

## Public Policy Analytics MUSA 508 / CPLN 592

Lecture – Fridays, 10:15am–11.45am; Meyerson Hall, B3 Michael Fichman Lab (Section 402) – 12pm–1.30pm – Meyerson Hall, B3 Matt Harris Lab (Section 403) – 12pm–1.30pm – Meyerson Hall, B4

## Instructors:

Name	Email	Office Hours	
Ken Steif, Ph.D	ksteif@upenn.edu	TBA	
Michael Fichman, MCP	mfichman@upenn.edu	Tues 10-12, Wed. 2-4, Calend.ly	
Matt Harris	matthew.harris@micadatalabs.com	Monday TBA, Calend.ly	

## TAs:

Name	Email	Office Hours
Anna Duan	annaduan@sas.upenn.edu	M/W 8:15-9:15PM, Tues 2-5, Calend.ly
Syndey Goldstein, MCP	sydng@upenn.edu	M/W 12-2PM, Calend.ly

## **Course Resources:**

Piazza page - <a href="https://piazza.com/class/krap1pcyj696n1">https://piazza.com/class/krap1pcyj696n1</a>
Course text - <a href="https://urbanspatial.github.io/PublicPolicyAnalytics/">https://urbanspatial.github.io/PublicPolicyAnalytics/</a> or <a href="https://www.routledge.com/Public-Policy-Analytics-Code-and-Context-for-Data-Science-in-Government/Steif/p/book/9780367507619">https://www.routledge.com/Public-Policy-Analytics-Code-and-Context-for-Data-Science-in-Government/Steif/p/book/9780367507619</a>

**Course Description:** This course teaches advanced spatial analysis and an introduction to data science/machine learning in the urban planning and public policy realm. The class focuses on real-world spatial analysis applications and, in combination with introductory machine learning, provides students a modern framework for efficiently allocate limited resources across space. We will learn novel approaches for ensuring our models do not discriminate against communities/people of color as well as communities of different income levels.

The format of the class includes weekly lectures/in-class demos and labs. There are six required assignments, including two projects. Prerequisites include either CPLN503, the summer GIS course or prior experience with GIS in a formal setting. Having experience in R and the 'tidyvserse' is helpful.

**Grading:** The grading breakdown is as follows: 50% for homework; 20% for project 1; 20% for project 2; and 10% for participation. Your homework/project grade is dependent on your ability to motivate your analysis and communicate empirical results to a non-technical audience using maps and data visualization. Your participation grade is a function of both in-class participation and Piazza participation.

**Homework:** Homework is due on the dates indicated on the schedule below. Late homework will be accepted but penalized. Please prepare all homework as an R Markdown. You are encouraged welcome to work in groups, but **you must submit a homework assignment that is uniquely yours.** Your willingness to ask and answer questions on Piazza will count towards your participation grade. Homeworks will be graded on a scale of 1 (revise and resubmit) to 3 (superlative).

**How we will assess you:** Your grade in this course is not a function of your ability to write efficient or even 'clean' code. It is dependent on your ability to communicate technical Planning and public policy analytical concepts to non-technical decision-makers. This is done through clear and concise writing, and most important, through data visualization.

Some students will come in with significant experience coding in R, others, not so much. Admittedly, newcomers will have a more significant learning curve, but I urge you to form pairs and project teams with students who have diverse skillsets. Piazza is a great way to meet new people.

**Readings**: The course text, <u>Public Policy Analytics</u>, includes all the code you will need for the course. A useful supplemental text on 'tidy' coding in R is <u>R for Data Science</u>. For some great intuition on data visualization and 'ggplot', check out <u>Data Visualization: A Practical Introduction</u>.

**Academic Integrity:** Teamwork is essential in this class. However, please ensure that the work you turn in is uniquely your own. Your ability to copy/paste and hack together code from the readings and from the internet is essential, particularly for newcomers, but plagiarism of any kind is strictly prohibited. Do not copy and paste policy-related narrative. If you have a question about these issues, just ask one of the professors. We will adhere to the University's <a href="Code of Academic Integrity">Code of Academic Integrity</a>.

**Software**: This course will be taught using R and R Studio.

Date	Lecture	Lab	Readings	Assignment due
		Introduction to the Tidyverse, Tidycensus, sf & ggplot		
3-Sep	Introduction to Public Policy Analytics		Read book <u>introduction</u>	
10-Sep	CANCELLED	Collaborating on data science projects w/ Github w/ rmarkdown & sf	Ch 1: Indicators for Transit Oriented Development	
17-Sep	Why start with Indicators?	Building indicators, in-class assignment work - TOD Assignment		
24-Sep	Data-driven Comprehensive Planning	Geoprocessing	Ch 2: Planning Urban Growth Area expansion	TOD Assignment (By beginning of class)
1-Oct	Project ${\bf 1}$ - Intro to geospatial modeling – Predicting home prices in Boulder, CO.	Intro to machine learning; regression; cross-validation.	Ch 3: Intro to Geospatial Machine Learning	
8-Oct	Project 1 - Intro to geospatial modeling — Predicting home prices in Boulder, CO.	Spatial autocorrelation; generalizability across space (race and class)	Ch 4: Geospatial ML; modeling the spatial process	Predicting home prices midterm project (Due Oct 13th before midnight)
15-Oct	NO CLASS			
22-Oct	Geospatial risk prediction – Predictive policing. See additional resources <u>here</u> .	Algorithmic bias across space; Racist training data?	Ch 5: Predictive Policing	
29-Oct	Churn Prediction	Logistic regression; confusion metrics; cost/benefit analysis	Ch 6: Bounce to Work!	Risk prediction (By beginning of class)
5-Nov	Data privacy, Disparate Impact, algorithmic fairness & predicting recidivism	Algorithmic bias across races; More racist training data; Memo-writing	Ch 7: Recidivism	Housing subsidy algorithm (By beginning of class)
12-Nov	Space/time rideshare trip prediction	time lags; gganimate; purrr	Ch 8: Predicting ride share demand	
19-Nov	Data governance - Dennis Culhane talk/Final Project work	Final Project	Final project	Either recidivism memo OR bike share prediction (By beginning of class)
26-Nov	No Class - Thanksgiving	Final Project	Final project	
3-Dec	Developing a GitHub.io portfolio / Final project work		Final project	
10-Dec	Final presentations - No lab			Final project (Final deliverable date TBD)