

# CS 488/588 – Applied Data Science with Python

**Semester:** Fall 2024 (August 19 – December 13)

**Credit Hours:** 3

**Instructor:** Alex Vakanski, [vakanski@uidaho.edu](mailto:vakanski@uidaho.edu)

**Office Location:** TAB 311, Idaho Falls Center

## Course Delivery Methods:

- Virtual meetings (live meetings, students participate through Zoom)
- Classroom (live meetings, video link from Idaho Falls)
- Online (recorded Zoom videos of lectures available to students to watch after the classes)

## Course Description

The course introduces students to Python tools and libraries that are commonly used by organizations for managing the various phases in the life cycle of data science projects. The content is divided into four main themes. The first theme reviews the fundamentals of Python programming. The second theme focuses on data engineering and explores Python tools for data collection, exploration, and visualization. The next theme covers model engineering and includes topics related to model design, selection, and evaluation for image processing, natural language processing, and time series analysis. The last theme introduces Data Science Operations (DSOps) and encompasses techniques for model serving, performance monitoring, diagnosis, and reproducibility of data science projects deployed in production. Throughout the course, students will gain hands-on experience with various Python libraries for data science workflow management. Additional work is required for graduate credit.

## Learning Outcomes

Upon the completion of the course, the students should demonstrate the ability to:

1. Attain proficiency with commonly used Python frameworks for managing the life cycle of data science projects.
2. Develop pipelines for integrating data from multiple sources, designing predictive models, and deploying the models.
3. Apply Python tools for data collection, analysis, and visualization, such as NumPy, Pandas, Matplotlib, and Seaborn, to real-world datasets.
4. Implement machine learning algorithms for image processing, natural language processing, and time series analysis using Python-based frameworks, such as Scikit-Learn, Keras, TensorFlow, and PyTorch.
5. Understand the principles of model selection and evaluation, including hyperparameter tuning, cross-validation, and regularization.
6. Understand the primary characteristics of current Python libraries for deployment, continuous integration, and monitoring of data science projects.
7. Deploy data science projects as web applications using Flask, FastAPI, and Django, and to cloud servers using Microsoft's Azure platform.

## Prerequisites

The course requires to have basic programming skills in Python. While having knowledge of data science methods would be advantageous, it is not mandatory.

## Textbooks

There are no required textbooks for this course.

## Grading

Student assessment will be based on 6 homework assignments (worth 60 pts), 3 quizzes (worth 30 marks), and class participation and engagement (worth 10 marks).

<i>Assessment Component</i>	<i>Marks</i>
Assignments (x6)	60
Quizzes (x3)	30
Class participation	10
<i>Total</i>	<i>100</i>

## Tentative Schedule

<u>Date</u>	<u>Topics, Tests, Assignments</u>
<b>Week 1:</b> Aug. 19	<b>Introduction</b> <b>Lecture 1: A Short History and Current State of Artificial Intelligence</b> <b>Python Review</b> <b>Lecture 2: Python Data Types</b> <ul style="list-style-type: none"> <li>– Numbers, Strings, Lists, Dictionaries, Tuples, Sets, Other Types</li> </ul> <b>Tutorial 1:</b> Jupyter Notebooks
<b>Week 2:</b> Aug. 26	<b>Python Review</b> <b>Lecture 3: Statements, Files</b> <ul style="list-style-type: none"> <li>– Conditional statements, loop statements</li> <li>– File handling, opening, writing, appending, iterating through a file, storing objects in files</li> </ul> <b>Lecture 4: Functions, Iterators, Generators</b> <ul style="list-style-type: none"> <li>– Function definition, function call, argument passing, namespace and scope</li> <li>– Iterators, generators, the 'yield' statement</li> </ul> <b>Tutorial 2:</b> Terminal and Command Line
<b>Week 3:</b> Sep. 2	<b>Python Review</b> <b>Lecture 5: Object-Oriented Programming</b> <ul style="list-style-type: none"> <li>– Class definition, attributes, methods, inheritance, polymorphism, special methods</li> </ul> <b>Lecture 6: Exceptions, Modules, Packages</b>

	<ul style="list-style-type: none"> <li>– Error types, error handling with 'try', 'except', 'finally', 'raise', 'assert'</li> <li>– Creating and using modules, module namespaces, organizing modules into packages</li> </ul> <p><b>Tutorial 3:</b> Python IDEs, Visual Studio Code</p> <p><u><a href="#">Due: Homework 1</a></u></p>
<b>Week 4:</b> Sep. 9	<p><b>Data Engineering Pipelines</b></p> <p><b>Lecture 7: NumPy for Array Operations</b></p> <ul style="list-style-type: none"> <li>– Array creation, indexing, broadcasting and vectorization, reshaping arrays, linear algebra with NumPy</li> </ul> <p><b>Lecture 8: Data Manipulation with Pandas</b></p> <ul style="list-style-type: none"> <li>– Importing data, indexing and slicing, merging, sorting, exporting data, dealing with missing values</li> </ul> <p><b>Tutorial 4:</b> Virtual Environments</p> <p><u><a href="#">Due: Homework 2</a></u></p>
<b>Week 5:</b> Sep. 16	<p><b>Data Engineering Pipelines</b></p> <p><b>Lecture 9: Data Visualization with Matplotlib</b></p> <ul style="list-style-type: none"> <li>– State-based approach, customizing plot appearance, saving figures, subplots, plotting functions, object-oriented approach</li> </ul> <p><b>Lecture 10: Data Visualization with Seaborn</b></p> <ul style="list-style-type: none"> <li>– Creating statistical plots, visualizing relationships between variables, customizing plot styles</li> </ul> <p><b>Tutorial 5:</b> Web Scraping</p> <p><b>Quiz 1</b></p>
<b>Week 6:</b> Sep. 23	<p><b>Data Engineering Pipelines</b></p> <p><b>Lecture 11: Data Exploration and Preprocessing</b></p> <ul style="list-style-type: none"> <li>– Exploratory data analysis, preprocessing numerical data, preprocessing categorical data</li> </ul> <p><b>Lecture 12: Databases and SQL</b></p> <ul style="list-style-type: none"> <li>– Intro to SQLite, creating databases, querying, sorting, filtering, joining tables, conditional expressions, grouping data, subqueries</li> </ul> <p><b>Tutorial 6:</b> Google Colab</p>
<b>Week 7:</b> Sep. 30	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 13: Scikit-Learn Library for Data Science</b></p> <ul style="list-style-type: none"> <li>– Supervised and unsupervised learning algorithms, regression, classification, clustering</li> <li>– <i>k</i>-Nearest Neighbors, Support Vector Machines, Logistic Regression, Decision Trees, Random Forest, Naïve Bayes, Stochastic Gradient Descent</li> </ul> <p><b>Tutorial 7:</b> Image Processing with Python</p> <p><u><a href="#">Due: Homework 3</a></u></p>
<b>Week 8:</b> Oct. 7	<p><b>Quiz 2</b></p>

<b>Week 9:</b> Oct. 14	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 14: Ensemble Models</b></p> <ul style="list-style-type: none"> <li>– Voting, bagging, boosting, stacking ensembles</li> </ul> <p><b>Lecture 15: Artificial Neural Networks</b></p> <ul style="list-style-type: none"> <li>– Activation functions, losses, backpropagation, fully-connected layers, classification and regression</li> </ul> <p><b>Tutorial 8:</b> TensorFlow</p>
<b>Week 10:</b> Oct. 21	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 16: Convolutional Neural Networks with Keras-TensorFlow, PyTorch</b></p> <ul style="list-style-type: none"> <li>– Convolutional layers, models for image classification, transfer learning and fine-tuning pretrained models</li> </ul> <p><b>Lecture 17: Model selection, Hyperparameter Tuning, Callbacks</b></p> <ul style="list-style-type: none"> <li>– Grid search, cross-validation, model evaluation, callbacks for model monitoring</li> </ul> <p><b>Tutorial 9:</b> PyTorch</p>
<b>Week 11:</b> Oct. 28	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 18: Neural Networks with PyTorch</b></p> <ul style="list-style-type: none"> <li>– Data loaders, model definition, training, and evaluation, custom datasets, model saving and loading</li> </ul> <p><b>Lecture 19: Natural Language Processing</b></p> <ul style="list-style-type: none"> <li>– Preprocessing text data, tokenization, representation of word groups, sequence model approach, word embeddings</li> </ul> <p><b>Tutorial 10:</b> TensorFlow Datasets</p> <p><a href="#">Due: Homework 4</a></p>
<b>Week 12:</b> Nov. 4	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 20: Transformer Networks</b></p> <ul style="list-style-type: none"> <li>– Self-attention mechanism, multi-head attention, positional encoding, encoder and decoder blocks, fine-tuning pretrained models</li> </ul> <p><b>Lecture 21: NLP with Hugging Face</b></p> <ul style="list-style-type: none"> <li>– Pipelines for NLP tasks, tokenizers, datasets, models</li> </ul> <p><b>Tutorial 11:</b> Project Monitoring with CometML, Weights and Biases</p>
<b>Week 13:</b> Nov. 11	<p><b>Model Engineering Pipelines</b></p> <p><b>Lecture 22: Diffusion Models for Text-to-Image Generation</b></p> <ul style="list-style-type: none"> <li>– Denoising diffusion models, text encoder, latent diffusion models, generating images with Stable Diffusion</li> </ul> <p><b>Lecture 23: Large Language Models</b></p> <ul style="list-style-type: none"> <li>– Fine-tuning a pretrained model, low-rank adaptation (LoRA), prefix tuning, prompt tuning</li> </ul> <p><b>Tutorial 12:</b> Git and Version Control</p> <p><a href="#">Due: Homework 5</a></p>

<b>Week 14:</b> Nov. 25	<b>Deployment Pipelines</b> <b>Lecture 24: Introduction to Data Science Operations (DSOps)</b> <ul style="list-style-type: none"> <li>– DS project life cycle, levels of automation, model deployment, model serving in production</li> </ul> <b>Lecture 25: Deploying Projects as Web Applications</b> <ul style="list-style-type: none"> <li>– Intro to Flask, FastAPI, Django, creating RESTful API to serve data and model predictions, models for tabular data and image classification</li> </ul> <b>Tutorial 13:</b> GitHub Actions
<b>Week 15:</b> Dec. 2	<b>Deployment Pipelines</b> <b>Lecture 26: Deploying Projects to the Cloud</b> <ul style="list-style-type: none"> <li>– Intro to Azure Machine Learning, no-code ML, code-based ML, training deep learning models with Azure ML</li> </ul> <b>Lecture 27: Reproducible Projects, Docker Containers, Kubernetes</b> <ul style="list-style-type: none"> <li>– Intro to Docker and containerization, building and managing Docker containers, Kubernetes for container orchestration and management</li> </ul> <a href="#">Due: Homework 6</a>
<b>Week 16:</b> Dec. 9	<b>Quiz 3</b>

### Academic Integrity

Students are expected to adhere to the highest academic standards of honesty and integrity. At UI, we assume students will do their own work. Plagiarism—passing off someone else's work as your own, without citing the source—should not be tolerated. This includes direct copying, rephrasing, and summarizing, as well as taking someone else's idea and putting it in different words. The best avenue for avoiding plagiarism issues is to fully cite all sources used for preparing assignments, texts, exams.

### Learning Civility

In any environment in which people gather to learn, it is essential that all members feel as free and safe as possible in their participation. To this end, it is expected that everyone in this course will be treated with mutual respect and civility, with an understanding that all of us (students, instructors, professors, guests, and teaching assistants) will be respectful and civil to one another in discussion, in action, in teaching, and in learning.

Should you feel our classroom interactions do not reflect an environment of civility and respect, you are encouraged to meet with your instructor during office hours to discuss your concern. Additional resources for expression of concern or requesting support include the Dean of Students office and staff (208-885--6757), the Uofl Counseling & Testing Center's confidential services (208-885-6716), the Uofl Office of Equity and Diversity (208-885-2468), or the Office of Civil Rights and Investigations (208-885-4285).

### Center for Disability Access & Resources (CDAR)

University of Idaho is committed to ensuring an accessible learning environment where course or instructional content are usable by all students and faculty. If you believe that you require disability-related academic adjustments for this class (including pregnancy-related disabilities), please contact Center for Disability Access and Resources (CDAR) to discuss eligibility. A current accommodation letter from CDAR is required before any modifications, above and beyond what is otherwise available for all other students in this class will be provided. Please be advised that disability-related academic adjustments are not retroactive. CDAR is located at the Bruce Pitman Building, Suite 127. Phone is 208-885-6307 and e-mail is [cdar@uidaho.edu](mailto:cdar@uidaho.edu). For a complete listing of services and current business hours visit <https://www.uidaho.edu/current-students/cdar>.

### **Inclusivity Statement**

As a professor/course instructor at the University of Idaho, I acknowledge the importance of diversity and inclusion and how these attributes contribute to the promotion of a positive educational experience. It is my intent to facilitate a healthy, productive, and safe learning environment where diverse thoughts, perspectives, and experiences are welcomed, and individuals' identities (including, but not limited to: race, sex, class, sexual orientation, gender identity, ability, religious beliefs, etc.) are valued and honored. I recognize that as an educator, it is my responsibility to take the initiative to continually learn about diverse perspectives and identities; therefore, if at any point during the course, you feel uncomfortable or concerned, I am more than willing to discuss suggestions, feedback, and anything else that might improve the general effectiveness of this course.

### **Healthy Vandals Policies**

Please visit the [University of Idaho COVID-19 webpage](#) often for the most up-to-date information about the UofI's response to Covid-19.

### **Vandal Food Pantry**

The [Vandal Food Pantry](#) is a free resource stocked weekly with food, grocery bags, and various hygiene items. Its eight locations across campus are accessible during building hours and open to all. Please take what you need.

### **Green Dot Safety Program**

What's Your Green Dot? It's up to all of us to make a safer campus. Vandal Green Dot is a program that helps students learn about the power of the bystander, how to recognize potentially risky situations, and realistic ways to intervene. Together we can bring down the number of people being hurt by interpersonal violence on our campus. No one has to do everything, but everyone has to do something! Learn more and get involved by visiting [UI's Green Dot Safety Program](#) or emailing [greendot@uidaho.edu](mailto:greendot@uidaho.edu).

### **Help and Resources**

#### *Student Resources*

The University of Idaho provides student support to ensure a successful learning experience.

- [Student Resources Webpage](#)
- [SI-PASS \(Peer Assisted Study Sessions\)](#) SI-PASS provides regularly scheduled, peer-led study sessions for difficult courses.

### *Library Help*

The Uofl Library website has many databases that will help you find relevant and reliable books, articles, images, and more. Don't hesitate to contact a librarian for research assistance.

- [Uofl Library Website](#)
- [Help - Reference Services](#)
- [Help for Distance Ed Students](#)

### *Technology Help*

The Uofl Student Technology Center provides many technology related services to students.

- PHONE: 208-885-HELP (208-885-4357)
- Technology Help Email: [support@uidaho.edu](mailto:support@uidaho.edu)
- [Technology Help Website](#)

### *Writing Support*

The Uofl Writing Center provides one-on-one assistance to student writers and other members of the campus community.

- PHONE: 208-885-6644
- Writing Center Email: [writingcenter@uidaho.edu](mailto:writingcenter@uidaho.edu)
- [Writing Center Website](#)

### **Uofl Moscow Land Acknowledgement**

Uofl Moscow is located on the homelands of the Nimiipuu (Nez Perce), Palus (Palouse) and Schitsu'umsh (Coeur d'Alene) tribes. We extend gratitude to the indigenous people that call this place home, since time immemorial. Uofl recognizes that it is our academic responsibility to build relationships with the indigenous people to ensure integrity of tribal voices.