

ADVANCED TOPICS OF MACHINE LEARNING

Course Description

Author: Eng. Carlos Andrés Sierra, M.Sc.
cavirguezs@udistrital.edu.co

Full-time Adjunct Professor
Computer Engineering Program
School of Engineering
Universidad Distrital Francisco José de Caldas

2026-I



Outline

- 1 You don't know who I am
- 2 Course Overview
- 3 Grading & Rules
- 4 Bibliography



Outline

1 You don't know who I am

2 Course Overview

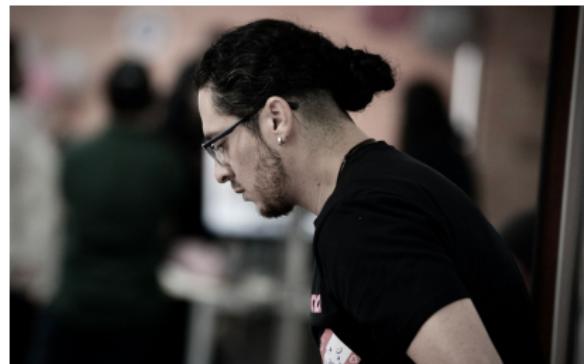
3 Grading & Rules

4 Bibliography



Academic Experience

- Computer Engineer, M.Sc. in Computer Engineering, and researcher for 16 years. *Evol. Comp.
Art. Life*
- 8 years as **full-time associate professor** at colleges, in Computer Engineering programs.
- 3 years as **lecturer professor** for both colleges and government STEM programs.
- Speaker at **IEEE** events and colleges in Colombia, Brazil, and Bolivia.



Non-academic Experience



- PyCon Colombia and Python Bogotá co-organizer.
- 3 years as **software engineer** for several **tech companies** in Colombia.
- 3 years as **Technical Leader** of Machine Learning and Data Science at a USA startup.
- 1.5 years as ML Ops Engineer for a **Fintech** company in LATAM.
- Currently, **Senior Engineering Manager** of Data Engineering and Machine Learning at Blend 360.



Outline

1 You don't know who I am

2 Course Overview

3 Grading & Rules

4 Bibliography



Overview

This course is designed to introduce undergraduate students to advanced machine learning techniques as part of the foundation for becoming an experienced AI engineer capable of developing intelligent solutions.

~5 years ago
The course starts with a comprehensive study of reinforcement learning and generative models. Then, it transitions into advanced unsupervised learning and deep learning optimization techniques. Finally, we will focus on online learning, transfer learning, and model interpretability for both research and industry applications.

Classes will consist of theoretical lectures, algorithm implementation sessions, and cutting-edge research projects. Also, you must complete some readings from advanced ML research papers and deep learning. In addition, there will be a semester-long research project, one final course test, and four practical challenges.



Overview

This course is designed to **introduce undergraduate students** to **advanced machine learning techniques** as part of the foundation for becoming an experienced *AI engineer* capable of developing *intelligent solutions*.

The course starts with a **comprehensive study** of **reinforcement learning** and **generative models**. Then, it transitions into **advanced unsupervised learning** and **deep learning optimization techniques**. Finally, we will focus on **online learning**, **transfer learning**, and **model interpretability** for both **research** and **industry applications**.

Classes will consist of **theoretical lectures**, **algorithm implementation sessions**, and **cutting-edge research projects**. Also, you must complete some readings from *advanced ML research papers* and *deep learning*. In addition, there will be a **semester-long research project**, **one final course test**, and **four practical challenges**.



Overview

This course is designed to **introduce undergraduate students** to **advanced machine learning techniques** as part of the foundation for becoming an experienced *AI engineer* capable of developing *intelligent solutions*.

The course starts with a **comprehensive study** of **reinforcement learning** and **generative models**. Then, it transitions into **advanced unsupervised learning** and **deep learning optimization techniques**. Finally, we will focus on **online learning**, **transfer learning**, and **model interpretability** for both **research** and **industry applications**.

Classes will consist of **theoretical lectures**, **algorithm implementation sessions**, and **cutting-edge research projects**. Also, you must complete some readings from **advanced ML research papers** and **deep learning**. In addition, there will be a **semester-long research project**, **one final course test**, and **four practical challenges**.



Goals

The main goal of this course is to provide students with advanced concepts and cutting-edge techniques for intelligent systems design capable of analyzing data, optimizing processes, and improving human-machine interactions.

At the end of this course you should be able to implement advanced ML models including reinforcement learning, generative models, and transfer learning techniques. Also, you should be able to develop complete AI solutions that solve real-world problems using state-of-the-art algorithms and optimization techniques.



Goals

The **main goal** of this course is to **provide students with advanced concepts and cutting-edge techniques** for **intelligent systems design** capable of analyzing data, optimizing processes, and improving **human-machine interactions**.

At the end of this course you should be able to **implement advanced ML models** including **reinforcement learning, generative models, and transfer learning techniques**. Also, you should be able to **develop complete AI solutions** that solve real-world problems using **state-of-the-art algorithms** and optimization techniques.



Prerequisites

Cost

This is an advanced course, so you must have some knowledge in:

- **Programming in Python** (advanced level), including libraries like NumPy, Pandas, and Scikit-learn.
- Mathematics fundamentals: linear algebra, calculus, and probability theory.
- Statistics and statistical inference: hypothesis testing, distributions, and regression analysis.
- Basic Machine Learning concepts: supervised/unsupervised learning, model evaluation, and cross-validation.
- Git usage, Jupyter Notebooks, and experience with ML frameworks like TensorFlow or PyTorch.
- Recommended: Experience with Docker containers and research paper reading.



Prerequisites

This is an advanced course, so you must have some knowledge in:

- **Programming** in **Python** (advanced level), including libraries like NumPy, Pandas, and Scikit-learn.
- **Mathematics** fundamentals: linear algebra, calculus, and probability theory.
- **Statistics** and statistical inference: hypothesis testing, distributions, and regression analysis.
- **Basic Machine Learning** concepts: supervised/unsupervised learning, model evaluation, and cross-validation.
- **Git** usage, **Jupyter Notebooks**, and experience with ML frameworks like **TensorFlow** or **PyTorch**.
- **Recommended:** Experience with Docker containers and research paper reading.



Prerequisites

This is an advanced course, so you must have some knowledge in:

- **Programming** in **Python** (advanced level), including libraries like NumPy, Pandas, and Scikit-learn.
- **Mathematics** fundamentals: **linear algebra**, **calculus**, and **probability theory**.
- **Statistics** and **statistical inference**: hypothesis testing, distributions, and regression analysis.
- **Basic Machine Learning** concepts: supervised/unsupervised learning, model evaluation, and cross-validation.
- **Git** usage, **Jupyter Notebooks**, and experience with ML frameworks like **TensorFlow** or **PyTorch**.
- **Recommended**: Experience with Docker containers and research paper reading.



Prerequisites

This is an advanced course, so you must have some knowledge in:

- **Programming** in **Python** (advanced level), including libraries like NumPy, Pandas, and Scikit-learn.
- **Mathematics** fundamentals: **linear algebra**, **calculus**, and **probability theory**.
- **Statistics** and **statistical inference**: hypothesis testing, distributions, and regression analysis.
- **Basic Machine Learning** concepts: supervised/unsupervised learning, model evaluation, and cross-validation.
- **Git usage**, **Jupyter Notebooks**, and experience with ML frameworks like **TensorFlow** or **PyTorch**.
- **Recommended**: Experience with Docker containers and research paper reading.



Prerequisites

This is an advanced course, so you must have some knowledge in:

- **Programming** in **Python** (advanced level), including libraries like NumPy, Pandas, and Scikit-learn.
- **Mathematics** fundamentals: **linear algebra**, **calculus**, and **probability theory**.
- **Statistics** and **statistical inference**: hypothesis testing, distributions, and regression analysis.
- **Basic Machine Learning** concepts: supervised/unsupervised learning, model evaluation, and cross-validation.
- **Git** usage, **Jupyter Notebooks**, and experience with ML frameworks like **TensorFlow** or **PyTorch**.
- **Recommended**: Experience with Docker containers and research paper reading.



Syllabus

Period	Topic	Time
Period I	Introduction to Machine Learning	2 sessions
	Reinforcement Learning Fundamentals	7 sessions
	Generative and Probabilistic Models	5 sessions
	Advanced Unsupervised Learning I	3 sessions
	Course Project Catch-Up	2 sessions
Period II	Advanced Unsupervised Learning II	4 sessions
	Deep Learning	4 sessions
	Online Learning	2 sessions
	ML Interpretability Techniques	3 sessions
	Final Test	1 session
Period III	Project Dissertation	2 sessions



Outline

1 You don't know who I am

2 Course Overview

3 Grading & Rules

4 Bibliography



Grades Percentages

35%

Period	Item	Percentage
Period I	Mini-Projects	20%
	Project Catch-Up	15%
Period II	Mini-Projects	15%
	Course Test	20%
Period III	Paper + Poster	15%
	Report + Implementation	10%
	Presentation	5%

45%



Don't hate the player, hate the game

- All assignments must be submitted on time, and in English.
Grammar and spelling will **not** be evaluated.
- Copying and pasting from the internet are **forbidden**. Please **develop** your own ideas and solutions.
- Class attendance is not mandatory. If you **miss** classes, you must study independently.
- No cell phones, smartwatches, WhatsApp, Tinder, or smart devices.
Just you and your brain. Pay attention in class.
- Communication with me must be via email or **Slack**. I will **not** answer any questions via WhatsApp.



Code of Conduct

- Always be **respectful** to your **classmates** and to me. You must be **kind** to everyone inside (*and outside*) the classroom.
- There is **(no best programming language, tool, or technology)** There are only **better** or **worse** solutions.
- You must be **honest** with your work. If you **(don't know something,** just **ask** me. I will be **glad** to help you.
- You must be **responsible** with your work. If you don't submit **on time** please **don't complain**.
- You must not be **disruptive** or **negatively affect** the **classroom environment**. If you do, I will **ask you** to **leave** the classroom.



Outline

- 1 You don't know who I am
- 2 Course Overview
- 3 Grading & Rules
- 4 Bibliography



Bibliography

Recommended bibliography:

- **Reinforcement Learning: An Introduction**, by Richard S. Sutton & Andrew G. Barto.
- **Deep Learning**, by Ian Goodfellow, Yoshua Bengio, & Aaron Courville.
- **Generative Deep Learning**, by David Foster.
- **Pattern Recognition and Machine Learning**, by Christopher M. Bishop.
- **The Elements of Statistical Learning**, by Trevor Hastie, Robert Tibshirani, & Jerome Friedman.
- **Interpretable Machine Learning**, by Christoph Molnar.
- **Online Learning and Online Convex Optimization**, by Shai Shalev-Shwartz.
- **Bayesian Reasoning and Machine Learning**, by David Barber.



Outline

- 1 You don't know who I am
- 2 Course Overview
- 3 Grading & Rules
- 4 Bibliography



Thanks!

Questions?



My Profile: www.linkedin.com/in/casierrav

