Announcement @May.31

- Programming Assignment 6
 - Due: June. 14, 11:59 pm
- Homework 6
 - Due: June. 12, 11:59 pm

Project Presentation and Submission

- Project presentation
 - Time: In class, June. 7 & 9 (Tue & Thu)
 - Each group has 8 minutes for presentation
 - Schedule:
 - ▶ June 7: group 1-9
 - ▶ June 9: group 10-17

- Report & Code
 - Due: 11:59pm, June. 9 (Thu)
 - Submission link at BB -> Project -> Project Submission
 - Format: PDF, academic paper, no page requirement (Report); Code
 - Each group only needs to submit one report and code

Final Exam

- Time
 - ▶ 8:00-9:30am, June. 16 (Thu.)
- Location
 - Online: Blackboard + Tencent Meeting
- Format
 - ▶ Open-book (纸质材料)
 - ▶ 15 multiple-choices, 2 problems
- Grade
 - ▶ 25% of the total grade
- ▶ F2018 final exam paper is available at:
 - ▶ Blackboard menu → Previous Exams → Fall 2018 Final Exam

Final Review

Disclaimer

- ▶ Topics covered in this review may not appear in the exam.
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Probabilistic temporal models

- Markov models
 - Markov assumption, Transition model
- Hidden Markov models
 - Transition model (states) + emission model (evidence)
 - Filtering: $P(X_t|e_{1:t})$
 - Forward algorithm
 - Most likely explanation: $argmax_{x_{1:t}}P(x_{1:t}|e_{1:t})$
 - Viterbi algorithm
- Dynamic Bayes networks
- Approximate inference by particle filtering
 - ▶ Propagate forward → Weight → Resample

Markov Decision Processes

- Markov Decision Process
 - States S, Actions A, Transitions P(s'|s,a), Rewards R(s,a,s')
- Quantities:
 - Policy, Utility, Values, Q-Values
- Solve MDP
 - Bellman equation
 - Value iteration
 - Policy iteration
 - Policy evaluation + Policy improvement

Reinforcement Learning

- Reinforcement learning
 - MDP without knowing T and R
 - Offline planning vs. online learning
- Model-based learning
- Model-free learning
 - Policy evaluation: Temporal Difference Learning
 - Exponential moving average
 - Computing q-values/policy: Q-Learning
- Exploration vs. Exploitation
 - Random exploration, exploration function
- Approximate Q-Learning
 - Feature-based representation of states

Supervised machine learning

- To learn an unknown target function f from labeled examples
- Classification (f with discrete output value)
 - Naïve Bayes
 - All attributes are independent given class
 - Generalization and overfitting, smoothing
 - Perceptron (linear classifier), neural networks
- Regression (f with continuous output value)
 - ▶ Linear regression, minimizing summed squared error

Unsupervised machine learning

- K-means
 - Clustering
 - Iteration:
 - Assign each data instance to closest center
 - Assign each center to the average of its assigned data points
- Expectation-Maximization
 - Learning Mixture of Gaussians
 - Iteration:
 - ▶ E-step: Compute label distribution of each data point
 - M-step: Update each Gaussian based on its (proportionately) assigned points



Natural Language Parsing

- Context-free grammars
 - ▶ Terminals, Non-terminals, Start symbol, Production rules
 - Rules may have probabilities
 - Sentence generation/parsing
- Parsing: CYK
 - Convert to Chomsky normal form
 - Dynamic programming: bottom-up table filling
 - Probabilistic CYK: Bottom-up computation of probabilities of best partial parses
- Regular grammars
- Dependency parsing
 - Graph-based parsing
 - DG vs. CFG

Good luck in your final exam ©

The Road Forward

- Related Courses
 - Undergraduate
 - SI151 Optimization and machine learning
 - CS150 Database and Data Mining
 - CS172 Computer vision I
 - Graduate
 - CS280 Deep Learning
 - CS282 Machine Learning
 - ▶ CS243 Introduction to Algorithmic Game Theory
 - ▶ CS272 Computer Vision II
 - CS283 Robotics
 - CS284 Simultaneous Localization and Mapping
 - ▶ SI232 Subspace Learning
 - ▶ SI252 Reinforcement Learning
 - More to come...

The Road Forward

- Learning recent developments in AI from top conferences
 - AI: IJCAI, AAAI
 - Caution: not top in ML, NLP, CV
 - ▶ ML: NIPS, ICML, ICLR
 - ▶ NLP: ACL, EMNLP, NAACL
 - CV: CVPR, ICCV, ECCV
 - Other: KDD, AAMAS, UAI, SIGIR, ...
- A good (but not perfect) way to judge a publication venue
 - Google Scholar Metrics

That's all!



