THE UNIVERSITY of TENNESSEE LIKNOXVILLE

MATHEMATICAL, NUMERICAL, BAYESIAN, AND CAUSAL PROBLEM SOLVING

MSE 494/MSE510

Instructor: Sergei V. Kalinin

TA: Sheryl Sanches

Times and locations: 9:45-11:00

TR, Ferris Hall 510



Symbols, numbers, data

- Numbers
- Patterns
- Magnitude
- Shapes and forms
- Symbols
- Algorithms
- Logic
- Correlations
- Prior Knowledge
- Causality
- Workflows



Materials science



Chemistry

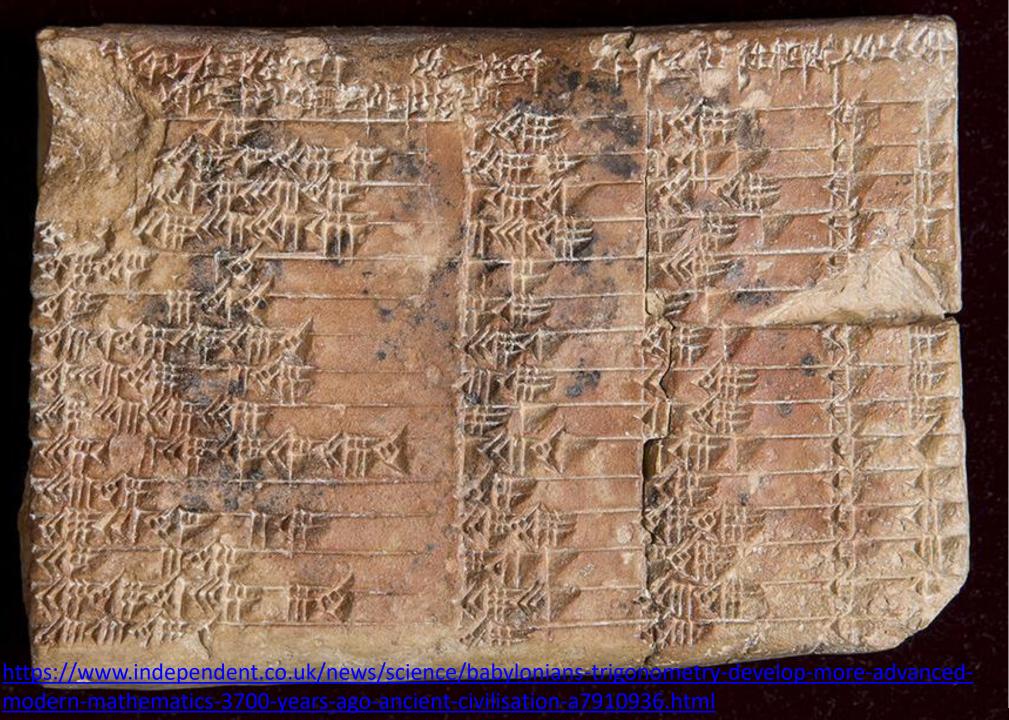


Physics



Chemical engineering





- Taxes:
 - Grain allotments
 - Workers
 - Weights of silver
- Geometry
- Pythagoras theorem
- Linear, quadratic, and cubic equations
- Hexadecimal system





India and China



The Tsinghua Bamboo Slips, containing the world's earliest decimal multiplicati on table, 305 BC during the Warring States period



The Nine Chapters on the Mathematical Art, one of the earliest surviving mathematical texts from China (2nd century AD)

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TABLE SHOWING THE PROGRESS OF NUMBER FORMS
                 IN INDIA
                   9 10 20 30 40 50 00 76 80 90 100 200 1000
                                         c. 250 BCE
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                                    AIN C. 50 BCE
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                                         c. 150 CE
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https://en.wikipedia.org/wiki/History of mathematics



- Algebra
- Algorithms
- Early calculus
- Trigonometry
- Cryptography
- Frequency analysis



Medieval Europe

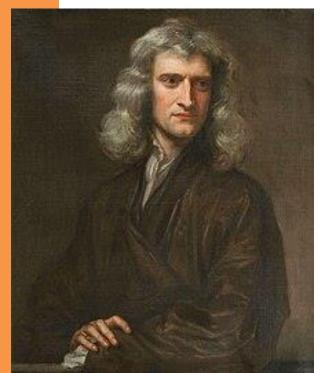
Primary drivers:

- Manufacturing
- Taxes
- Philosophy
- Theology

https://www.discovermiddle ages.co.uk/medievallife/medieval-life-and-times



Renaissance



Isaac Newton (1642 –1726)



Leonhard Euler (1707 – 1783)



Friedrich Gauss (1777 –1855)



Joseph Fourier (1768 –1830)

https://en.wikipedia.org/wiki/Isaac_Newton
https://en.wikipedia.org/wiki/Leonhard_Euler
https://en.wikipedia.org/wiki/Carl_Friedrich_Gauss
https://en.wikipedia.org/wiki/Joseph_Fourier

- Mechanics
- Electromagnetism
- Heat conduction

From Renaissance to XX Century





3. 21 cm-Bronzemörser in Belagerungslaffete mit Schießrädern.

alamy

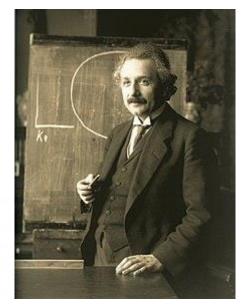
Image ID: BJW8AE ww.alamy.com Gödel proved that the world of pure mathematics is inexhaustible; no finite set of axioms and rules of inference can ever encompass the whole of mathematics; given any finite set of axioms, we can find meaningful mathematical questions which the axioms leave unanswered. I hope that an analogous Situation exists in the physical world. If my view of the future is correct, it means that the world of physics and astronomy is also inexhaustible; no matter how far we go into the future, there will always be new things happening, new information coming in, new worlds to explore, a constantly expanding domain of life, consciousness, and memory.

Freeman Dyson

From Lecture 1, 'Philosophy', in a series of four James Arthur Lectures, 'Lectures on Time and its Mysteries' at New York University (Autumn 1978).

Printed in 'Time Without End: Physics and Biology in an Open Universe', Reviews of Modern Physics (Jul 1979), **51**, 449.

The Quantum Era







Bohr



Noether



Schwinger



Feynman

- Quantum mechanics
- Relativity
- Quantum electrodynamics
- Discovery of spin
- ... and so on

For a physicist mathematics is not just a tool by means of which phenomena can be calculated, it is the main source of concepts and principles by means of which new theories can be created.

Freeman Dyson

In 'Mathematics in the Physical Sciences', *Scientific American* (Sep 1964), **211**, No. 3, 129.

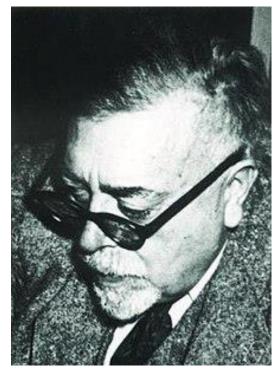
Science and Engineering of Big Numbers



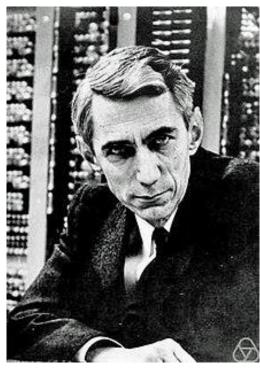
John von Neumann (1903 – 1957)



Eugene Wigner (1902 – 1995)



Norbert Wiener (1894 – 1964)



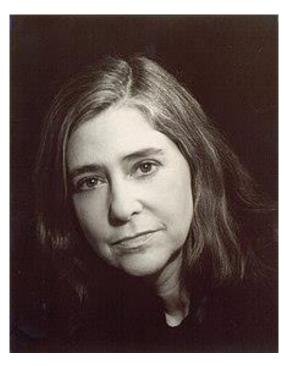
Claude Shannon (1916 – 2001)

https://en.wikipedia.org/wiki/John von Neumann https://en.wikipedia.org/wiki/Eugene Wigner https://en.wikipedia.org/wiki/Norbert Wiener https://en.wikipedia.org/wiki/Claude Shannon

Science and Engineering of Big Numbers



Grace Hopper (1906 –1992)



Margaret Hamilton (1936 -)

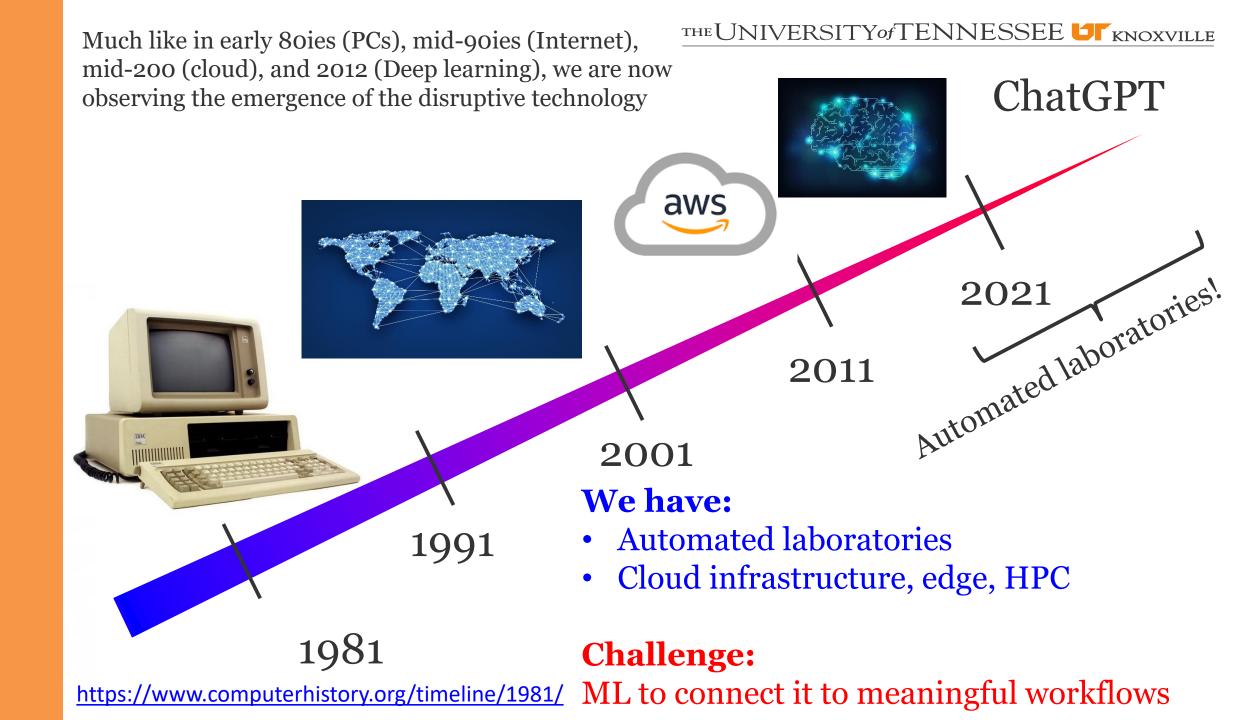


Katherine Johnson (1918 –2020)

https://en.wikipedia.org/wiki/Grace Hopper

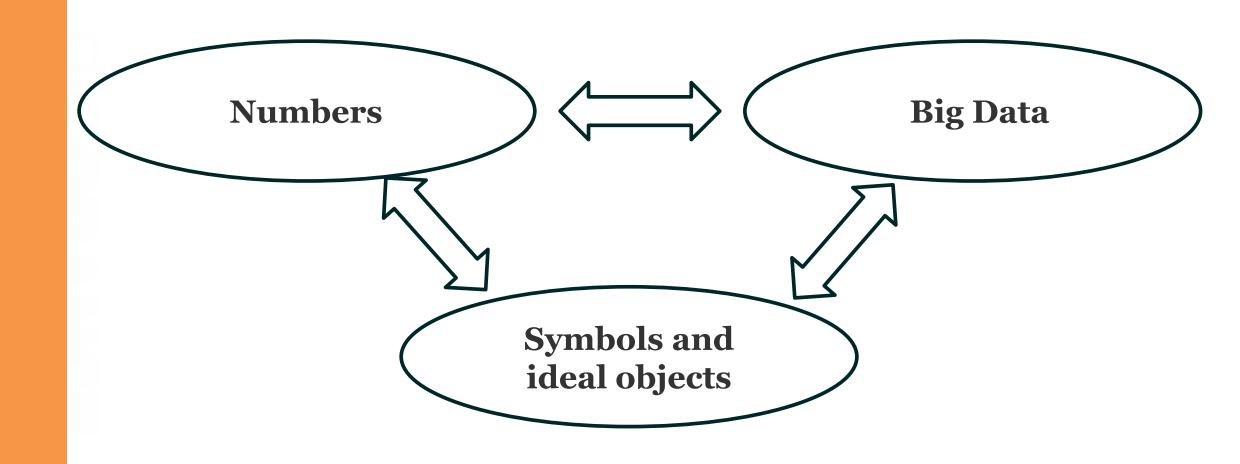
https://en.wikipedia.org/wiki/Margaret Hamilton (software engineer)

https://en.wikipedia.org/wiki/Katherine Johnson



New addition: big data

We are now entering the new era of science when in addition to classical mathematics and physics tools we have new intuition and methods of big data



Course Information

Faculty Contact Information:

Instructor: Prof. Sergei V. Kalinin,

Office: 314 IAMM

E-mail: sergei2@utk.edu

Teaching Assistant: Sheryl Sanchez, <u>ssanch18@vols.utk.edu</u>

Instructor Availability:

Please don't hesitate to email me with updates, questions, or concerns. I will typically respond within 24 hours during the week and 48 hours on the weekend. I will notify you if I will be out of town and if connection issues may delay a response.

Meeting Time: 9:45-11:00 TR, Ferris Hall 510

The lectures and materials will be posted on Canvas and at GitHub: https://github.com/SergeiVKalinin/MSE Spring2024

Office Hours:

Friday 1:30 - 3:00 PM are open for 1:1 meetings to discuss any course related item. Can also be made by scheduling via email.

Course Outline

- 1. Mathematical methods in physical science
- 2. Introduction into Python and SciPy ecosystems
- 3. Classical numerical methods with Python
- 4. Regression methods (linear, functional fits, symbolic)
- 5. Bayesian data analysis
- 6. Causal methods
- 7. Gaussian Processes, Bayesian Optimization, and active learning
- 8. Gaussian processes meet physics

Value Proposition

- 1. Achieve proficiency in scientific Python, tapping into its diverse mathematical applications.
- 2. Gain the ability to solve both foundational and advanced equations with confidence.
- 3. Deep dive into data analysis, mastering tools that empower data-driven decisions.
- 4. Grasp the nuances of Bayesian methods, learning how to weave together data with prior knowledge.
- 5. Explore the intricate landscape of causal analysis within the ML spectrum.
- 6. Develop a solid understanding of probabilistic techniques for decision-making in uncertain scenarios.

Prerequisites

To be successful in this course you will need a general background in materials science. Python or similar programming experience, while not essential, will be extremely useful. Students without any prior programming experience should expect to spend extra time outside of class learning basic skills.

Outcomes

- 1. This course aims to provide students with the skills needed to link physics, numerical methods, and big data
- 2. Students should learn how to combine intuition from mathematics, physics, and machine learning methods
- 3. The course is designed to provide students with basic knowledge of numerical methods, causal analysis, and Bayesian methods
- 4. Preparedness for the Future of Science and Industry: With insights into automated labs, large language models in scientific workflows, and federated tools and workflows, students will be prepared for the future of industry. These skills are increasingly important as companies automate processes and incorporate AI into their workflows. This knowledge can help students stand out in the job market and be prepared for the careers of the future.

This and that

Learning Environment:

The class will be delivered as in-person lectures. The Jupyter notebooks, code libraries, and videos provided. Weekly programming exercises will be assigned via Google Colabs and those students wishing to interact with the instructor in person should attend office hours.

Use of ChatGPT:

Strongly encouraged both for programming and written assignments. However, the students have to be aware of the limitations of the generative models.

Grading & Policies:

•	Midterms (2)	30%
•	Homeworks	45%
•	Final Project	25%

Reference Materials

I will provide copies of lecture notes, presentations, and Colabs on GitHub and Canvas. There is no specific textbook for the course, and we will take material from a variety of sources including:

- Andrew Bird et al, Python Workshop, https://www.packtpub.com/product/the-python-workshop/9781839218859
- Oswaldo Martin, Bayesian Analysis with Python Second Edition, https://subscription.packtpub.com/book/data/9781789341652/
- Alexander Molak, Causal Inference and Discovery in Python, https://subscription.packtpub.com/book/data/9781804612989/

Homework 1:

- Create new Colab, https://colab.google/
- Chapter 1-4 and 10, Python Workshop.

Homework, midterm, and finals format

- All homeworks, midterms, and finals will be in the Google Colab format
- Use the code for programming exercises and markdown fields for text responses
- Share in the "comment" or "editor" modes
- The Colabs should save all graph outputs
- The Colabs should be able to run from the beginning to end (e.g. if I restart the runtime and run all)
- Submit to sergei2vk@gmail.com

Homework 1:

- Create new Colab, https://colab.google/
- Chapter 1-4 and 10, Python Workshop.

Welcome aboard!