# Stats 115: Probabilistic Decision Making

### Lectures

When : Winter 2020, Tuesdays & Thursdays, 2pm – 3:15pm Where : Royce 162

#### **Sections**

When : Wednesdays, 5-5:50pm

Where : WGYOUNG 1044

### **Instructor Info**

| Professor    | : Tao Gao (Dept. of Statistics)  |
|--------------|--|
| Office:      | : Math & Science Building 8917   |
| Email        | : tao.gao@stat.ucla.edu (must include " [Stats 115] " in the title of your emails) |
| Office Hours | : Tuesday 3:30 – 4:30pm, or by appointment   |

## **Teaching Assistant**

| ТА           | : Kaiwen Jiang                 |
|--------------|--------------------------------|
| Office:      | : Math & Science Building 8349 |
| Email        | : kaiwenj@g.ucla.edu           |
| Office Hours | : Monday 5-6pm                 |

## **Course Description**

Decision matters. At every moment, numerous decisions are made by humans and machines. What is a "rational" decision ? How to make a rational decision when information is incomplete, when the future is uncertain ? Probabilistic decision making offers a unified framework for solving these challenges. Learn how to build models that can infer, predict and act.

## **Justification**

Making decisions is at the core of both cognitive sciences and artificial intelligence. Yet most current courses on probability focus on analyzing data, testing hypotheses, instead of making decisions and taking actions. This course will introduce the mathematical foundation of rational decision making under uncertainty. It will reveal the deep connections between probabilistic inference and decision theory.

### **Course Objectives**

(1) Formulating a decision making problem as probabilistic inference.

(2) Deriving algorithms that can make decisions under uncertainty.

(3) Implementing code that execute inference and decision.

## **Prerequisites**

Taking this course without the recommended prerequisites is possible, but you will need my approval first.

- Introduction to Probability (Stats 100A)
- Experience with Python (or equivalent programming languages)

## <u>Grading</u>

- Two Exams (50%)
- Eight weekly assignments (50%)

## Late Policy

All assignments must be turned in via CCLE on time. We will allow a total of five late days cumulatively – no explanation required. We will not make any additional allowances for late assignments: the late days are intended to provide for exceptional circumstances, and students should avoid using them unless absolutely necessary. Any assignments that are submitted late (with insufficient late days remaining) will not be graded.

| Week | Date     | Title  |
|------|----------|--|
| 1    | Jan 7    | Introduction: Decision Under Uncertainty                   |
|      | Jan9     | Utility theory: Quantifying Rationality                    |
| 2    | Jan 14   | Probability: Quantifying Uncertainty                       |
|      | Jan 16   | Making Simple Rational Decisions                           |
| 3    | Jan 21   | Bayesian Probabilistic Inference                           |
|      | Jan 23   | The Value of Information                                   |
| 4    | Jan 28   | Bayesian Networks: Causality and Independence              |
|      | Jan 30   | Exact Inference in Bayesian Networks                       |
| 5    | Feb 4    | Review and QA  |
|      | Feb 6    | Mid-term Exam  |
| 6    | Feb 11   | Decision Networks: Making Decisions with Bayesian Networks |
|      | Feb 13   | Markov Process: Probabilistic Reasoning Over Time          |
| 7    | Feb 18   | Presidents' Day Break                                      |
|      | Feb 20   | Markov Decision Process                                    |
| 8    | Feb 25   | Value Iteration  |
|      | Feb 27   | Q-learning: Model-free Reinforcement Learning              |
| 9    | March 3  | Monte-Carlo Tree Search: Model-based Planning              |
|      | March 5  | Alpha-Zero: Integrating Model-based and Mode-free Methods  |
| 10 - | March 10 | Review and QA  |
|      | March 12 | Final Exam   |

### **Topics & Tentative Schedule**

### <u>Readings</u>

- Artificial Intelligence: A Modern Approach (third edition), Chapters 13-17.
- Reinforcement Learning: An Introduction (second edition)
  <u>http://incompleteideas.net/book/bookdraft2017nov5.pdf</u>