Theoretical Foundations of Computer Vision

New Proposed Course CIS 6930 (Proposed Official Listing CAP 6XXX)
First Offered, Spring 2003
Department of Computer Science, Florida State University

Class time and location
Tuesday and Thursday, 6:35-8:00PM, LOV 103.

Instructor
- Instructor: Xiuwen Liu
- Email: liux@cs.fsu.edu (strongly preferred)
- Home page: http://www.cs.fsu.edu/~liux
- Office: 166 Love Building (LOV) Phone: (850) 644-0050
- Office Hours: Tuesday and Thursday 9:30-11:00AM and by appointments

Course Home Page
http://www.cs.fsu.edu/~liux/courses/vision2/index.html
This web page contains the up-to-date information related to this course such as news, announcements, assignments, lecture notes, useful links to resources that are helpful to this class. You are required to visit this web site on a regular basis.
Besides the course home page, an email mailing list and news group will also be established and used to post news and updates.

Course Rationale
Computer vision has evolved into an important field to understand human visual information processing and to design machine vision systems that can interact with their environment flexibly with tremendous civil and military applications. During the last two decades, computer vision has matured with commonly accepted theoretical frameworks to formulate and approach vision problems. The course is a second course in computer vision. With the assumption that students know and understand well elements of vision problems and algorithms, it formulates the vision in a mathematical framework and within the formulations to discuss different approaches and the relationships among those approaches. Thus, it is a critical advanced course for students who are interested in research in computer vision.

Course Description
This course covers theoretical foundations of computer vision. By formulating computer vision as a statistical inference process, computational approaches to vision are presented and analyzed systematically. Topics include Marr’s computational vision paradigm, regularization theory, feature extraction principles, classification algorithms, Bayesian inference framework for vision, pattern theory, and visual learning theories. It concludes with open issues and research directions in computer vision.

Prerequisites
CAP 5415 – Principles and Algorithms for Computer Vision, or permission of the instructor.
Course Objectives

Upon successful completion of this course of study a student:

- Knows how to formulate vision problems mathematically and understands different approaches to vision within this formulation.
- Understands the Marr’s paradigm and its limitations.
- Understands the elements of pattern theory and knows how to formulate and solve typical vision problems in pattern theory.
- Understands the Bayesian inference framework and knows to formulate and solve typical vision problems through Bayesian inference.
- Understand basic principles in visual learning and issues in visual learning.
- Understands relationships among different approaches to vision, their equivalence conditions and limitations.
- Understands issues in a generic vision system and knows potential computational approaches.
- Knows the research issues in computer vision.
- Has some experience with creative research in computer vision.
- Is well prepared to carry out research in computer vision.

Textbook and Class Materials

“Perception as Bayesian Inference” (David C. Knill and Whitman Richards, Cambridge University Press, 1996), recommended.

In addition to the textbooks, papers and notes from the literature will be distributed along the lectures, including the following journals and conference proceedings:

- IEEE Transactions on Pattern Analysis and Machine Intelligence.
- International Journal on Computer Vision
- Computer Vision and Image Understanding
- Proceedings of the International Conference on Computer Vision.
- Proceedings of the European Conference on Computer Vision

Student Responsibilities

Attendance is required for this class. Unless you obtain prior consent of the instructor, missing classes will be used as bases for attendance grading. In case that it is necessary to skip a class,
students are responsible to make up missing covered materials. Participation of in-class discussions and activities is also required. All submitted assignments and projects must be done by the author. It is a violation of Academic Honor Code to submit other’s work and the instructor and TA of this course take the violations very seriously.

**Assignments and Projects**

Exercises will be given to help you understand the basic concepts and techniques and need not to be turned in. There will be a term programming project, which can be done in any programming language including Matlab, Java, and C/C++. The project must involve some creativity and novelty. Based on the student’s research interest, a research paper, which can be a literature review or a survey on a particular topic, will be also assigned. There will be a final exam.

**Grading Policy**

Grades will be determined as follows:

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<th>Assignment</th>
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<tr>
<td>Attendance</td>
<td>10 %</td>
<td>Term project</td>
<td>25 %</td>
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<tr>
<td>Class participation</td>
<td>10 %</td>
<td>Research Paper</td>
<td>15 %</td>
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<td>Presentations</td>
<td>15 %</td>
<td>Final exam</td>
<td>25 %</td>
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Grading will be based on the weighted average as specified above and the following scale will be used (suppose the weighted average is $S$ in 100 scale)

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<td>$83 \leq S &lt; 87$</td>
<td>B</td>
<td>$70 \leq S &lt; 73$</td>
<td>C-</td>
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**Late Penalties**

Assignments are due at the beginning of the class on the due date. Assignments turned in late, but before the beginning of the next scheduled class will be penalized by 10 %. Assignments that are more than one class period late will NOT be accepted.

**Submission and Return Policy**

All tests/assignments/projects/homework will be returned as soon as possible after grading but no later than two weeks from the due date.

**Tentative Schedule**
• Week 1: Introduction
     o Computational characteristics of human visual information processing.
     o Generic constraints for an effective vision system.
     o General introduction to the mathematical frameworks for computer vision.
     o Problems and goals of computer vision.
• Week 2: Mathematical formulations of computer vision problems.
     o Typical structures in images.
     o Models and representations for typical structures.
     o Vision as a computational process.
• Weeks 3-4: Marr’s computational vision paradigm.
     o Vision as a feedforward inference process.
     o Regularization theory for computer vision.
     o Feature extraction principles for computer vision.
         ▪ Minimum description length criterion.
         ▪ Maximum entropy criterion.
         ▪ Redundancy reduction criterion.
         ▪ Descriptive models in computer vision.
     o Classification algorithms.
         ▪ Bayesian decision theory.
• Weeks 5-6: Computer vision as Bayesian Inference.
     o Representations in a vision system.
     o Seeing as an inference process.
     o Bayesian inference framework for vision.
• Weeks 7-8: Grenander’s Pattern Theory.
     o Mathematical framework to represent visual patterns.
     o Deformation structures of visual patterns.
     o “Analysis as synthesis” paradigm.
     o Generative models for computer vision.
• Weeks 9-10: Visual learning.
     o Computational approaches to visual learning.
     o Visual learning as dimension and redundancy reduction.
• Weeks 11-13: A unified framework for computer vision.
     o Relationships among different approaches, including equivalence conditions,
    limitations, advantages and disadvantages of existing approaches.
     o Issues in designing a generic vision system.
     o Integration of Marr’s paradigm and Grenander’s pattern theory.
         ▪ Vision as inference with hierarchical models.
• Week 14: Implementation issues in computer vision.
     o Computational complexity of vision algorithms.
     o Optimization techniques for computer vision systems.
     o Monte-Carlo Markov chain techniques for high dimensional problems.
     o Hardware implementation issues.
• Week 15: Current and future research directions in computer vision.
     o Open issues in computer vision.
     o Challenges of developing generic computer vision systems.
**Academic Honor Code**

Programming assignments/written assignments/quizzes/exams are to be done individually, unless specified otherwise. It is a violation of the Academic Honor Code to take credit for the work done by other people. It is also a violation to assist another person in violating the Code (See the FSU Student Handbook for penalties for violations of the Honor Code.). The judgment for the violation of the Academic Honor Code will be done by the TA, the instructor and a third part member (another faculty member in the Computer Science Department not involved in this course). Once the judgment is made, the case is closed and no arguments from the involved parties will be heard. Examples of cheating behaviors include:

- Discuss the solution for a homework question.
- Copy programs for programming assignments.
- Use and submit existing programs/reports on the world wide web as written assignments.
- Submit programs/reports/assignments done by a third party, including hired and contracted.

**Penalty for violating the Academic Honor Code:** A 0 grade for the particular homework/project/exam and a reduction of one letter grade in the final grade for all parties involved. A report will be sent to the department head for further administrative actions.

**Accommodation for Disabilities**

Students with disabilities needing academic accommodations should: 1. Register with and provide documentation to the Student Disability Resource Center (SDRC); 2. Bring a letter to the instructor from the SDRC indicating you need academic accommodations. This should be done within the first week of class. *This syllabus and other class materials are available in alternative format upon request.*