

# SYDE 675

# Pattern Recognition

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Electrical and Computer Engineering

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# Course Info

- Instructor: Ali Ayub ([agayub@uwaterloo.ca](mailto:agayub@uwaterloo.ca))
- TAs: Yuxian Huang ([y675huan@uwaterloo.ca](mailto:y675huan@uwaterloo.ca))
- Website: <https://sites.google.com/view/aliayub/teaching/syde-675-winter-2024>
- Updates and news will be posted on the course website and LEARN.
- Website:
  - Slides, notes, assignments, logistics
- LEARN
  - Submissions and grades
- Office Hours (E7, 5402). Wed 12:00 – 1:00 PM (or by appt)

# Course Goals

- To introduce the fundamental principles of pattern recognition such as pattern representation, distance measures, probability measures, model learning, etc.
- To learn how to choose appropriate technique for a given problem and verify the capabilities of the chosen technique.
- To focus on using learned concepts for applications such as optical character recognition, speech recognition, robot vision, medical imaging, remote sensing, satellite image analysis, etc.
- To learn how to do research and write a scientific paper.

# Topics Covered

- Introduction, Pattern Recognition Problem Definition, Machine Learning, Bias-Variance Trade-off
- Distance Measures for Pattern Classification
- Bayesian Decision Theory
- Probabilistic Methods for Classification
- Parametric and Nonparametric Density Estimation
- Parametric and Nonparametric Clustering
- Classical supervised learning techniques
- Feature Extraction, Feature Selection, Feature Engineering
- Deep Learning

# Textbook

No textbook. Following, however, are some excellent resources:

- [An Introduction to Pattern Recognition and Machine Learning](#), by Paul Fieguth
- Marc Peter Deisenroth, A. Aldo Faisal, and Cheng Soon Ong. [Mathematics for Machine Learning](#). Cambridge University Press, 2020. **freely available online**
- Aston Zhang, Zack C. Lipton, Mu Li and Alex J. Smola. [Dive into Deep Learning](#). 2019. **freely available online**
- Ian Goodfellow, Yoshua Bengio and Aaron Courville. [Deep Learning](#) (2016). **freely available online**

# Workload

- Between 20 25 lectures, each ~1 hour
- Three assignments, roughly every two and a half weeks
- Each one worth 18%,  $3 \times 18\% = 54\%$  total
- Submit on LEARN
  - LaTeX typesetting is recommended

# Policies

- Do your own work!
  - High-level discussion: Great!
  - Copying/sharing code/solutions: Bad! (0% in the assignment/project)
  - Acknowledge sources
  - Please refer to Policy 71 on student academic discipline
- **No late submissions** unless you have a legitimate reason
  - Travelling, being busy with other stuff, or simply forgetting to submit are not considered legitimate reasons
- Regrade requests: within 1 week of grades released

# Project

- Remaining 46%
- You need at least 50% in the project to pass the course.
- Research project:
  - Ideal: A novel and interesting project submittable to an ML conference
  - Project proposal: 10% of the project grade
  - Project presentation: (last week of classes) 10% of the project grade
  - Writeup: 80% of the project grade
- The projects will be done in groups of 2 or 3 students, and generally involve implementing a method from a paper, and extending that method in a novel or interesting way. If you desire some inspiration, a list of example papers will be posted.

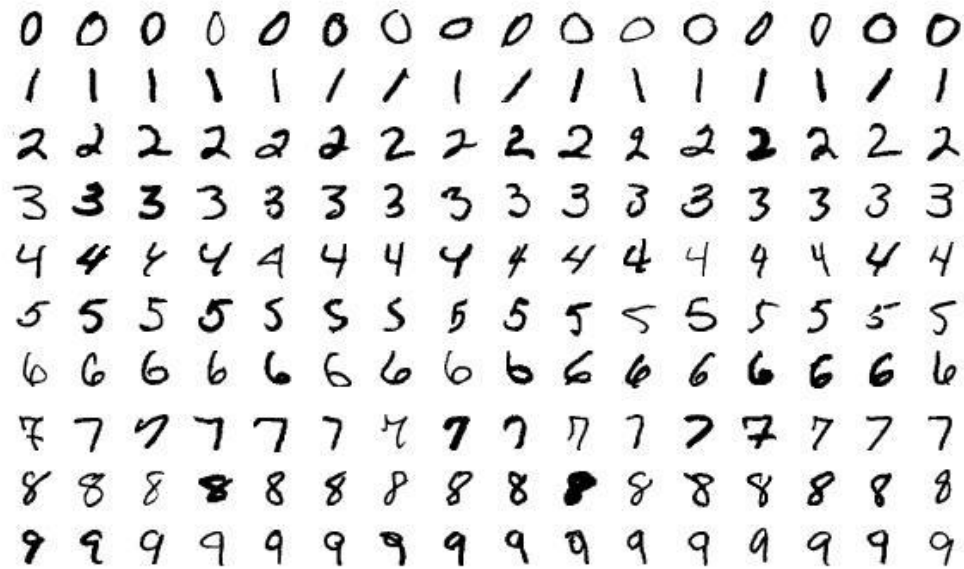


# Project Ideas

- Some general categories of project ideas:
  - Literature survey: include the problem definition and cite at least 10 papers that you plan to survey
  - Empirical evaluation: include the problem definition that you want to study, and cite at least 5-8 papers that you plan to review
  - Algorithm design: include the problem definition and a justification for why the current approaches are not satisfactory for the given problem. Cite at least 5 papers related to the chosen problem

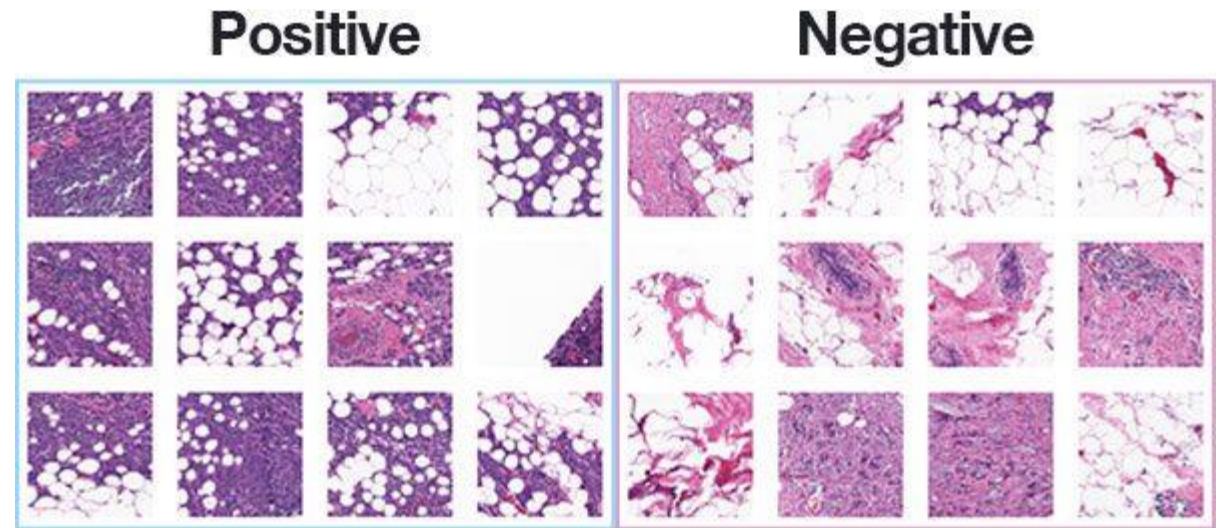
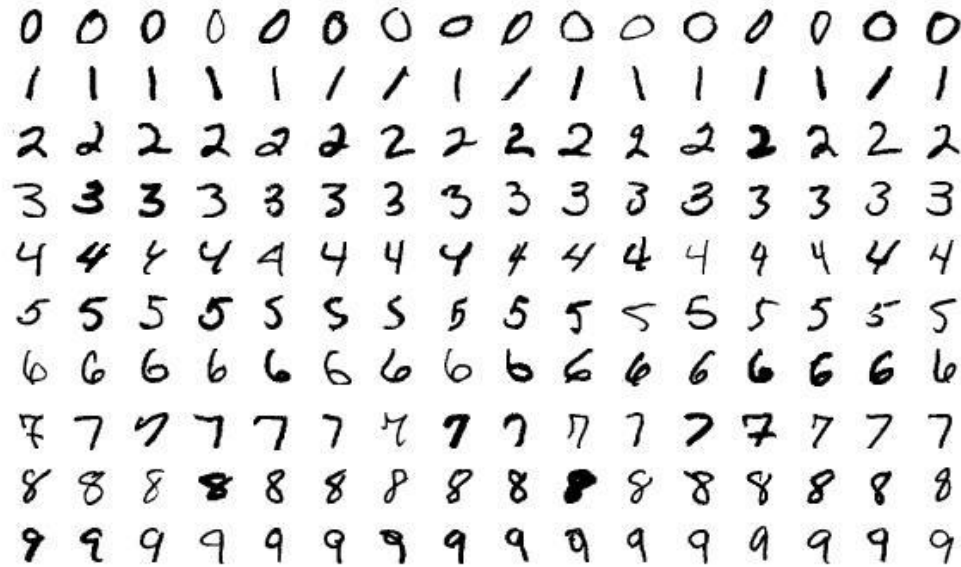
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- Applications: image classification, speech recognition, medical image analysis, etc.



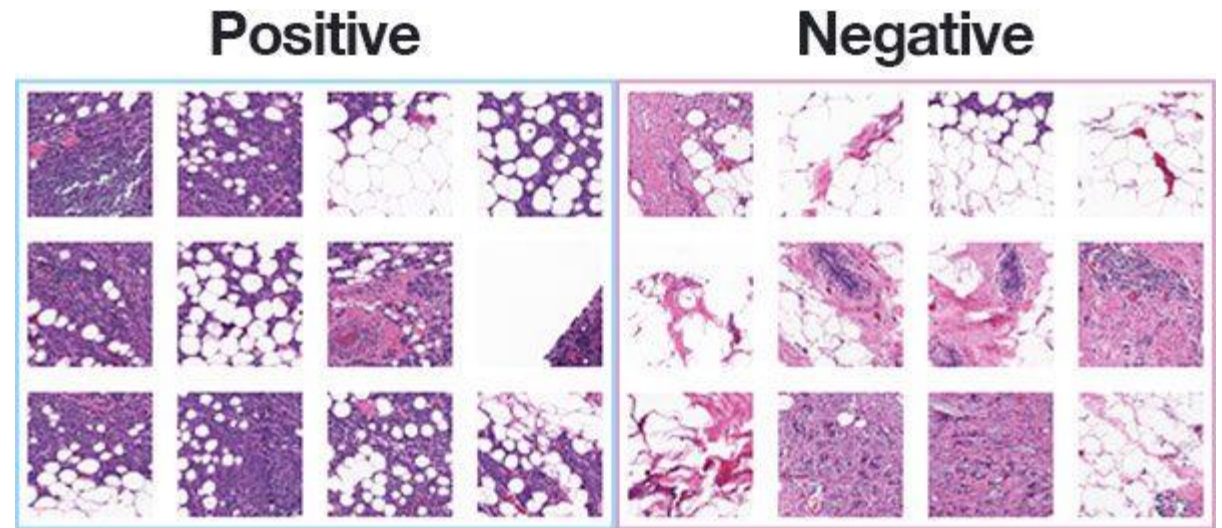
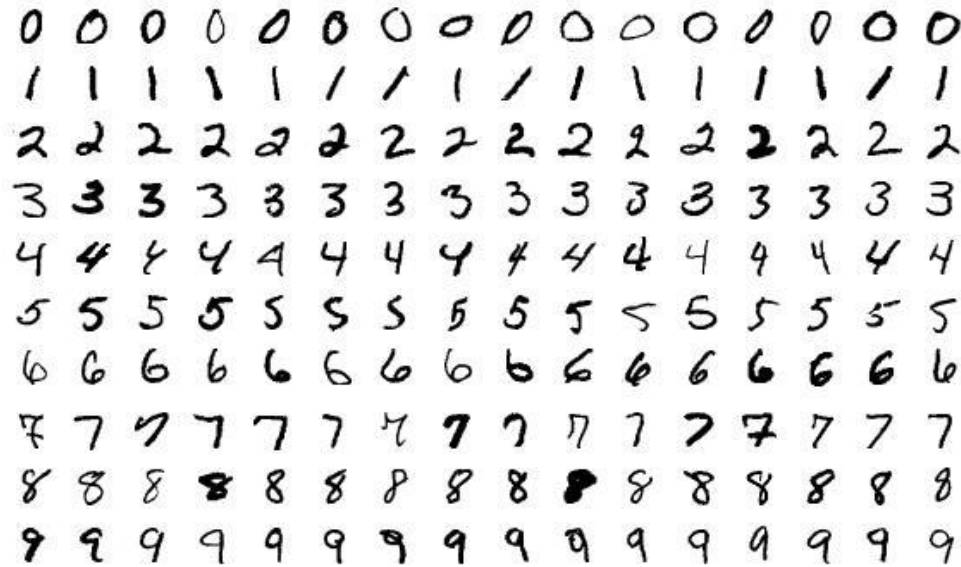
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  - Goal: the function  $f$  is able to perform predictions/inferences on some unseen data that was not in the training examples (test data).

Questions?

