COURSE ANNOUNCEMENT FOR MACHINE LEARNING (\$ 2021)

Instructor

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Lecture language

- Lecturing language: English
- However, if you have any question, you can ask me in Chinese
 - But, I will answer the questions in English (along with the translation of the problems)

Grading

- □ MT 35%
- □ Final 40%
- □ Homework 25%
- □ Term Project (Bonus) 10%

Exam

- □ MT/Final are open-book exams
- Problems may need paper-and-pen calculation
- □ The exam time is 3 hours (more than enough)
- Exam dates: 4/23 & 6/25
- Do not bring too many materials, as you may not have enough time to both read books and answer questions

Homework

- Homework assignments mix programming-based problems with non-programming problems
- Homework is due on next lecture meeting after announced
 - For example, if homework one is announce on Feb. 5, it will be due on Feb. 12
- Homework solutions will be given after the due dates

Project

- Grading is based on completeness and difficulty level
- Each term member will receive his/her points based on his/her contribution
- Project grading
 - Project presentation: 60%
 - □ Project report: 40%
- □ Read the following blog for challenging problems: https://www.analyticsvidhya.com/blog/2015/06/start-journey-kaggle/

Textbook

- No official textbook is assigned
- Lecture PPT files will be made available
- Reference book
 - Peter Harrington, Machine Learning in Action, 2012
 - Charu C. Aggarwal, Neural Networks and Deep Learning, Springer, 2018
 - E. Alpaydin, Introduction to machine learning, 3rd ed
- Please DO NOT distribute my PPT over internet
 - Copyright issues

Lecture style

- □ Follow PPT files
- Skip most of detailed math, presenting only the main concepts
- Provide numerical examples (mainly searched over the Internet) to illustrate how to implement some algorithms

Time & place

- Lecture Day: Friday
- □ Time: 9:10 AM to 12:00 PM
- □ Place: Tech building 1322 (科研1322)

- Week 1
 - Class announcement
 - Introduction to Al
 - introduction to machine learning
- □ Week 2
 - Basics of supervised learning
 - Example: k-NN
 - Classification and regression
 - VC dimension
 - Bayesian decision theory

- Week 3
 - Naive Bayes classifiers
 - ML and MAP estimation
 - Multivariate methods
- □ Week 4
 - Density estimation
 - □ Gaussian mixture model (GMM)

- □ Week 5:
 - Feature selection
 - Brief intro to Hidden Markov model (HMM)
- Week 6: Holiday
- □ Week 7
 - Decision trees
 - ID3
 - **C4.5**
 - Random forest
 - Basics of optimization

- □ Week 8
 - Support vector machine (SVM)
 - Multi-class classification
- □ Week 9: MT
- □ Week 10
 - MT sol
 - Boosting methods
 - Adaboost
 - Theories & more

- Week 11
 - Boosting methods (cont'd)
 - Clustering algorithms
 - Dimension reduction techniques: PCA
- □ Week 12
 - Dimension reduction techniques: ICA, FA, LDA
 - Back propagation

- Week 13
 - Cost functions
 - Writing Keras programs
 - Convolutional neural networks
 - Depthwise spatial convolution
- □ Week 14
 - YOLO
 - Training neural networks
 - Visualization

- Week 15
 - LSTM
 - Autoencoder and GAN
- Week 16
 - Ensemble learning
 - Conducting experiments
 - Design and analysis of experiments (Ng's Basics of ppt)
 - Reinforcement learning basis

- □ Week 17
 - Q learning and SARSA
 - DQN & DDQN
 - Intro to Policy-based learning
 - AC3
 - PPO
- □ Week 18: Final exam

Why to study so many tools

- Neural networks with deep learning algorithms are good for massive training examples, but sometimes we do NOT have (e.g., medical imaging)
- It is like a chef who needs various kinds of knifes



Why to study so many tools

- Almost all existing methods have shortcomings
- When encountering problems, we can switch to another methods
- Exercise: Find out shortcomings of some well-known methods
 - Neural networks (particularly CNN)
 - Decision tree (C4.5)
 - Adaboost