

CS-466/566: Math for AI

Module 01: Course Introduction

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The University of Alabama

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Welcome

Welcome to CS-466/566: Math for AI!

Instructor Information

Instructor: Dr. Mahmoud Mahmoud
Institution: The University of Alabama
Email: mmahmoud@ua.edu
Office Hours: By appointment

Course Website

You can find all course materials, announcements, and updates on the [Blackboard](#).

Prerequisites

A background in programming is recommended for this course. We will cover the basics from scratch.

Academic Integrity

All students are expected to adhere to the university's academic integrity policy. Collaboration is encouraged, but all submitted work must be your own.

About Me

Experience

- **Associate Professor, CS, The University of Alabama (Current)**
- **Associate Professor, ECE, NC A&T State University (2024-2025)**
- **Assistant Professor, ECE, NC A&T State University (2019-2024)**

Research & Achievements

- **Interests:** Trustworthy AI, Cybersecurity, and Privacy in CPS.
- **Fellowship:** Summer Faculty Fellow at Fermi National Accelerator Laboratory.
- **Editor:** Editor IEEE Transactions on Dependable and Secure Computing
- **Funding:** Active external funding from **NSF, NASA, DOE, CIA**, etc.



What do you think this course is about?

- Is this a traditional pure mathematics course?
 - Answer: **✗** not exactly
- Is this a pure machine learning course?
 - Answer: **✗** not exactly
- Is this a programming course?
 - Answer: **✗** not exactly
- Answer: The course spans the **four pillars of machine learning**: **✓**
 - Linear Algebra
 - Calculus
 - Probability & Statistics
 - Optimization

Machine Learning and AI Everywhere!

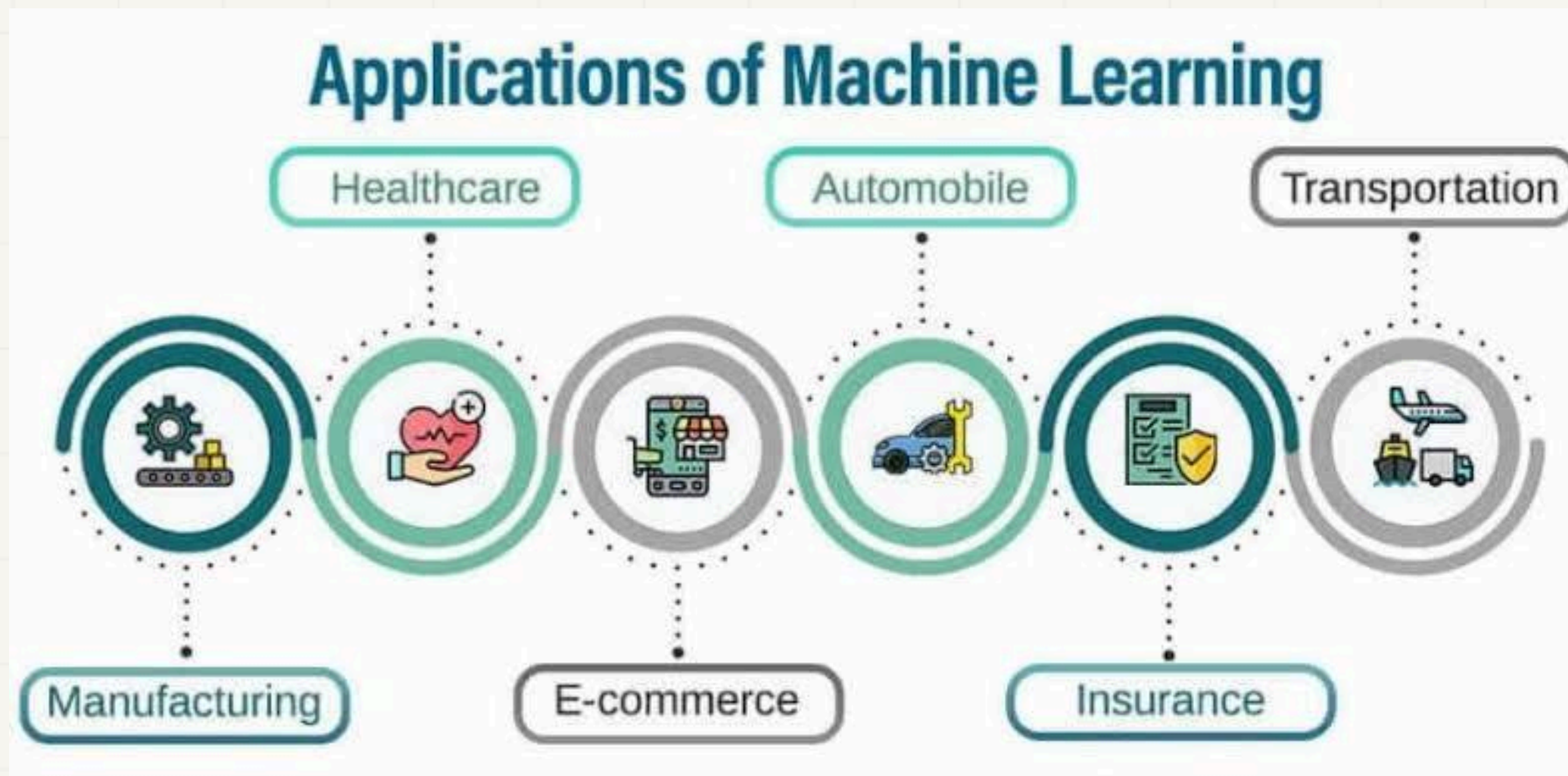


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Course Structure and Content

Week	Tuesday	Thursday	Topic	Source
1	—	Jan 8	Course Introduction + Motivation; What is Math for ML?	Syllabus
2	Jan 13	Jan 15	The Geometry of Data	Ch 1–2
3	Jan 20	Jan 22	Computational Linear Algebra	Ch 3–4
4	Jan 27	Jan 29	Systems & Eigen-Geometry	Ch 5–6
5	Feb 3	Feb 5	Dimensionality Reduction & Factorization	Ch 7
6	Feb 10	Feb 12	Graphs & Stochastic Systems	Ch 8 + Supp.
7	Feb 17	Feb 19	Calculus Foundations for ML	Ch 9–11
8	Feb 24 (Q&A)	Feb 26 (Midterm)	Midterm Review & Exam	—
9	Mar 3	Mar 5	The Engine of Learning (Derivatives)	Ch 12
10	Mar 10	Mar 12	Single-Variable Optimization & Integration	Ch 13–14
—	Mar 17	Mar 19	Spring Break – No Class	—
11	Mar 24	Mar 26	Multivariable Calculus	Ch 15–16
12	Mar 31	Apr 2	High-Dimensional Optimization	Ch 17
13	Apr 7	Apr 9	Probability Spaces	Ch 18
14	Apr 14	Apr 16	Random Variables & Distributions	Ch 19
15	Apr 21	Apr 23 (Last Class)	Expectation & Information Theory	Ch 20 + Supp.

** Timeline and content are subject to change depending on the pace of the class.**

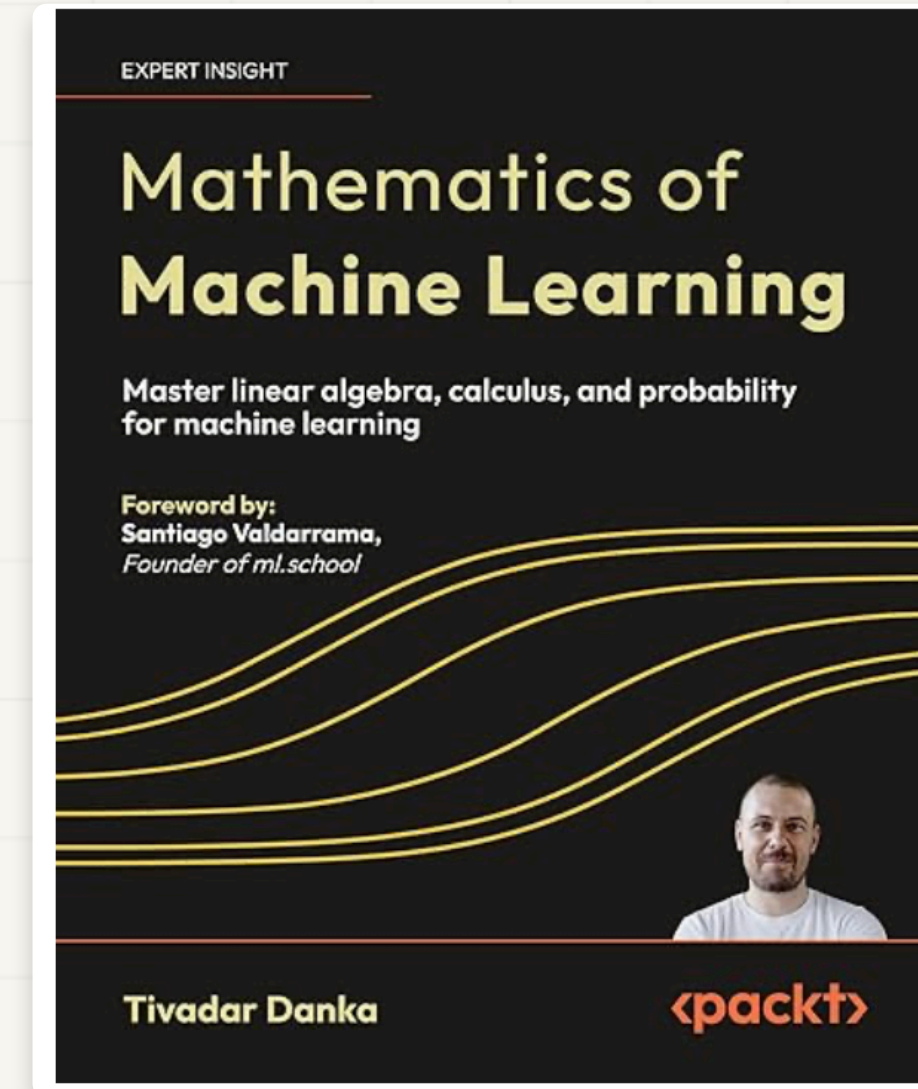


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Resources and Python Help

- **Recommended Books (No book required):**
 - [Mathematics of Machine Learning](#) (Tivadar Danka). *Master linear algebra, calculus, and probability for machine learning.*
 - [Mathematics for Machine Learning](#) (Deisenroth, Faisal, Ong).
- **Visual Learning:**
 - [3Blue1Brown](#) (Linear Algebra, Calculus, Neural Networks).
- **Python Libraries:**
 - Core: [NumPy](#), [SciPy](#), [Matplotlib](#).
 - Autodiff & DL: [PyTorch](#), [JAX](#).



What you need to know?

- **Some Machine Learning Knowledge**
 - This is *not* a pure ML/Programming/Math course.
- **Knowledge of Python**
 - Preferably Python for ML.
 - Experience with [Jupyter Notebooks](#) or [Google Colab](#) is helpful.
- **Knowledge of statistics and probability is strongly recommended!**



Class Expectations

- **Emphasis on interaction — lots of it!**
- I encourage questions from you throughout the class.
- Don't hesitate — ask immediately whenever something isn't clear.
- You might get a bonus point for your questions!
- Please stay engaged: avoid chatting with friends, working on other courses, or distracting others during class.

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Assessments and Course Resources

Component	CS 466	CS 566
Labs / Programming Assignments	45%	45%
Quizzes & Homework	15%	15%
Midterm Exam	15%	10%
Final Exam	25%	15%
Graduate Project	—	15%
Total	100%	100%

Course Resources

- All papers available per lecture.
- All lecture notes available per lecture.
- **No required book but recommended resources are available.**

Programming Assignments

- 4–5 assignments throughout the semester covering:
 - Linear Algebra
 - Calculus
 - Probability
 - Optimization
- All submissions via Blackboard.
- Assignments are individual.
- Programming + Report.
- Google Colab recommended.

Quizzes & Homework

- Weekly homework assignments.
- **All submissions via Blackboard.**
- **Assignments are individual.**
- Math style questions.
- **Due at specified deadlines.**
- **10% penalty per day late.**
- **Max 3 days late.**
- **>3 days = Zero.**
- **No extensions except emergencies.**

Class Contributions (+5%)

- Pop Attendance (1 point)
 - > 50 : 1 pt | > 25 : 0.5 pts | < 25 : 0 pts
- Knowledge Quiz (1 point)
 - Short quizzes to check understanding.
- Participation (3 points)
 - Active discussion.
 - Ask/answer questions.
 - Sticky note for contributions.

Grading Scheme

CS 466

Grade	Description
A+	95-100
A	90-94
A-	88-90
B+	85-87
B	83-84
B-	80-82
C+	77-79
C	73-76
C-	70-72
D+	67-69
D	63-66
D-	60-62
F	Below 60

CS 566

Grade	Description
A	90-100
B	80-89
C	70-79
D	60-69
F	Below 60

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Midterm and Final Exams

Overview

- The **midterm and final exams** together are worth **40%** (CS 466) / **25%** (CS 566) of your final grade.
- Exams will feature “**old style**” questions—expect clear, direct problems that test your understanding, not tricky or ambiguous items.

Expected Dates

- **Midterm Exam:** *Week 7* (exact date to be announced)
- **Final Exam:** *During the official final exam period* (date set by the university)

Grading Policy

- If needed, the **highest mark achieved on the exam will be considered the full mark.**
- All students' scores may be **scaled relative to the top score** to ensure fairness.

What to Expect

- Questions will focus on core concepts, problem-solving, and your ability to apply what you've learned.

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Project Expectations (15%) and Due Dates

Project Overview

- Projects are individual.
- **Project Theme:** Choose a core mathematical concept (e.g., Optimization, Matrix Factorization, Probability) applied to AI, provide a proposal, implement the concept from scratch (or a novel variation), and write a research paper.

Key Milestones & Deadlines

- **Project Presentations:** (Most likely Video Recording)
 - Short in-class presentations with peer feedback.
- **Final Deliverables Due:** End of Week 16 (Dec 5)
 - Written report/paper (5–8 pages) and code submission.

Expectations

- Clear problem statement and motivation.
- Demonstration of understanding of course concepts.
- Well-documented code and reproducible experiments.
- Insightful analysis and discussion of results.
- Proper citation of all sources and related work.

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Academic Integrity - Collaboration Guidelines

- **Collaboration is encouraged!**
- **Over the line**
 - Showing your work to a friend – **OK**
 - Sending your work to a friend – **Not OK**
 - Finding sources, ideas, examples – **OK**
 - Copying text, ideas, code – **Not OK**
 - Using ChatGPT/LLM to help with part of the code – **OK** (In most cases)
 - Copying all code from ChatGPT/LLM – **Not OK**
 - Using whatever ML/DL/AI library in Python – **OK**
- **Sometimes, you will be asked to explain your code**

Academic Integrity - Zero Tolerance Policy

- **Zero tolerance!**
 - Immediate failure for the assignment
 - If repeated: failure of the course and referral to the Office of Academic Integrity
 - **I do what I say. So don't do it!**
- **Random search**
 - Testing code for statistical similarity; renaming variable names is not going to fool us!
 - Other technologies

Academic Integrity - Final Warning

- **Reports/assignments/other tasks are your work, so never ever copy from others**
 - Or do not give it to others
 - If detected, you will get a zero
 - No exceptions
- **PLEASE don't violate**
- **No exception will be provided**

Thank You!

