Boston University Metropolitan College

BOSTON UNIVERSITY

Course Title: Data Science with Python
Course Number: BU MET CS 577 A1 (Spring 2026)
Course Format: On Campus

Instructor Name: Eugene Pinsky   epinsky@bu.edu

Computer Science Department, Metropolitan College, Boston University

1010 Commonwealth Avenue, Room 327, Boston, MA 02215

[Course Times:](#)   Wednesday 6:00 – 8:45 p.m.
                Office Hours: Fridays 11-1 or by appointment
                685-725 Comm Ave CAS 313

Teaching Assistants: (TBA)

Course Description

There is a growing need for specialists with backgrounds in Python who can apply data science methods to practical problems at their workplaces. Working in data science requires an understanding of many interdisciplinary concepts and involves data mining and the application of various methods.

The proposed course is designed to fill this need. Students will learn major Python tools, machine learning classifiers, and techniques for data analysis. There are weekly assignments and mini-projects on topics covered in class. These assignments will help build necessary statistical, visualization, and other data science skills for effective use of data science in various applications, including finance, time series analysis, and recommendation systems. In addition, students will choose a topic for a final project and present it on the last day of class.

The proposed course can be taken by students with not exclusively computer science backgrounds who have basic knowledge of Python.

Books

Required:

"Python for Data Analysis", by W. McKinney, O'Reilly Publishing, ISBN-13: 978-1491957660, purchased from Barnes & Noble

Recommended:

Machine Learning and Artificial Intelligence: Concepts, Algorithms and Models by Reza Rawassizadeh, ISBN-13 979-8992162103

"Python Data Analysis" by Armando Fandango, Packt Publishing, ISBN-13: 978-1787127487

"Python Data Science Handbook" by Jake VanderPlas, O'Reilly Publishing, ISBN-13: 978-1491912058

Courseware

Blackboard
Course Notes

Additional materials will be added to the "From Your Professor" section under the group discussion section.

BU Community Public Health Policies

All students returning to campus must follow all the compliance requirements as specified in the student link.

Class Policies

14 weekly programming assignments submitted through blackboard on-line. Late homework is accepted with 50% penalty. In-class quizzes. Final projects are submitted through blackboard on-line. Students will present their projects on the last day of class. Both quizzes and final are closed-book and taken in-class

Note: This is an on-campus class. Class attendance is **mandatory**. No make-ups of quizzes. Unexcused absences would result in lower grades.

Academic Conduct Code – "Cheating and plagiarism will not be tolerated in any Metropolitan College course. They will result in no credit for the assignment or examination and may lead to disciplinary actions. Please take the time to review the Student Academic Conduct Code:

Academic conduct code as specified below:

> http://www.bu.edu/met/metropolitan_college_people/student/resources/conduct/code.html.

> NOTE: [This should not be understood as a discouragement for discussing the material or your particular approach to a problem with other students in the class. On the contrary – you should share your thoughts, questions and solutions]

Grading Criteria:

35% homework, 20% quizzes, 30% final, 15% final project

Class Meetings, Lectures & Assignments:

The course is divided into 6 modules (each module is about 2 weeks).

Module	Topic	Readings Due
1	Review of Python, Numpy and data analysis libraries	Chapters 1,2 Course notes
2	Pandas, Matplotlib & Seaborn, error metrics, model selection trade-offs	Chapter 4, 5, 8 Course notes
3	Supervised learning and decision boundaries. Logistic regression and nearest neighbor classifiers. Parameter Estimation with gradient descent	Course notes
4	Linear and polynomial models for prediction. Linear regression and classification. Parameter estimation	Course notes
5	Bayes rule and Naïve Bayesian Classification. Decision trees. Ensemble learning with random forest classifiers	Course notes
6	Large-margin classification and kernels. Support Vector Machines. Unsupervised learning. k-means clustering	Course notes
7	Course review, project presentations and final exam	Course notes