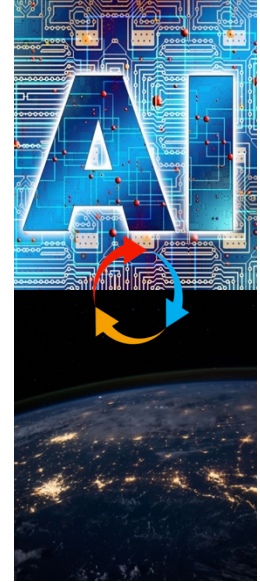


# GEOG 4591/6591: Introduction to Geospatial Artificial Intelligence

## Course Description

The rapid increase in high-quality data, advanced machine learning algorithms, and the availability of fast hardware have largely contributed to a renewed interest in Artificial Intelligence (AI) from academia, industry, and the general public in recent years. From a geospatial point-of-view, GeoAI, as an interdisciplinary field of Geography/GIScience and AI, advocates the idea of developing and utilizing AI techniques to address a variety of problems related to both the natural environment and our human society. In this course, we will AI and machine learning methods and will focus on their applications on geospatial problems. We will start from the fundamental techniques related to preparing GIS data for machine learning models, and then move to discussing machine learning concepts and how to use machine learning to address spatial problems. This course has an emphasis on both theoretical concepts and hands-on skills, and we will use a high-level programming language (e.g., Python) to implement machine learning models.

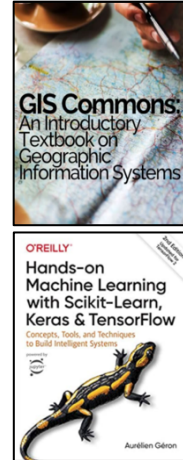


## Course Information

- **Course number:** GEOG 4591/6591
- **Credits:** 3
- **Lectures:** TR, 11:10-12:25, Geography-Geology Building, Room 311
- **Instructor:** Dr. Gengchen Mai
  - **Email:** [gengchen.mai25@uga.edu](mailto:gengchen.mai25@uga.edu)
  - **Office hours:** Tuesday 10:00 – 11:00
  - **Office:** Geography-Geology Building, Room 312
- **Teaching Assistant:** Hao Yang
  - **Email:** [hy96161@uga.edu](mailto:hy96161@uga.edu)
  - **Office hours:** Tuesday 14:00 – 15:00
  - **Office:** Geography-Geology Building, Room 321B

## Textbook

- Michael Schmandt (2009): GIS Commons: An Introductory Textbook on Geographic Information Systems (free textbook at <https://giscommons.org/>; referred as “GIS” in the syllabus)
- Aurélien Géron (2019): Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems (ebook available from UB library; referred as “ML” in the syllabus)



## Course Website

The website of this course is available on UGA ELC.

## Prerequisites

This course does not have a formal prerequisite yet. However, before taking this upper level course, you may consider taking the following courses first:

- GIS background: Geog4470/6470, Geog4300/6300 or equivalent
- Basic programming background: CSCI 1301, C1360, Geog4590/6590, or equivalent

## Software Environment

- Anaconda (a python-based software suite)
- Jupyter Notebook.

They have been installed on lab computers for you to use. However, students are also expected to download and install them on their personal computers. Both are free packages.

## Tentative Course Schedule\*

Week	Date	Topics	Reading
Week 1	08/17	- Course introduction I	Python 1, 2, 3
Week 2	08/22 08/24	- Course introduction I - Python, Jupyter Notebook, and Google Colab	Python 5, 6, 7
Week 3	08/29 08/31	- Python fundamentals - Lab 1: Use geopy to calculate geodesic distances between locations	Python 8, 9, 10
Week 4	09/05 09/07	- Python fundamentals II - Spatial data models & formats - Lab 1: Continue ( <b>Paper #1 Presentation</b> )	GIS Concepts (Ch. 1)
Week 5	09/12 09/14	- Vector data analysis - Raster data analysis ( <b>Paper #2 Presentation</b> )	GIS Data Model: Vector (Ch. 1)
Week 6	09/19 09/21	- Lab 2: vector data analysis - Lab 2: Continue ( <b>Paper #3 Presentation</b> )	GIS Data Model: Raster (Ch. 1)
Week 7	09/26 09/28	- Machine learning fundamentals - Lab 3: raster data analysis ( <b>Paper #4 Presentation</b> )	ML Preface, 1
Week 8	10/03 10/05	- Integrating GIS and ML in geospatial projects - Lab 3: Continue ( <b>Paper #5 Presentation</b> )	ML 4
Week 9	10/10 10/12	- Regression - <b>ClimaX geo-foundation model talk by Guest Speaker, Tung Nguyen, from UCLA CS [No class on 10/12]</b> - <b>Project Proposal Due</b>	ML 2
Week 10	10/17 10/19	- Lab 4: Making predictions using GIS, regression, and machine learning - Lab 4 Continue & Project Consultant ( <b>Paper #6 Presentation</b> )	ML 6
Week 11	10/24 10/26	- Decision tree and random forest I ( <b>Paper #7 Presentation</b> ) - Decision tree and random forest II ( <b>Paper #8 Presentation</b> )	ML 7
Week 12	10/31 11/02	- Lab 5: Integrating GIS and random forest for land use classification on remote sensing imagery	ML 9

		- Lab 5 Continue & Project Consultant ( <b>Paper #9 Presentation</b> )	
Week 13	11/07 11/09	- Geospatial clustering I - Geospatial clustering II ( <b>Paper #10 Presentation</b> )	ML 9
Week 14	11/14 11/16	- Neural Network ( <b>Paper #11, #12 Presentation</b> ) - Embedding and Location Representation Learning - Course Summary ( <b>Paper #13 Presentation</b> )	
Week 15	11/21 11/23	<b>Thanksgiving – No classes</b>	
Week 16	11/28 11/30	- Lab 6: Geospatial clustering with geotagged photo location data - Lab 6: Continue & Project Consultant	
Week 17	12/05	- <b>Final Project Presentation</b>	

**\*Course project paper is due on Dec 10, 2023, 12:00 pm (at noon).**

## Course Requirements

### 1. Lab assignments (40%)

There will be 6 lab assignments in total led by TA. You will learn how to use different machine learning models in different geospatial projects.

### 2. Student research paper presentation (10%) and participation (5%)

Each student will read one scientific paper on relevant topic and present it in class (**8 min presentation + 2 min Q&A**). You will receive a paper list early in the semester to choose the paper to present and the date of presentation:

- Undergraduate: each student can pick a general GeoAI review paper
- Graduate: each student can pick a GeoAI technical paper

Please submit the **PDF file of your presentation**. To share with the class, please also **post the original article on a designated discussion board**. The due date is your presentation day. For technical papers, your presentation should introduce the research problem, research objective, methodology, and findings/results addressed in the article, as well as your critique if any. The presentation will be graded by the instructor.

### 3. Invited Talk Summary (5%)

There will be 2+ invited talks at UGA GeoAI talk series. All students are required to attend at least one of them and submit a **one-page summary (12 font, 1.5 line space) of one invited talk** after it.

#### 4. Course project (40%)

The course project can be about using any GeoAI models on a specific geospatial research problem. The project needs to be conducted by:

- A team of no more than 3 undergraduate students;
- Or one graduate student.

#### Submission:

- 1) Write a project proposal (**1-2 pages, dual in Week 9**). The instructor will provide feedback on the proposed project idea. **(10%)**
- 2) Finish the proposed project and submit a scientific paper of it (**4-12 pages including references**). Submit the paper in ELC. **(20%)**
- 3) Make a final presentation of the project. **(10%)**

Both project proposal and final paper should use [IJGIS word template](#) or [LaTeX template](#). Please **submit a single PDF for each submission**. Note that **if you are planning to submit your paper to a journal and has a idea of the venue, you can directly use the journal template you are targeting at.**

The presentation will be graded by all other students, teaching assistant, and the instructor. The weight for each group is - **all other students (40%), teaching assistant (30%), and the instructor (30%)**. The **maximum and minimum score from students will not be considered.**

#### Grade weighting scheme

Component	Weight
Lab assignment (6)	40%
Research Paper Presentation (1)	10%
Participation	5%
Invited Talk Summary (1)	5%
Course project proposal	10%

<b>Course project paper</b>	20%
<b>Course project presentation</b>	10%

Final Letter Grades: The final letter grade will be determined according to the scale below.

<b>A : [90,100]</b>	<b>C : [74, 77)</b>
<b>A-: [88,90)</b>	<b>C-: [70, 74)</b>
<b>B+: [86, 88)</b>	<b>D+: [67, 70)</b>
<b>B : [83, 86)</b>	<b>D : [64, 67)</b>
<b>B-: [80, 83)</b>	<b>D-: [60, 64)</b>
<b>C+: [77, 80)</b>	<b>F: &lt;60</b>

**Due Dates:** All Assignments must be completed on time. Submittal of assignments after due dates is accepted but with a penalty as 10% of the percentage grade for each day they are late (note: anytime passing the due time will be counted as late for one day, and anytime passing the first late day will be counted as late for two days, and so forth). Submissions that are late for more than 5 days will not be accepted. Late in-class assignments will not be accepted. In each case, exceptions are possible only with documentation of a medical or family emergency.

**Academic Honesty:** Cheating and plagiarism will automatically earn zero (0) points for the assignment or exam. All academic work must meet the standards contained in “A Culture of Honesty.” Each student is responsible to inform themselves about those standards before performing any academic work.