

# Announcement @May.31

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



# Project Presentation and Submission

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## ▶ 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

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- ▶ 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

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

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- ▶ 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:  $\text{argmax}_{x_{1:t}} P(x_{1:t} | e_{1:t})$ 
    - ▶ Viterbi algorithm
- ▶ Dynamic Bayes networks
- ▶ Approximate inference by particle filtering
  - ▶ Propagate forward → Weight → Resample



# Markov Decision Processes

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- ▶ 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

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- ▶ 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

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- ▶ 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

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- ▶ 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

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- ▶ 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

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- ▶ 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...

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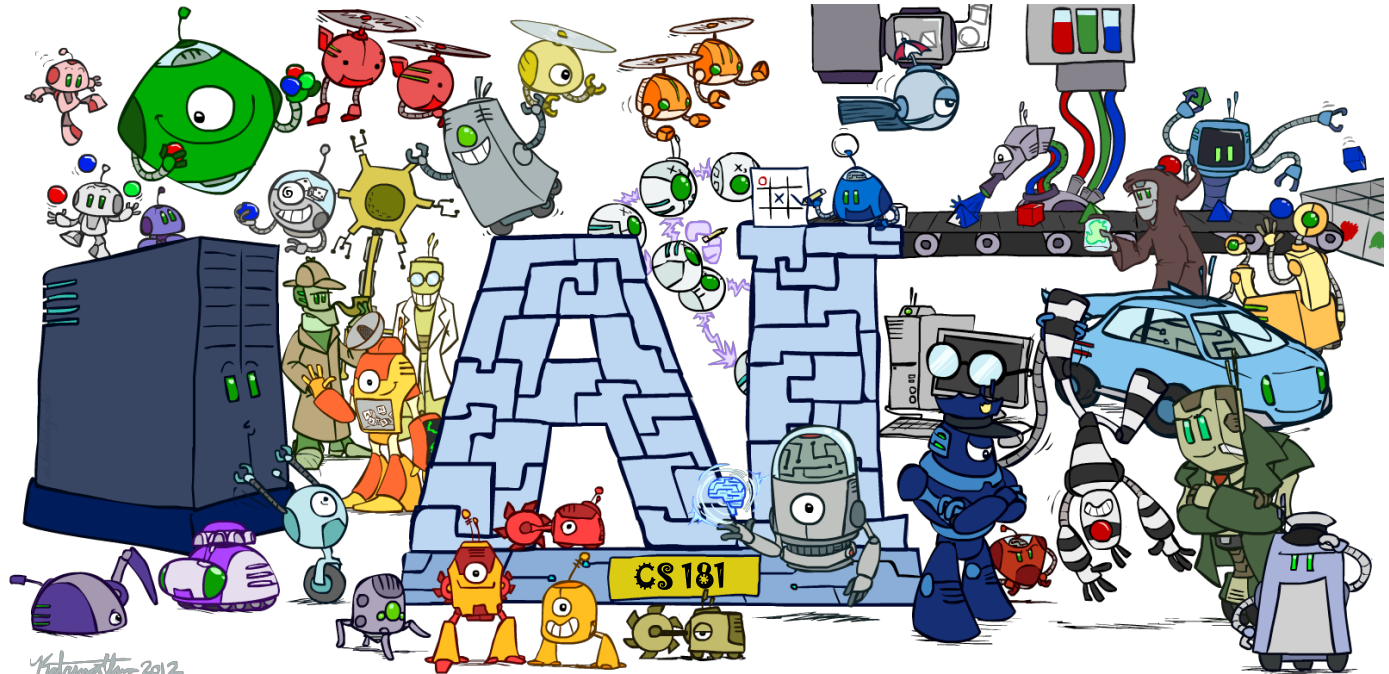
# The Road Forward

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- ▶ 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!



Kahnman 2012