

CS 7800/4810:  
Data Str & Alg for Scalable Computing  
Spring 2026

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no  
smartphones



no  
laptop



**Why?**

there is enough evidence that laptops and phones slow you down

# Ask questions



... and answer my questions.

Our main **goal** is to have **interesting discussions** that will help to gradually understand the material

**(it's ok if not everything is clear, as long as you have questions!)**

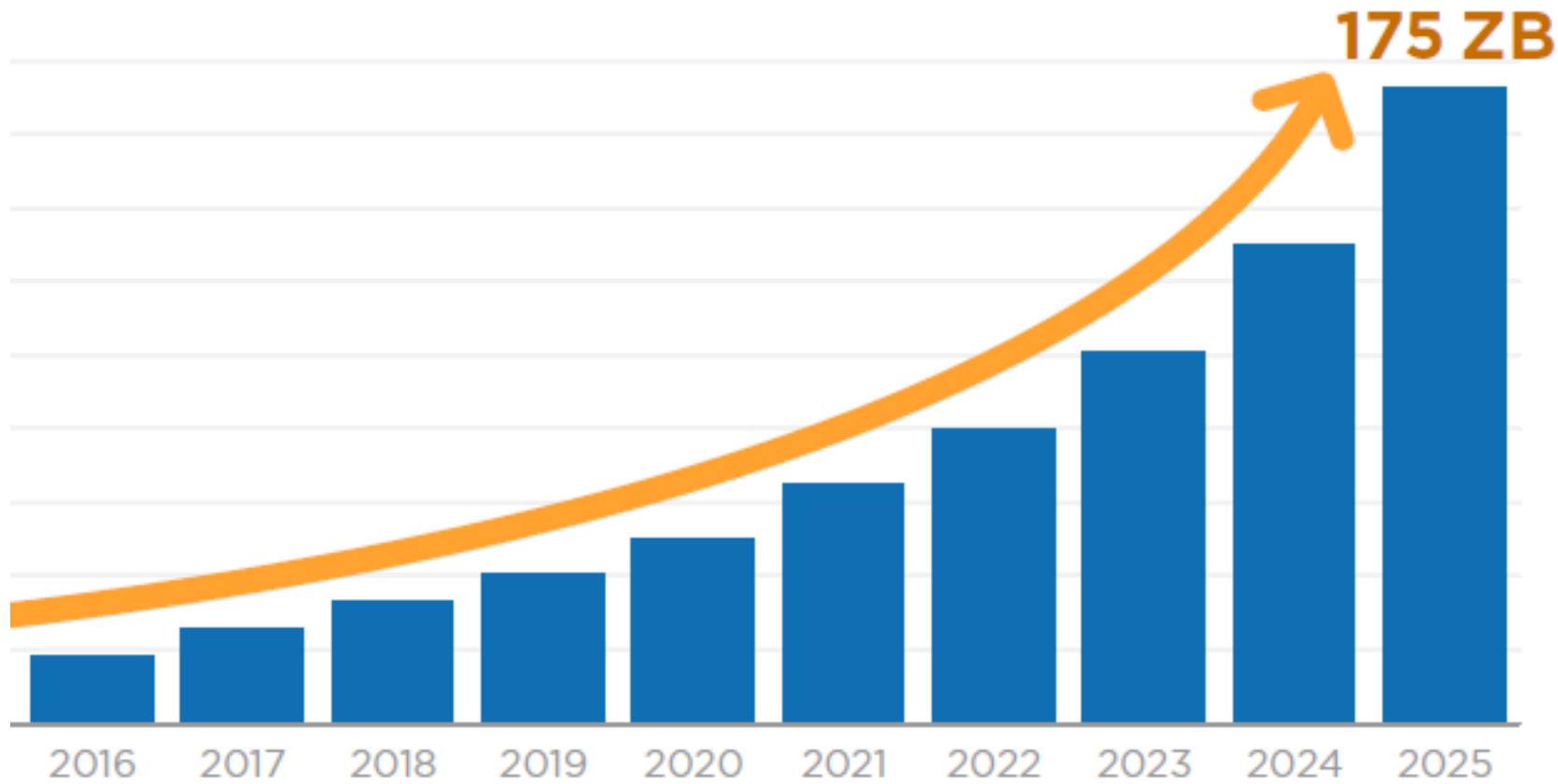
# Today's agenda

- Course logistics overview
- Why scalable computing?



I want you to speak up!  
[and you can always interrupt me]

# Modern data challenges



IDC says 175 ZB will be created by 2025 (image courtesy IDC)

# Data is the new oil!



Courtesy: <https://www.economist.com/>

David Parkins

**But oil has to be refined and extracted to be usable**

**Our job is to develop refinement machinery to extract *information* from data!**

# Course objectives

- Learn about advanced data structures and algorithms to solve massive-scale data processing/analysis problems.
- Next-generation challenges in data systems.
- Students will become proficient in:
  - Advanced data structures and algorithms
  - Implementing high-performance data structures & algorithms
  - Modern hashing and approximation for machine learning applications
  - Building/analyzing scalable algorithms (disk-based & distributed)

# Course topics

- Compact trees
- Succinct data structure
- Hashing/Hash tables
- Filters and sketches
- Cardinality estimation
- Locality sensitive hashing
- Approximate neighbor search (Vector databases)
- External memory algorithms
- Distributed hash tables

# Background

- I assume you have already taken undergrad/grad Data Str & Alg course (e.g., CS 3000 and 5800) or similar.
- You are comfortable with basic data structures and algorithms and writing C/C++ code (not a hard constraint).
- We will discuss modern variations to classical data structures and algorithms that are designed for massive-scale data.
- Things that we will not cover:  
Basic data structures, algorithms, asymptotic analysis, recursion.

# Course logistics

- Course policies + Schedule

Refer to canvas

- Course website

<https://khoury.northeastern.edu/home/pandey/courses/cs7800/spring26/index.html>

- Academic honesty

- Refer to [Northeastern Academic Integrity Policy](#).
- If you are not sure, ask me.
- I am serious. DO NO PLAGIARISE.

# What is plagiarism

- Listening while someone dictates a solution.
- Basing your solution on any other written solution.
- Copying another student's code or sharing your code with any other student.
- Searching for solution online (e.g., stack overflow, Github, ChatGPT).

# What is collaboration

- Asking questions on Piazza.
- Working together to find a good approach for solving a problem.
  - Students with similar understanding of the material.
- A high-level discussion of solution strategy.
- If you collaborate with other students, declare it upfront

# Instructor office hours

- Before class in my office
  - Mon Wed 1:30 PM – 2:30 PM
  - WVH 478
- Things that we can talk about:
  - Issues on projects
  - Paper clarification/discussions
  - Getting involved in a research project
  - Help with your research

# Teaching assistant

- TA: Yuvaraj Chesetti
  - Office hours: ---
  - 3<sup>rd</sup> year PhD student
- **Research on:**
  - Hash tables
  - Adaptive filters
  - Learned indexes
- **Interests**
  - Music, Sports, Video games



# Instructor

- Research:
  - Large-scale data systems
  - Computational biology
  - Graph processing
  - GPUs
- Previous:
  - Research Scientist, VMware Research
  - Postdoc: CMU/UC Berkeley
- Interests:
  - Outdoors (Running/Hiking/Biking /Skiing /Swimming/...)
  - Sports (Cricket/Soccer/Racket sports)



Val de Gardena Dolomites Italy

What's the longest hike you have finished?

# Course rubric

- Programming assignment
- Project
- Final exam
- Class participation and scribe

# Scribing lectures

- Use the **latex template** to scribe
- Each student may have to scribe 1-2 lectures, depending on class size.
- Pick a date and send an email to the TA. First-come first-served.
- Submit scribe notes (pdf + source).
- Scribe notes are due **by 9pm on the day after lecture**.

# Assignments

- Assignment will include a combination of:
  - Small programming tasks
  - Benchmarking and writing report
- Do all development on your local machine.
  - Can also use Khoury machines/Explorer cluster
- Do all benchmarking using Khoury machines/Explorer cluster

# Project

- Each group (3 people) will choose a project that is:
  - Relevant to the materials discussed in class.
  - Requires a significant theory/programming effort from all team members.
  - Unique (i.e., two groups cannot pick same idea).
  - Approved by me.
- We will provide sample project topics.
- The project will have two milestones.

# Assignments/Projects

- The assignment will be done individually
- The project will be done in a groups of 2 to 3 students
  - You should form groups based on talking to other students
  - Otherwise, we will form groups randomly

# Plagiarism warning

- These projects must be all of your own code.
- You may **not** copy source code from other groups or the web.
- Plagiarism will **not** be tolerated.  
See [Northeastern Academic Integrity Policy](#) for additional information.

# Grade breakdown

- Assignment 20%
- Final project 40%
- Class participation 20%
- Final (usually take home) 20%

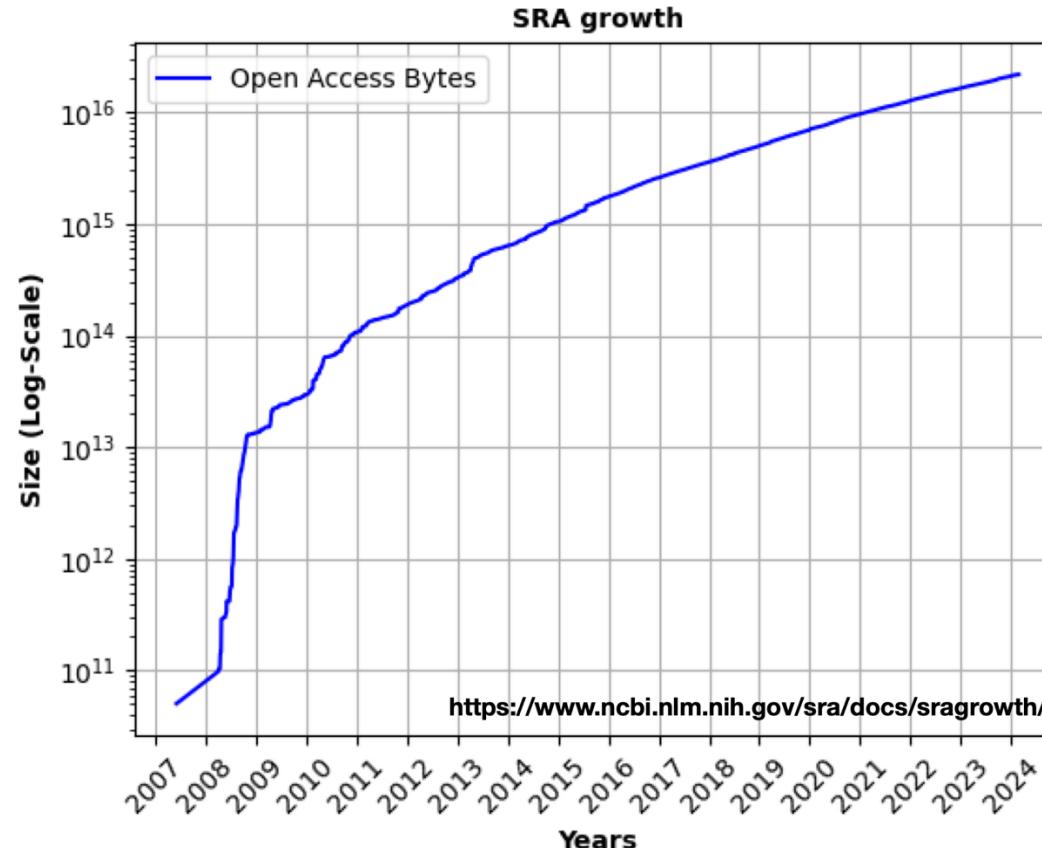
# Course mailing list

- Online discussion through Canvas
- If you have a technical question about the projects, please use Canvas
  - Don't email me or TAs directly

All non-assignment/non-project questions should be sent to me.

# Sequence read archive (SRA) search

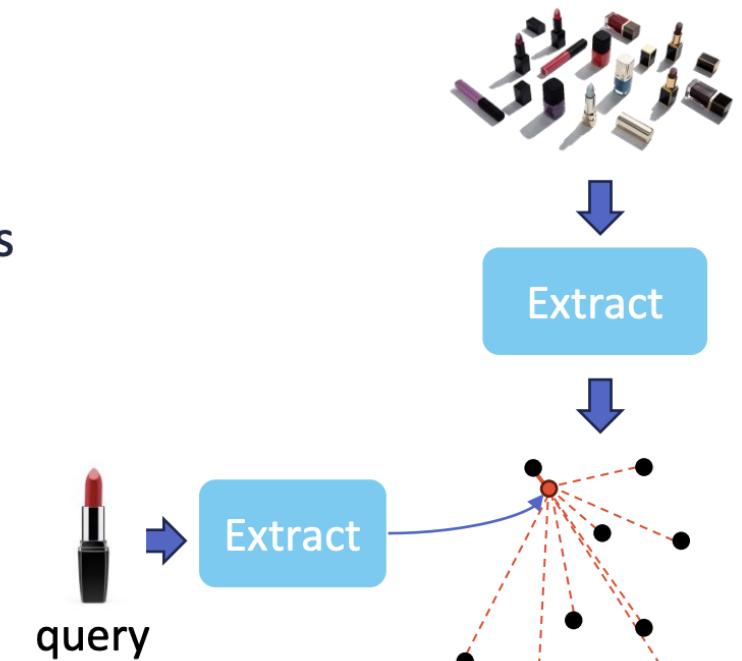
SRA contains a lot of diversity information



What if I find, e.g., a new disease-related gene, and want to see if it appeared in other experiments?

# Visual product search

- Take photo → find matching product
- General idea:
  - Extract feature vectors (**embeddings**) from product images
  - Store in some DB
  - At runtime: extract image features
  - ... then find nearest neighbours
- Example: JD.com [Li, Middleware'18]
  - **100B** products, **1B** daily updates
  - Requirement: support fast update
  - Requirement: query fresh data



# Problems

- $N = 100B = 100,000,000,000$  vectors
- Each vector is large:  $D = \sim 1,000$  floats
- **Problem 1:** storing  $N$  vectors for fast access: **400 terabytes**
  - Too much for RAM
- **Problem 2:** finding nearest neighbour:
  - Distance to  $N$  vectors =  $O(ND)$  multiply-adds  $\rightarrow N*D = 100T$
  - Even at 20 TFLOPS, **5 second latency** per query (ignoring other costs)
- “Put it in a database and index?”
  - Index what? DB indices designed for individual attributes, not ANN search on vectors
  - Not clear how to shard vectors

# Next lecture

- Compact trees