Held on the traditional, ancestral, and unceded territory of the Musqueam people.
Big Data Phenomenon

• We are collecting and storing data at an unprecedented rate.

• Examples:
  – News articles and blog posts.
  – YouTube, Facebook, and WWW.
  – Credit cards transactions and Amazon purchases.
  – Gene expression data and protein interaction assays.
  – Maps and satellite data.
  – Large hadron collider and surveying the sky.
  – Phone call records and speech recognition results.
  – Video game worlds and user actions.
Machine Learning

• What do you do with all this data?
  – Too much data to search through it manually.
• But there is valuable information in the data.
  – Can we use it for fun, profit, and/or the greater good?
• **Machine learning**: use computers to automatically detect patterns in data and make predictions or decisions.
• Most useful when:
  – Don’t have a human expert.
  – Humans can’t explain patterns.
  – Problem is too complicated.
Machine Learning vs. Statistics

• Machine learning (ML) is very similar to statistics.
  – A lot of topics overlap.

• But ML places more emphasis on:
  1. Computation and large datasets.
  2. Predictions rather than descriptions.
  4. Models that work across domains.

• The field is growing very fast:
  – 2018 NeurIPS Sold out in ~11 minutes.
    • 13000 registrations in 2020.
  – Influence of $$$ too.
Applications

- Spam filtering.
- Credit card fraud detection.
- Product recommendation.
- Motion capture.
- Machine translation.
- Speech recognition.
- Face detection.
- Object detection.
- Sports analytics.
- Cancer subtype discovery.
Applications

• Gene localization/functions/editing.
• Personal Assistants.
• Medical imaging.
• Self-driving cars.
• Scene completion.
• Image search and annotation.
• Artistic rendering.
• Physical simulations.
• Image colourization.
• Source separation
• Game-playing.
Next Topic: Course Registration
CPSC 340 and CPSC 440 (and 532M and 540)

- There are two “core” ML classes: CPSC 340 and CPSC 440.
  - You can take CPSC 340 for grad credit as CPSC 532M (though not this year).
  - You can take CPSC 440 for grad credit as CPSC 540.
  - Structured as one full-year course: 440 starts where 340 ends.

- We have also have CPSC 330 (focuses on using machine learning methods).
CPSC 340 and CPSC 440 (and 532M and 540)

• CPSC 340/532M (the other course):
  – Introductory course on data mining and ML.
  – Emphasis on applications and core ideas of ML.
    • And how to implement methods, not just how to use them.
  – Most useful techniques that you can apply to your research/work.

• CPSC 440/540 (this course):
  – Sequel course covering topics that do not appear in 340.
  – More “stuff that requires more time/background” than “advanced ML”.
    • Assumes strong background on fundamental ML concepts.
    • Assumes stronger math/CS background.
You should take CPSC 340/532M First!!!

• If you can only take one class, take the other class (CPSC 340/532M).
  – 340/532M covers the most useful methods and ideas.
  – If you take 440/540 first, you’ll miss half the story and a lot will seem random.
    • This is not an intro course and there is not a lot of review in 440/540.
      – So 440/540 is missing a lot important topics.
  – 440/540 is NOT an “advanced” version of 340/532M.
    • It just covers the methods that require more advanced math/CS background to use.

• It is much better to do CPSC 340/532M first:
  – Many people have taken 340/532M *after* 440/540 (not recommended).
  – A few people took 440/540 then 340/532M then *sat in on 440/540 again,*
    (REALLY not recommended).
  – In industry/research, BIG mistakes are usually related to the (340) fundamentals!
    • Not because of an “advanced” machine learning error.
I’m assume you already know the below “typical” topics (covered in 340):
- Calculus in matrix notation, including derivation of normal equations for least squares.
- IID assumption, complexity vs. generalization trade-off, ensemble methods, and cross-validation.
- Probabilistic classifiers, maximum likelihood, and MAP estimation.
- Radial basis functions, how to show a function is convex.
- Stochastic gradient descent, softmax loss, and L1-regularization.
- PCA and collaborative filtering.

You will get lost very quickly if you don’t know this material.
- You should already be able to write code implementing all of the above ideas.

CPSC 440/540 Course Outline:
- Density estimation, Bayesian methods, graphical models, mixtures and latent variables.
Prerequisites (A lot of Math and CS)

Examples of CS concepts you should know:
- writing/debugging complex programs, binary trees, hash functions, graphs, big-O, randomized algorithms, dynamic programming, NP-completeness.

Examples of math concepts you should know:
- matrix algebra, norms, gradients, random variables, expectations, minimizing quadratic functions, random vectors.
Auditing

• **Auditing 540**, an excellent option:
  – Pass/fail on transcript rather than grade.
  – Do 1 assignment or write a 2-page report on one technique from class or attend > 90% of classes.
  – But please do this officially:
    • [http://students.ubc.ca/enrolment/courses/academic-planning/audit](http://students.ubc.ca/enrolment/courses/academic-planning/audit)

• Auditing 440:
  – There are a limited number of “seats”.
  – If these are full, we won’t allow auditors.
  – If these are not full, see above for how we’ll deal with 540 auditors.
Next Topic: Lectures
Lectures

• Classes will be online until at least January 24th (link on Canvas, mailed to waiting list).
  – At that time, we may move to in-person instruction.

• All slides will be posted online (before lecture, and final version after).

• Please ask questions: you probably have similar questions to others.
  – I may deflect to the next lecture or Piazza for certain questions.

• Be warned that the course will move fast and cover a lot of topics:
  – Big ideas will be covered slowly and carefully.
  – But a bunch of other topics won’t be covered in a lot of detail.

• Isn’t it wrong to have only have shallow knowledge?
  – In this field, it’s better to know many methods than to know 5 in detail.
    • This is called the “no free lunch” theorem: different problems need different solutions.
    • If you know why something is important, and the core ideas, you can fill in details later.
Warning regarding teaching quality

• This is only the second time that CPSC 440 has been offered.
  – And I didn’t have as much time to prepare as I wanted.
    • It will still rely a lot on my old CPSC 540 material, which is not always “undergrad friendly”.
  – So the course isn’t as “put together” as you might like.
    • Won’t be as smooth as a course that has been offered many times.
      – Switching from PowerPoint to Beamer and back, assignments may not sync well with lectures, some course material will be underdeveloped, and so on.
  – I’m also not a teaching faculty member.
    • I run one of the typically-largest labs in CS (e.g., I have 6 PhD students and CS median is 1).
    • I try but may not be as available/good as some the full-time teachers.

• Don’t expect a high grade without a high effort.
  – We cover a lot of material and my assignments are LONG. This is not an “easy” class.

• If these things are going to bother you, it might be better to take this course later and/or take a different course this term.
Textbook and Other Optional Reading

• No textbook covers all course topics.
• The closest is Kevin Murphy’s “Machine Learning”.
  – But we’re using a very different order.

• For each lecture:
  – I’ll give relevant sections from this book.
  – I’ll give other related online material.
• There is a list of related courses on the webpage.
Textbook and Other Optional Reading

• Other good machine learning textbooks:
  – All of Statistics (Wasserman).
  – Elements of Statistical Learning (Hastie et al.).
  – Pattern Recognition and Machine Learning (Bishop).

• Good textbook covering needed mathematical background:
  – Mathematics for Machine Learning (Deisenroth, Faisal, Ong).
  – Available online.

• Good textbooks on specialized topics from this course:
  – Probabilistic Graphical Models (Koller and Friedman).
  – Deep Learning (Goodfellow et al.).
  – Bayesian Data Analysis (Gelman).
Bonus Slides

• I will include a lot of “bonus slides”.
  – May mention advanced variations of methods from lecture.
  – May overview big topics that we don’t have time for.
  – May go over technical details that would derail class.

• You are **not expected to learn** the material on these slides.
  – But you may find them interesting or useful in the future.

• I’ll use a different colour of background on bonus slides.
  – I often include “post-lecture” bonus slides after the “Summary” slide.
Getting Help

- We will use Piazza for course-related questions:
  - Link on Canvas.
  - Private posts asking about general information will be made public without asking.
- Weekly or almost-weekly Tutorials:
  - Run by TAs covering related material, mainly to help with assignments.
  - They are after class Monday/Wednesday, optional, and starting next week.
- Instructor and TA office-hours:
  - Schedule on Canvas (starting this week).
- Teaching Assistants:
  - Ali Behrouz
  - Raghav Goyal
  - Lironne Kurzman
  - Alan Miligan
  - Betty Shea (Head TA)
  - Yilin Yang
Next Topic: Grades and Course Work
50%: Assignments

• 50% of your course grade will be based on 4 assignments:
  – Written answers, math, and Julia programming.
  – No, you can’t do the assignments in Python, R, Matlab, and so on.
    • Julia is free and way faster than Python/R/Matlab.
    • Assignments have prepared code, and multiple languages would be hard on Tas.
    • It’s important to know how to learn a new language (you won’t always use the same language).

• Due at midnight on due dates:
  – First assignment due Friday of next week.
    • NO LATE DAYS WILL BE CONSIDERED FOR THE FIRST ASSIGNMENT (even if you “didn’t know you would be registered”).
    • Subsequent assignments due every ~3-4 weeks, with more flexibility.

• Start early, the assignments are a lot of work:
  – Previous students estimated that each assignment takes 6-25 hours:
    • This was heavily correlated with satisfying prerequisites.
    • Please look through the assignment in previous offerings to see length/difficulty.

• Doing assignments in groups:
  – Assignment 1 should be done on your own.
  – Assignments 2-4 can be done in groups of 1 to 2 (but hand in one assignment for the group).
50%: Choose Your Own Adventure

- 50% of your grade will be based on a combination of the following:
  1. Final exam.
  2. Course project.
  3. Example lecture and assignment.

- You do not have to do all three of the above.
  - 440 students: 50% based on the maximum grade across these 3 options.
    - Have to do at least 1 out of the 3.
  - 540 students: 25% based on the maximum, 25% based on 2nd-best.
    - Have to do at least 2 out of the 3.
Final Exam

• Final exam details:
  – Scheduled by UBC, date currently unknown.
    • If you plan to take it, don’t make travel plans before April 27th.
  – If in person: closed book, three pages of double-sided “cheat sheets”.
  – If online: 24-hours to take exam.

• There will be two types of questions:
  – ‘Technical’ questions requiring things like pseudo-code or derivations.
    • On topics covered in assignments (similar to assignment questions).
  – ‘Conceptual’ questions testing understanding of key concepts.
    • All lecture slide material except “bonus slides” is fair game here.
Course Project and Example Lecture/Assignment

• Course projects and example lecture/assignment:
  – More details coming later in the term.
  – Should be done in groups of 2-3.
  – Project scope will be smaller than projects in most classes.
  – You can use any language.
  – Will (probably) be due on the last day of exams.
  – You can use different groups for different projects/assignments.
    • Mixing and matching 440 and 540 students is also ok.
Late Assignment Policy

• You can submit one of Assignments 2-4 up to one week late.

• Example:
  – Assignment 3 is due on a Friday at midnight.
  – You can hand it in anytime before the following Friday.
    • You do not need to “declare” this, there is no penalty and no questions asked.
    • But then your group cannot hand in Assignment 4 late.

• For groups, can hand in late if at least one hasn’t handed in late already.

• No other extensions will be considered.
  – Also, since you can submit more than once, you have no excuse not to submit something preliminary by the deadline.
Cheating and Plagiarism

• Read about UBC’s policy on “academic misconduct” (cheating):

• When submitting assignments, acknowledge all sources:
  – Put “I had help from Lucy on this question” on your submission.
  – Put “I got this from another course’s answer key” on your submission.
  – Put “I copied this from the Coursera website” on your submission.
  – Otherwise, this is plagiarism (course material/textbooks are ok with me).

• At Canadian schools, this is taken very seriously.
  – Could receive 0 in course, be expelled from UBC, or have degree revoked.
Summary

• Machine learning:
  – Automatically detecting patterns in data to help make predictions and/or decisions.

• CPSC 440:
  – Advanced/difficult 2nd or 3rd+ course on this topic.
  – Also called CPSC 540 for grad students.
  – “Sequel” class to CPSC 340 (not an “advanced” version of it).

• Course admin:
  – These slides are the syllabus!
  – Check here before asking course admin question so I don’t do this:

• Next time: estimating COVID-19 prevalence.
UBC provides resources to support student learning and to maintain healthy lifestyles but recognizes that sometimes crises arise and so there are additional resources to access including those for survivors of sexual violence. UBC values respect for the person and ideas of all members of the academic community. Harassment and discrimination are not tolerated nor is suppression of academic freedom. UBC provides appropriate accommodation for students with disabilities and for religious and cultural observances. UBC values academic honesty and students are expected to acknowledge the ideas generated by others and to uphold the highest academic standards in all of their actions. Details of the policies and how to access support are available here: https://senate.ubc.ca/policies- resources-support-student-success
Quotes from people who probably should have taken CPSC 340:

- “I did Coursera [or other online class] and have done well in Kaggle competitions.”
  - Neither of these cover calculus in matrix notation or MLE and MAP estimation.
- “I’ve used SVMs, PCA, and L1-regularization in my work.”
  - Sure, but do you know how to implement them from scratch?
- “I’ve seen most of the 340 topics before.”
  - Sure, but at what level of detail and do you know how to implement them from scratch?
- “I want to apply machine learning in my research.”
  - Great! Take 340 to learn how the most useful tools work and also what can go wrong.
- “I took a machine learning course at my old school.”
  - 340 is more broad/advanced than at most schools (talk to me if unsure).
- “I’ve already learned about deep learning, so can I skip the basic stuff?”
  - When something goes wrong, you are going to want to understand the fundamentals.
- “I took CPSC 540 with you. I wish I would have taken CPSC 340 first.”
  - From a really-smart person who was working in a machine learning research job at the time.