

ITP-259: Basics of Artificial Intelligence

Units: 4 Term: Spring 2025 Time:

Location:

Instructor: Dr. Martin Prescher Contact Info: <u>prescher@usc.edu</u> Office Hours:

- In-Person:
- Virtual:
- Virtual Office hour time slots need to be booked using this link (a Zoom link will be sent upon booking):

https://calendly.com/prescher-usc/ai-office-hours-virtual-zoom

It is recommended to also book time slots for in-person office hours to 'jump the line'. Otherwise it is first come first serve.

Learning Assistants:

Learning Labs: Time and location posted on Brightspace

IT Help: Viterbi IT Hours of Service: Monday – Friday, 8:30 a.m. – 5:00 p.m. Contact Info: DRB 205 (213) 740-0517 <u>engrhelp@usc.edu</u>

Course Description

Learn the basics of artificial intelligence. Explore how machines can be programmed to learn. Appreciate the state-of-the-art as well as the challenges in AI. Understand the gains made in artificial narrow intelligence (ANI).

Catalogue Description

Concepts and techniques in artificial intelligence. Approaches in artificial narrow intelligence. Applications of AI that are already changing society. Ethical and social questions in AI.

Learning Objectives

After completing the course students will be able to:

- Define artificial intelligence and its use cases
- Describe latest advances in artificial intelligence
- Explain the basics of specific AI techniques and their applications
- Think critically about the impact (benefits and challenges) of AI on the world

Prerequisite(s): ITP 115 or ITP 116 and ITP-216

Course Notes

Grading Type: Letter grade

Lecture slides, assignments, readings, announcements, and other class information will be posted on Blackboard.

Outside of class time, course discussion will take place on Piazza.

Technological Proficiency and Hardware/Software Covered

Platforms, packages, and tools that will be used for the class will be provided to students free of charge. The following software will be used in the class. Students should have access to a computer.

- Python packages Keras, TensorFlow, scikit-learn https://www.tensorflow.org/
- Pycharm https://www.jetbrains.com/pycharm/
- Google Colab <u>https://colab.research.google.com/</u>
- Recast.ai chatbot <u>https://cai.tools.sap/</u>
- NLTK <u>https://www.nltk.org/</u>

Required Readings and Supplementary Materials

Readings will be assigned from the following sources. All books are available online through USC Safari Books. <u>https://libraries.usc.edu/databases/safari-books</u>

- Getting Started with Artificial Intelligence, by Tom Markiewicz, Josh Zheng ("MZ")
 - Publisher: O'Reilly Media, Inc.
 - Release Date: October 2020
- Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, 2nd Edition, by Aurélien Géron ("AG")
 - Publisher: O'Reilly Media, Inc.
 - Release Date: September 2019
- Hands-On Artificial Intelligence for Beginners, by Patrick D. Smith ("PS")
 - Publisher: Packt Publishing
 - Release Date: October 2018

Description and Assessment of Assignments, Grading Breakdown

The Course Schedule on Page 4 describes assignment description and due dates.

Assignment	% of grade
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Homework	30
Exam 1	30
Exam 2	30
Final Project	10
TOTAL	100

Homework will be assessed on timeliness, completeness, thoroughness, and conclusions drawn. Each component of the assessment will have points associated with it. This will vary for each assignment and will be listed in the homework instructions. Grader will provide feedback for each component of the assessment through Blackboard.

Homework instructions will be posted on Blackboard with a due date, objective, expectations, data files, resources, and deliverables. In most cases, the deliverable will be answers in a Word document along with source code and any other output files.

Assignment Submission Policy

Students are responsible for completing individual assignments by stated deadlines. Assignments must be turned in to Blackboard. Assignments turned in late will have 25% of the total points deducted from the graded score for each late day. You will have 3 "grace days" for the semester. i.e. no late penalty on HW for a cumulative 3 days. No grace days are available for Project.

Grading Timeline

Assignments Grading will typically be completed 7 days after the deadline for submission. Any variations will be announced in class or on blackboard. All grading discrepancies must be resolved within a week of grade release.

For grade disputes, contact your grader first. If resolution is not reached, contact the instructor.

Additional Policies

No make-up exams (except for documented medical or family emergencies or religious observance) will be offered. If student will not be able to attend an exam due to an athletic game or other valid reason, then they must coordinate with the instructor before the exam is given. Student may arrange to take the exam before they leave, with an approved university personnel during the time they are gone, or within the week the exam is given. If students do not take an exam, they will receive a 0 for the exam.

If students need accommodations authorized by OSAS (*Office of Student Accessibility and Services*), notify the instructor at least two weeks before the exam. This will allow time for arrangements to be made.

Course Content Distribution and Synchronous Session Recordings Policies

USC has policies that prohibit recording and distribution of any synchronous and asynchronous course content outside of the learning environment.

Recording a university class without the express permission of the instructor and announcement to the class, or unless conducted pursuant to an Office of Student Accessibility Services (OSAS) accommodation. Recording can inhibit free discussion in the future, and thus infringe on the academic

freedom of other students as well as the instructor. (<u>Living our Unifying Values: The USC Student</u> <u>Handbook</u>, page 13).

Distribution or use of notes, recordings, exams, or other intellectual property, based on university classes or lectures without the express permission of the instructor for purposes other than individual or group study. This includes but is not limited to providing materials for distribution by services publishing course materials. This restriction on unauthorized use also applies to all information, which had been distributed to students or in any way had been displayed for use in relationship to the class, whether obtained in class, via email, on the internet, or via any other media. (Living our Unifying Values: The USC Student Handbook, page 13).

Policy on the use of AI generators

In this course, I encourage you to use artificial intelligence (AI)-powered programs to help you with assignments that indicate the permitted use of AI. You should also be aware that AI text generation tools may present incorrect information, biased responses, and incomplete analyses; thus they are not yet prepared to produce text that meets the standards of this course. To adhere to our university values, you must cite any AI-generated material (e.g., text, images, etc.) included or referenced in your work and provide the prompts used to generate the content. Using an AI tool to generate content without proper attribution will be treated as plagiarism and reported to the Office of Academic Integrity. Please review the instructions in each assignment for more details on how and when to use AI Generators for your submissions.

Course Schedule

Exam 1 (Midterm): **Oct 16,** during class time Exam 2 (final): **Dec 4,** during class time

1 Course introduction What is Al? 2 History of Al,1950 – present Landmark achievements Al categories – ANI, AGI, ASI Where is Al today? 3 Turing test Basics of Al approaches 4 Rule-based approach to Al Heuristics approach to Al Supervised Al Nuspervised Al Neural networks 7 Perceptron Deep neural networks 8 Training DNN Midterm 9 Keras, TensorFlow Computer vision 10 CNN Convolutional neural networks 11 Training CNN CNN for image classification 12 Natural language processing NLP contd. 13 Tokenization Embeddings 14 Bias in Al/ML, Ethics in Al. Guest speaker. Recurrent Neural Networks Review Exam II FileNal Proiet due	Week	Topics/Daily Activities
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Statement on Academic Conduct and Support Systems

Academic Integrity:

The University of Southern California is a learning community committed to developing successful scholars and researchers dedicated to the pursuit of knowledge and the dissemination of ideas. Academic misconduct, which includes any act of dishonesty in the production or submission of academic work, compromises the integrity of the person who commits the act and can impugn the perceived integrity of the entire university community. It stands in opposition to the university's mission to research, educate, and contribute productively to our community and the world.

All students are expected to submit assignments that represent their own original work, and that have been prepared specifically for the course or section for which they have been submitted. You may not submit work written by others or "recycle" work prepared for other courses without obtaining written permission from the instructor(s).

Other violations of academic integrity include, but are not limited to, cheating, plagiarism, fabrication (e.g., falsifying data), collusion, knowingly assisting others in acts of academic dishonesty, and any act that gains or is intended to gain an unfair academic advantage.

The impact of academic dishonesty is far-reaching and is considered a serious offense against the university. All incidences of academic misconduct will be reported to the Office of Academic Integrity and could result in outcomes such as failure on the assignment, failure in the course, suspension, or even expulsion from the university.

For more information about academic integrity see <u>the student handbook</u> or the <u>Office of Academic</u> <u>Integrity's website</u>, and university policies on <u>Research and Scholarship Misconduct</u>.

Please ask your instructor if you are unsure what constitutes unauthorized assistance on an exam or assignment, or what information requires citation and/or attribution.

Students and Disability Accommodations:

USC welcomes students with disabilities into all of the University's educational programs. <u>The Office of Student Accessibility Services</u> (OSAS) is responsible for the determination of appropriate accommodations for students who encounter disability-related barriers. Once a student has completed the OSAS process (registration, initial appointment, and submitted documentation) and accommodations are determined to be reasonable and appropriate, a Letter of Accommodation (LOA) will be available to generate for each course. The LOA must be given to each course instructor by the student and followed up with a discussion. This should be done as early in the semester as possible as accommodations are not retroactive. More information can be found at <u>osas.usc.edu</u>. You may contact OSAS at (213) 740-0776 or via email at <u>osasfrontdesk@usc.edu</u>.

Support Systems:

Counseling and Mental Health - (213) 740-9355 – 24/7 on call

Free and confidential mental health treatment for students, including short-term psychotherapy, group counseling, stress fitness workshops, and crisis intervention.

<u>988 Suicide and Crisis Lifeline</u> - 988 for both calls and text messages – 24/7 on call

The 988 Suicide and Crisis Lifeline (formerly known as the National Suicide Prevention Lifeline) provides free and confidential emotional support to people in suicidal crisis or emotional distress 24 hours a day, 7 days a week, across the United States. The Lifeline is comprised of a national network of over 200 local crisis centers, combining custom local care and resources with national standards and best practices. The new, shorter phone number makes it easier for people to remember and access mental health crisis services (though the previous 1 (800) 273-8255 number will continue to function indefinitely) and represents a continued commitment to those in crisis.

<u>Relationship and Sexual Violence Prevention Services (RSVP)</u> - (213) 740-9355(WELL) – 24/7 on call Free and confidential therapy services, workshops, and training for situations related to gender- and power-based harm (including sexual assault, intimate partner violence, and stalking).

Office for Equity, Equal Opportunity, and Title IX (EEO-TIX) - (213) 740-5086

Information about how to get help or help someone affected by harassment or discrimination, rights of protected classes, reporting options, and additional resources for students, faculty, staff, visitors, and applicants.

Reporting Incidents of Bias or Harassment - (213) 740-5086 or (213) 821-8298

Avenue to report incidents of bias, hate crimes, and microaggressions to the Office for Equity, Equal Opportunity, and Title for appropriate investigation, supportive measures, and response.

The Office of Student Accessibility Services (OSAS) - (213) 740-0776

OSAS ensures equal access for students with disabilities through providing academic accommodations and auxiliary aids in accordance with federal laws and university policy.

USC Campus Support and Intervention - (213) 740-0411

Assists students and families in resolving complex personal, financial, and academic issues adversely affecting their success as a student.

Diversity, Equity and Inclusion - (213) 740-2101

Information on events, programs and training, the Provost's Diversity and Inclusion Council, Diversity Liaisons for each academic school, chronology, participation, and various resources for students.

<u>USC Emergency</u> - UPC: (213) 740-4321, HSC: (323) 442-1000 – 24/7 on call

Emergency assistance and avenue to report a crime. Latest updates regarding safety, including ways in which instruction will be continued if an officially declared emergency makes travel to campus infeasible.

<u>USC Department of Public Safety</u> - UPC: (213) 740-6000, HSC: (323) 442-1200 – 24/7 on call Non-emergency assistance or information.

Office of the Ombuds - (213) 821-9556 (UPC) / (323-442-0382 (HSC)

A safe and confidential place to share your USC-related issues with a University Ombuds who will work with you to explore options or paths to manage your concern.

Occupational Therapy Faculty Practice - (323) 442-2850 or otfp@med.usc.edu

Confidential Lifestyle Redesign services for USC students to support health promoting habits and routines that enhance quality of life and academic performance.