

Course: Data Acquisition & Management (AIM 5001)

Credits: 3 Credits / Graduate

Pre/Coreqs: N/A

Instructor: James Topor

Instructor Contact: james.topor@yu.edu

COURSE OVERVIEW

Data Acquisition and Management focuses on the data structures, data design patterns, algorithms, methods, and best practices for the pre-modeling phases of data science workflows, including problem formulation, gather, analyze, explore, model, and communicate, analytics programming focuses on the gather, analyze, and explore workflow steps. This comprises the "data wrangling" work which is where most data scientists spend the majority of their time. Because data science is iterative, this preparatory work informs the modeling phase. Often, the creation and validation of new models requires going back for additional data, different data transformations, and exploration of data distributions. In short, every effective data scientist needs to master analytics programming. Course topics include reading from or writing to databases, text files, and the web; shaping data into "tidy" data frames, exploratory data analysis, data imputations, feature engineering, and feature scaling.

COURSE LEARNING OUTCOMES

By the end of this course, students will be able to:

- Obtain data from structured and unstructured data sources.
- Transform, modify and explore data as needed to support and validate modeling operations.
- Engineer data features based on business and modeling constraints
- Perform basic Exploratory Data Analysis
- Create high quality explanatory narratives and visualizations in support of reproducible analytical work

REQUIRED MATERIALS

- Larry Rockoff, The Language of SQL, 2nd Edition. Addison-Wesley (2017).
- S. Juba, A. Volkov, Learning PostgreSQL 11, 3rd Edition. Packt Publishing (2019).
- Wes McKinney, Python for Data Analysis, 2nd edition, O'Reilly (2018).
- Mark Pilgrim, Dive Into Python 3, freely available web-based content: http://diveintopython3.problemsolving.io/table-of-contents.html
- Zheng, Alice and Casari, Amanda, Feature Engineering for Machine Learning: Principles and Techniques for Data Scientists. O'Reilly (2018).

Web-based readings and videos on related topics will also be assigned.

Relevant Software, Hardware, or Other Tools:

We will make use of the <u>PostgreSQL</u> relational database platform, the <u>MongoDB</u> NoSQL database platform, and the <u>Neo4J</u> Graph Database platform. We will make use of Python via the freely available <u>Anaconda</u> environment, including <u>Jupyter Notebooks</u> and the <u>Spyder IDE</u>. Students are also welcome to use Google's <u>Colab</u> platform where feasible. Details for obtaining and installing the appropriate software will be provided in the course materials. All of the software will work on (or from) both PCs and Macs.



ASSIGNMENTS & GRADING

Approach to Assignments. All Python-based projects and assignments are to be written in IPython (Jupyter) notebooks and submitted via Canvas. Non Python-based assignments will be submitted directly within the AIM 5001 Canvas portal.

Evaluation Criteria. All course projects will be evaluated like work assignments from a demanding employer. The primary evaluation basis is adherence to the deliverables stated in each assignment's functional requirements. To achieve a top grade, students must also adhere to best practices for software engineering principles, including reproducibility; following appropriate coding guidelines; and DRY. Furthermore, assignments must be clearly and concisely written using proper English language grammar and should present relevant supporting text in a logical flow. Presentations should include an appropriate level of detail for their intended audience.

Assignments	Grading
Discussions / Weekly Response Assignments (13 x 10 Pts) The fourteen module-specific discussions will focus primarily on use cases related to the topics covered within the associated module. Students will prepare short responses to discussion questions, which will be used to prompt group discussion.	6%
Assignments (11 x 100pts) On most weeks when projects are not due, there will be short-form ("miniproject") assignments to help reinforce the current learning material. These assignments may include completing tasks using course analytical tools. Some assignments may require working in small groups.	38%
Projects (2 x 100pts) Students will work individually and in teams on two data acquisition and management projects. At the end of the course, each student will have a portfolio of increasingly complex projects ready to show an employer.	18%
Midterm Exam (100 Points) Students will address a series of practical challenges derived from the content of AIM 5001 Course Modules 1 through 7.	7%
Final Project (150pts) and Presentation (50pts) Working individually or as part of a small team, students will create a formal proposal that specifies one or more research questions to be answered based on data students have chosen to work with. They will then attempt to answer the research questions described in their proposal using the skills they've developed during the semester. Students will present their final projects to their peers for feedback.	20%
Final Exam (100 Points) Students will address a series of practical challenges derived from the content of AIM 5001 Course Modules 1 through 14.	11%

- All projects and assignments, unless otherwise noted, are due end of day on Sundays.
- Each week's materials will be made available via Canvas no later than the previous Friday at 6:00 a.m. ET.
- Course Completion Requirements: As a prerequisite to passing this course, you must complete all four projects (including the final), and make the final presentation during



the final class session. Failure to either submit any one of the four projects or present your final project will preclude you from achieving a passing grade in this course. Please note that completion of the four projects is not the sole determinant of whether you will receive a passing grade: however, failure to submit any one of the four will prevent you from achieving a passing grade.

- **Discussion / Response Assignments**: While this material is important, please note that this work only makes up 6% of your grade. Please do the readings, and participate in the discussions and any discussion-related group assignments. If you have limited time for the course, please remember to invest the majority of your efforts in completing the projects and assignments. The assignments merit close attention because they will help you to be successful on the projects.
- Reproducibility Requirement, Testing Requirement, But Not Perfection! Students are
 responsible for providing all code and data so that your work can be reproduced by
 others. If you turn in code that does not run, you will not receive credit, unless you also
 include an explanatory note at the time of submission. At the same time, you don't need
 to turn in perfect code. Generous partial credit will be given for deliverables that are
 timely, tested, and reproducible.
- Policy on Sharing and "Stealing" Code. In this course, you may collaborate and you may take base code from whatever sources you wish. But you must document what you started with, and what you added, so you are graded on your own contributed work! Furthermore, use of any automated/GPT tools for any aspect of your assignment, discussion, project, or exam work will be treated as CHEATING, and is strictly prohibited. Failure to provide proper citations for any third party components of the content you submit will be treated as a violation of the Katz School's Student Code of Conduct and will be treated accordingly.
- Late work policy. Please note: <u>Assignments, discussion responses, exams, and projects cannot be accepted after their due dates / submit deadlines for any reason</u>. Any assignment, discussion, exam, or project that is not submitted before its associated deadline will automatically be assigned a grade of **ZERO**. <u>Late submits will be neither reviewed nor graded</u>. You will enhance your chances for success in this class if you start early, and turn in your work on time (even if it's not perfect!).
- Students that complete all work in a satisfactory and timely manner will earn a maximum grade of A-. To earn a grade of A in *Data Acquisition & Management*, you'll need to demonstrate work above and beyond what is expected.



GRADING SCALE:

ACADEMIC GRADES				ADMINI	
Quality of Performance	Letter Grade	Range %	GPA/Quality Pts.		GRADE
Excellent - work is of exceptional	А	94- 100	4		G
quality	A- 90 - 3.7	3.7		1	
Good - work is above average	B+	87 - 89.9	3.3		L
Satisfactory	В	83 - 86.9	3		W
Below Average	B-	80 - 82.9	2.7		
	C+	77 - 79.9	2.3		Note that
Poor		70 - 76.9	2		grades F
Failure	F	< 70	.000		

How This Course Works:

Class sessions are held every week on *Mondays from 5.30 p.m. to 7:30 p.m. ET*, with the exception of Katz School official holidays. You are strongly encouraged to attend these weekly classes since each will include opportunities for hands-on learning via discussions and case studies as well as a presentation / demonstration of many of the concepts you will need to use for any assignment or project due that week. You are also required to bring your laptop to these Live Sessions as this will serve to facilitate the hands-on learning segments. Class dates can be found in the Course Schedule shown on the following page.

Office Hours can be scheduled by appointment. If you need extra help and are willing to invest the time and effort to be successful, your instructor will make time available to help you. **But...**you should not be asking for extra help on a project or assignment the day before or the day it is due, since this will indicate that you are not investing the time and effort needed to be successful in the course.

You are encouraged to ask questions on Canvas where other students will be able to benefit from your inquiries. For the most part, you can expect your instructor to respond to questions asked either via email or via Canvas within one business day.



KATZ SCHOOL CLASS ATTENDANCE POLICY

For programs that meet in face-to-face format, students are expected to attend all scheduled classes in their entirety. Students who fail to fulfill this requirement will receive an academic penalty appropriate for the course work missed.

Students may not miss 20% (3 classes) or more of their scheduled class. If a student misses 20% or more of a course during the semester, they will receive a final grade of "F." This grade will be reflected on the student's official university transcript.

For programs within clinical components students may not miss 20% or more of any course, clinical or not. At the Katz School, this pertains only to students in the Speech Language Pathology program. If a student misses 20% or more of a course during the semester, they will receive a final grade of "F." This grade will be reflected on the student's official university transcript.

If the student is absent because of a disability which is documented with the Office of Disability Services at Yeshiva, falls ill or there are other extenuating circumstances, the student must inform the instructor in advance. The instructor may require appropriate documentation to make any exception to this policy.



COURSE SCHEDULE

Students should expect to spend <u>a minimum</u> of 9 hours each week outside of the classroom sessions on the materials, assignments, discussions, and projects required for this course.

Module	TOPIC	SCHEDULE OF MAJOR ASSIGNMENTS	
Module 1 Jan 16 – Jan 28	Intro to SQL + PostgreSQL	M1 Assignment	
Class: Mon Jan 22			
Module 2	SQL Aggregation & Grouping + Principles of Database	M2 Assignment	
Jan 29 – Feb 4	Design		
Class: Mon Jan 29	S S		
Module 3	Python Basics: Syntax, Data Types, Objects, Control	M3 Assignment	
Feb 5 – Feb 11	Flow		
Class: Mon Feb 5			
Module 4	Python Data Structures, Comprehensions, & Functions	M4 Assignment	
Feb 12 – Feb 18			
Class: Mon Feb 12			
Module 5	Regular Expressions	M5 Assignment	
Feb 19 – Feb 25			
Class: Mon Feb 19			
Module 6	NumPy: Numerical Python	M6 Assignment	
Feb 26 – Mar 3	,	<u> </u>	
Class: Mon Feb 26	** Final Project Requirements Distributed **		
Module 7	Pandas Series & Dataframe Objects	Project 1 Due	
Mar 4 – Mar 10	·		
Class: Mon Mar 4			
Module 8	Exploratory Data Analysis + Creating Visualizations in	M8 Assignment	
Mar 11 – Mar 17	Python		
Class: Mon Mar 11	Midterm Exam (via Canvas Thurs 3pm – Fri 3pm)	Midterm Exam	
Module 9	Working with Web Data	M9 Assignment	
Mar 18 – Mar 24		** 1st Draft of Final Project	
Class: Mon Mar 18		Proposal Due Sun Mar 24**	
Module 10	Data Preparation & Feature Engineering	Project 2 Due	
Mar 25 – Mar 31		,	
Class: Mon Mar 25			
Module 11	Data Reshaping & Aggregation in Pandas	M11 Assignment	
	Dala keshaping & Aggregation in Fandas		
Apr 1 – Apr 7		** Final Project Proposal	
Class: Mon Apr 1		Due Sun Apr 7	
Module 12	Text Mining	M12 Assignment	
Apr 8 – Apr 14			
Class: Mon Apr 8		1 1 1 1 1 1	
Module 13	Graph Databases: Neo4J	M13 Assignment	
Apr 15 – Apr 21			
Class: Mon Apr 15			
Apr 22 – Apr 30	*** NO CLASSES: PASSOVER BREAK ***		
Module 14	NOSQL Databases: MongoDB	Final Exam	
W May 1 – May 5	*** Class session to be held Wed May 1 12.30pm ***	I III LAGIII	
Class: **W May 1	FINAL EXAM (Online via Canvas, Thur 3pm – Fri 3pm)		
Module 15	Final Project Presentations + Writeups Due	** Final Project Writeups	
May 6 - May 8	Third Tojour Tojour Andrews Trinoops Doo	Due Monday May 6**	
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ONLINE LEARNING POLICIES

Online Learning Formats

Your course consists of two online learning formats:

- **Synchronous Learning**: While our class sessions will be held on campus, we may occasionally be required to meet via a live real time session using Zoom (webinar system). During these sessions, we will be able to see and talk with each other. Attendance is required.
- **Asynchronous Learning**: Pre-created content such as videos, assignments, links and articles. There will also be the use of community and collaboration tools like discussion boards and group tools. These sessions are not in real-time but rather involve engagement over the course of each week.

Online Learning Engagement Policy

A successful online class only happens when there is an active community. Students are required to attend both the weekly live synchronous sessions and participate in other community building activities such as the discussion boards.

Netiquette

Netiquette is a set of rules for behaving properly in an online course. Often the anonymity of online courses can cause a lapse in judgement when learners are excited or passionate about a subject. This can lead to statements that could be demeaned as offensive. You are all adults and are treated as such. However, it is still important to talk about these issues. The following bullet points cover some basics communicating in an online course:

- Be sensitive to the fact that there will people with different cultural and linguistic backgrounds, as well as different political and religious beliefs.
- Use good taste when composing your responses in Discussion Forums. Swearing and profanity is also part of being sensitive to your classmates and should be avoided.
- Don't use all capital letters when composing your responses as this is considered "shouting" on the Internet and is regarded as impolite or aggressive.
- Be respectful of your others' views and opinions. Avoid "flaming" (publicly attacking or insulting) them as this can cause hurt feelings and decrease the chances of getting all different types of points of view.
- Be careful when using acronyms. If you use an acronym it is best to spell out its meaning
 first, then put the acronym in parentheses afterward, for example: Frequently Asked
 Questions (FAQs). After that you can use the acronym freely throughout your message.
- Use good grammar and spelling (avoid using text messaging shortcuts).
- If you aren't sure what someone meant, consider asking for clarification.
- Remember that your peers are not required to respond to your specific post, so don't be offended if your question goes unanswered.

UNIVERSITY POLICIES & RESOURCES



ACCESSIBILITY AND ACCOMMODATIONS

The Office of Disability Services collaborates with students, faculty and staff to provide reasonable accommodations and services to students with disabilities. Students with disabilities who are enrolled in this course and who will be requesting documented disability-related accommodations should make an appointment with the Office of Disability Services, (646) 592-4132, rkohn1@yu.edu, during the first week of class. Once you have been approved for accommodations, please submit your accommodation letter to ensure the successful implementation of those accommodations. For more information, please visit: http://yu.edu/Student-Life/Resources-and-Services/Disability-Services/

ACADEMIC INTEGRITY

The submission by a student of any examination, course assignment, or degree requirement is assumed to guarantee that the thoughts and expressions therein not expressly credited to another are literally the student's own. Evidence to the contrary will result in appropriate penalties.

Academic integrity is a set of responsibilities and standards to facilitate high academic quality and rigor with the purpose of clarifying expectations and student conduct. The submission by a student of any coursework, or degree requirement is assumed to guarantee that the thoughts and expressions therein not expressly credited to another are literally the student's own. Examples of violations on academic integrity are, but not limited to:

- Cheating
- Plagiarism
- Dishonesty
- Assisting or attempting to assist another student in an act of academic dishonesty
- Providing papers, essays, research, or other work to aid another student in Intentional Misrepresentation
- Engaging in unauthorized cooperation with other individuals in completing assignments or examinations
- Submitting the same assignment, in part or whole, in more than one course, whether at YU or another institution, without prior written approval from both faculty members.

These Academic Integrity principles are incorporated within the Student Code of Conduct/Code of Ethics, with which all students are required to adhere to at all times.

For more information, visit http://yu.edu/registrar/grad-catalog/

Disciplinary Actions for Violations of Academic Integrity Standards

In this course, disciplinary actions for violations of the Student Code of Conduct/Code of Ethics are as follows:

- For the first incident: A grade of zero will be assigned for the assignment, project, or discussion. The student's final grade for the course will also be reduced by a full letter grade, i.e., if, at the end of the semester, the student would have otherwise earned a grade of A-, that grade will be lowered to a grade of B-.
- For the second incident: Student will be assigned a final grade of 'F' for the course, and no further assignment/project/discussion work will be accepted from the student for the balance of the semester for any reason.



YU Refund Policy

You should be aware of the universities refund policy. Please review this information: https://www.yu.edu/osf/undergraduate-accounts/withdrawal.

Academic Calendar

You should review the academic calendars, including add/drop dates. Please review this information: https://www.yu.edu/registrar/grad-calendar.

STUDENT SUPPORT SERVICES

Katz School offers academic support through the Learning Hub. This support service includes writing, academic integrity (APA format), English as a Second Language, and general academic tutoring. For more information, please contact katz@yu.edu.

If you need any additional help, please visit Student Support Services: http://yu.edu/academics/services/