ICSI 436/536
Introduction to Machine Learning
(Spring 2020)
Class Meeting Time: 10:15AM - 11:35AM
Location: LC3C

INSTRUCTOR

<table>
<thead>
<tr>
<th>Instructor’s name</th>
<th>Siwei Lyu</th>
</tr>
</thead>
<tbody>
<tr>
<td>Instructor’s title</td>
<td>Professor</td>
</tr>
<tr>
<td>Office location</td>
<td>UAB 413</td>
</tr>
<tr>
<td>Office hours</td>
<td>MW 11:00am - 12:00pm</td>
</tr>
<tr>
<td>E-mail address</td>
<td><a href="mailto:slyu@albany.edu">slyu@albany.edu</a></td>
</tr>
</tbody>
</table>

TEACHING ASSISTANTS

<table>
<thead>
<tr>
<th>TA</th>
<th>Yi Wei</th>
</tr>
</thead>
<tbody>
<tr>
<td>TA’s office location</td>
<td>UAB415</td>
</tr>
<tr>
<td>TA’s office hours</td>
<td>T 1-3pm</td>
</tr>
<tr>
<td>TA’s email addresses</td>
<td><a href="mailto:ywei2@albany.edu">ywei2@albany.edu</a></td>
</tr>
</tbody>
</table>

REQUIRED TEXTBOOK

We do not use a textbook, all should be based on the lecture slides of the instructor

COURSE DESCRIPTION / OVERVIEW

Machine learning is an important and rapid growing branch of artificial intelligence. The aim of machine learning is to design algorithms that can extract information from environment automatically and improve their ability to perform the intended task. This course starts with a high level overview of general problems in machine learning, followed by a review of mathematical backgrounds and numerical optimization methods that are essential for machine learning algorithms, after that several important topics in machine learning will be covered.

PREQUISITES
the prerequisite to this class is very important, and lack of knowledge of these subjects will make difficult
to make positive progress in the class. There will be an entrance exam to test the readiness of this class.
• Linear Algebra (AMAT 220 or equivalent)
• Multivariate calculus (AMAT 214 or equivalent)
• Discrete probability (AMAT 367 or equivalent)
• Numerical methods (CSI 401 or equivalent).

COREQUISITES

None

LEARNING OBJECTIVES / OUTCOMES

At the completion of the course the student will:

• understand key concepts and algorithms in machine learning;
• develop a fundamental understanding of machine learning algorithms and tools;
• be able to apply such algorithms to practical applications.

The topics that will be covered in this course are provided at the end of the syllabus.

TENTATIVE LIST OF TOPICS TO BE COVERED

1. dimension reduction methods:
   1. principal component analysis, multi-dimensional scaling and ISO-MAP
2. classification methods:
   1. ROC and AUC, types of loss functions
   2. linear discriminant analysis, k-nearest neighbor classifier, and logistic regression
3. regression methods:
   1. least squares regression, ridge regression, and 11 regularized least squares regression
4. clustering methods:
   1. k-means clustering and EM algorithm
5. support vector machines for classification and regression
6. deep learning: artificial neural networks and other hierarchical models

COURSE WEBSITE AND BLACKBOARD

Blackboard will be used to provide essential course materials, the most current syllabus, and assignment
documents and no separate course website will be maintained. However, this is not an online course and
class attendance and participation is essential and required.

ASSESSMENT AND POLICIES

The accomplishment of course objectives will be assessed by applying the concepts and tools for
engineering design in a combination of team and individual assignments/labs/projects, tests, and a final
project that includes research and design, a written component, and an oral presentation.

Exams: Two exams will be given. A portion of the class period preceding each exam will be utilized for a
review session. There is/is not a final exam during finals week.

Project / Labs / Assignment: Projects / labs / assignments will be assigned and will be conducted both out of
class and during lab period. They will be graded on a 10-point scale and will be totaled together to account for 30% of the final grade.

Final Project: A final project will/will not be required. The requirements for this assignment will be fully described in a Blackboard later in the course.

Grading
A final grade will be determined as a weighted average of these scores using the following weights:
- 20% in class quizzes
- 30% homework assignments
- 20% midterm exam
- 30% final project

Grading Scale
A: 100-95 points A-: 94-90 points
B+: 89-87 points B: 84-86 points B-: 80-83 points
C+: 79-76 points C: 75-70 points
D: 69-60 points E: 59 points and below

Students must complete all requirements in order to pass the course. A grade of incomplete will be given only when circumstances beyond the student's control cause a substantial amount of course work to be unfinished by the end of the semester. Whenever possible, the student is expected to make extra efforts to prevent this situation from occurring. The instructor will be the sole judge of whether an incomplete is warranted. Final grades are computed based on the above formulas and are NOT negotiable. Per department policy, “…students may not submit additional work or be re-examined for the purpose of improving their grades once the course has been completed and final grades assigned.” purpose of improving their grades once the course has been completed and final grades assigned.”

ATTENDANCE
Students are expected to attend every class and to arrive on time. Please DO NOT disrupt the class by entering late or leaving early without instructor approval. Attendance will be taken at every class meeting. Although it is not part of the final grade, consistently missing classes will cause degraded performance of your final grade: one “no show” without legitimate explanation that is caused by uncontrollable or unavoidable reasons in class will lead to a 30% reduction of total final grades; twice and more will lead to an F grade to the class. If you know that it will be difficult for you to consistently get to class on time and stay for the entire period, you should take this course at a time that better fits your schedule.
Computers may be used during class for note taking as long as the use is not disruptive or distracting. Also see http://www.albany.edu/health_center/medicalexcuse.shtml.

LATE HOMEWORK TURNINS
Homework turned in before or on the specified due date and time, in class or submitted through Blackboard, depending on the circumstance, are eligible for 100% of the grade. If you choose to turn in after the due date and time passes, for the first 24 hour period after the due date and time, your assignment will be eligible for 50% of the full grade; after that, your assignment will be eligible for 0% of the full grade.
WITHDRAWAL WITHOUT PENALTY

Please pay attention to the drop date, which is the last date you can drop this course with no financial consequence. After that, you should consult the university’s liability schedule (http://www.albany.edu/studentaccounts/liability.php) to consider dropping from this class. This may happen when you have to miss many assignments for foreseeable scenarios. IMPORTANT: It is your responsibility to take such an action by this date, and don’t wait until it’s too late to see us when you get in trouble.

INCOMPLETE AND EXTRA CREDIT POLICY

As per the Undergraduate Bulletin, the grade of Incomplete (I) will be given "only when the student has nearly completed the course requirements but because of circumstances beyond the student's control the work is not completed." A student granted an incomplete will make an agreement specifying what material must be made up, and a date for its completion. The incomplete will be converted to a normal grade on the agreed upon completion date based upon whatever material is submitted by that time. The instructor will be the sole judge of whether an incomplete is warranted. IMPORTANT: Incomplete will not be given to students who have not fulfilled their classwork obligations, and who, at the end of the semester, are looking to avoid failing the course. There will be no extra credit work. All students will be expected to complete, and be graded on, the same set of assignments.

NON-CLASS RELATED USE OF TECHNOLOGY

Use of electronic devices (cell phone, tablets, personal laptop computers) for non-class purposes while the class is in session is not allowed. If this is violated in a consistent manner after initial warning is issued by the instructor, the student involved will be treated as unexcused missing the day’s class.

RESPONSIBLE COMPUTING

Students are required to read the University at Albany Policy for the Responsible Use of Information Technology (https://www.albany.edu/its/its_policies.htm). Students will be expected to apply the policies discussed in this document to all computing and electronic communications in the course.

STUDENTS WITH DISABILITIES

Reasonable accommodations will be provided for students with documented physical, sensory, systemic, cognitive, learning and psychiatric disabilities. If you believe you have a disability requiring accommodation in this class, please notify the Director of the Disability Resource Center (Campus Center 130, 442-5490). That office will provide the course instructor with verification of your disability, and will recommend appropriate accommodations. For further information refer to the University’s Disclosure Statement regarding Reasonable Accommodation found at the bottom of the document at the following website: http://www.albany.edu/disability/docs/RAP.doc. This website can be reached by following the link under “Reasonable Accommodation Policy” at the following webpage http://www.albany.edu/disability/faculty-staff.shtml.

ACADEMIC HONESTY AND OVERALL REGULATIONS

Every student has the responsibility to become familiar with the standards of academic integrity at the University. Faculty members must specify in their syllabi information about academic integrity, and may refer students to this policy for more information. Nonetheless, student claims of ignorance, unintentional error, or personal or academic pressures cannot be excuses for violation of academic integrity. Students are
responsible for familiarizing themselves with the standards and behaving accordingly, and UAlbany faculty are responsible for teaching, modeling and upholding them. Anything less undermines the worth and value of our intellectual work, and the reputation and credibility of the University at Albany degree. Plagiarism and other acts of academic dishonesty will be punished. Read the Standards of Academic Integrity and policies in the University Bulletin (https://www.albany.edu/undergraduate_bulletin/regulations.html).

CAUTION AND A STRONG WORD OF WARNING!!!! Plagiarism and other acts of academic dishonesty will be punished. Students are expected to submit original work. While you may discuss a problem with another student, the work you submit must be your own. Any student who submits copied work or any student that provides work for copying will earn a zero grade for that assignment. If there is more than one copying incident, the student will be graded an E for the class. As per college policy, cheating activity, including cheating in exams, quizzes, projects, etc., WILL be written up in a Violation of Academic Integrity Report (VAIR) reported to the college administration, which includes the Computer Science Chair, the College of Engineering and Applied Sciences Dean, and the Vice Provost of Undergraduate Studies. This will become a part of your permanent record. Multiple incidents will result in being expelled from the college.

SEXUAL VIOLENCE PREVENTION AND RESPONSE POLICIES

Title IX of the Education Amendments of 1972 is a federal civil rights law that prohibits discrimination on the basis of sex in federally funded education programs and activities.

The SUNY-wide Sexual Violence Prevention and Response Policies prohibit offenses defined as sexual harassment, sexual assault, intimate partner violence (dating or domestic violence), sexual exploitation, and stalking. The SUNY-wide Sexual Violence Prevention and Response Policies apply to the entire University at Albany community, including students, faculty, and staff of all gender identities. The University at Albany provides a variety of resources for support and advocacy to assist individuals who have experienced sexual offenses.

Confidential support and guidance can be found through the Counseling Center (518-442-5800, https://www.albany.edu/counseling_center/), the University Health Center (518-442-5454, https://www.albany.edu/health_center/), and the Interfaith Center (518-489-8573, https://www.albany.edu/spirituality/onCampus.shtml). Individuals at these locations will not report crimes to law enforcement or university officials without permission, except for in extreme circumstances, such as a health and/or safety emergency. Additionally, the Advocates at the University at Albany’s Advocacy Center for Sexual Violence are available to assist students without sharing information that could identify them (518-442-CARE, https://www.albany.edu/advocacycenter/).

Sexual offenses can be reported non-confidentially to the Title IX Coordinator within The Office for Equity and Compliance (518-442-3800, https://www.albany.edu/equity-compliance/, Building 25, Room 117) and/or the University Police Department (518-442-3131, http://police.albany.edu/).

Please note, faculty members are considered “responsible employees” at the University at Albany, meaning that they are required to report all known relevant details about a complaint of sexual violence to the University’s Title IX Coordinator, including names of anyone involved or present, date, time, and location.
The following schedule of lecture topics and reading assignments is preliminary and may be changed as the semester progresses. The final schedule and specific assignments will be provided in Blackboard.

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<thead>
<tr>
<th>date</th>
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<th>topic</th>
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<tbody>
<tr>
<td>1/23</td>
<td>information &amp; general introduction</td>
<td>1/30</td>
<td>review of calculus</td>
</tr>
<tr>
<td>1/28</td>
<td>review of LA (1)</td>
<td>2/6</td>
<td>robust and RWLLSE, online learning &amp; recursive LLSE</td>
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<tr>
<td>2/4</td>
<td>regression &amp; LLSE</td>
<td>2/6</td>
<td>robust and RWLLSE, online learning &amp; recursive LLSE</td>
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<tr>
<td>2/11</td>
<td>review of Python</td>
<td>2/13</td>
<td>model selection &amp; LOO LSE</td>
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<tr>
<td>2/18</td>
<td>penalty &amp; segmented LLSE</td>
<td>2/20</td>
<td>regularization, ridge LLSE &amp; LASSO</td>
</tr>
<tr>
<td>2/25</td>
<td>classification &amp; discriminative LLSE</td>
<td>2/27</td>
<td>clustering &amp; multi-modal LLSE</td>
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<tr>
<td>3/3</td>
<td>dimension reduction &amp; total LLSE</td>
<td>3/5</td>
<td>review of LA (2)</td>
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<tr>
<td>3/10</td>
<td>overview of ML algorithms</td>
<td>3/12</td>
<td>mid-term exam</td>
</tr>
<tr>
<td>3/17</td>
<td>no class (spring break)</td>
<td>3/19</td>
<td>no class (spring break)</td>
</tr>
<tr>
<td>3/24</td>
<td>clustering: k-means</td>
<td>3/26</td>
<td>clustering: graph cut</td>
</tr>
<tr>
<td>3/31</td>
<td>dimension reduction: PCA</td>
<td>4/2</td>
<td>kernel methods</td>
</tr>
<tr>
<td>4/7</td>
<td>classification: AUC</td>
<td>4/9</td>
<td>dimension reduction: LDA</td>
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<tr>
<td>4/14</td>
<td>SVM formulation</td>
<td>4/16</td>
<td>SVM theory</td>
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<tr>
<td>4/21</td>
<td>SVM algorithm</td>
<td>4/23</td>
<td>neural networks</td>
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<tr>
<td>4/28</td>
<td>no class (CURCE use of room)</td>
<td>4/30</td>
<td>deep learning</td>
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<tr>
<td>5/5</td>
<td>deep learning applications</td>
<td>TBA</td>
<td>final exam</td>
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