Lecture 1 What is Machine Learning?

EE4563/ EL9123: INTRODUCTION TO MACHINE LEARNING

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Learning Objectives

- ☐ Provide examples of machine learning used today
- ☐ Given a new problem, qualitatively describe how machine learning can be used
 - Formulate a potential machine learning task
 - Identify the data needed for the task
 - Identify objectives
- □Classify a machine learning task:
 - Supervised vs. unsupervised, regression vs. classification
- ☐ For supervised learning, identify the predictors and target variables
- ☐ Determine the role of expert knowledge in the task vs. data-driven learning





Outline

- What is Machine Learning?
- ☐ Types of machine learning algorithms
 - Classification
 - Regression
 - Unsupervised learning
- ■Why the hype today?
- ■Some slides from:
 - A. Zisserman, "Machine Learning Introduction"
 - Alpaydin, "Introduction to Machine Learning"





What is Machine Learning?

☐ Learn to improve algorithms from data.

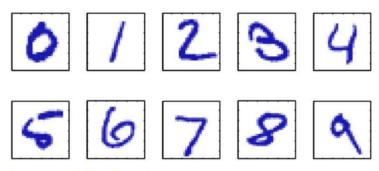
■Why?

- Human expertise does not exist (navigating on Mars),
- Humans are unable to explain their expertise (speech recognition)
- Solution changes in time (routing on a computer network)
- Solution needs to be adapted to particular cases (user biometrics)





Example 1: Digit Recognition



Images are 28 x 28 pixels

- ☐ Recognize a digit from the image
- Want a function $f(x) \in \{0,1,...,9\}$, x is a 28 x 28 matrix
 - Takes input as an image
 - Returns the estimated digit



"Expert-Driven" Approach

- ☐ Use your knowledge about digits
- ☐ Construct simple rules
- ■Example: It is a digit 7 if:
 - There is a single horizontal line
 - There is a single vertical line
 - The vertical line is above the horizontal line
- ☐ Very difficult to make work in practice
 - Lacks robustness
- ☐ Cannot easily code our knowledge
 - Hard to translate to simple mathematical formula





















Images are 28 x 28 pixels

```
def detect_vert_edge(image):
    ...
def detect_horiz_edge(image):
    ...
def is_above(image):
    ...

def classify(image):
    ...
    pos_v = detect_vert_edge(image)
    pos_h = detect_horiz_edge(image)
    ...
    if is_above(pos_v, pos_h) and ...:
        digit = 7
    ...
```

return digit





ML Approach: Supervised Learning

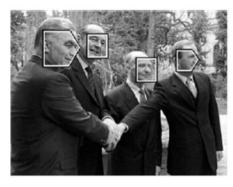
- ☐ Learn the function from data
 - No manual coding!
- ☐ Start with training data
 - ∘ Ex: 6000 examples of each digit
- \square Learn a classifier f(x)
 - Should match label well on training data
 - Generally fit parameters in the function
- \square Given new data x use function to guess digit
- □<u>Current systems</u> get <0.21% errors
 - (as of 1/20/2018)
- ☐ First commercial application:
 - Used by USPS for recognizing zip codes on letters

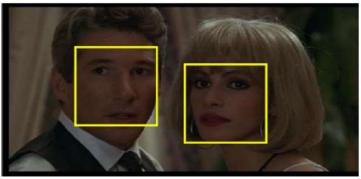
Training examples

Each sample must be labeled by hand who knows truth



Example 2: Face Detection





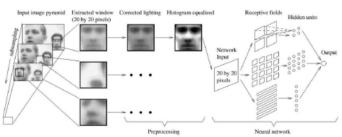
- ☐ Also a supervised learning problem
- ☐ For each image region, determine if
 - Face or non-face



Training Data

- ☐ Typical early face recognition datasets:
- **□**5000 faces
 - All near frontal
 - Vary age, race, gender, lighting
- ☐ 10⁸ non faces
- ☐ Faces are normalized (scale, translation)
- "functions" that work well may be very complex
- ☐ Many more datasets are available now:
 - See http://www.face-rec.org/databases/
 - You can use this for your project!



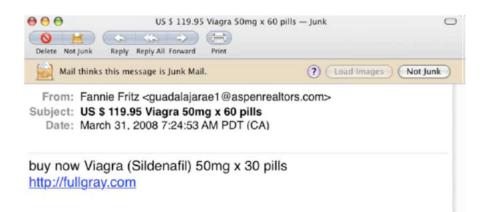


Rowley, Baluja and Kanade, 1998





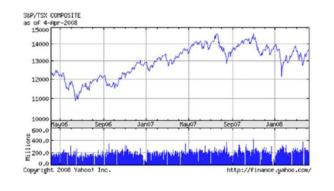
Example 3: Spam Detection



- □Classification problem:
 - Is email junk or not junk?
- ☐ For ML, must represent email numerically
 - Common model: bag of words
 - Enumerate all words, i = 1, ..., N
 - Represent email via word count x_i = num instances of word i
- ☐ Challenge:
 - Very high-dimensional vector
 - System must continue to adapt (keep up with spammers)



Example 4: Stock Price Prediction



- □Can you predict the price of a stock?
- ■What variables would you use?
- ■What is a non-machine learning approach?



Machine Learning in Many Fields

- ☐ Retail: Market basket analysis, Customer relationship management (CRM)
- ☐ Finance: Credit scoring, fraud detection
- ☐ Manufacturing: Control, robotics, troubleshooting
- ☐ Medicine: Medical diagnosis
- ☐ Telecommunications: Spam filters, intrusion detection
- ☐ Bioinformatics: Motifs, alignment
- ■Web mining: Search engines
- **...**





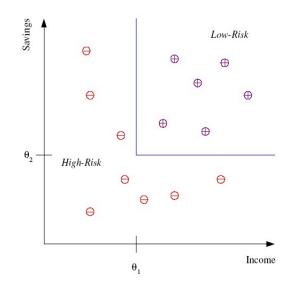
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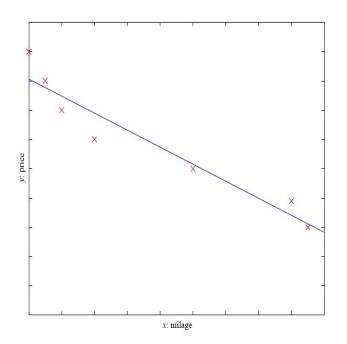
Classification

- ☐ Example: Credit score
- □ Determine if customer is high-risk or low-risk
- Select some features:
 - Example: income & savings
 - Represent as a vector $x = (x_1, x_2)$
- ☐ Learn a function from features to target
 - Use past training data
 - Need to get this data
- ☐ The function on the right is an example of a decision tree.



Regression

- \square Target variable y is continuous-valued
- ■Example:
 - Predict y = price of car
 - From x = mileage, size, horsepower, ..
 - Can use multiple predictors
- ☐ Assume some form of the mapping
 - Ex. Linear: $y = \beta_0 + \beta_1 x$
 - \circ Find parameters β_0 , β_1 from data



Regression Example

Machine Learning Repository

Center for Machine Learning and Intelligent Systems

Diabetes Data Set

Download: Data Folder, Data Set Description

File Names and format:

- (1) Date in MM-DD-YYYY format
- (2) Time in XX:YY format
- (3) Code
- (4) Value

The Code field is deciphered as follows:

- 33 = Regular insulin dose
- 34 = NPH insulin dose
- 35 = UltraLente insulin dose
- 48 = Unspecified blood glucose measurement
- 57 = Unspecified blood glucose measurement 58 = Pre-breakfast blood glucose measurement
- 59 = Post-breakfast blood glucose measurement
- 60 = Pre-lunch blood glucose measurement
- 61 = Post-lunch blood glucose measurement
- 62 = Pre-supper blood glucose measurement
- 63 = Post-supper blood glucose measurement
- 64 = Pre-snack blood glucose measurement
- 65 = Hypoglycemic symptoms
- 66 = Typical meal ingestion
- 67 = More-than-usual meal ingestion
- 68 = Less-than-usual meal ingestion
- 69 = Typical exercise activity
- 70 = More-than-usual exercise activity
- 71 = Less-than-usual exercise activity

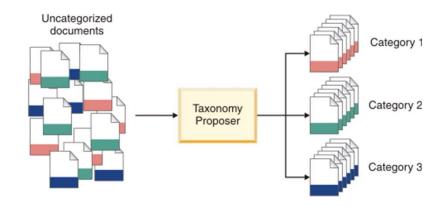
- ☐ Predict blood glucose level
- ☐ Many possible predictors:
 - Recent past levels
 - Insulin dose
 - Time of last meal
 - ٥ ...
- ☐ Check out data in:

https://archive.ics.uci.edu/ml/datasets/Diabetes



Unsupervised Learning

- Learning "what normally happens"
- ■No output
- □ Clustering: Grouping similar instances
- Example applications
 - Customer segmentation
 - Image compression: Color quantization
 - Bioinformatics: Learning motifs



Example: Document classification

http://www.ibm.com/support/knowledgecenter

/SSBRAM_8.7.0/com.ibm.classify.ccenter.doc/

c_WBG_Taxonomy_Proposer.htm





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What ML is Doing Today?

- ☐ Autonomous driving
- ■Jeopardy
- ☐ Very difficult games: Alpha Go
- Machine translation
- ☐ Many, many others...







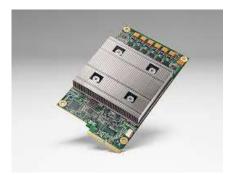




Why Now?

- ☐ Machine learning is an old field
 - Much of the pioneering statistical work dates to the 1950s
- ■So what is new now?
- ☐Big Data:
 - Massive storage. Large data centers
 - Massive connectivity
 - Sources of data from Internet and elsewhere
- ☐ Computational advances
 - Distributed machines, clusters
 - GPUs and hardware





Google Tensor Processing Unit (TPU)





Top Journals

- □ Journal of Machine Learning Research <u>www.jmlr.org</u>
- ☐ Machine Learning
- Neural Computation
- Neural Networks
- □ IEEE Trans on Neural Networks and Learning Systems
- □ IEEE Trans on Pattern Analysis and Machine Intelligence
- □ Journals on Statistics/Data Mining/Signal Processing/Natural Language Processing/Bioinformatics/...





Top Conferences

- □ International Conference on Machine Learning (ICML)
- ☐ European Conference on Machine Learning (ECML)
- Neural Information Processing Systems (NIPS)
- ☐ Uncertainty in Artificial Intelligence (UAI)
- □ Computational Learning Theory (COLT)
- ☐ International Conference on Artificial Neural Networks (ICANN)
- ☐ International Conference on AI & Statistics (AISTATS)
- ☐ Knowledge Discovery and Data Mining (KDD)
- □ International Conference on Computer Vision and Pattern Recognition (CVPR)
- ☐ International Conference on Computer Vision (ICCV)
- ☐ European Conference on Computer Vision (ECCV)





Exercise

- ☐ Break into small groups
- ☐ Take a field that interests you:
 - Ex. Driving a car, understanding social networks, finding a good date, recommend a movie to watch, ...
- □ Identify a specific task that can be done with machine learning
 - What is the objective of the task?
 - What is the data you need?
 - What type of ML problem is this? Classification, regression, ...
 - How would your approach compare to an expert-driven method?



