

Artificial Intelligence

CSCI 440

SEMESTER (fall 2016)

PROFESSOR/CLASS INFORMATION

Shieu-Hong Lin

<u>(Course) Title:</u> Artificial Intelligence <u>Term</u>: fall, 2016 <u>Location</u>: LIB 141 <u>E-Mail</u>: shieu-hong.lin@biola.edu <u>Office Hours</u>: M to Th 8:30-10:30am <u>MW 11:30~1:00pm</u> <u>Meeting with Professor</u>: **Make Appt via Email** <u>Office Location</u>: Grove 8 <u>Course Code/#</u>: CSCI 440 <u>Class Days/Time</u>: M W 1:30-2:45pm <u>Credit Hours/Units</u>: 3 <u>Office Phone</u>: 562 903-4741

<u>Dept. Website</u>: http://csci.biola.edu <u>Admin Assistant</u>: Jerrianne Smith, x4741

COURSE DESCRIPTION

Concepts and techniques of artificial intelligence, representation, search strategies, control, communication and perception, and applications. Offered other year.

COURSE ALIGNMENT WITH PROGRAM LEARNING OUTCOMES

This upper-division course is an elective for Computer Science majors. Successful completion of this course (see next section) will prepare students to demonstrate a developing proficiency toward the accomplishment of PLOs: (i) modeling, analysis, and problem solving, (ii) Knowledge of the theory and practice of computing, and (iii) integration of faith and learning.

COURSE OBJECTIVES AND STUDENT LEARNING OUTCOMES

By the completion of this course including class participation, class assignments (referred to as "Tasks"), class readings and group interaction, the following objectives and learning outcomes will be assessed and demonstrated:

IDEA Objective #4: Developing specific skills, competencies, and points of view needed by professionals in the field most closely related to this course (Essential emphasis).

STUDENT LEARNING OUTCOMES (The learner will demonstrate that he or she has satisfactorily fulfilled IDEA Objective #4 by being able to):

- Gain a broad understanding of Artificial Intelligence research and the applications in speech recognition, machine learning and data mining, search, and automatic reasoning.
- Establish the in-depth understanding of the mathematical and algorithmic framework underlying the very important subject of machine learning, including a series of hands-on assignments on hidden Markov models.
- Cultivate the problem solving capability based on the in-depth understanding of machine learning or other selected subject through hands-on study projects using the related tools and programming environments.

REQUIRED TEXTS

• Peter Flach, *Machine_Learning*, Cambridge University Press, 2012.

LEARNING TASKS (Assignments) & ASSESSMENT (Grading)

Description and Weighting of Assignments:

Task 1: Weekly Reading and Progress Report

Due Date: Thursday of the week (15 assignments) Weighting: 15% Possible Points: 4 points each. Effort (2 points): Report the (i) a numerical estimate of the amount of time he/she spent for the reading, (ii) a numerical percentage regarding the percentage of stuff in the reading actually read and understood, and (iii) whether the student has come to the class this week.

Reflection on the reading (2 points):

The student need to put down 1 to 2 paragraphs of his/her thoughts about the reading such as new insight you gained, interesting things encountered, questions of things you don't understand, and so forth.

Assessment:

For the effort part,

the student is expected to (i) have attended the class this week at least once (0.5 point), and (ii) have either gained a good understanding of 80% or more of the contents or have spent at least three hours in the reading (1.5 points).

For the reflection part,

the student is expected to show substantial evidence of understanding or effort of trying to understand the contents in the reading.

Task 2: Homework, Labs, and Programming Assignments

Due Date: Thursday of the week **Weighting**: 45% **Possible Points**: 6 points each for programming assignments. 1-6 points for each Lab and Homework depending on the requirement in it.

Description: Hands-on programming assignments require the student to incrementally develop modeling, analytical, and programming skills based on the concepts of artificial intelligence learned in the class.

Integrity rules for programming assignments:

- Peer discussion is encouraged: Peer discussion is encouraged to cultivate an open learning environment in the class, but you should carefully read the guidelines below to avoid any dishonest behavior and never step over the guidelines explicitly described in the following.
- Never use code written by others: Any copy-and-paste of code from other people's programs or from websites is viewed as cheating and you will get 0 points for the assignment.
- Never circulate your code to others: Peer discussion of code shown on the screen is
 acceptable for debugging purpose and for explanation of ideas. But you should never
 pass around your code (electronically or on paper) to others except for the TA and the
 instructor. Violating this rule is viewed as cheating in the class and the provider will
 receive 0 points for the assignment.
- Never provide false or exaggerated results of test cases: You need to describe results of test cases in the self-evaluation report. Providing false or exaggerated results of test cases in the report is viewed as cheating and you will receive 0 points for the assignment.
- **Demonstrate the credibility of your authorship of the work**: When you submit your code as your own work for points, you should make sure that you are able to explain your code and reconstruct your code from scratch without any outside help when requested. If you are not able to do that on your own when requested, you will get 0 points for the assignment and there will be an investigation.
- **Consequence of cheating in the class**: Cheatings end in 0 points for the assignments followed by discipline actions described in the student handbook.

Assessment: The student needs to submit (all related .cpp and .h files) together with a self-evaluation report. The self-evaluation report should describe results from sufficient test cases that are verified by a peer reviewer. We'll grade each programming assignment in a 0-6 scale based on the following rubric.

- 0. Nothing done or missing the self-evaluation report or missing the integrity review in the report.
- 1. Source code is completed but the code fails to compile successfully.
- 2. Source code can compile and do something required, but has serious bugs or miss a couple of key features.
- 3. Source code can compile and do most of the features required, but has many minor bugs or miss a key required feature.

- 4. Source code can compile and do all the features required, nearly fully functional, only a couple of minor bugs.
- 5. Source code can compile and do all the features required, fully functional, no bugs.
- 6. In addition to the points received above, get one more point if

a. the self-evaluation report contains sufficient descriptions of test cases used (0.25 point), and
b. the self-evaluation report indicates the results of the test cases were verified by a peer reviewer (0.25 point), and
c. the source code is well indented and commented to make it visually very readable (0.5

point).

Task 3: Study Project

Weighting: 15%

Description: The student conducts a study project on assigned subjects to develop their own problem-solving approach based on the machine learning concepts learned in the clas.

Assessment: The submitted work done for the study project will be examined to determine whether a satisfactory level of study has been done based on the breadth, depth, and clarity of the submitted work.

Task 4: Exams (Quizzes, midterm and final exams)

Due Date: Exams in the midterm and final exam week

Weighting: 25%

Description: The exams have both written components which mainly test the conceptual understanding of AI and machine learning, and the programming components which test the skills of implementation.

Assessment: The written component will be graded based on the answers provided while the programming component will be graded based on the same rubric provided for the weekly programming assignments.

CLASS INFORMATION

1. Class Attendance:

Attendance: You are expected to attend the class regularly. Missing the class may seriously hamper your understanding of many key concepts and programming skills critically needed in your programming assignments. Class attendance is counted toward points for the weekly reading report.

2. Turning in Assignments:

Assignments are expected to be electronically submitted under the Canvas system. Due dates are all on Mondays. The submission link under Canvas may remain open for 2 more days after the due date as grace period.

3. Late Policy:

1 point will be deducted for late submission within 2 days of the due date while the submission link is still open. **You will receive no points after the submission on canvas is closed** unless it is something like a serious health issue with statements from the doctor as proof.

8. Computation of Final Grade:

| Total | 100% |
|--------------------------------------|------|
| Exams | 25 % |
| Study Project | 15% |
| Homework and Programming Assignments | 45% |
| Weekly Progress Report | 15 % |

9. Final grades will be awarded on the following point system:

| А | 93% |
|----|-------------------|
| A- | 90% |
| B+ | 87% |
| В | 84% |
| B- | 80% |
| C+ | 77% |
| С | 74% |
| C- | 70% |
| D+ | 67% |
| D | 64% |
| D- | 60% to pass class |
| | |

GENERAL INFORMATION

1. The GPA System used by the University Registrar's Office is:

| A = 4.0 | B = 3.0 | C = 2.0 | D = 1.0 |
|-----------|-----------|-----------|-----------|
| | | | |
| A- = 3.66 | B- = 2.66 | C- = 1.66 | D- = 0.66 |
| | | | |
| B+ = 3.33 | C+ = 2.33 | D+ = 1.33 | F = 0.0 |

2. Method of Instruction:

The following methods of instruction will be included in this course:

- 1. Lecture
- 2. Written Reports
- 3. Programming Assignments
- 4. Labs
- 5. Reading
- 3. Posting of Grades:

Grades for individual assignments will be posted under Biola's Canvas system. To access the records online, log on to <u>http://canvas.biola.edu/</u> to make sure the records are accurate.

Tentative Schedule

- Weeks 1-2 Bayesian network as probabilistic models for artificial intelligence
- Weeks 3-6 Speech recognition and spelling recognition using HMMs
- Weeks 7-10 Machine learning: concepts and key techniques
- Weeks 11-12 Search, knowledge representation, and the applications
- Weeks 13~15 Selected topics and projects
- Week 15 Project presentation