

# Syllabus

## DCP4300/URP6931 Artificial Intelligence in the Built Environment

### 2026 Spring

#### Instructor

Shenhai Wang, Ph.D. ([Google Scholar](#))

Director of [Urban AI Lab @UFL](#)

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Tuesday afternoon: 1:55pm – 4:55pm

Location: TBD

Office hour: TBD

#### Course description

This course introduces how to investigate the built environment with artificial intelligence (AI). The course consists of three modules with Module 1 focused on computational basics, Module 2 on diverse data in built environment, and Module 3 on AI techniques. Specifically, Module 1 introduces Python, GitHub, Google Colab, among other computational toolboxes. Module 2 presents diverse data sources, including the conventional data, such as census and travel survey, and the unconventional data, such as imagery, networks, and texts. Module 3 introduces four dominant AI models, including artificial neural networks (ANN) for numeric data, convolutional neural networks (CNN) for imagery, graph neural networks (GNN) for spatial networks, and LSTM and large language models (LLM) for texts. Students will learn the major Python packages, e.g., Pandas, GeoPandas, PyTorch, thus processing and analyzing the conventional and unconventional data structures. Through reading materials, students will also learn AI as a general analytical framework for economic development, urban mobility, sustainability, energy consumption, design, and community development. This course focuses on enhancing students' hands-on experiences in AI for built environment, and empowering designers, planners, engineers, and data scientists to leverage AI to dissect cities and tackle enduring challenges in the built environment.

#### Course prerequisites

No prerequisite course is needed. However, students need to take Practicum AI at UFL as a concurrent requirement. The prior Python coding skills are encouraged, and the prior knowledge in probability, statistics, and linear algebra can also facilitate your learning experiences. If you have questions, please contact the instructor to discuss your qualification. This course is the prerequisite for the Intermediate Urban Analytics in Spring 2025.

#### Coding

This course uses Python as the programming language. Students are expected to use [Google Colab](#) in the lab sessions. You don't need to install Python and other libraries on your laptop because Colab provides a standardized coding platform. The course will teach Python coding modules including [Numpy](#), [Pandas](#), [Matplotlib](#), [Statsmodels](#), [GeoPandas](#), [Scikit-Learn](#), and [PyTorch](#).

#### Textbook

No textbook is required. The following textbooks are recommended as references.

1. VanderPlas, J. (2016). Python data science handbook: Essential tools for working with data. O'Reilly Media, Inc. GitHub link [here](#).
2. Bishop, C. M. (2006). Pattern recognition and machine learning. GitHub link [here](#).

3. Goodfellow, I., Bengio, Y., & Courville, A. (2016). Deep learning: MIT press Cambridge.

## Course schedule

Week	Dates	Lectures	Lab sessions	Psets	Project
1	Jan 13	Class overview			
<b>Module 1. Computational Basics</b>					
2	Jan 20	Python, Colab, and data sources	Lab 02. Python in Colab		
3	Jan 27	Python, GitHub, and data processing	Lab 03. Data collection and processing		
<b>Module 2. Diverse Data Sources</b>					
4	Feb 03	Data description and visualization	Lab 04. Matplotlib, Seaborn, and Pandas		
5	Feb 10	Correlational analysis and urban application I: Gasoline consumption and cities	Lab 05. Gasoline consumption and cities	pset 1 out	
6	Feb 17	Spatial network and visualization	Lab 06. Visualizing property values		Idea guidance out;
7	Feb 24	Urban imagery and text data	Lab 07. Downloading imagery with Google API	pset 1 due	
8	Mar 03	Midterm presentation for project ideas	NA	pset 2 out	Idea due;
<b>Module 3. AI Methods and Applications</b>					
9	Mar 10	Modeling basics and artificial neural networks	Lab 09. Predicting property values		
10	Mar 24	Convolutional neural networks for urban imagery	Lab 10. Dissecting satellite imagery	pset 2 due	Proposal guideline out
11	Mar 31	Semantic segmentation and generative AI	Lab 11. Semantic segmentation for street-view images	pset 3 out	
<b>Module 4. Final Project</b>					
12	April 07	Proposal presentation	NA		Proposal presentation; proposal due. Final report guidance out.
13	April 14	Everything everywhere all at once: a comprehensive review	NA	pset 3 due	
15	April 21	Final Presentation			Final presentation and final report due

Note: This schedule is **subject to changes**.

## Four Required Reading Tasks

1. Week 5 (**Required**). Newman, P. W., & Kenworthy, J. R. (1989). Gasoline consumption and cities: a comparison of US cities with a global survey. *Journal of the American Planning Association*, 55(1), 24-37.

2. Week 9 (**Required**). Wang, S., Mo, B., Hess, S., & Zhao, J. (2021). Comparing hundreds of machine learning classifiers and discrete choice models in predicting travel behavior: an empirical benchmark. arXiv preprint arXiv:2102.01130.
3. Week 10 (**Required**). Jean, N., Burke, M., Xie, M., Davis, W. M., Lobell, D. B., & Ermon, S. (2016). Combining satellite imagery and machine learning to predict poverty. *science*, 353(6301), 790-794.
4. Week 11 (**Required**). Xue, J., Jiang, N., Liang, S., Pang, Q., Yabe, T., Ukkusuri, S. V., & Ma, J. (2022). Quantifying the spatial homogeneity of urban road networks via graph neural networks. *Nature Machine Intelligence*, 4(3), 246-257.

## Course communication

The instructor can be reached through Canvas inbox or by email. Expect a response within 48 hours, excluding holidays and weekends. General questions can be posted to the Canvas class website discussion board.

## Details in Course Schedule

Practicum AI provides a wonderful overview for using Python, Colab, GitHub, and other tools in computation. It is highly recommended for the students to read the materials.

- Link to Practicum AI: <https://practicumai.org/>
- GitHub repository: <https://github.com/PracticumAI>

## Module 1. Computational basics

- Google Colab: Google Tutorial about Colab features
  - [https://colab.research.google.com/notebooks/basic\\_features\\_overview.ipynb](https://colab.research.google.com/notebooks/basic_features_overview.ipynb)
- GitHub: Getting started with GitHub
  - <https://docs.github.com/en/get-started/quickstart/hello-world>
- Python basics.
  - 30-min crash course: <https://github.com/srebalaji/python-crash-course>
  - 4-hour crash course: <https://github.com/Python-Crash-Course/Python101>
  - MIT Open Course for Python Programming: <https://ocw.mit.edu/courses/6-0001-introduction-to-computer-science-and-programming-in-python-fall-2016/>
- Zotero.
  - Website: <https://www.zotero.org/>
- Overleaf
  - Tutorials: <https://www.overleaf.com/learn/latex/Tutorials>

## Module 2: Diverse Data Sources

- Python statsmodels for regressions
  - Getting started: <https://www.statsmodels.org/stable/gettingstarted.html>
  - Linear regression example: <https://www.statsmodels.org/stable/regression.html>
- Python Pandas and Matplotlib for data description and visualization
  - Pandas and Matplotlib: Chapters 3 and 4 in VanderPlas, J. (2016) Python data science handbook: Essential tools for working with data.
  - Python tutorials: Chapters 3 and 4 in the GitHub repository: <https://github.com/jakevdp/PythonDataScienceHandbook/tree/master/notebooks>
- Data Sources.
  - American Community Survey: <https://www.census.gov/programs-surveys/acs>
  - Census Geography and TIGER Lines: <https://www.census.gov/geographies/mapping-files/time-series/geo/tiger-line-file.html>

- Readings
  - **(Required)** Newman, P. W., & Kenworthy, J. R. (1989). Gasoline consumption and cities: a comparison of US cities with a global survey. *Journal of the American Planning Association*, 55(1), 24-37.

### Module 3: AI Applications

- Python Scikit-learn for machine learning.
  - Scikit learn tutorial: <https://scikit-learn.org/stable/>
  - Supervised learning tutorial in Scikit-learn: [https://scikit-learn.org/stable/supervised\\_learning.html#supervised-learning](https://scikit-learn.org/stable/supervised_learning.html#supervised-learning)
  - Unsupervised learning tutorial in Scikit-learn: [https://scikit-learn.org/stable/unsupervised\\_learning.html](https://scikit-learn.org/stable/unsupervised_learning.html)
  - GitHub tutorials for Chapters 4, 5, and 7 for Bishop, C. M. (2006): <https://github.com/ggerm/prml>
- Python PyTorch for deep learning.
  - Official webpage: <https://pytorch.org/>
  - PyTorch examples: [https://pytorch.org/tutorials/beginner/pytorch\\_with\\_examples.html](https://pytorch.org/tutorials/beginner/pytorch_with_examples.html)
- Readings
  - **(Required)** Wang, S., Mo, B., Hess, S., & Zhao, J. (2021). Comparing hundreds of machine learning classifiers and discrete choice models in predicting travel behavior: an empirical benchmark. *ArXiv:2102.01130*.
  - **(Required)**. Xue, J., Jiang, N., Liang, S., Pang, Q., Yabe, T., Ukkusuri, S. V., & Ma, J. (2022). Quantifying the spatial homogeneity of urban road networks via graph neural networks. *Nature Machine Intelligence*, 4(3), 246-257.
  - **(Required)**. Jean, N., Burke, M., Xie, M., Davis, W. M., Lobell, D. B., & Ermon, S. (2016). Combining satellite imagery and machine learning to predict poverty. *science*, 353(6301), 790-794.
  - (Optional). Hinton, G., Vinyals, O., & Dean, J. (2015). Distilling the knowledge in a neural network. *arXiv preprint arXiv:1503.02531*.
  - (Optional). Wang, Q., Wang, S., Zheng, Y., Lin, H., Zhang, X., Zhao, J., & Walker, J. (2024). Deep hybrid model with satellite imagery: How to combine demand modeling and computer vision for travel behavior analysis? *Transportation Research Part B: Methodological*, 179, 102869.
  - (Optional). Gebru, T., Krause, J., Wang, Y., Chen, D., Deng, J., Aiden, E. L., & Fei-Fei, L. (2017). Using deep learning and Google Street View to estimate the demographic makeup of neighborhoods across the United States. *Proceedings of the national academy of sciences*, 114(50), 13108-13113.
  - (Optional). Zheng, Y., Lin, Y., Zhao, L., Wu, T., Jin, D., & Li, Y. (2023). Spatial planning of urban communities via deep reinforcement learning. *Nature Computational Science*, 3(9), 748-762.

### Grading

Your grade consists of the follow three components.

Components	Total points	Percentage of final grade
Course participation	100	10%
Problem sets (3)	100 each	45%
Project	100	45%

### Course participation (10 pts)

Attendance and participation in the class are required. Attendance and participation grade will be computed in proportion to the number of presences. Students are also highly encouraged to get engaged in the class discussions. The university policy can be found [here](#).

### **Problem sets (45 pts)**

Problem sets are designed to help you learn how to apply the analytical tools to cities. Students are allowed to work in groups, as long as each group is comprised of no more than three people and each member submits their own written answers.

1. Pset 1 (15 pts) – Data description and visualization.
2. Pset 2 (15 pts) – Diverse data structures.
3. Pset 3 (15 pts) – AI applications.

### **Project: three stages (45 pts)**

1. Idea (5 pt). Limit to 1 page.
2. Proposal (10 pt). Limit to 3 pages.
3. Final paper (30 pt). Limit to 8 pages.

### **Grading scale**

The following table is used as **an example only**.

	<b>Percent Grade</b>	<b>4.0 Scale</b>		<b>Percent Grade</b>	<b>4.0 Scale</b>
A	94-100	4.0	C	73-76	2.0
A-	90-93	3.67	C-	70-72	1.67
B+	87-89	3.33	D+	67-69	1.33
B	83-86	3.0	D	65-66	1.0
B-	80-82	2.67	E/F	Below 65	0.0
C+	77-79	2.33			

The grading follows the university policy [here](#).

### **Late Submission**

It is important to meet deadlines. All work must be completed and submitted by the designated date and time on Canvas. However, life is sometimes uncertain. Therefore, you are allowed to submit your assignments late, but with **3 points deducted for every 24 hours**. For example, if an assignment is submitted 1 hour after the deadline, its full grade will automatically drop from 15 points to 12 points. This policy applies to both Psets and the milestones of the project.

## **Other UF policies and resources**

### **Accommodating Students with Disabilities**

Students requesting accommodation for disabilities must first register with the Dean of Students Office (<https://disability.ufl.edu/>). The Dean of Students Office will provide documentation to the student who must then provide this documentation to the instructor when requesting accommodation. You must submit this documentation prior to submitting assignments or taking the quizzes or exams. Accommodations are not retroactive, therefore, students should contact the office as soon as possible in the term for which they are seeking accommodations.

### **Academic integrity and UF honor code**

Academic honesty and integrity are fundamental values of the University community. UF students are bound by The Honor Pledge which states, “We, the members of the University of Florida community, pledge to hold ourselves and our peers to the highest standards of honor and integrity by abiding by the Honor Code. On all work submitted for credit by students at the University of Florida, the following pledge

is either required or implied: “On my honor, I have neither given nor received unauthorized aid in doing this assignment.” Furthermore, you are obligated to report any condition that facilitates academic misconduct to appropriate personnel. If you have any questions or concerns, please consult with the instructor or TAs in this class.

## **Course evaluation**

Students are expected to provide professional and respectful feedback on the quality of instruction in this course by completing course evaluations online via GatorEvals. Guidance on how to give feedback in a professional and respectful manner is available [here](#). Students will be notified when the evaluation period opens and can complete evaluations through the email they receive from GatorEvals, in their Canvas course menu under GatorEvals, or via this [link](#). Summaries of course evaluation results are available [here](#).

## **Other academic resources**

- For On-Campus URP Students: Graduate Coordinator contact information: Laura Dedenbach, [laurajd@ufl.edu](mailto:laurajd@ufl.edu), 352-294-1493.
- Career Connections Center: Reitz Union Suite 1300, 352-392-1601. Career assistance and counseling services [career.ufl.edu/](http://career.ufl.edu/).
- Library Support: various ways to receive assistance with respect to using the libraries or finding resources. [cms.uflib.ufl.edu/ask](http://cms.uflib.ufl.edu/ask)
- Teaching Center: Broward Hall, 352-392-2010 or to make an appointment 352- 392-6420. General study skills and tutoring. [teachingcenter.ufl.edu/](http://teachingcenter.ufl.edu/)
- Writing Studio: 2215 Turlington Hall, 352-846-1138. Help brainstorming, formatting, and writing papers. [writing.ufl.edu/writing-studio/](http://writing.ufl.edu/writing-studio/)
- For issues with technical difficulties for E-learning in Canvas, please contact the UF Computing Help Desk at 352-392-4357 or via e-mail at [helpdesk@ufl.edu](mailto:helpdesk@ufl.edu)
- Student Complaints: [sccr.dso.ufl.edu/policies/student-honor-code- student-conduct-code/](http://sccr.dso.ufl.edu/policies/student-honor-code- student-conduct-code/)

## **Health and wellness**

- U Matter, We Care: If you or someone you know is in distress, please contact [umatter@ufl.edu](mailto:umatter@ufl.edu), 352-392-1575, or visit [umatter.ufl.edu/](http://umatter.ufl.edu/) to refer or report a concern and a team member will reach out to the student in distress.
- Counseling and Wellness Center: Visit [counseling.ufl.edu/](http://counseling.ufl.edu/) or call 352-392-1575 for information on crisis services as well as non-crisis services.
- Student Health Care Center: Call 352-392-1161 for 24/7 information to help you find the care you need or visit [shcc.ufl.edu/](http://shcc.ufl.edu/).
- University Police Department: Visit [police.ufl.edu/](http://police.ufl.edu/) or call 352-392-1111 (or 9-1-1 for emergencies).
- UF Health Shands Emergency Room / Trauma Center: For immediate medical care call 352-733-0111 or go to the emergency room at 1515 SW Archer Road, Gainesville, FL 32608; [ufhealth.org/emergency-room-trauma-center](http://ufhealth.org/emergency-room-trauma-center)