Computer Science EN.601.465 / EN.601.665
Natural Language Processing
Fall, 2021 (4 credits)

Instructor
Professor Jason Eisner <jason@cs.jhu.edu>
Office hours: After class in the classroom, or by appointment
Office Zoom link: https://wse.zoom.us/my/jeisner1

Teaching Assistants
TA: Zhichu (Brian) Lu <zlu39@jhu.edu>
CA: Chenyu (Heidi) Zhang <czhan105@jhu.edu>
CA: Jingyu (Jack) Zhang <jzhan237@jhu.edu>

Office hours: TBA. See the class website, http://cs.jhu.edu/~jason/465.
You can reach all the course staff at once via a private post on Piazza, or via cs465-staff@cs.jhu.edu.

Meetings
Monday, Wednesday, Friday, 3:00–4:15 pm
Tuesday, 6:00–7:30 pm (recitation)
Class Zoom link: https://wse.zoom.us/j/95850009212
Class Zoom passcode: Available from https://meetinginfo.jhu.edu/. Will also be emailed to registered and waitlisted students before the first class.
This class is in the “flexible time slot” MWF 3-4:30. Please keep the entire slot open. Class will usually end at 4, followed by office hours in the classroom from 4-4:30 (stick around to get your money’s worth!). However, class will sometimes run till 4:15 in order to keep up with the syllabus. I’ll try to give advance notice of these “long classes,” which among other things make up for no-class days when I’m out of town.
We also run a once-per-week recitation led by the prof or the TA. This session is normally held in the Tuesday slot, and will focus on solving problems together. It is meant as an efficient and cooperative way to study for an hour: it reinforces the past week’s class material without adding to your homework load. If you come to discussion session as recommended, you won’t be startled by the exam style—the discussion problems are taken from past exams and are generally interesting.
We’ll occasionally “flip the classroom,” to make better interactive use of our synchronous class meeting times. This means that you’ll watch a lecture video ahead of time on your own, and we’ll instead use the meeting time for Q&A, discussion, enrichment, and/or collaborative problem-solving.

Remote Participation
In-person attendance is currently limited to 49 students because of the campus Covid policy. Preference for these seats goes to students who are enrolled for a grade. Some students may not be able to attend in person:
enrolled students who are unable to travel to Baltimore at this time
enrolled students who are quarantining
waitlisted students who are still hoping to join the course
auditors or unofficial auditors

We hope to allow these students to join the live lectures by Zoom, although it’s not yet clear whether the A/V setup will let them ask questions during class. We will also make all course materials available to them, including recordings of live and pre-recorded lectures, practice problems from recitation (with solutions), homework handouts, and reading handouts.

Textbooks
This course does not follow any textbook. The material is covered by the lectures and videos and the reading handouts that are associated with the homework projects, supplemented by in-class discussion. However, the following textbooks are at an appropriate level. They are usually a good reference to get another written perspective on the material. The website will suggest optional readings from time to time.

- Dan Jurafsky & James Martin (2020), *Speech and Language Processing (3rd ed.*)*. A draft of the 3rd edition is online [here](#).
- Jacob Eisenstein (2019), *Introduction to Natural Language Processing*.
- Chris Manning & Hinrich Schütze (1999), *Foundations of Statistical Natural Language Processing*. A full PDF is available via the JHU library at [this link](#).

Online Resources
Everything you need will be linked from the class homepage, [http://cs.jhu.edu/~jason/465](http://cs.jhu.edu/~jason/465).

Go there now! Make sure to sign up for the Piazza site, and make sure to watch the assigned lecture videos. (And of course, come to class, do the homeworks, take the exams.)

Course Information

- **Catalog description**: This course is an in-depth overview of techniques for processing human language. How should linguistic structure and meaning be represented? What algorithms can recover them from text? And crucially, how can we build statistical models to choose among the many legal answers?

  The course covers methods for trees (parsing and semantic interpretation), sequences (finite-state transduction such as morphology), and words (sense and phrase induction), with applications to practical engineering tasks such as information retrieval and extraction, text classification, part-of-speech tagging, speech recognition and machine translation. There are a number of structured but challenging programming assignments. [Applications]

- **Prerequisites**
  - Data Structures (601.226)
  - Python
  - Basic familiarity with partial derivatives, matrix multiplication, and probabilities

The class aims to be fairly self-contained and teach you everything else you need, presenting it from an NLP perspective. That includes relevant aspects of automata (600.271), probability (553.420/620 or 553.310/311), and machine learning (601.475/675, 601.482/682, ...). So those courses are not formal prerequisites. That said, it may be helpful to have had prior exposure to those concepts.

- **Elective** (Applications)
Course Goals
This course is designed to introduce you to some of the problems and methods of natural language processing, and their relation to linguistics and statistics. At the end you should agree (I hope!) that language is subtle and interesting; feel some ownership over some of NLP’s formal and statistical techniques; and be able to understand research papers in the field. (Caveat: understanding recent research papers may sometimes additional background in machine learning.)

Not to mention all this.

Specific Outcomes for this course:

- Show sensitivity to linguistic phenomena and an ability to model them with formal grammars.
- Understand and carry out proper experimental methodology for training and evaluating empirical NLP systems.
- Be able to manipulate probabilities, construct statistical models over strings and trees, and estimate parameters using supervised and unsupervised training methods.
- Be able to design, implement, and analyze NLP algorithms.

This course will address the following CSAB Criterion 3 Student Outcomes. Graduates of the program will have an ability to:

1. Analyze a complex computing problem and to apply principles of computing and other relevant disciplines to identify solutions.
2. Design, implement, and evaluate a computing-based solution to meet a given set of computing requirements in the context of the program’s discipline.
3. Function effectively as a member or leader of a team engaged in activities appropriate to the program’s discipline.

Course Topics
Please see the class website for a list of topics and resources: http://cs.jhu.edu/~jason/465.

Course Expectations & Grading
We may revisit this breakdown as we get a sense of what is best for our Covid-addled learning situation. Any changes will be announced.

- ≈ 5% participation
- 50% homeworks (see lateness policy)
- 15% midterm exam
- 30% final exam

Participation includes your contributions to an interesting and useful class discussion, whether synchronously during class sessions, or via public posts or replies on our Piazza discussion website. This includes asking questions, of course. We’ve found that participation has a substantial effect on the final grade.

Homeworks will be submitted via Gradescope. We expect to have about 7 homework projects, which focus on different skills you’ll need to do NLP:

1. writing linguistic grammars
2. manipulating probability formulas
3. implementing and properly evaluating supervised models
4. designing and implementing combinatorial algorithms
5. working with formal meaning representations
6. training unsupervised neural models for structured prediction
7. constructing models algebraically
We plan to have two exams—a midterm exam and a final exam. These will have some challenging questions. In the recitations, you’ll get to work in small groups to solve interesting problems from past exams.

Late Homework Policy
Since emergencies sometimes arise, I’ll allow you a total of up to 10 late days during the term. One day = 24 consecutive hours. (Thus, if you are 12 hours late, that counts as 1 whole day.) The full policy and advice on how to use it are at http://cs.jhu.edu/~jason/465/late-policy.html.

Key Dates
Please see the class website: http://cs.jhu.edu/~jason/465.

Assignments & Readings
Please see the class website: http://cs.jhu.edu/~jason/465.
The material below is standard for CS Department syllabi in Fall 2021, except for portions in this color.

**Ethics**

The strength of the university depends on academic and personal integrity. In this course, you must be honest and truthful, abiding by the *Computer Science Academic Integrity Policy*:

Cheating is wrong. Cheating hurts our community by undermining academic integrity, creating distrust, and fostering unfair competition. The university will punish cheaters with failure on an assignment, failure in a course, permanent transcript notation, suspension, and/or expulsion. Offenses may be reported to medical, law or other professional or graduate schools when a cheater applies.

Violations can include cheating on exams, plagiarism, reuse of assignments without permission, improper use of the Internet and electronic devices, unauthorized collaboration, alteration of graded assignments, forgery and falsification, lying, facilitating academic dishonesty, and unfair competition. Ignorance of these rules is not an excuse.

Academic honesty is required in all work you submit to be graded. Except where the instructor specifies group work, you must solve all homework and programming assignments without the help of others. For example, you must not look at anyone else’s solutions (including program code) to your homework problems. However, you may discuss assignment specifications (not solutions) with others to be sure you understand what is required by the assignment.

If your instructor permits using fragments of source code from outside sources, such as your textbook or on-line resources, you must properly cite the source. Not citing it constitutes plagiarism. Similarly, your group projects must list everyone who participated.

Falsifying program output or results is prohibited.

Your instructor is free to override parts of this policy for particular assignments. To protect yourself: (1) Ask the instructor if you are not sure what is permissible. (2) Seek help from the instructor, TA or CAs, as you are always encouraged to do, rather than from other students. (3) Cite any questionable sources of help you may have received.

On every exam, you will sign the following pledge: "I agree to complete this exam without unauthorized assistance from any person, materials or device. [Signed and dated]". Your course instructors will let you know where to find copies of old exams, if they are available.

In this class (NLP), some homeworks will indicate that they do allow collaboration. For a collaborative homework, you are expected to do the work *together*, not divide it up: if you didn’t work on a question, you don’t deserve credit for it! Your solutions should emerge from collaborative discussions with the whole group present (whatever “present” means in online learning).

In this class (NLP), practice exam problems will be provided for you. You should not otherwise make use of homeworks or exams from previous years.

Report any violations you witness to the instructor.

You can find more information about university misconduct policies on the web at these sites:

- For undergraduates:  

- For graduate students:  
  [http://e-catalog.jhu.edu/grad-students/graduate-specific-policies/](http://e-catalog.jhu.edu/grad-students/graduate-specific-policies/)
Personal Wellbeing

- If you are sick, in particular with an illness that may be contagious, notify me by email but do not come to class. Rather, visit the Health and Wellness: 1 East 31 Street, 410-516-8270. See also http://studentaffairs.jhu.edu/student-life/support-and-assistance/absences-from-class/illness-note-policy/
- All students with disabilities who require accommodations for this course should contact me at their earliest convenience to discuss their specific needs. If you have a documented disability, you must be registered with the JHU Office for Student Disability Services (385 Garland Hall; 410-516-4720; http://web.jhu.edu/disabilities/) to receive accommodations.
- If you are struggling with anxiety, stress, depression or other mental health related concerns, please consider visiting the JHU Counseling Center. If you are concerned about a friend, please encourage that person to seek out our services. The Counseling Center is located at 3003 North Charles Street in Suite S-200 and can be reached at 410-516-8278 and online at http://studentaffairs.jhu.edu/counselingcenter/

Classroom Climate

As your instructor, I am committed to creating a classroom environment that values the diversity of experiences and perspectives that all students bring. Everyone here has the right to be treated with dignity and respect. I believe fostering an inclusive climate is important because research and my experience show that students who interact with peers who are different from themselves learn new things and experience tangible educational outcomes. Please join me in creating a welcoming and vibrant classroom climate. Note that you should expect to be challenged intellectually by me, the TAs, and your peers, and at times this may feel uncomfortable. Indeed, it can be helpful to be pushed sometimes in order to learn and grow. But at no time in this learning process should someone be singled out or treated unequally on the basis of any seen or unseen part of their identity.

If you ever have concerns in this course about harassment, discrimination, or any unequal treatment, or if you seek accommodations or resources, I invite you to share directly with me or the TAs. I promise that we will take your communication seriously and to seek mutually acceptable resolutions and accommodations. Reporting will never impact your course grade. You may also share concerns with the Department Head (Randal Burns, randal@cs.jhu.edu), the Director of Undergraduate Studies (Joanne Selinski, joanne@cs.jhu.edu), the Assistant Dean for Diversity and Inclusion (Darlene Saporu, dsaporu@jhu.edu), or the Office of Institutional Equity (oie@jhu.edu). In handling reports, people will protect your privacy as much as possible, but faculty and staff are required to officially report information for some cases (e.g. sexual harassment).