

PHYSICS 410/510:
IMAGE ANALYSIS
WITH APPLICATIONS IN PHYSICS

PROF. RAGHUVeer PARTHASARATHY
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Room with Eye – Maurice Tabard (1930)

COURSE DESCRIPTION

In a large, diverse, and ever-expanding variety of fields in science and technology, we wish to extract quantitative information from images. These images may be captured by microscopes, telescopes, cell phone cameras, medical imaging devices, or other instruments, but nearly all share the characteristic of being arrays of numbers that reflect the local intensity of some sort of signal. Methods of computational image analysis are therefore very broadly applicable throughout science and technology, and we'll see some of the scientific insights they enable.

Learning Goals

Our goals will be to become familiar with techniques, algorithms, and approaches for computational image analysis; to practice and improve skills in writing code to implement image analysis methods; to learn how image analysis is governed by principles of optics and statistics; and to understand various applications of computational image analysis in the sciences.

Prerequisites and programming languages

Programming. Students should have a good grasp of some high-level programming language and should be comfortable with manipulating arrays and writing functions. I *very strongly* recommend Python or MATLAB; all of my examples will use these, and I don't expect to be able to provide useful assistance in other languages. Moreover, Python and MATLAB have useful features like logical indexing and excellent toolboxes / libraries of basic image processing functions.

The vast majority of the work in this course involves programming.

If you're unsure of whether your programming skills are adequate, please see the "Homework 0" assignment posted here: <https://pages.uoregon.edu/raghu/TeachingFiles/Homework0.pdf> . Some Python exercises I wrote a few years ago, mainly for someone who would be taking this course, are at https://docs.google.com/document/d/1rGo98lf3_Bk1yQyF7lcZVAw5OFZ-myivWd2gId0vzAc/edit?usp=sharing .

Other background. Students should have a basic understanding of diffraction and properties of light, and a good grasp of calculus, and matrix arithmetic. Other topics covered in the standard lower-level physics curriculum. Non-physics students can probably brush up on diffraction before the course with a few hours of watching YouTube videos and doing practice exercises. (Maybe these? <https://www.youtube.com/watch?v=1bHipDSHVG4>, <https://www.youtube.com/watch?v=CRhsPoTOHIU>, https://www.youtube.com/watch?v=sTa-Hn_eisw .)

What this course isn't

There are a lot of excellent software packages out there, both open source and commercial. Though we'll touch on some of them, this is not a class on using those tools. Rather, we focus on writing code ourselves. This is immensely useful *even if* one uses others' software packages, because (i) writing code to transform one's data into forms that can be input into other software is often necessary or at least useful; (ii) the variety of scientific questions one can ask far exceeds the variety of pre-made analysis methods that can address them, so writing one's own tools is often necessary to tackle questions of interest; (iii) familiarity with common image processing methods is useful for planning experiments, to better assess what tasks are easy or difficult; (iv) writing code yourself can give a deep understanding of the capabilities and, especially, the pitfalls associated with various algorithms, which can be important to realize even if later using canned software.

Students should also be aware that this course will cover less physics than a typical physics course, less computer science than a typical computer science course, etc. It won't follow as much of a conceptual arc as most courses, and instead will be a collection of topics with which to become acquainted through exercises. Despite all this, people who have taken the course before (either in its official or "informal" versions) seem to have liked it and found it useful.

Topics

The list below is roughly in chronological order, but not exactly. Some topics, like “microscopes and cameras” and “images and ethics” will be interspersed among others.

Introduction / fundamentals / thresholding

In which we consider basic properties of digital images and apply thresholds to enhance contrast

Spatial filtering

In which we examine various local filters

The point-spread function, simulated images, noise

In what way is an image different from the reality it represents? How is noise manifested, and how does that limit what we can learn from an image?

Microscopes and cameras

We look at the physics that underlies aspects of image formation, and therefore image analysis

Sub-pixel localization

How can we determine the location of a point source (such as a single molecule, a colloidal particle, or a star)? In addition to being the key issue underlying super-resolution microscopy, this is applicable to many “tracking” problems. We’ll also learn how statistical principles constrain the accuracy of localization.

Maximum Likelihood Estimation

In which we learn and apply a bit of statistics

De-noising

Can we remove noise from images? How? Why?

Morphological image processing

In which we familiarize ourselves with a useful “toolkit” of image processing operations

Hough Transformation

A clever, useful approach for finding and analyzing geometric features.

Image segmentation

One of the most important, and hardest, problems: what are the “objects” in an image, and how can we identify them – not just their centers, but their size and shape? We’ll consider approaches including “region based” methods, especially watershed segmentation.

Deconvolution

Can we reconstruct a higher resolution image than the one we have?

Machine learning methods (an introduction)

Can we get the computer to do our thinking for us? How? A *very limited* introduction to a very large topic

Images and ethics

When is it ok to “manipulate” images? How do we detect fraud in research papers? Does machine learning introduce different ethical concerns than other methods?

Logistics

CLASS TIMES	TuTh 2:00 - 3:50 pm, Willamette 318 We'll probably use the second half of Tuesday's classes (3:00-3:50 pm) as an optional office hours / question period. We will also likely end a bit early on Thursdays because of the Physics colloquium (4pm).
INSTRUCTOR	Professor Raghuv​eer Parthasarathy (Par-tha-sa-ra-thē) Email: raghu@uoregon.edu Office: 362 Willamette Hall
TEACHING ASSISTANT	Tom Gorordo Email: tgorordo@uoregon.edu Office: 463 Willamette Hall
OFFICE HOURS	Prof. Parthasarathy: Monday 1:00-1:50 pm, Wed. 1:30-2:20 pm; WIL 362 Tom Gorordo: Tues. 4:00-4:50 pm, Wed. 4:00-4:50 pm; WIL 463 <i>Office hour times may change, both by request (if particular times are not good for many students) and due to scheduling conflicts that arise.</i>
CANVAS	We will be using Canvas in this course distribute and collect course materials. Log on to canvas.uoregon.edu using your DuckID. If you have questions about accessing and using Canvas, visit the Canvas support page . Canvas and Technology Support also is available at: 541-346-4357 & livehelp.uoregon.edu If you face Internet access challenges, visit Information Services' web page on going remote . If sharing files via Canvas is cumbersome, we may make use of Dropbox, OneDrive, or other services.
TEXTBOOK	There is no required textbook. The lectures and supplemental readings will be sufficient. The following are useful books: <ul style="list-style-type: none"> • <i>Digital Image Processing</i>, 4th edition, by Rafael C. Gonzalez, Richard E. Woods (Pearson; 2018) • <i>Digital Image Processing Using MATLAB</i>, 2nd edition, by Rafael C. Gonzalez, Richard E. Woods, and Steven L. Eddins (Gatesmark Publishing; 2009) • <i>Foundations of Computer Vision</i> by Antonio Torralba, Phillip Isola and William T. Freeman (MIT Press; 2024). A very recent book, with excellent descriptions of machine learning / neural networks. Available in print and as an e-book from UO Libraries.

[...]

Course structure and assignments

<p>HOMEWORK</p>	<p>There will be homework assignments approximately every week. You are encouraged to discuss the questions with others, but of course, the work you submit and the code you write should be your own.</p> <p>Assignments will be submitted online, via Canvas.</p> <p>The assignments will mostly involve writing programs. In many cases, the result will be an image or graph that you submit, along with text, and it will not be necessary to submit your code. In some cases, we will ask for the code itself. Don't hesitate to ask if instructions are unclear.</p> <p>Be prepared to explain and discuss your homework results in class.</p> <p>Each student has a total of four free late days to use throughout the term. Each late day allows you to turn in a single homework assignment up to 24 hours past the deadline with no penalty. There is no restriction on the number of free late days you may use per assignment. You do not need to ask to use free late days; simply turn in the assignment when you are done, and we will note that you used a free late day when grading. Once you have exhausted your free late days, no late homework will be accepted.</p>
<p>CLASS PARTICIPATION</p>	<p>I'm a strong proponent of active learning, rather than having classes filled with monotonous lecturing, as a tool for fostering learning and engagement. In this course especially, active participation is crucial – discussing what worked and didn't work in exercises, or other observations, allows everyone to benefit from each other's experiences, and this is a topic that is very much learned through experience. Therefore, there is a class participation grade. You will be expected to contribute to discussions, and I will gently let you know if you are consistently not contributing.</p>
<p>FINAL PROJECT</p>	<p>There will be a final project that involves investigating and implementing an image analysis method, submitting a written report on it, and giving a presentation to the class. This will be done in groups, with details described later in the course.</p>
<p>EXAMS</p>	<p>There will be no exams.</p>
<p>GRADING</p>	<p>The various grade components and their weights for the final grade are:</p> <ul style="list-style-type: none"> • <i>Weekly Homework Assignments: 70%</i> • <i>Class participation: 10%</i> • <i>Final Project: 20%</i> <p>Overall Grade:</p> <p>A=[90, 100]%; B=[80,90)%; C=[70,80)%; D=[60,70)%; F<60%.</p>
<p>410/510 GRADING</p>	<p>This is both an undergraduate (410) and graduate (510) course. Homework assignments and the requirements for the final project will be slightly different for the two courses.</p>

Other policies

ACADEMIC INTEGRITY	Students are expected to abide by university policies on academic honesty, avoiding unauthorized help on assignments and examinations, the use of sources without acknowledgment plagiarism, fabrication, and cheating of all types. The Student Conduct Code (https://dos.uoregon.edu/conduct) provides definitions of these terms and explanations of the university policy on the subject. I take academic misconduct very seriously, as it is disrespectful to your fellow students, your instructor, and society. I will report misconduct to the Office of Student Conduct and Community Standards—consequences can include failure of the course.
STUDENTS WITH DISABILITIES	All of us at the University of Oregon are working to create inclusive learning environments. Please notify me if there are aspects of the instruction or design of this course that result in disability-related barriers to your participation. You are also encouraged to contact the Accessible Education Center at 541-346-1155 or uoac@uoregon.edu .
CHANGES TO THE SYLLABUS	Course requirements, deadlines, and grading percentages are subject to change, but any changes will be discussed and explained. I will be mindful of the impacts any unfolding events may be having on you. I encourage you to talk with me about what you are experiencing so we can work together to help you succeed in this course.