Introduction to Computational Thinking and Data Science

DSCI 549  

Term: Spring 2024
Units: 4
Time: Tuesdays 2pm-5:20pm
Location: OHE 136 and online

Instructor: Dr. Ning Wang
Office Hours: arranged by appointment only via email
Office hours location: Zoom meeting room https://usc.zoom.us/j/96268416493
Contact Info: nwang@ict.usc.edu

TA: TBD
Nishi Doshi, nishimit@usc.edu  Harsh Parikh, harshjin@usc.edu

Catalogue Course Description
Introduction to data analysis techniques and associated computing concepts for non-programmers. Topics include foundations for data analysis, visualization, parallel processing, metadata, provenance, and data stewardship.

Expanded Course Description
This course will teach non-programmers to think in computing terms about modern topics, and to approach real-world phenomena through data science. The course will enable students to:

- Acquire computational thinking skills that will enable students to represent and reason about complex problems in the digital arena
- Understand different kinds of data in terms of their possibilities and limitations to approach complex problems cast in terms of the emerging field of data science
- Become data science scholars through best practices in data documentation and dissemination

The course is intended for students in disciplines outside of computer science, so no prior experience with computer science is assumed. The course topics will be particularly relevant to students interested in physical sciences and social sciences.

This class will include eight homework assignments, a midterm and a final exam.

Learning Objectives
This course teaches non-programmers to think in computing terms about modern topics, and to approach real-world phenomena through data science. The course introduces different kinds of data and
corresponding approaches to data analysis, including geospatial data, time series, networks, and multimedia data. Students learn to run multi-step analysis through a graphical workflow interface, and will experience first-hand complex concepts in data science such as parallel computing, provenance, and visualization. Students also learn to use ontologies and logic representations to capture metadata and other knowledge about complex data. The course includes practical lessons to use workflow and ontology development toolkits, as well as best practices for data stewardship and dissemination.

**Prerequisite(s):** None  
**Co-Requisite(s):** None  
**Recommended Preparation:** Mathematics and logic undergraduate courses.

### Software and Supplementary Readings
All required software is freely available for students to install on their personal computers or to access through a web interface.

There is no textbook. Students can find all the supplementary readings online. Supplementary readings include:


### Description and Assessment of Homework Assignments
There will be 8 homework assignments. The homework assignments include a class project that will be developed by the students independently in 3 separate stages, getting feedback from the instructor at each stage. The assignments must be submitted individually and students will receive individual scores. Students may NOT work in groups to complete the tasks. The homework assignments are expected to take 6-8 hours. Each assignment is graded on a scale of 0-100. The homework topics are listed in the Course Schedule.

### Assignment Submission Policy
Homework assignments are due at 11:59pm on the due date and should be submitted in Desire2Learn (D2L). Homework will be accepted up to 1 week late as long as the student requested a late submission ahead of time, and in that case the assignment will be graded at 20% less than the possible points for the
assignment. After one week, the assignment will not be graded. Exceptions will only be made with a note from a professional: for illness or family caregiving due to illness, religious observances, USC athletic event.

**Syllabus and Class Schedule**

<table>
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<tr>
<th>Week</th>
<th>Topic</th>
<th>Material Covered</th>
<th>Homework assigned</th>
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<td><strong>Section I: Introduction to Computational Thinking and Data Science</strong></td>
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| Week 1   | Computational thinking and data science | • What is computational thinking  
• Computational thinking for reasoning and analysis  
• What is data science  
• Data scientists  
• The context of data science | HW1: Project part 1 – Finding data |
|          | Data                             | • What is data  
• What is not (yet) data  
• Time series data  
• Networked data  
• Geospatial data  
• Text data  
• Labeled and annotated data  
• Big data |                                   |
| Week 2   | Data analysis software           | • Programs for data analysis  
• Inputs and Outputs  
• Program Parameters  
• Programming Languages  
• Programs as Black Boxes  
• Algorithms versus software |                                   |
|          | Multi-step data analysis as workflows | • Building workflows by composing software  
• Pre-processing and post-processing data  
• Workflows for data analysis  
• Workflow inputs and parameters  
• Executing workflows  
• Exploring data through workflows  
• Workflows in practice |                                   |
| **Section II: Data Analysis** |                                   |                                   |                                          |
| Week 3   | Logic and probability for statistics | • Basic probability for statistics  
• Logic for statistics  
• Null hypothesis significance testing  
• Sampling distributions | HW2: Exploring data analysis workflows |
|          | Basic statistics                  | • Descriptive statistics |                                   |
| Week 4 (Jan 30) | **Data analysis tasks (I)** | • Inferential statistics  
  o T-tests  
  o ANOVAs  
  o Chi-squared tests  
  o Correlation  
  • Data analysis tasks in data mining, statistics, and machine learning  
  • Supervised learning  
  o Classification tasks  
  o Classification algorithms  
  o Evaluation of classifiers |
| --- | --- | --- |
| **Data analysis tasks (II)** | • Unsupervised learning  
  o Clustering  
  o Pattern detection  
  o Anomaly detection |
| Week 5 (Feb 6) | **Data analysis tasks (III)** | • Unsupervised learning  
  o Clustering  
  o Pattern detection  
  o Anomaly detection  
  • Supervised learning  
  o Classification tasks  
  o Classification algorithms  
  o Evaluation of classifiers  
  • Simulation and prediction  
  • Causality  
  o Probabilistic graphical models  
  o Bayesian networks  
  o Causal models  
  • Reinforcement Learning  
  • Networks  
  o Network structure  
  o Dynamic networks  
  o Scale-free networks |
| **Data analysis tasks (IV)** | • Networks  
  o Network structure  
  o Dynamic networks  
  o Scale-free networks |
| **Exam Review** | • Exam Review |
| **MIDTERM EXAM** | • Midterm exam |
| **Analyzing different kinds of data (I)** | • Time series  
  o Collecting time series data  
  o Pre-processing time series data  
  o Event detection  
  • Granger causality |
| **Analyzing different kinds of data (II)** | • Analyzing text data  
  o Pre-processing text  
  o Document classification  
  o Document clustering  
  o Topic detection |

**HW3: Analyzing data with workflows**

**HW4: Processing Different Types of Data**
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| **Week 10 (Mar 12)** | Data visualization | • Quality of visualizations  
• Major types of visualizations  
• Time series visualizations  
• Geospatial visualizations  
• Multi-dimensional spaces  
• Network visualizations |
| **Week 11 (Mar 19)** | User experience, user interfaces, user studies (I) | • UX/UI Design Principles  
• AB testing  
• User study design |
| **Week 12 (Mar 26)** | Analysis for experiments | • Advanced analysis for experiments  
• Appropriate statistical tests |

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<th>Section V: Data analysis at scale</th>
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| **Week 12 (Mar 26)** | Causal claims from user studies | • Correlational research  
• Comparing correlational research to experiments  
• Ensuring internal validity |
| **Parallel and distributed computing for big data (I)** | • Cost of computation  
• Divide and conquer  
• Speedup with parallel processing  
• Limits of speedup: Critical path  
• Amdahl’s law  
• When problems are not parallelizable |

**HW5: Project part 2 – Design of data analysis approach**

**HW6: Data analysis and research methods**

**Section VI: Metadata**
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<th>HW7: Project part 3 – Final report</th>
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<td>• Multi-core computing</td>
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<td>• Parallel programming languages</td>
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<td>Semantic metadata</td>
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<td>• Basic metadata versus semantic metadata</td>
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<td>• Domain metadata and ontologies</td>
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<td>Week 14 (Apr 9)</td>
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<td>• Taxonomies and class inheritance</td>
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<td>• The Semantic Web</td>
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<td>Provenance and standards</td>
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<td>• Quality control</td>
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• Data integration

Privacy and ethics
• Privacy
• Sensitive data
• Anonymization
• Research ethics

Week 16
Final Exam Review
• Final Exam Review

Final Exam
Thursday, May 2nd, 2024, 2:00-4:00 PM PST

Grading Breakdown
Homework: There will be eight homework assignments throughout the course (see description above).
Midterm Exam: A short answer closed book midterm exam will cover all of the material up to that point.
Final Exam: A short answer closed book final exam will cover all of the material covered in the class.

Grading Schema:
Homework assignments 50%
Midterm: 25%
Final: 25%

Total 100%

Grades will range from A through F. The following is the breakdown for grading:

94 - 100 = A 74 – 76.99 = C
90 – 93.99 = A- 70 – 73.99 = C-
87 – 89.99 = B+ 67 – 69.99 = D+
84 – 86.99 = B 64 – 66.99 = D
80 – 83.99 = B- 60 – 63.99 = D-
77 – 79.99 =C+ 59.99 and below = F

Academic Conduct and Support Systems

Honor Code
In response to recommendations made by the Academic Integrity Task Force to the Dean, the USC Viterbi School of Engineering now has an Honor Code. The Code was developed by Viterbi students, and its text is as follows:

Engineering enables and empowers our ambitions and is integral to our identities. In the Viterbi community, accountability is reflected in all our endeavors.

Engineering+ Integrity.
Engineering+ Responsibility.
Engineering+ Community.
Think good. Do better. Be great.

These are the pillars we stand upon as we address the challenges of society and enrich lives.

During your time here at Viterbi, please know that academic and personal resources are available to help:
• The student-driven and student-written Honor Code is here: http://viterbi.usc.edu/academics/integrity/.

• An introductory video is posted at https://myviterbi.usc.edu/ under the link "Academic Integrity Introduction" and serves as a reminder of the school’s emphasis in maintaining a high level of academic integrity.

• Master’s and PhD students can contact the GAPP office in OHE 106 (https://gapp.usc.edu/) for other helpful resources.

• The Viterbi Academic and Resource Center (VARC) (http://viterbi.usc.edu/students/undergrad/varc) has a variety of services available.

Academic Integrity
The Viterbi School takes academic integrity violations seriously. Most of the violations that have been reported in the past fall into four categories: unauthorized collaboration, plagiarism, code sharing, and cheating on an exam. Specifically:

• Unauthorized collaboration - Unauthorized collaboration on a project, homework or other assignment. (section 11.14.B) All homework assignments must be individually developed. Students that collaborate on assignments will be referred to the Academic Integrity Coordinator.

• Plagiarism - Presenting someone else’s ideas as your own, either verbatim or recast in your own words - is a serious academic offense with serious consequences.

• Code sharing - Obtaining for oneself or providing for another person a solution to homework, a project or other assignment, without the knowledge and expressed consent of the instructor. (section 11.14.A)

• Cheating in an exam - This may involve a number of violations, such as looking at class notes during the exam, looking at other student's exam, "texting" with other students during the exam. See the section titled Two Exams for a list of specific violations.

Please note that these are only the basic violations that we have encountered in the past, and there are many more. Please familiarize yourself with the discussion of plagiarism in SCampus in Section B.11.00, Behavior Violating University Standards and Appropriate Sanctions available at https://scampus.usc.edu/b/11-00-behavior-violating-university-standards-and-appropriate-sanctions/.

All academic integrity violations will be referred to the Academic Integrity Coordinator of the Viterbi School of Engineering. The process for adjudicating these cases is available in SCampus, Part B, Section 13.

Other Misconduct
Other forms of academic dishonesty are equally unacceptable. See additional information in SCampus and university policies on scientific misconduct, http://policy.usc.edu/scientific-misconduct/.

Discrimination, sexual assault, and harassment are not tolerated by the university. You are encouraged to report any incidents to the Office of Equity and Diversity http://equity.usc.edu/ or to the Department of Public Safety http://capsnet.usc.edu/department/department-public-safety/online-forms/contact-us. This is important for the safety whole USC community. Another member of the university community - such as a friend, classmate, advisor, or faculty member - can help initiate the report, or can initiate the report on behalf of another person. The Center for Women and Men http://www.usc.edu/student-affairs/cwm/ provides 24/7 confidential support, and the sexual assault resource center webpage http://sarc.usc.edu describes reporting options and other resources.

Support Systems
A number of USC’s schools provide support for students who need help with scholarly writing. Check with your advisor or program staff to find out more. Students whose primary language is not English should check with the American Language Institute http://dornsife.usc.edu/ali which sponsors courses and workshops specifically for international graduate students. The Office of Disability Services and Programs
http://sait.usc.edu/academicsupport/centerprograms/dsp/home_index.html provides certification for students with disabilities and helps arrange the relevant accommodations. If an officially declared emergency makes travel to campus infeasible, USC Emergency Information http://emergency.usc.edu/ will provide safety and other updates, including ways in which instruction will be continued by means of blackboard, teleconferencing, and other technology.

Diversity
The diversity of the participants in this course is a valuable source of ideas, problem solving strategies, and engineering creativity. The instructors encourage and support the efforts of all of our students to contribute freely and enthusiastically. As members of an academic community, it is our shared responsibility to cultivate a climate where all students and individuals are valued and where both they and their ideas are treated with respect, regardless of their differences, visible or invisible.

Students with Disabilities
Any student requesting academic accommodations based on a disability is required to register with Disability Services and Programs (DSP) each semester. A letter of verification for approved accommodations can be obtained from DSP. Please be sure the letter is delivered to me as early in the semester as possible. DSP is located in STU 301 and is open 8:30 a.m. - 5:00 p.m., Monday through Friday. Website and contact information for DSP: http://sait.usc.edu/academicsupport/centerprograms/dsp/home_index.html, (213) 740-0776 (Phone), (213) 740-6948 (TDD only), (213) 740-8216 (FAX), ability@usc.edu.

Emergency Preparedness/Course Continuity in a Crisis
In case of a declared emergency if travel to campus is not feasible, USC executive leadership will announce an electronic way for instructors to teach students in their residence halls or homes using a combination of Blackboard, teleconferencing, and other technologies.