Fall 2020

CS 301-001: Introduction to Data Science

Senjuti Basu Roy

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CS 301, Introduction to Data Science, Fall 2020

Instructor: Senjuti Basu Roy

Webex link:
https://njit.webex.com/njit/j.php?MTID=m599e64f3ec3fe8d10f7237da6797bf09

Class sessions: Monday/Wednesday, 12:30 – 1:50 pm
Instructor Office: GITC 4302
Office Hours: Monday/Wednesday (2 pm – 3 pm) /by appointment on (use the same Webex link)
(email: senjutib@njit.edu)

Teaching Assistant: TBD, email: TBD, office hour: M/W 3-4 pm (GITC 4111)

Course Overview: The course is an introductory data science course that focuses on how to develop principled analytics and implementations on a variety of large data sets. The course will be designed in two parts - during the first part of the course, we will focus on learning models, formalism, and algorithmic techniques that are popular in data science and heavily used in practice. In the second part of the course, students will be introduced to data science tools, such as performing data analysis with Excel and Python. Extra attention will be paid to strengthen theoretical as well as development/programming skills of the students in performing data analyses using real world small and large-scale datasets.

The primary objectives of the course are:

- Establishing quantitative view and mastering scientific approaches for analyzing large scale datasets.
- Learn data science algorithms and applications.
- Implementation skill of data science algorithms.
- Understand analysis, metrics, visualization and navigation of results
- Learn how to use existing data science tools (e.g., python, excel)

The outcomes of the course are:

a. An ability to perform predictive modeling in various data science applications;
b. An ability to perform semi-supervised modeling in various data science applications;
c. An ability to preform correlation and clustering analysis in various data science applications;
d. An ability to perform analysis on noisy, high dimensional, large scale datasets;
e. An ability to develop end to end solutions to real world data science problems –formalizing the problem, identifying appropriate modeling
techniques, and developing solutions;
f. An ability to implement real world large scale data science problems and evaluate its outcome in a principled manner.

**Prerequisite: CS 114, Math 333**

**Textbooks**

<table>
<thead>
<tr>
<th>Book Title</th>
<th>Author</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data Mining: Practical Machine Learning Tools and Techniques</td>
<td>Frank &amp; Witten</td>
</tr>
<tr>
<td>Applied Statistics for Engineers and Scientists (3E 14)</td>
<td>Devore &amp; Farnum</td>
</tr>
</tbody>
</table>

**Reference Books:**

<table>
<thead>
<tr>
<th>Book Title</th>
<th>Author</th>
</tr>
</thead>
<tbody>
<tr>
<td>Introduction to Data Science: A Python Approach to Concepts, Techniques and Applications</td>
<td>Laura Igual, et. al</td>
</tr>
</tbody>
</table>

**Course Content:**

There will be several aspects to this course:

1. Instructor lectures

2. Individual assignments: There will be four homework assignments, involving both theory and implementation.

4. Final project: A class project will be assigned and will be completed individually. Additionally, each student must present his or her final project during the presentation week.

5. Midterm and Final Exam

6. Class participation

**Course Schedule: (may change)**

<table>
<thead>
<tr>
<th>Dates</th>
<th>Schedule</th>
</tr>
</thead>
<tbody>
<tr>
<td>Week 1</td>
<td>Introduction to Data Science, Understanding different types of data</td>
</tr>
<tr>
<td>Week 2</td>
<td>Statistical techniques related to data science</td>
</tr>
<tr>
<td>Week 3</td>
<td>Analyzing outcomes of data science processes</td>
</tr>
</tbody>
</table>
Week 4    Data fitting and outlier analyses
Week 5    Data discretization
Week 6    Dimensionality reduction
Week 7    **Midterm**, Assessing predictive models
Week 8    Linear supervised algorithms
Week 9    Non-linear supervised algorithms
Week 10   Semi-supervised algorithms
Week 11   Correlation and clustering techniques
Week 12   Search algorithms for intelligent systems
Week 13   Emerging topics in data science
Week 14   Student Presentation
Week 15   **Final Exam**
Grading Scheme*:

I reserve the right to make small adjustments to grade weights, or to add small assignments as the need arises.

<table>
<thead>
<tr>
<th>Item</th>
<th>Grade (% of final grade)</th>
<th>Due date</th>
</tr>
</thead>
<tbody>
<tr>
<td>Homework Assignments (total 4)</td>
<td>7.5*4=30%</td>
<td>Home works would be posted on Fridays. It would be due on Fridays after 1-week.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>HW-1 posted: Sept 18</td>
</tr>
<tr>
<td></td>
<td></td>
<td>HW-2 posted: Oct 9</td>
</tr>
<tr>
<td></td>
<td></td>
<td>HW-3 posted: Nov 6</td>
</tr>
<tr>
<td></td>
<td></td>
<td>HW-4 Posted: Nov 27</td>
</tr>
<tr>
<td>Final Project+ presentation</td>
<td>15%</td>
<td>Posted: Nov 9</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Due: December 9 Presentation TBD</td>
</tr>
<tr>
<td>Midterm Exam</td>
<td>20%</td>
<td>October 12</td>
</tr>
<tr>
<td>Final Exam</td>
<td>30%</td>
<td>TBD</td>
</tr>
<tr>
<td>Class participation and attendance</td>
<td>5%</td>
<td></td>
</tr>
</tbody>
</table>

All exams are open book and notes.

*20% late submission penalty if submitted within one day of deadline. Beyond that, late submission is not entertained. Unless otherwise stated, all submissions are due by 11:59 pm eastern on the due date.

An extra credit assignment (could also be a quiz or an exam) may be given to make up for 5% of the course total. The exact date of the extra credit is not decided yet.

Grading Scale
A <= 20%, B+<= 20%, B <= 20%, C+/C/D/F/W the rest of the class
Honor and Policy

*Academic Integrity is the cornerstone of higher education and is central to the ideals of this course and the university. Cheating is strictly prohibited and devalues the degree that you are working on. As a member of the NJIT community, it is your responsibility to protect your educational investment by knowing and following the academic code of integrity policy that is found at:* [http://www5.njit.edu/policies/sites/policies/files/academic-integrity-code.pdf](http://www5.njit.edu/policies/sites/policies/files/academic-integrity-code.pdf).

*Please note that it is my professional obligation and responsibility to report any academic misconduct to the Dean of Students Office. Any student found in violation of the code by cheating, plagiarizing or using any online software inappropriately will result in disciplinary action. This may include a failing grade of F, and/or suspension or dismissal from the university. If you have any questions about the code of Academic Integrity, please contact the Dean of Students Office at dos@njit.edu”*