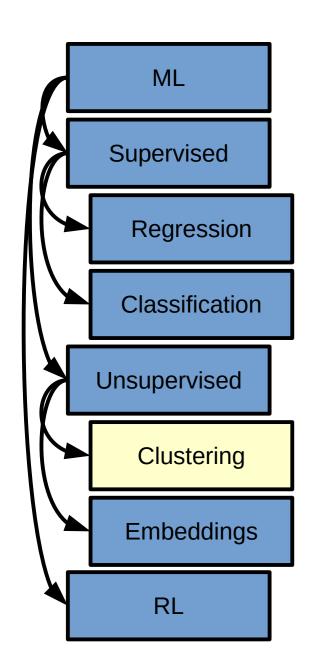
Core technology: Choosing what to trade, and when

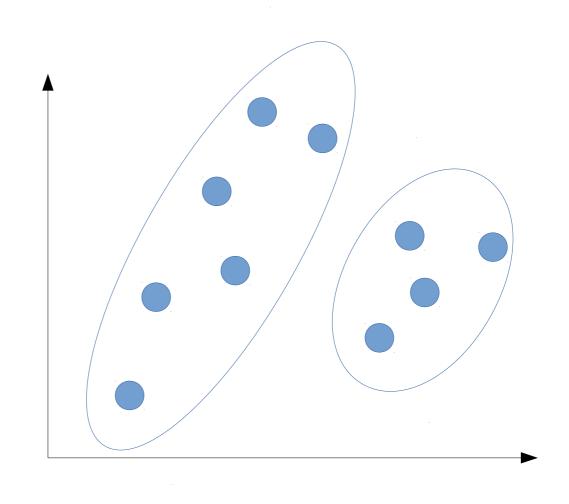
Machine Learning Taxonomy





Terminology: Clustering





Example: News

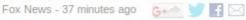
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An intensive manhunt was underway Monday for three inmates who pulled a "Shawshank"-style escape through a hole in their California jail cell -- and, who may have ties to notorious Vietnamese street gangs and Iran.

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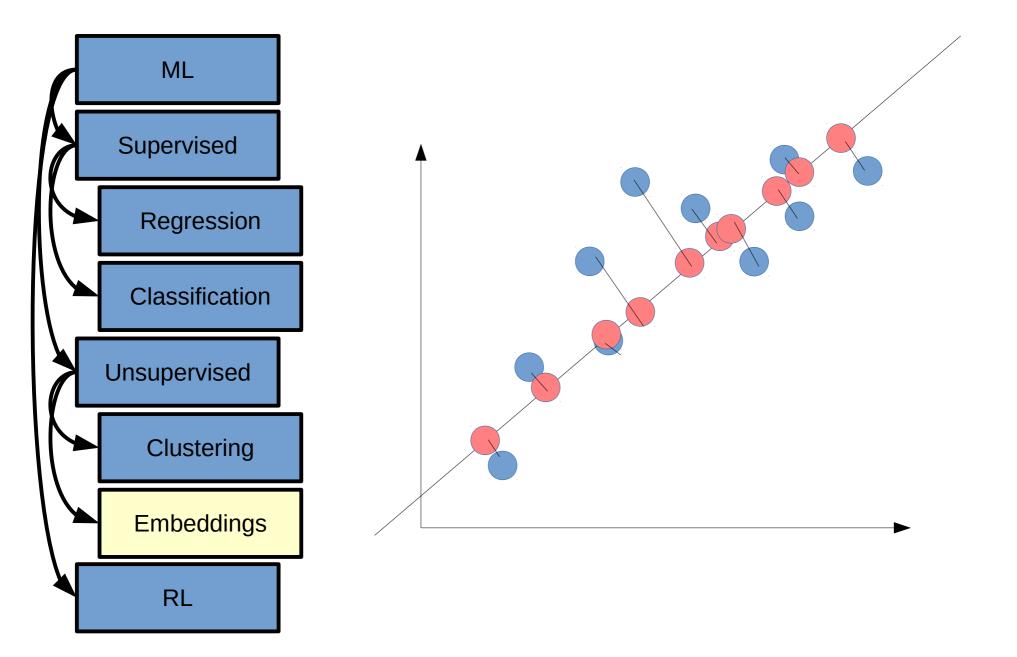




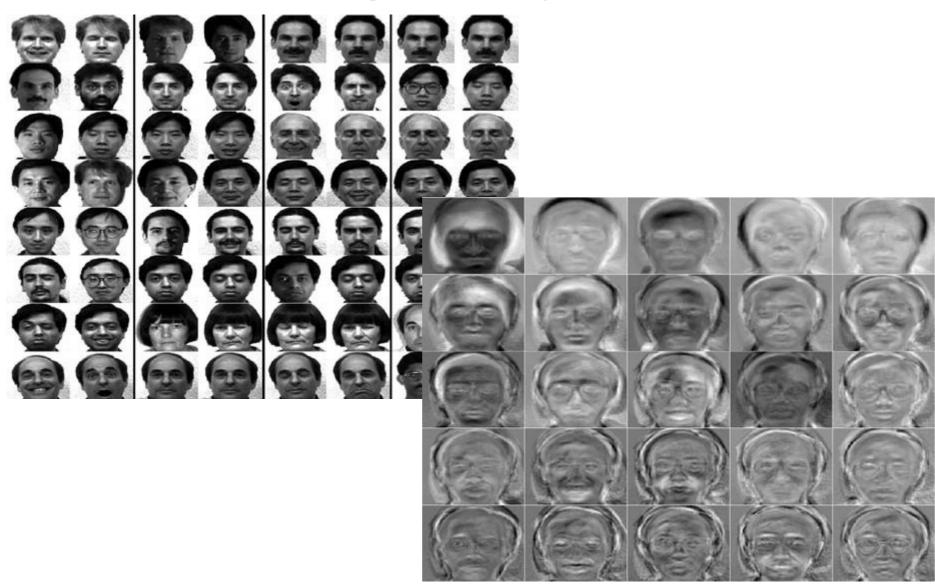




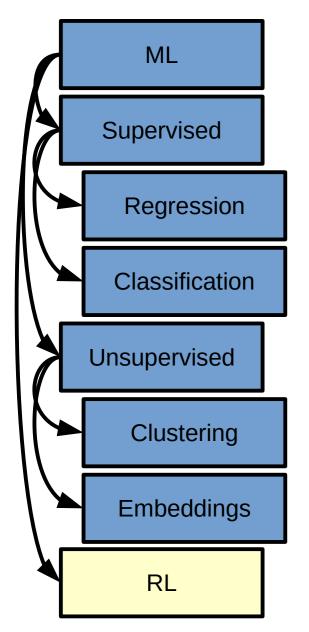
Terminology: Embedding

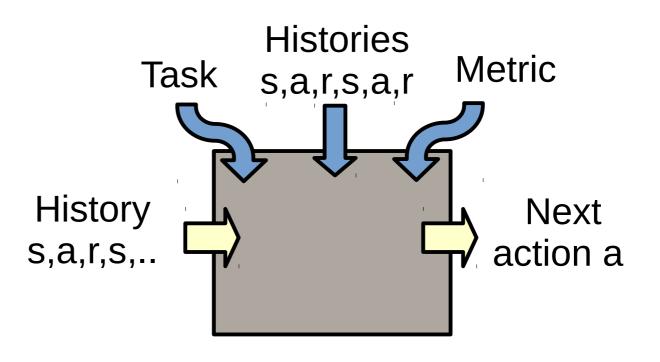


Example: Eigenfaces

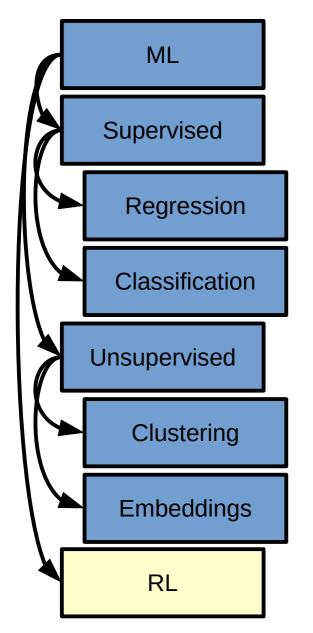


Terminology: Reinforcement Learning



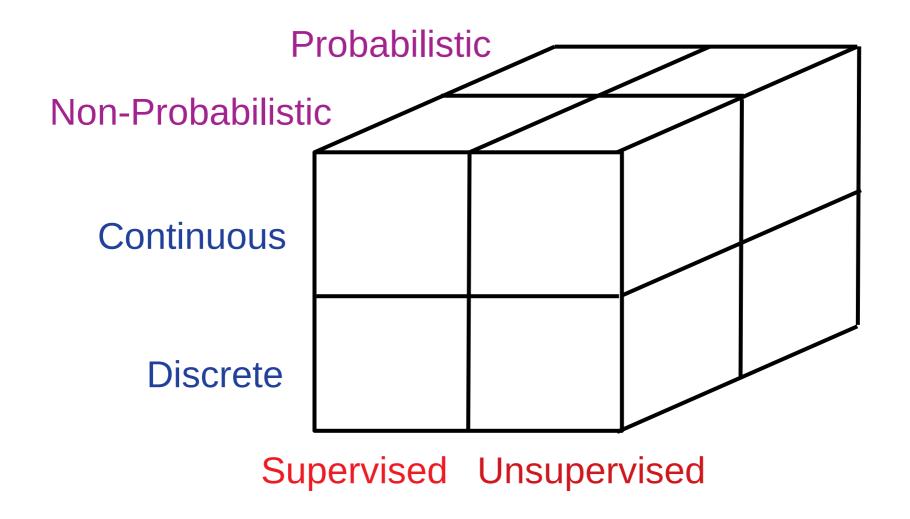


Terminology: Reinforcement Learning





Peter Kormushev, Imperial College



+ Graphical models, reinforcement learning (some overlap with CS182)

Course Structure

Structure of the Course

- Start with some basic regression problems.
- Discuss model selection and evaluation.
- Continue with the "cube"++
 - Supervised: regression, classification
 - Unsupervised: classification, embeddings
 - Reinforcement Learning

Lectures will include real world stories and concept exercises; we'll also have an embedded ethics component

Logistics

- Book: Bishop; also v0.0 notes at https://github.com/harvard-ml-courses/cs181textbook
- Homeworks: smaller exercises, toy problems
 - MUST be in LaTeX
 - Distributed over Github
- Practicals: real data, compete in teams
- Sections: math review, then flipped classroom
- Piazza: Clarification and Content Tags
- Canvas: Announcements, Gradebook
- Staff organization: Instructor (content), Course manager (logistics), Head TF, TFs

Details on the website

https://harvard-ml-courses.github.io/cs181-web

Read the syllabus!

FAQs

- Can I sit in/audit/take the course? [Yes!]
- Can I do simultaneous enrollment? [Maybe]
 - "I will not take advantage of any 1-1 compensatory instruction. I further understand that I, and only I, am solely responsible for anything from lecture, including announcements that are may not be posted anywhere else."
- Do I need all the prereqs? [Maybe]
 - What you need: programming, statistics, calculus, and linear algebra
 - check out https://mml-book.com/

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Problem 1.1:
Take the gradient with respect to the vector of weights *w*

. . .

Let's get technical! Nonparametric Regression