

ECE 586: Vector Space Methods

Course Introduction Video

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- Engineering uses science and mathematics to **invent, design, and build** things that solve problems and achieve practical goals
- From a mathematical perspective, engineering models the world using **vectors** and analyzes it using logic, linear algebra, and optimization
- This class is essentially an applied math class designed for graduate engineers in **data science, signal processing, and quantum computing**
- Its goal is to **refresh and extend your undergraduate knowledge** of linear systems, vector spaces, matrices, and optimization

Major Topics

- Logic and Set Theory
- Metric Spaces and Topology
 - Practical applications include the analysis of iterative algorithms
- Linear Algebra: Normed and Inner-Product Spaces
 - Practical applications include spaces of functions and Markov chains
- Approximation and Projection
 - Practical applications include function approximation and machine learning
- The Four Fundamental Subspaces and Singular Value Decomposition
 - Practical applications include dimensionality reduction
- Optimization in Vector Spaces
 - Practical applications include duality bounds in convex optimization

Example Problem 1

- Approximation Error in Linear Systems
 - Suppose you know a vector b satisfies $Ax = b$ for an invertible matrix A
 - To compute x , you could multiply by A^{-1} to get $A^{-1}b = A^{-1}(Ax) = x$
 - But, what if A is not known perfectly?
 - Let's assume we know $\hat{A} = A + E$ where the error matrix E is "small"
 - Now, we can compute $\hat{x} = \hat{A}^{-1}b$. But, what can we say about $x - \hat{x}$?
- The mathematical study of "closeness" is called **Topology**
 - To answer the above question, one can define distances between objects
 - Using these distances, we can define a constant $c(\hat{A})$ such that
$$\text{distance}(x, \hat{x}) \leq c(\hat{A}) \text{ distance}(A, \hat{A}) = c(\hat{A}) \text{ length}(E)$$

Example Problem 2

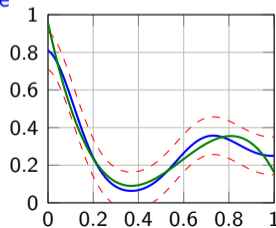
- Approximation of Functions by Simpler Functions

- Suppose you want to find a 3rd-order polynomial that gives a “good” approximation of a function $f(x)$ on the interval $[0, 1]$.
- A key question is, “What do we mean by good?” max error? avg error?
- A standard approach is to minimize the mean squared error and this gives

$$\min_{a_0, a_1, a_2, a_3} \int_0^1 \left[f(x) - (a_0 + a_1x + a_2x^2 + a_3x^3) \right]^2 dx$$

- Given $f(x)$ and the ability to integrate, one can solve this via brute force
 - This course will show there is also an simple intuitive approach
- Solution via Orthogonality and Linear Algebra
 - Compute a_0, a_1, a_2, a_3 using 4 linear equations ($k = 0, 1, 2, 3$) given by

$$\int_0^1 (a_0 + a_1x + a_2x^2 + a_3x^3)x^k dx = \int_0^1 f(x)x^k dx.$$



Example Problem 3

- Convex Optimization

- Consider a real function $f(x)$ on the real interval $[a, b]$
- A **chord** of f is a line connecting $(x, f(x))$ and $(y, f(y))$
- $f(x)$ is **convex** if all chords lie above the function, i.e. for $\lambda \in [0, 1]$,

$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$$

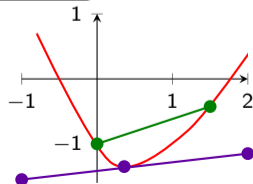
- Convexity implies all local minima have the same **minimum value**

Q: Can we compute a simple lower bound on the minimum value?

- Solution via Tangent Line

- Yes, a convex function also **lies above any tangent line** and

$$\min_{x \in [a, b]} f(x) \geq \min_{x \in [a, b]} [f(x_0) + (x - x_0)f'(x_0)]$$



Course Overview and Philosophy

- First Theme: Standard definitions and vocabulary
 - Engineering has many subfields where new ideas are described using **abstract mathematics**
 - A key goal of this course is to **teach the vocabulary** necessary to read and understand papers and textbooks in these subfields
- Second Theme: Standard theorems and their proof
 - Our focus is on **mathematical foundations** of engineering analysis
 - Proofs **illustrate mathematical reasoning** and show why results are true
 - Emphasis is on general techniques that **extend to advanced problems**
- Assignments
 - Written homework **builds knowledge** of definitions and techniques
 - Computer assignments **demonstrate practical applications**
 - These are like training for a sport; each workout provides a small gain but **consistent effort leads to significant improvement** over time

- Course Elements
 - Syllabus: Lists lecture/video topics by date along with HW
 - Course Notes: Detailed description of the course content
 - Flip Videos: Slide-based introduction to course notes 😊
 - Lectures: Q/A about videos/notes + group work on example problems
- My Recommendations
 - Watch the flip video first and then read the matching course notes
 - Afterward, try the homework while referring to the course notes as needed
 - If you're stuck >30min on a question, ask for a hint (via e-mail or web)
 - Visit office hours to get additional help as needed
 - Make sure to watch the video before the class meeting!