

## **ENGR 1330- D01: Computational Thinking with Data Science**

### **Important Guidelines for Face-to-Face and Hybrid Classes (not this class):**

- If Texas Tech University campus operations are required to change because of health concerns related to the COVID-19 pandemic, it is possible that courses will move to a fully online delivery format. [Note: this course is already fully online]. Should that be necessary, students will be advised of technical and/or equipment requirements, including remote proctoring software.
- Policy on absences resulting from illness: We anticipate that some students may have extended absences. To avoid students feeling compelled to attend in-person class periods when having symptoms or feeling unwell, a standard policy is provided that holds students harmless for illness-related absences (see Section A below).

### **Important Guidelines for Online and Distance Classes (this class):**

To attend class, students need to have:

- access to a computer with webcam and microphone for remote delivery of the class. The computer should meet the Whitacre College of Engineering laptop requirements, which can be found at:  
<https://www.depts.ttu.edu/coe/dean/engineeringitservices/buyingtherightcomputer.php>
- reliable fast Internet access to participate in lectures and laboratory sessions
- a TTU email account for course related communication
- the following software installed: Zoom video conferencing tool, web browser, Microsoft Office, Adobe Reader

Online exams and quizzes within this course require online proctoring. Therefore, students will be required to have a webcam (USB or internal) with a microphone when taking an exam or quiz. Students understand that this remote recording device is purchased and controlled by the student and that recordings from any private residence must be done with the permission of any person residing in the residence. To avoid any concerns in this regard, students should select private spaces for the testing. Students with concerns may discuss location of an appropriate space for the recordings with their instructor or advisor. Students must ensure that any recordings do not invade any third-party privacy rights and accept all responsibility and liability for violations of any third-party privacy concerns. Setup information will be provided prior to taking the proctored exam. For additional information about online proctoring, you can visit the online proctoring student FAQ

### **A. Illness-Based Absence Policy**

If at any time during this semester you feel ill, in the interest of your own health and safety as well as the health and safety of your instructors and classmates, you are encouraged not to attend face-to-face class meetings or events. Please review the steps outlined below that you should follow to ensure your absence for illness will be excused. These steps also apply to not participating in synchronous online class meetings if you feel too ill to do so and missing specified assignment due dates in asynchronous online classes because of illness.

1. If you are ill and think the symptoms might be COVID-19-related:

- a. Call Student Health Services at (806) 743-2848 or your health care provider. During after-hours and on weekends, contact TTU COVID-19 Helpline at (806) 743-2911.
  - b. Self-report as soon as possible using the [Dean of Students COVID-19 webpage \(https://www.depts.ttu.edu/dos/COVID-19Absence.php\)](https://www.depts.ttu.edu/dos/COVID-19Absence.php). This website has specific directions about how to upload documentation from a medical provider and what will happen if your illness renders you unable to participate in classes for more than one week.
  - c. If your illness is determined to be COVID-19-related, all remaining documentation and communication will be handled through the Office of the Dean of Students, including notification of your instructors of the time you may be absent from and may return to classes.
  - d. If your illness is determined not to be COVID-19-related, please follow steps 2.a-d below.
2. If you are ill and can attribute your symptoms to something other than COVID-19:
- a. If your illness renders you unable to attend face-to-face classes, participate in synchronous online classes, or miss specified assignment due dates in asynchronous online classes, you are encouraged to contact either Student Health Services at (806) 743-2848 or your health care provider. Note that Student Health Services and your own and other health care providers may arrange virtual visits.
  - b. During the health provider visit, request a “return to school” note.
  - c. E-mail the instructor a picture of that note.
  - d. Return to class by the next class period after the date indicated on your note.

Following the steps outlined above helps to keep your instructors informed about your absences and ensures your absence or missing an assignment due date because of illness will be marked excused. You will still be responsible to complete within a week of returning to class any assignments, quizzes, or exams you miss because of illness.

**Course Instructor:**

Instructor : Dr. Tanja Karp

Email : [tanja.karp@ttu.edu](mailto:tanja.karp@ttu.edu)

Office hours : directly after laboratory session and by appointment via email

**Teaching assistants:**

Teaching assistants will be available remotely from 6 – 9 pm Central Time US to answer questions regarding course content and / or lab assignments. A Zoom link will be shared on the Blackboard page for this class.

**Class meetings:**

Lecture: Recorded lectures can be watched. Links to the lecture recordings will be posted on Blackboard.

**Lab meetings:**

Lab meetings are scheduled weekly for Tuesdays and Thursdays 8 – 9:20 am Central Time US and will be conducted via Zoom. Students are strongly encouraged to participate in these lab

meetings, during which they will perform hands-on assignments. Students receive course credit for attending the labs and for the submitted work.

However, students may also opt to perform the lab assignments outside the assigned lab meeting times. Students who decide to do so will have to perform the asynchronous lab assignment option. Students may change their modality of lab participation on a weekly basis.

All lab assignment details (synchronous and asynchronous) and deadlines will be posted on Blackboard.

### **Catalog Course Description:**

Introduces Python programming, its relevant modules and libraries, and computational thinking for solving problems in Data Science. Students will learn data science approaches to importing, manipulating, and analyzing data as well as modeling and visualizing real-world data sets in various science and engineering disciplines.

- 3 credit hours comprising of lectures and hands-on lab sessions.
- This course provides a hands-on learning of principles of programming and data science by introducing Python programming, its relevant modules and libraries, and computational thinking for solving problems in data science. Students will learn data science approaches to importing data, manipulating data, and analyzing it as well as modeling and visualizing real-world data sets in various science and engineering disciplines.

### **Textbook:**

Ani Adhikari and John DeNero, *Computational and Inferential Thinking, The Foundations of Data Science*, Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International (CC BY-NC-ND 4.0). **Link:** <https://www.inferentialthinking.com/chapters/intro>.

### **Course Content:**

- Computational thinking for problem-solving: Logical problem solving, decomposition, pattern recognition, abstraction, representation, algorithm design, and generalization.
- Python Programming: Variables, constants, data types, data structures, strings, math operators, Boolean operators, expressions, program constructs, functions, loop, I/O files, modules, and database.
- Data science fundamentals:
  - ✓ *Experimental setup*: Importing and formatting data sets, displaying data, data pre-processing.
  - ✓ *Introductory statistical analysis with Python*: Elementary statistics, randomness, sampling, probability distribution, confidence intervals, hypothesis testing, and A/B testing.
  - ✓ *Basic data analysis, visualization, and machine learning*: Data pre-processing, basic supervised/unsupervised learning, performance evaluation metrics.

## Learning Outcomes:

On completion of the course, students should be able to:

- Implement basic Python programs using computational thinking concepts.
- Apply Python programming constructs and libraries to solve relevant data science problems.
- Perform fundamental data analytics and basic visualization using Python script language.

## ABET Student Outcomes

- Engineering:
  - ✓ An ability to identify, formulate, and solve complex engineering problems by applying principles of engineering, science, and mathematics.
  - ✓ An ability to acquire and apply new knowledge as needed, using appropriate learning strategies.
- Computer Science:
  - ✓ Analyze a complex computing problem and to apply principles of computing and other relevant disciplines to identify solutions.
  - ✓ Design, implement, and evaluate a computing-based solution to meet a given set of computing requirements in the context of the program's discipline.

## Resources/Tools

Platforms for Python Programming

- **Anaconda platform** (<https://www.anaconda.com/>): Anaconda distribution is an open-source Data Science Distribution Development Platform. It includes Python 3 with over 1,500 data science packages making it easy to manage libraries and dependencies. Available in Linux, Windows, and Mac OS X.
- **Jupyter** (<https://jupyter.org/>): JupyterLab is a web-based interactive development environment for Jupyter notebooks, code, and data. JupyterLab is flexible: configure and arrange the user interface to support a wide range of workflows in data science, scientific computing, and machine learning.

Modules for Python Programming

- **Math module** (<https://docs.python.org/3/library/math.html>): Gives access to the mathematical functions defined by the C standard e.g. factorial, gcd, exponential, logarithm.
- **Operator module** (<https://docs.python.org/3/library/operator.html>): Helps in exporting a set of efficient functions corresponding to the intrinsic operators of Python. For example, the operator `add(x,y)` is equivalent to the expression `x+y`.

Python Modules for Data Science

- **Scipy module** (<https://www.scipy.org/>): A Python-based ecosystem of open-source software for mathematics, science, and engineering. Some of the core packages are:
  - **Numpy**: Provides n-dimensional array package
  - **Scipy**: Fundamental for scientific computing (e.g. linear algorithm, optimization)
  - **Matplotlib**: Visualizations/2D plotting

- **IPython:** Enhanced interactive console
- **Pandas:** Data structures and data analysis
- **Scikit-learn module** (<https://scikit-learn.org/stable/>): A library for machine learning in Python. It is a simple and efficient tool for predictive data analysis. It is built on NumPy, SciPy, and matplotlib modules.

**Course Schedule (Tentative – Lectures will be recorded)**

Week	Lecture	Lab
1	<p><b><i>Introduction to Computational Thinking with Data Science:</i></b></p> <ul style="list-style-type: none"> <li>- Computational thinking</li> <li>- Programming principles</li> <li>- Data science</li> </ul> <p><b><i>Programming Principles:</i></b></p> <ul style="list-style-type: none"> <li>- Data types (int, float, string, bool)</li> <li>- Variables, operators, expressions, basic I/O</li> <li>- String functions and operations</li> <li>- Data structures: array, list, tuple, set, dictionary</li> <li>- Conditional statements</li> </ul>	<p><b><i>Preparation:</i></b></p> <ul style="list-style-type: none"> <li>- Environment set up – Jupyter notebook;</li> <li>- Computational thinking examples;</li> </ul> <p><b><i>Introduction to Python:</i></b></p> <ul style="list-style-type: none"> <li>- Data types (e.g. int, float, string, bool)</li> <li>- Expressions</li> <li>- Data structures</li> <li>- Conditional statements</li> </ul>
2	<p><b><i>Programming Principles:</i></b></p> <ul style="list-style-type: none"> <li>- Loops</li> <li>- Functions</li> <li>- Variable Scope</li> <li>- Class and objects</li> <li>- File handling</li> </ul>	<p><b><i>Introduction to Python:</i></b></p> <ul style="list-style-type: none"> <li>- Loops</li> <li>- Functions</li> <li>- Variable Scope</li> <li>- Class and objects</li> <li>- File handling</li> </ul>
3	<p><b><i>Data Representation and Operations:</i></b> Python library: NumPy</p> <ul style="list-style-type: none"> <li>- Data representation: arrays, vectors, matrices</li> <li>- Data operations: indexing, math functions</li> </ul>	<p><b><i>Exercises on:</i></b></p> <ul style="list-style-type: none"> <li>- Python programming principles</li> <li>- NumPy</li> </ul>
4	<p><b><i>Data Query and Manipulation:</i></b> Python Library: Pandas</p> <ul style="list-style-type: none"> <li>- Data frame: create, index, read/write to file, summarize statistics, and fill and drop values</li> </ul> <p><b><i>Data Display:</i></b> Python Libraries: Matplotlib</p> <ul style="list-style-type: none"> <li>- Data Display for line charts, bar charts, box plot, scatter plot, and histograms</li> </ul> <p><b><i>Review:</i></b> NumPy, Pandas, Matplotlib</p>	<p><b><i>Exercises on:</i></b></p> <ul style="list-style-type: none"> <li>- Pandas</li> <li>- Data display</li> </ul>
5	<p><b><i>Midterm 1</i></b> <b><i>Data Modeling -Statistical Approach:</i></b></p>	<p><b><i>Exercises on:</i></b></p> <ul style="list-style-type: none"> <li>- Causality and simulation</li> </ul>

	<ul style="list-style-type: none"> <li>- Establishing causality</li> <li>- Randomness: iteration, simulation, probabilities</li> </ul>	<ul style="list-style-type: none"> <li>- Probabilities</li> </ul>
6	<p><b>Data Modeling -Statistical Approach:</b></p> <ul style="list-style-type: none"> <li>- Sampling and empirical distributions</li> </ul> <p><b>Hypothesis testing:</b></p> <ul style="list-style-type: none"> <li>- General concept and examples of assessing models</li> <li>- Comparing proportions, type1 &amp; type2 errors, p-value.</li> </ul>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- Sampling</li> <li>- Hypothesis testing</li> </ul>
7	<p><b>Statistical Analysis:</b></p> <ul style="list-style-type: none"> <li>- Comparing two samples: A/B Testing</li> <li>- Confidence intervals</li> </ul>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- A/B testing</li> <li>- Confidence intervals</li> </ul>
8	<p><b>Statistical Analysis:</b></p> <ul style="list-style-type: none"> <li>- Interpreting confidence intervals</li> <li>- Center and spread</li> <li>- Normal distribution</li> </ul>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- Confidence intervals</li> <li>- Center and spread</li> <li>- Normal distribution</li> </ul>
9	<p><b>Statistical Analysis:</b></p> <ul style="list-style-type: none"> <li>- Sample means</li> <li>- Review</li> </ul> <p><b>Midterm 2</b></p>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- Sample means</li> </ul>
10	<p><b>Data Modeling - Machine Learning Approach:</b></p> <ul style="list-style-type: none"> <li>- Correlation</li> <li>- Linear regression</li> <li>- Least squares</li> </ul> <p><b>Issue final projects + presentation template</b></p>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- Correlation</li> <li>- Least squares</li> <li>- Computing residuals</li> </ul>
11	<p><b>Data Modeling - Machine Learning Approach:</b></p> <ul style="list-style-type: none"> <li>- Residuals</li> <li>- Regression inference</li> <li>- Evaluation metrics: accuracy, error</li> </ul>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- Computing residuals</li> <li>- Regression</li> <li>- Regression with evaluation</li> </ul>
12	<p><b>Classification:</b></p> <ul style="list-style-type: none"> <li>- Supervised learning</li> <li>- Nearest neighbor</li> </ul> <p><b>Classification Evaluation and Making Decisions:</b></p> <ul style="list-style-type: none"> <li>- Confusion matrix, precision, recall, accuracy, F-score</li> <li>- Making decisions</li> </ul> <p><b>Review</b></p>	<p><b>Exercises on:</b></p> <ul style="list-style-type: none"> <li>- KNN</li> <li>- KNN with evaluation</li> </ul>
13	<p><b>Midterm 3</b></p> <p><b>Final Project Q&amp;A</b></p>	<p><b>Work on Final Project</b></p>
14	<p><b>Final Project Presentations</b></p> <p><b>Review</b></p>	

**Course Assessment and Grading Criteria:**

There will be three midterm exams and one comprehensive final project for the course. In addition, lab participation, quizzes, and assignments will also be given credits that will contribute to the final grade. **If the assignments and the final project are submitted late, they will not receive any credit.** Students will be assessed based on the following criteria:

Assessment methods	Weight (%)
Midterm-1	14
Midterm-2	14
Midterm-3	14
Lab participation	6
Quizzes	12
Assignments	25
Final project	15
Overall total	100

The modality of each assessment method will be specified on Blackboard.

**Note** - Students are encouraged to discuss the grading of assignments with the instructor. Grade adjustments will only be made if the student discusses the assignment with the instructor within one week after receiving a grade for the assignment. All grades will be posted on Blackboard.

**Grading:**

The following grading scale will be used:

- A (90.0 - 100)
- B (80.0 – 89.9)
- C (70.0 – 79.9)
- D (55.0 – 69.9)
- F (<55.0)

**Late Assignment Policy:**

Late submissions will not be accepted unless the student has a valid excuse due to illness.

**Academic Integrity:**

Academic integrity is taking responsibility for one’s own class and/or course work, being individually accountable, and demonstrating intellectual honesty and ethical behavior. Academic

integrity is a personal choice to abide by the standards of intellectual honesty and responsibility. Because education is a shared effort to achieve learning through the exchange of ideas, students, faculty, and staff have the collective responsibility to build mutual trust and respect. Ethical behavior and independent thought are essential for the highest level of academic achievement, which then must be measured. Academic achievement includes scholarship, teaching, and learning, all of which are shared endeavors. Grades are a device used to quantify the successful accumulation of knowledge through learning. Adhering to the standards of academic integrity ensures grades are earned honestly. Academic integrity is the foundation upon which students, faculty, and staff build their educational and professional careers. [Texas Tech University (“University”) Quality Enhancement Plan, Academic Integrity Task Force, 2010]. Academic Misconduct is explained in Part IB of the [Texas Tech Student Handbook](#).

**Religious Holy Day Statement:**

“Religious holy day” means a holy day observed by a religion whose places of worship are exempt from property taxation under Texas Tax Code §11.20. A student who intends to observe a religious holy day should make that intention known to the instructor prior to the absence. A student who is absent from classes for the observance of a religious holy day shall be allowed to take an examination or complete an assignment scheduled for that day within a reasonable time after the absence. A student who is excused may not be penalized for the absence; however, the instructor may respond appropriately if the student fails to complete the assignment satisfactorily.

**Americans with Disabilities Act (ADA):**

Any student who, because of a disability, may require special arrangements in order to meet the course requirements should contact the instructor as soon as possible to make any necessary arrangements. Students should present appropriate verification from Student Disability Services during the instructor’s office hours. Please note: instructors are not allowed to provide classroom accommodations to a student until appropriate verification from Student Disability Services has been provided. For additional information, please contact Student Disability Services in West Hall or call 806-742-2405.

**Civility in the Classroom Statement:**

Texas Tech University is a community of faculty, students, and staff that enjoys an expectation of cooperation, professionalism, and civility during the conduct of all forms of university business, including the conduct of student–student and student–faculty interactions in and out of the classroom. Further, the classroom is a setting in which an exchange of ideas and creative thinking should be encouraged and where intellectual growth and development are fostered. Students who disrupt this classroom mission by rude, sarcastic, threatening, abusive or obscene language and/or behavior will be subject to appropriate sanctions according to university policy. Likewise, faculty members are expected to maintain the highest standards of professionalism in all interactions with all constituents of the university ([www.depts.ttu.edu/ethics/matadorchallenge/ethicalprinciples.php](http://www.depts.ttu.edu/ethics/matadorchallenge/ethicalprinciples.php)).