## Machine Learning: Course Overview

CS 760@UW-Madison







- typically the class was limited to 30
- we've allowed ~100 to register
- the waiting list full
- unfortunately, many on the waiting list will not be able to enroll
- but CS760 will be offered in the next semester!



#### Teaching team



- Yingyu Liang email: yliang@cs.wisc.edu office hours: Thu 4-5pm office: BBCollaborate Ultra on Canvas
- TAs: Xufeng Cai, Diwanshu Jain email: xcui74@wisc.edu, djain23@wisc.edu office hours: Thu 8:30-9:30pm (Cai), TBD (Jain) office: BBCollaborate Ultra on Canvas
- Graders: Shah Nisarg, Sharma Vedang
- More Information on course website: <u>http://pages.cs.wisc.edu/~yliang/cs760\_spring21/</u>

#### Attending lecture synchronously



If you can join CS 760 M/W/F at 2:30pm CT:

- (Optional) download slides
  - Course website  $\rightarrow$  Schedule  $\rightarrow$  [Slides]
- Sign in to Canvas
- Join BBCollaborate Ultra session "Lecture"
- Lecture block 1:
  - Watch video
    - Can stay in BBCollaborate Ultra
    - If that fails, stream in Kaltura Gallery
    - If that fails, download from Kaltura Gallery
  - Q&A: Ask questions by "raise hands", discuss
  - Quizzes: Short ungraded quizzes to check concepts
- Lecture block 2 (same as 1)
- Lecture block 3 (same as 1)
- After class, check or post questions on Piazza

#### Attending lecture asynchronously



If you cannot join CS 760 M/W/F at 2:30pm CT:

- (Optional) download slides
- Sign in to Canvas
- Watch 3 lecture videos in Kaltura Gallery
- After class, watch BBCollaborate Ultra recordings to listen to Q&A/Quizzes
  - BBCollaborate Ultra  $\rightarrow$  menu  $\rightarrow$  Recordings
  - Quizzes: download from course website, next to slides
- Check or post questions on Piazza

#### Where to find content



- Canvas private materials that should not be shared
  - Videos
  - Assignments
  - Grades
- Course website public materials
  - Slides
  - Schedule
  - Policies
- Piazza
  - Discussion, questions
  - Announcements



- we'll have ~30 lectures in total, just like a standard TR class
- can push the lectures forward (finish early, leave time for review)
- see the schedule on the course website



- a variety of learning settings: supervised learning, unsupervised learning, reinforcement learning, etc.
- a broad toolbox of machine-learning methods: decision trees, nearest neighbor, neural nets, Bayesian networks, SVMs, etc.
- some underlying theory: bias-variance tradeoff, PAC learning, mistake-bound theory, etc.
- experimental methodology for evaluating learning systems: cross validation, ROC and PR curves, hypothesis testing, etc.



- 1. Understand what a learning system should do
- 2. Understand how (and how well) existing systems work

Emphasize on understanding, laying the foundation for future research in machine learning.

If you just want to **use** machine learning, but do not plan to do **research** in machine learning, better to take:

- CS540
- STAT 451
- ECE/CS/ME 532

#### Course requirements



- 7-8 homework assignments: 60%
  - programming
  - computational experiments (e.g., measure the effect of varying parameter *x* in algorithm *y*)
  - some written exercises
  - post on Canvas; submit your solutions on Canvas
  - will drop the lowest scored one in calculating the final score
- Midterm Exam: 20%
- Final Exam: 20%

#### Expected background



- CS 540 (Intro to Artificial Intelligence) or equivalent
- good programming skills
- probability/statistics
- linear algebra
- calculus, including partial derivatives

#### **Programming languages**



for the programming assignments, you can use

C C++ Java Perl Python R Matlab

- Highly suggest: Python
- programs must be callable from the command line and must run on the CS lab machines (this is where they will be tested during grading!)

#### **Course readings**



#### Optional but recommend <u>one</u> of the following books

- Pattern Recognition and Machine Learning. C. Bishop. Springer, 2011.
- *Machine Learning: A Probabilistic Perspective*. K. Murphy. MIT Press, 2012.
- Understanding Machine Learning: From Theory to Algorithms. S. Shalev-Shwartz, S. Ben-David. Cambridge University press, 2014.



#### Course readings



- the books can be found online or at Wendt Commons Library
- additional readings will come from online articles, surveys, and chapters
- will be posted on course website

# Machine Learning Examples



#### What is machine learning?



 "A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P, if its performance at tasks in T as measured by P, improves with experience E."

----- Machine Learning, Tom Mitchell, 1997



#### What is machine learning?



- the study of algorithms that improve their performance P at some task T with experience E
- to have a well-defined learning task, we must specify: < *P*, *T*, *E* >

#### ML example: image classification





indoor

outdoor

#### ML example: image classification



- T: given new images, classify as indoor vs. outdoor
- P: minimize misclassification costs
- *E* : given images with indoor/outdoor labels

#### ML example: spam filtering



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#### ML example: spam filtering



- T: given new mail message, classify as spam vs. other
- P: minimize misclassification costs
- *E* : previously classified (filed) messages

#### ML example: predictive text input





#### ML example: predictive text input

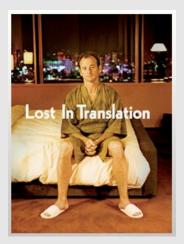


- *T* : given (partially) typed word, predict the word the user intended to type
- *P* : minimize misclassifications
- E: words previously typed by the user
   (+ lexicon of common words + knowledge of keyboard layout)

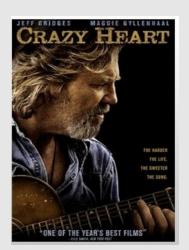
domain knowledge

#### ML example: Netflix Prize

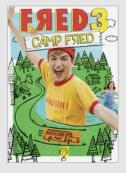
















Our best guess for Mark: ★★★★☆☆

#### ML example: Netflix Prize



- *T* : given a user/movie pair, predict the user's rating (1-5 stars) of the movie
- *P* : minimize difference between predicted and actual rating
- *E* : histories of previously rated movies (user/movie/rating triples)

### ML example: autonomous helicopter





video of Stanford University autonomous helicopter from http://heli.stanford.edu/

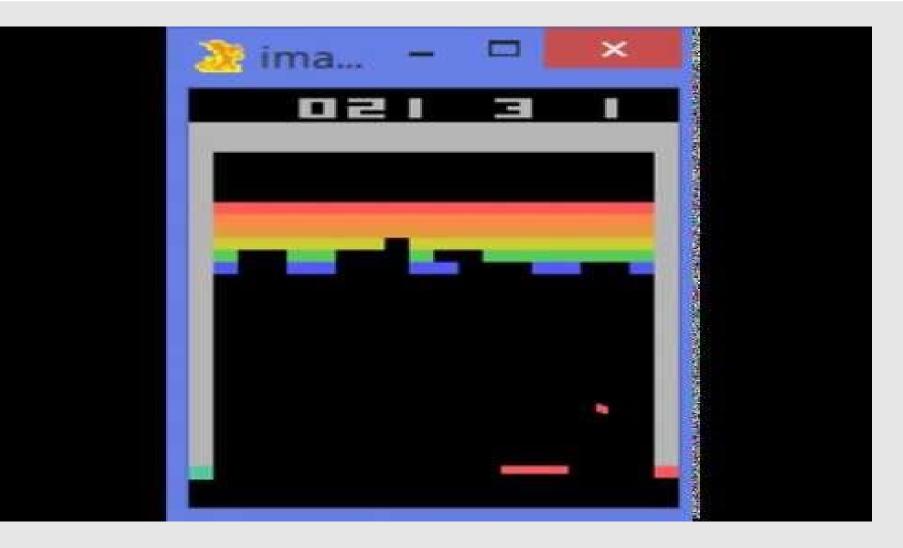
## ML example: autonomous helicopter



- *T* : given a measurement of the helicopter's current state (orientation sensor, GPS, cameras), select an adjustment of the controls
- *P* : maximize reward (intended trajectory + penalty function)
- *E* : state, action and reward triples from previous demonstration flights

#### ML example: Atari Breakout





Google DeepMind's Deep Q-learning playing Atari Breakout

From the paper "Playing Atari with Deep Reinforcement Learning", by Volodymyr Mnih, Koray Kavukcuoglu, David Silver, Alex Graves, Ioannis Antonoglou, Daan Wierstra, Martin Riedmiller

#### ML example: AlphaGo







# Assignments



### **Reading assignment**



• read

• article by Jordan and Mitchell on course website

• course website:

http://pages.cs.wisc.edu/~yliang/cs760\_spring21/

#### HW1: Background test



- posted on Canvas
- will set up how to submit the solutions on Canvas
- If you find many questions intimidating, we suggest you drop the course and take it again in the future when you are more prepared.

#### Background test



#### Topics

- linear algebra
- calculus
- probability
- big-O notations for algorithm analysis
- basic programming skills

#### Test example



Consider the matrix X and the vectors  $\mathbf{y}$  and  $\mathbf{z}$  below:

$$X = \begin{pmatrix} 9 & 8 \\ 7 & 6 \end{pmatrix} \qquad \mathbf{y} = \begin{pmatrix} 9 \\ 8 \end{pmatrix} \qquad \mathbf{z} = \begin{pmatrix} 7 \\ 6 \end{pmatrix}$$

1. Is X invertible? If so, give the inverse, and if no, explain why not.

2. If  $y = \tan(z)x^{6z} - \ln(\frac{7x+z}{x^4})$ , what is the partial derivative of y with respect to x?

#### Test example



Match the distribution name to its probability density / mass function. Below,  $|\mathbf{x}| = k$ . (f)  $f(\boldsymbol{x};\boldsymbol{\Sigma},\boldsymbol{\mu}) = \frac{1}{\sqrt{(2\pi)^k \boldsymbol{\Sigma}}} \exp\left(-\frac{1}{2}(\boldsymbol{x}-\boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\boldsymbol{x}-\boldsymbol{\mu})\right)$ (g)  $f(x; n, \alpha) = {n \choose x} \alpha^x (1 - \alpha)^{n-x}$  for  $x \in \{0, \dots, n\}; 0$ otherwise (h)  $f(x; b, \mu) = \frac{1}{2b} \exp\left(-\frac{|x-\mu|}{b}\right)$ (a) Laplace (i)  $f(\boldsymbol{x}; n, \boldsymbol{\alpha}) = \frac{n!}{\prod_{i=1}^{k} \alpha_i^{x_i}} \prod_{i=1}^{k} \alpha_i^{x_i}$  for  $x_i \in \{0, ..., n\}$  and (b) Multinomial (c) Poisson  $\sum_{i=1}^{k} x_i = n; 0$  otherwise (d) Dirichlet (j)  $f(x; \alpha, \beta) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$  for  $x \in (0, +\infty)$ ; 0 oth-(e) Gamma erwise (k)  $f(\boldsymbol{x}; \boldsymbol{\alpha}) = \frac{\Gamma(\sum_{i=1}^{k} \alpha_i)}{\prod_{i=1}^{k} \Gamma(\alpha_i)} \prod_{i=1}^{k} x_i^{\alpha_i - 1}$  for  $x_i \in (0, 1)$  and  $\sum_{i=1}^{k} x_i = 1; 0$  otherwise (1)  $f(x;\lambda) = \lambda^x \frac{e^{-\lambda}}{x!}$  for all  $x \in Z^+$ ; 0 otherwise

#### Test example



Draw the regions corresponding to vectors  $\mathbf{x} \in \mathbb{R}^2$  with the following norms:

- 1.  $||\mathbf{x}||_1 \le 1$  (Recall that  $||\mathbf{x}||_1 = \sum_i |x_i|$ )
- 2.  $||\mathbf{x}||_2 \le 1$  (Recall that  $||\mathbf{x}||_2 = \sqrt{\sum_i x_i^2}$ )
- 3.  $||\mathbf{x}||_{\infty} \leq 1$  (Recall that  $||\mathbf{x}||_{\infty} = \max_{i} |x_{i}|$ )

#### Useful resources for background



Probability

• Lecture notes:

http://www.cs.cmu.edu/~aarti/Class/10701/recitation/prob\_review.pdf

Linear Algebra:

- Short video lectures by Prof. Zico Kolter: <u>http://www.cs.cmu.edu/~zkolter/course/linalg/outline.html</u>
- Handout associated with above video: <u>http://www.cs.cmu.edu/~zkolter/course/linalg/linalg\_notes.pdf</u>
- Book: Gilbert Strang. Linear Algebra and its Applications. HBJ Publishers.

Big-O notation:

- <u>http://www.stat.cmu.edu/~cshalizi/uADA/13/lectures/app-b.pdf</u>
- <u>http://www.cs.cmu.edu/~avrim/451f13/recitation/rec0828.pdf</u>